UNIMODAL POLYNOMIALS AND LATTICE WALK ENUMERATION WITH EXPERIMENTAL MATHEMATICS

by

BRYAN EK

A dissertation submitted to the School of Graduate Studies Rutgers, The State University of New Jersey in partial fulfillment of the requirements for the degree of Doctor of Philosophy Graduate Program in Mathematics Written under the direction of Doron Zeilberger and approved by

New Brunswick, New Jersey May, 2018

ABSTRACT OF THE DISSERTATION

Unimodal Polynomials and Lattice Walk Enumeration with Experimental Mathematics

By BRYAN EK Dissertation Director: Doron Zeilberger

The main theme of this dissertation is retooling methods to work for different situations.

I have taken the method derived by O'Hara and simplified by Zeilberger to prove unimodality of q-binomials and tweaked it. This allows us to create many more families of polynomials for which unimodality is not, a priori, given. I analyze how many of the tweaks affect the resulting polynomial.

Ayyer and Zeilberger proved a result about bounded lattice walks. I employ their generating function relation technique to analyze lattice walks with a general step set in bounded, semi-bounded, and unbounded planes. The method in which we do this is formulated to be highly algorithmic so that a computer can automate most, if not all, of the work. I easily recover many well-known results for simpler step sets and discover new results for more complex step sets.

Acknowledgements

I would like to thank my advisor Doron Zeilberger for his direction. He provided several fun projects to work on that resulted in this dissertation. I would also like to thank Cole Franks for his edits and suggestions. And thank you to Michael Saks for his discussion and comments on my work that provided many insights for further analysis. Thank you to my committee for serving as such. Thank you to my friends for the welcome diversions. And a big thank you to my whole family for supporting me along this journey!

This research was funded by a SMART Scholarship: USD/R&E (The Under Secretary of Defense-Research and Engineering), National Defense Education Program (NDEP) / BA-1, Basic Research.

Dedication

Dedicated to my wife Kaitlin.

Table of Contents

\mathbf{A}	bstra	.ct			•••				•	 •		•	•	•••		•	 •	 •			ii
A	cknov	wledge	ments		• •							•	•			•		 •			iii
D	edica	tion .							•		• •	•	•	• •			 •	 •	•		iv
Li	st of	Tables	5						•		• •		•					 •	•		viii
Li	st of	Figure	es								• •		•				 •	 •	•		ix
1.	Intr	oducti	on		•••				•	 •	• •	•	•				 •	 •	•	•	1
2.	Uni	modal	Polyno	omials					•									 •	•		4
	2.1.	Introd	uction .		• •							•	•			•		 •			4
		2.1.1.	Motiva	tion .	•••													 •			4
		2.1.2.	Symme	etric and	d Un	imo	dal											 •			6
		2.1.3.	Partitie	ons								•••									10
		2.1.4.	Chapte	er Orgai	nizat	ion			•									 •			10
	2.2.	Maple	Program	n								•••					 •				11
	2.3.	q-bino	mial Pol	ynomia	ls .																12
	2.4.	Origin	al Recu	rrence																	13
	2.5.	Altere	d Recur	rence .																	19
		2.5.1.	Restric	ted Par	titio	ns .															19
		2.5.2.	Adjust	ed Initi	al Co	ondi	tion	s.													35
		2.5.3.	Adjust																		35
	2.6.	OEIS																			37
	2.7.		ision an																		39

3.	Latt	tice Wa	alk Enumeration	41
	3.1.	Introd	uction	41
		3.1.1.	Motivation	42
		3.1.2.	Definitions	43
		3.1.3.	Chapter Organization	45
	3.2.	Bound	led	47
		3.2.1.	Walking "Anywhere"	47
		3.2.2.	Returning to the <i>x</i> -axis	49
	3.3.	Semi-b	oounded	52
		3.3.1.	Arbitrary Lower Limit	57
		3.3.2.	Meanders	58
	3.4.	Guess-	and-Check Method	60
	3.5.	Algebr	raic to Recursive	62
		3.5.1.	Conversion	62
		3.5.2.	Searching	65
	3.6.	Unbou	nded	69
		3.6.1.	Walking "Anywhere"	69
		3.6.2.	Returning to the <i>x</i> -axis	70
		3.6.3.	Alternative Method	72
		3.6.4.	Selected Step Sets	73
		3.6.5.	Time Comparison	76
	3.7.	Asymp	ptotics	79
		3.7.1.	Discriminant	83
		3.7.2.	Meanders	84
	3.8.	Applic	eations	86
		3.8.1.	Combining Solutions	86
		3.8.2.	Weighted Walks	87
		3.8.3.	2-step Examples	88
	3.9.	Conclu	usion and Future Work	93

4.	Conclusion	•	 •	•••	•	•	•	•	•	•	•	 •	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	96
Re	eferences											 •																					97

List of Tables

2.1.	Unimodal Sequences from the OEIS	38
3.1.	Bounded Walk Enumeration	49
3.2.	Bounded Bridge Enumeration 1000 terms	51
3.3.	500 term Excursion Enumeration	56
3.4.	1000 term Excursion Enumeration	57
3.5.	1000 term Meander Enumeration	60
3.6.	Finding Excursion Minimal Polynomial $\mathcal{S} = \{[1, -2], [1, -1], [1, 0], [1, 1], [1, 2]\}$	61
3.7.	Finding Meander Minimal Polynomial $S = \{[1, -2], [1, -1], [1, 0], [1, 1], [1, 2]\}$	62
3.8.	Enumerating Excursions More Efficiently	63
3.9.	Enumerating Meanders More Efficiently	64
3.10.	Alternative Recurrence Step Set $\{[1,2], [1,-3]\}$	65
3.11.	Alternative Recurrence Step Set $\{[1, -2], [1, -1], [1, 0], [1, 1], [1, 2]\}$	66
3.12.	Alternative Recurrence Step Set $\{[1, -1], [3, -1], [1, 0], [3, 0], [2, 1], [1, 2], [2, 2]\}$	68
3.13.	Finding Minimal Polynomial $\mathcal{S} = \{[1, -2], [1, -1], [1, 0], [1, 1]\}$	76
3.14.	Finding Minimal Polynomial $S = \{[1, -2], [1, -1], [1, 0], [1, 1], [1, 2]\}$	77
3.15.	. 1000 term Unbounded Enumeration	78

List of Figures

2.1.	Contributions from Different Partitions to Produce $G(5,5)$	26
3.1.	Walk Examples	44

Chapter 1 Introduction

Humans have been counting for millennia, whether it was to keep track of how many berry bushes remained to gather from, how many people lived in their village, or how many days it had been since the water froze. A single person could simply count the bushes or people or days; the important jump for society was to communicate that number. One of the earliest methods was tally marks. In ancient Babylon 5,000 years ago, the abacus arose as a method of subtracting and adding numbers in shorthand. A classic tale of Gauss' intellect is how he solved the grade school busy-work problem of adding the numbers from 1 to 100, which can be generalized to adding any sequence of numbers with constant difference.

Zeilberger used recurrences and properties of unimodal polynomials inspired by O'Hara to show that the *q*-binomials are unimodal [Zei89a]. He and Ayyer also used generating function relations to prove a result about polymers bounded between plates [AZ07].

In addition to the important specific results, the previous examples provide something more consequential: the system of obtaining the result. To the cavemen, the amount of food left was probably their top priority, but what they began paved the way for more complex analysis¹ of our world. The methods allow for a solution to not just the original question or task, but with simple tweaks, can answer much more.

For many early questions, particularly Gauss' busywork, we can have a computer bruteforce a result. But while computers are faster, *much* faster, than us they still take nonzero time for even the simplest computations. Over the years, we have created machines that can compute faster and faster. But instead of computing faster, we want to compute smarter.

¹And eventually Complex Analysis.

We could go through the process of solving a system of 1,000 linear equations and variables by hand, but that would take forever. Computers are very good at doing simple things quickly. The important facet is the method and set-up process to allow the computer to do its thing.

This is the methodology of experimental mathematics: to use computers efficiently for mathematical investigations. The more data we gather, the easier it is to draw insights. Computing has come quite a long way for mathematics.

The main theme of this dissertation is retooling methods to work for different situations. As a part of this retooling, we have made two major Maple packages for demonstrations of the various devices.

We begin by delving into unimodal polynomials. The first chapter is all about how we can tweak a single recurrence that proves unimodality for the *q*-binomial coefficients into a source of many more unimodal polynomials. The recurrence provided by Zeilberger [Zei89b] and inspired by O'Hara [O'H90] is given and several adjustments are introduced. How these adjustments affect the final polynomial are explored for most of the chapter. The unimodal polynomial chapter was produced in conjunction with the Maple package **Gnk** available at

$math.rutgers.edu/\sim bte14/Code/Gnk/Gnk.txt.$

The package provides for quick enumeration of families of unimodal polynomials as well as a few functions for insight into the background work.

The second chapter is on the subject of lattice walks. Given a certain set of allowable steps in a 2D lattice (none of which move to the left), how many ways are there to walk between points in the plane? To tackle this question, I use generating function relations motivated by Ayyer and Zeilberger [AZ07]. For walks bounded above and below, we are guaranteed to find an exact expression for the generating function: a rational function. For more general walks, we are only guaranteed to find a polynomial for which the generating function is a root (in terms of formal power series). For some polynomials this means we can give an exact expression but, in general, the *minimal polynomial* is the best we, or anyone, can do. The lattice walk chapter was produced in conjunction with the Maple package **ScoringPaths** available at

$math.rutgers.edu/{\sim}bte14/Code/ScoringPaths/ScoringPaths.txt.$

Within this package are all of the procedures needed to generate the mentioned results automatically. There are several different ways we can enumerate the walks; this chapter discusses them, their pros and cons, and the specific methods of enumeration are included in **ScoringPaths**.

Chapter 2

Unimodal Polynomials

2.1 Introduction

This chapter is available as a stand-alone paper on arXiv.org, number 1711.11252. It has been submitted to the *Journal of Difference Equations and Applications*.

The q-binomial coefficients were conjectured to be unimodal as early as the 1850s, but it remained unproven until Sylvester's 1878 proof using invariant theory. In 1982, Proctor gave an "elementary" proof using linear algebra. Finally, in 1989, Kathy O'Hara provided a *combinatorial* proof of the unimodality of the q-binomial coefficients. Very soon thereafter, Doron Zeilberger translated the argument into an elegant recurrence. We introduce several perturbations to the recurrence to create a larger family of unimodal polynomials. We analyze how these perturbations affect the final polynomial and analyze some specific cases.

2.1.1 Motivation

"The study of unimodality and log-concavity arise often in combinatorics, economics of uncertainty and information, and algebra, and have been the subject of considerable research." [AAR00]

Intuitively, many sequences seem to be unimodal, but how does one prove that fact? We will review some methods of building unimodal sequences as well as a few lemmas that imply unimodality.

Knowing a sequence is unimodal allows for guaranteed discovery of the global extremum using an easy search algorithm. Unimodality is also useful for probability applications. Identifying a probability distribution as unimodal allows certain approximations for how far a value will be from its mode (Gauss' inequality [Gau23]) or mean (Vysochanskij-Petunin inequality [DFV80]).

Suppose we start with a set of nice combinatorial objects that satisfy property \mathcal{P} . How can we make the set larger and still satisfy property \mathcal{P} ? Or a slightly different property \mathcal{P}' ? The goal of this project is to use reverse engineering to obtain highly non-trivial, surprising theorems about unimodality.

Example 2.1. All of the following functions are not only polynomial for $n \ge 0, 0, 0, 4, 4$, respectively, but they are also unimodal.

$$\begin{split} P_{1}(n) &= \frac{1-q^{n+1}}{1-q}, \\ P_{2}(n) &= \frac{1-q^{2n+1}}{1-q}, \\ P_{3}(n) &= \frac{2q^{3n+2}-2q^{3n+1}+q^{3n}-q^{2n+1}-q^{n+1}+q^{2}-2q+2}{(1-q)^{2}}, \\ P_{4}(n) &= \left(5q^{4n+2}-5q^{4n+1}+3q^{4n-2}-3q^{4n-3}+4q^{4n-4}-4q^{2n+3}-4q^{2n-1}\right) \\ &\quad + 4q^{6}-3q^{5}+3q^{4}-5q+5\right) \middle/ (1-q)^{2}, \\ P_{5}(n) &= \left(2q^{5n+3}-4q^{5n+2}+7q^{5n+1}-5q^{5n}+5q^{5n-3}-5q^{5n-4}+4q^{5n-5}-4q^{5n-6}+6q^{5n-9}\right) \\ &\quad - 6q^{5n-10}+3q^{5n-11}-5q^{4n+2}+5q^{4n+1}-4q^{4n-2}+4q^{4n-3}-3q^{4n-4}-5q^{3n+4} \\ &\quad + 5q^{3n+3}-3q^{3n+2}-6q^{3n-4}+6q^{3n-5}-3q^{3n-6}+3q^{2n+9}-6q^{2n+8}+6q^{2n+7} \\ &\quad + 3q^{2n+1}-5q^{2n}+5q^{2n-1}+3q^{n+7}-4q^{n+6}+4q^{n+5}-5q^{n+2}+5q^{n+1} \\ &\quad - 3q^{14}+6q^{13}-6q^{12}+4q^{9}-4q^{8}+5q^{7}-5q^{6}+5q^{3}-7q^{2}+4q-2\right) \middle/ (-1+q)^{3}. \end{split}$$

Example 2.2. The following functions are all unimodal for $n \ge 0$.

$$\begin{split} Q_1(n) &= \frac{4\left(1-q^{n+1}\right)}{1-q}, \\ Q_2(n) &= 5 \frac{8q^{2n+4}-2q^{n+3}-2q^{n+2}+8q-3[(q^5+1)(q^n-(-q)^n)+(q^4+q)(q^n+(-q)^n)]}{2q\left(1-q\right)\left(1-q^2\right)}, \\ Q_3(n) &= \left(16\left[3-3q+16q^2-16q^{n+1}-16q^{n+3}-16q^{n+5}+16q^{2n+3}\right. \\ &\quad + 16q^{2n+5}+16q^{2n+7}-16q^{3n+6}+3q^{3n+7}-3q^{3n+8}\right] \\ &\quad + (1-(-1)^n)(-q)^{(3n-9)/2}\left[8q^{17}-8q^{16}+64q^{15}-9q^{14}-55q^{13}-60q^{11}+57q^{10}\right. \\ &\quad - 125q^9+125q^8-57q^7+60q^6+55q^4+9q^3-64q^2+8q-8\right] \\ &\quad + (1-(-1)^n)q^{(3n-9)/2}\left[8q^{17}-8q^{16}+64q^{15}+9q^{14}-73q^{13}-60q^{11}-65q^{10}\right. \\ &\quad - 3q^9+3q^8+65q^7+60q^6+73q^4-9q^3-64q^2+8q-8\right] \\ &\quad + (1+(-1)^n)(-q)^{(3n-6)/2}\left[12q^{14}-12q^{13}+64q^{12}-75q^{11}+74q^{10}-127q^9\right. \\ &\quad + 127q^5-74q^4+75q^3-64q^2+12q-12\right] \\ &\quad + (1+(-1)^n)q^{(3n-6)/2}\left[12q^{14}-12q^{13}+64q^{12}-53q^{11}-74q^{10}-q^9\right. \\ &\quad + q^5+74q^4+53q^3-64q^2+12q-12\right] \right) \Big/ 4(1-q)^2(1-q^6). \end{split}$$

 Q_3 is truly amazing. It appears quite unwieldy and at first glance one might doubt that it has real coefficients. But for any nonnegative integer n, $Q_3(n)$ is guaranteed to be a unimodal polynomial in $\mathbb{Z}[q]$. Try simplifying Q_3 assuming n is even/odd. One can obtain further simplifications assuming $n = 0, 1, 2, 3 \pmod{4}$.

2.1.2 Symmetric and Unimodal

We recall several definitions and propositions from Zeilberger [Zei89a] for the sake of completeness.

Definition 2.3 (Unimodal). A sequence $\mathcal{A} = \{a_0, \ldots, a_n\}$ is unimodal if it is weakly increasing up to a point and then weakly decreasing, i.e., there exists an index i such that $a_0 \leq a_1 \leq \cdots \leq a_i \geq \cdots \geq a_n$.

Definition 2.4 (Symmetric). A sequence $\mathcal{A} = \{a_0, \ldots, a_n\}$ is symmetric if $a_i = a_{n-i}$ for every $0 \le i \le n$.

A polynomial is said to have either of the above properties if its sequence of coefficients has the respective property.

Definition 2.5 (Darga). The darga of a polynomial $p(q) = a_i q^i + \cdots + a_j q^j$, with $a_i \neq 0 \neq a_j$, is defined to be i + j, i.e., the sum of its lowest and highest powers.

darga is equal to $2 \cdot C(p)$ used by Brent [BB90].

Example 2.6. $darga(q^2 + 3q^3) = 5$ and $darga(q^2) = 4$.

Proposition 2.7. The sum of two symmetric and unimodal polynomials of darga m is also symmetric and unimodal of darga m.

Proposition 2.8. The product of two symmetric and unimodal nonnegative¹ polynomials of darga m and m' is a symmetric and unimodal polynomial of darga m + m'.

Proof. A polynomial is symmetric and unimodal of darga m if and only if it can be expressed as a sum of "atomic" entities of the form $c(q^{m-r} + q^{m-r-1} + \cdots + q^r)$, for some **positive** constant c and integer $0 \le r \le \frac{m}{2}$. By Proposition 2.7, it is enough to prove that the product of two such atoms of dargas m and m' is symmetric and unimodal of darga m + m':

$$(q^{m-r} + \dots + q^r)(q^{m'-r'} + \dots + q^{r'}) = q^{m+m'-r-r'} + 2q^{m+m'-r-r'-1} + \dots + 2q^{r+r'+1} + q^{r+r'}.$$

Proposition 2.9. If p is symmetric and unimodal of darga m, then $q^{\alpha}p$ is symmetric and unimodal of darga $m + 2\alpha$.

One example is the binomial polynomial: $(1 + x)^m$. It is symmetric and unimodal of darga m.

Definition 2.10 (γ -nonnegative [Brä15]). If h(x) is a symmetric function (in its coefficients), then we can write $h(x) = \sum_{k=0}^{\lfloor m/2 \rfloor} \gamma_k x^k (1+x)^{m-2k}$. We call $\{\gamma_k\}_{k=0}^{\lfloor m/2 \rfloor}$ the γ -vector of h. If the γ -vector is nonnegative, h is said to be γ -nonnegative.

¹Nonnegative is necessary: $(-1 + x - x^2)^2 = 1 - 2x + 3x^2 - 2x^3 + x^4$.

One can use Propositions 2.7, 2.8, and 2.9 to prove that γ -nonnegative implies symmetric and unimodal of darga m.

Example 2.11. To prove the polynomials in Example 2.1 are actually unimodal, one can show

$$\begin{split} P_1(n) &= \sum_{i=0}^n q^i, \\ P_2(n) &= P_1(2n), \\ P_3(n) &= 2P_1(3n) + q^2 P_1(2n-2) P_1(n-2), \\ P_4(n) &= 5P_1(4n) + 3q^4 P_2(2n-4) + 4q^6 P_1(2n-4) P_2(n-4), \\ P_5(n) &= 2P_1(5n) + 5q^2 P_1(4n-2) P_1(n-2) + 4q^8 P_2(2n-6) P_1(n-4) \\ &+ 5q^6 P_1(3n-4) P_2(n-4) + 3q^{12} P_1(2n-6) P_3(n-6), \end{split}$$

and then use Propositions 2.7, 2.8, and 2.9 along with the property that $darga(P_k(n)) = nk$. One must initially assume $n \ge 0, 0, 2, 4, 6$, respectively, but the k = 3, 5 bounds can be lowered by checking smaller values of n manually.

Example 2.12. To prove the polynomials in Example 2.2 are in fact unimodal, one can show

$$Q_1(n) = 4 \sum_{i=0}^{n} q^i,$$

$$Q_2(n) = q^2 Q_2 (n-2) + 5Q_1 (2n),$$

$$Q_3(n) = q^6 Q_3 (n-4) + 3Q_1 (3n) + 4q^2 Q_1 (2n-2) Q_1 (n-2),$$

and then use induction on n for Q_2, Q_3 , with base cases $n = \{0, 1\}, \{0, 1, 2, 3\}$, respectively, and Propositions 2.7, 2.8, and 2.9 along with the property that $darga(Q_k(n)) = nk$.

The wonderful aspect about these proofs is that they can be easily verified by computer!

Being able to write a function in "decomposed" form allows us to quickly verify the unimodal nature of each part and hence the sum. However, the combined form generally does not lead to an obvious decomposition. This is the motivation behind using reverse engineering: to guarantee that the decomposed form exists.

Real-rootedness and Log-concavity

The following properties are greatly related to unimodality but turn out to not be applicable in our situation. They are included for completeness of discussion.

Definition 2.13 (Real-rootedness). The generating polynomial, $p_{\mathcal{A}}(x) := a_0 + a_1q + \dots + a_nq^n$, is called real-rooted if all its zeros are real. By convention, constant polynomials are considered to be real-rooted.

Definition 2.14 (Log-concavity). A sequence $\mathcal{A} = \{a_0, \ldots, a_n\}$ is log-concave (convex) if $a_i^2 \ge a_{i-1}a_{i+1}$ ($a_i^2 \le a_{i-1}a_{i+1}$) for all $1 \le i < n$.

There is also a notion of k-fold log-concave and infinitely log-concave with open problems that may be of interest to the reader [Brä15].

Proposition 2.15. The Hadamard (term-wise) product of log-concave (convex) sequences is also log-concave (convex).

Lemma 2.16 (Brändén [Brä15]). Let $\mathcal{A} = \{a_k\}_{k=0}^n$ be a finite sequence of nonnegative numbers.

- If $p_{\mathcal{A}}(x)$ is real-rooted, then the sequence $\mathcal{A}' := \{a_k / \binom{n}{k}\}_{k=0}^n$ is log-concave.
- If \mathcal{A}' is log-concave, then so is \mathcal{A} .
- If \mathcal{A} is log-concave and positive,² then \mathcal{A} is unimodal.

For a self-contained proof using less general (and possibly easier to understand) results, see Lecture 1 from Vatter's Algebraic Combinatorics class [Vat09], which uses the book A Walk Through Combinatorics [Bón06]. The converse statements of Lemma 2.16 are false. There are log-concave polynomials that are not real-rooted and there are unimodal polynomials which are not log-concave.

Brändén also references a proof by Stanley [Sta89] that:

Lemma 2.17. If A(x), B(x) are log-concave, then A(x)B(x) is log-concave. And if A(x) is log-concave and B(x) is unimodal, then A(x)B(x) is unimodal.

 $^{^{2}}$ Nonnegative is not sufficient: $\{1, 0, 0, 1\}$ is log-concave (and log-convex) but not unimodal.

It is not sufficient for A, B to be unimodal: $(3 + q + q^2)^2 = 9 + 6q + 7q^2 + 2q^3 + q^4$.³

2.1.3 Partitions

Definition 2.18 (Partition). A partition of k is a non-increasing sequence of positive integers $\lambda = [a_1, a_2, \dots, a_s]$ s.t. $\sum_{i=1}^s a_i = k$.

We will use frequency representation $\lambda = [a_1^{b_1}, a_2^{b_2}, \dots, a_r^{b_r}]$ s.t. $a_i > a_{i+1}$ and $b_i > 0$ to abbreviate repeated terms [And98].⁴ Note now that $\sum_{i=1}^r a_i b_i = k$. There is possible ambiguity as to whether x^y is x repeated y times, or x^y counted once. We always reserve exponentials in partitions to indicate repetition.

The size of a partition, denoted $|\lambda|$, will indicate the **number** of parts.⁵ In standard notation $|\lambda| = s$; in frequency notation $|\lambda| = \sum_{i=1}^{r} b_i$.

 d_i is the number of parts of size *i* in the partition. In frequency notation, $d_{a_i} = b_i$.

 $\lambda \vdash k$ indicates λ is a partition of k.

We will use λ to denote a partition. Unless otherwise stated, λ is a partition of k.

2.1.4 Chapter Organization

This chapter is organized in the following sections:

- Section 2: Maple Program: Briefly describes the accompanying Maple package.
- Section 3: q-binomial Polynomials: Describes the original interesting polynomials.
- Section 4: Original Recurrence: Introduces the recurrence that generates the q-binomial polynomials and simultaneously proves their unimodality. We briefly discuss properties of the recurrence itself.
- Section 5: Altered Recurrence: Modifications are injected into the recurrence that still maintain unimodality for the resulting polynomials. We examine the effects

³Stanley [Sta89] incorrectly writes the coefficient of q^3 as 1.

⁴ If $b_i = 1$, it is omitted.

⁵Many papers use $|\lambda| = k$ to denote what λ partitions.

various changes have. This section contains most of the reference to new unimodal polynomials.

Section 6: OEIS: Uses some of the created unimodal polynomials to enumerate sequences for the OEIS.⁶

Section 7: Conclusion and Future Work: Provides avenues for future research.

2.2 Maple Program

The backbone of this paper is based on experimental work with the Maple package **Gnk** available at

math.rutgers.edu/~bte14/Code/Gnk/Gnk.txt.

I will mention how functions are implemented throughout this paper using IMPLEMENTED As For help with any function, type **Help(function)**.

The most important function is **KOHgeneral**. It modifies the original recurrence in Eqn. (2.3) in several different ways. Using this function, one can change the recurrence call, multiply the summand by any manner of constant, restrict the types of partitions, or even hand-pick which partitions should be weighted most highly. The key part of these changes is that the solution to the recurrence remains symmetric and unimodal of darga nk.

math.rutgers.edu/~bte14/Code/Gnk/Polynomials/ contains examples of many non-trivial one-variable rational functions that are guaranteed to be unimodal polynomials (for $n \ge n_0$). They were created using the **KOHrecurse** function and restricting the partitions summed over in (2.3) to have smallest part $\ge 1, 2, 3, 4$ and distance $\ge 1, 2, 3, 4$. In decomposed form, they are clearly unimodal; but in combined form one is hard-pressed to state unimodality with certainty.

The "impressive" polynomials in Examples 2.1 and 2.2 were generated by **RandomThe-oremAndProof**. It uses the recurrence in Eqn. (2.3) with random multiplicative constants in each term. The parameters were (n, q, 5, 5, 0, false) and (n, q, 3, 5, 0, true), respectively.

⁶Online Encyclopedia of Integer Sequences [Nei].

Initially, $Q_3(n)$ was an even larger behemoth. This will typically happen when Maple solves a recurrence equation so going beyond K = 3 and *complicated* = *true* is not recommended unless one wants to spend substantial time parsing the polynomial into something much more readable. Or better yet, one could come up with a way for Maple to do this parsing automatically.

There are functions to test whether a polynomial is symmetric, unimodal, or both (isSymmetric(P,q), isUnimodal(P,q), isSymUni(P,q), respectively). As an extra take-away, generalPartitions outputs all partitions restricted to a minimum/maximum integer/size as well as difference⁷ and congruence requirements. One can also specify a finite set of integers which are allowed in genPartitions.

2.3 *q*-binomial Polynomials

Of particular interest among unimodal polynomials are the q-binomial polynomials $\begin{bmatrix}n\\k\end{bmatrix}_q$ IMPLEMENTED AS **qbin(n,k,q)**. The q-binomials are analogs of the binomial coefficients. In the limit $q \to 1^-$, we obtain the usual binomial coefficients. We parametrize q-binomial coefficients in order to eliminate the redundancy: $\begin{bmatrix}n\\k\end{bmatrix}_q = \begin{bmatrix}n\\n-k\end{bmatrix}_q$:

$$G(n,k) = G(k,n) = {\binom{n+k}{k}}_q = \frac{[n+k]_q!}{[n]_q![k]_q!} = \frac{(1-q^{n+1})\cdots(1-q^{n+k})}{(1-q)\cdots(1-q^k)},$$
(2.1)

IMPLEMENTED AS $\mathbf{Gnk}(\mathbf{n},\mathbf{k},\mathbf{q})$. $[n]_q = \frac{1-q^n}{1-q}$ is the q-bracket. $[n]_q! = \prod_{i=1}^n [i]_q$ denotes the q-factorial⁸ IMPLEMENTED AS $\mathbf{qfac}(\mathbf{n},\mathbf{q})$. It is non-trivial that G(n,k) is even a polynomial for integer n,k. Polynomial is proven since G(n,k) satisfies the simple recurrence relation

$$G(n+1,k+1) = q^{k+1}G(n,k+1) + G(n+1,k).$$
(2.2)

This recurrence also shows that G(n, k) is the generating function for the number of partitions in an $n \times k$ box. A partition either has a part of largest possible size $(q^{k+1}G(n, k+1))$ or it does not (G(n+1, k)). To prove this, one can simply check that G(n, k) does satisfy

⁷Consecutive parts in the partition differ by $\geq d$.

⁸The q-factorial is the generating polynomial for the number of inversions over the symmetric group S_n [Brä15].

this recurrence as well as the initial conditions G(n,0) = 1 = G(0,k).⁹ This argument also confirms G(n,k) = G(k,n).¹⁰

The question of unimodality for the *q*-binomial coefficients was first stated in the 1850s by Cayley and then proven by Sylvester in 1878 [Brä15]. The first "elementary" proof was given by Proctor in 1982. He essentially described the coefficients as partitions in a box (grid-shading problem) and then used linear algebra to finish the proof [Pro82].

Lemma 2.16 does not apply here; q-brackets and q-factorials are in the class of logconcave (use Lemma 2.17) but not real-rooted (for n > 1) polynomials while these G(n, k)polynomials are in the class of unimodal, but not necessarily log-concave polynomials. For example, $G(2,2) = 1 + q + 2q^2 + q^3 + q^4$ is unimodal but not log-concave. In fact, most q-binomial polynomials are not log-concave since the first (and last) 2 coefficients are 1. Explicitly: G(n, k) is log-concave $\Leftrightarrow k = 1$.

2.4 Original Recurrence

The symmetric and structured nature of G(n, k) and its coefficients is visually muddled by the following recurrence that Zeilberger [Zei89a] created to translate O'Hara's [O'H90] combinatorial argument¹¹ to mostly algebra.

$$G(n,k) = \sum_{\substack{(d_1,\dots,d_k);\sum_{i=1}^k id_i = k}} q^{k\left(\sum_{i=1}^k d_i\right) - k - \sum_{1 \le j < i \le k} (i-j)d_id_j} \prod_{i=0}^{k-1} G\left((k-i)n - 2i + 2\sum_{j=0}^{i-1} (i-j)d_{k-j}, d_{k-i}\right)$$
(2.3)

is IMPLEMENTED AS KOH(n,k,q). The product is part of the summand and the outside sum is over all partitions of k. d_i is the number of parts of size i in the partition. The

⁹The coefficients are symmetric since each partition of ℓ inside an $n \times k$ box corresponds to a partition of $nk - \ell$.

 $^{{}^{10}}G(n-k,k)$ also counts the number of subspaces of dimension k in a vector space of dimension n over a finite field with q elements.

¹¹Her argument is rewritten/explained by Zeilberger [Zei89b].

initial conditions 12 are

$$G(n < 0, k) = 0, \quad G(n, k < 0) = 0, \quad G(0, k) = 1, \quad G(n, 0) = 1, \quad G(n, 1) = \frac{1 - q^{n+1}}{1 - q}$$
(2.4)

By using Propositions 2.7, 2.8, and 2.9, and induction on the symmetric unimodality of G(a, b) for a < n or b < k, we can see (after a straightforward calculation) that the right hand side is symmetric and unimodal of darga nk. For each partition, the darga will be:

$$2\left[k\left(\sum_{i=1}^{k} d_{i}\right) - k - \sum_{1 \le j < i \le k} (i-j)d_{i}d_{j}\right] + \sum_{i=0}^{k-1} \left((k-i)n - 2i + 2\sum_{j=0}^{i-1} (i-j)d_{k-j}\right)d_{k-i} = nk.$$

Recall that $\sum_{i=1}^{k} i d_i = k$.

One is led to ask how long this more complicated recurrence will take to compute? What is the largest depth of recursive calls that will be made in Eqn. (2.3)? This is answered with a brute-force method in **KOHdepth**. **KOHcalls** returns the total number of recursive calls made. But is there an explicit answer?

We will use n', k' to denote the new n, k in the recursive call for a given partition. We sometimes also treat them as functions of the index: $n'(i) = (k-i)n - 2i + 2\sum_{j=0}^{i-1} (i-j)d_{k-j}$. Once a partition only has distinct parts, the recurrence will cease after a final recursive call, since each $k' \leq 1$. Therefore to maximize depth, we need to maximize using partitions with repeated parts and maximize the size of n', k'.

One method is to begin by using $\left\lfloor \left\lfloor \frac{k}{2} \right\rfloor^2 \right\rfloor$ with a 0,1 as needed and the recursive call with $i = \left\lceil \frac{k}{2} \right\rceil$. Then (since $k' = d_{k-i} = 2$) repeatedly use the partition [1²] and i = 1. This method calls

$$G\left(\left(k - \left\lceil \frac{k}{2} \right\rceil\right)n - 2\left\lceil \frac{k}{2} \right\rceil + 2\sum_{j=0}^{\lceil k/2 \rceil - 1} \left(\left\lceil \frac{k}{2} \right\rceil - j\right) d_{k-j}, d_{k-\lceil k/2 \rceil}\right)$$
$$= G\left(\left\lfloor \frac{k}{2} \right\rfloor n - 2\left\lceil \frac{k}{2} \right\rceil + 2\sum_{j=0}^{\lceil k/2 \rceil - 1} 0, d_{\lfloor k/2 \rfloor}\right)$$
$$= G\left(\left\lfloor \frac{k}{2} \right\rfloor n - 2\left\lceil \frac{k}{2} \right\rceil, 2\right).$$

 $^{{}^{12}}G(n,1)$ is explicitly given because Eqn. (2.3) only yields the tautology G(n,1) = G(n,1).

15

Then using the partition $[1^2]$, and i = 1, calls G(n'-2, 2). Thus, the total depth of calls is

$$\left\lceil \frac{\lfloor k/2 \rfloor n - 2 \lceil k/2 \rceil}{2} \right\rceil + 1 = \left\lceil \frac{\lfloor k/2 \rfloor n}{2} \right\rceil - \left\lceil \frac{k}{2} \right\rceil + 1$$

In fact

Lemma 2.19. If k = 1, 2 or if $k \ge 4$ is even (odd) and $n \ge 2$ ($n \ge 4$), then the maximum depth of recursive calls is

$$\left\lceil \frac{\lfloor k/2 \rfloor n}{2} \right\rceil - \left\lceil \frac{k}{2} \right\rceil + 1.$$

For k = 3, the maximum depth of recursive calls is $\left\lceil \frac{n}{4} \right\rceil$.

Proof. Base case: k = 1. There are no recursive calls:

$$\left\lceil \frac{\lfloor 1/2 \rfloor n}{2} \right\rceil - \left\lceil \frac{1}{2} \right\rceil + 1 = \left\lceil \frac{0 \cdot n}{2} \right\rceil - 1 + 1 = 0.$$

For k = 2, the only option that yields recursive calls is $[1^2]$ and i = 1: G(n - 2, 2). Thus, we will have

$$\left\lceil \frac{n}{2} \right\rceil = \left\lceil \frac{\lfloor 2/2 \rfloor n}{2} \right\rceil - \left\lceil \frac{2}{2} \right\rceil + 1$$

recursive calls, which matches with expected. For k = 3, the only recursive call that does not immediately terminate is $\lambda = [1^3]$ and i = 2. The call is

$$G\left((3-2)n-2\cdot 2+2\sum_{j=0}^{2-1}(2-j)d_{3-j},d_{3-2}\right)=G(n-4,3).$$

Thus, we will have $\left\lceil \frac{n}{4} \right\rceil$ recursive calls. The difference from other k arises because $\left\lfloor \frac{3}{2} \right\rfloor = 1$.

First consider k = even and n = 2. By Proposition 2.20, the depth is

$$1 = \frac{kn}{4} - \frac{k}{2} + 1 = \left\lceil \frac{\lfloor k/2 \rfloor n}{2} \right\rceil - \left\lceil \frac{k}{2} \right\rceil + 1$$

Now let $k \ge 4$ and $n \ge 3$ $(n \ge 4$ if k = odd). Assume true for (n', k') s.t. k' < k or k' = k and n' < n.

What if we repeat part $m, \ell \geq 2$ times? The recursive call uses i = k - m.

$$G\left((k-k+m)n-2(k-m)+2\sum_{j=0}^{k-m-1}(k-m-j)d_{k-j},d_m\right)$$
$$=G\left(mn-2k+2m+2\sum_{j=m+1}^{k}(j-m)d_j,\ell\right).$$

By the induction hypothesis, the depth of this call is (note $ml \leq k$)

$$\begin{split} 1 - \left\lceil \frac{\ell}{2} \right\rceil + \left\lceil \frac{\lfloor \ell/2 \rfloor \left(mn - 2k + 2m + 2 \sum_{j=m+1}^{k} (j-m) d_j \right) \right]}{2} \\ &\leq 1 - \left\lceil \frac{l}{2} \right\rceil + \left\lceil \frac{\lfloor l/2 \rfloor \left(mn - 2k + 2m + 2 \sum_{j=m+1}^{k} j d_j \right) \right]}{2} \\ &\leq 1 - \left\lceil \frac{\ell}{2} \right\rceil + \left\lceil \frac{\lfloor \ell/2 \rfloor \left(mn - 2k + 2m + 2(k - \ell m) \right)}{2} \right\rceil \\ &= 1 - \left\lceil \frac{\ell}{2} \right\rceil + \left\lceil \frac{\lfloor \ell/2 \rfloor \left(mn - 2k + 2m + 2(k - \ell m) \right)}{2} \right\rceil \\ &\leq 1 - \left\lceil \frac{\ell}{2} \right\rceil + \left\lceil \frac{\lfloor \ell/2 \rfloor \left(mn - 2k - 2\ell \right)}{4} \right\rceil \\ &\leq 1 - \left\lceil \frac{\ell}{2} \right\rceil + \left\lceil \frac{m\ell \left(n + 2 - 2\ell \right)}{4} \right\rceil . \end{split}$$

The upper bound is maximized when ℓ is minimized: $\ell = 2$, upper bound $= \left\lceil \frac{kn}{4} - \frac{k}{2} \right\rceil$. We now need to confirm that $\lambda = \left\lfloor \left\lfloor \frac{k}{2} \right\rfloor^2 \right\rfloor$ (with 1 if needed), $i = \left\lceil \frac{k}{2} \right\rceil$ achieves this upper bound. The depth is

$$1 - \left\lceil \frac{2}{2} \right\rceil + \left\lceil \frac{\lfloor 2/2 \rfloor \left(\lfloor \frac{k}{2} \rfloor n - 2k + 2 \lfloor \frac{k}{2} \rfloor \right)}{2} \right\rceil = \left\lceil \frac{\left(\lfloor \frac{k}{2} \rfloor n - 2 \lceil \frac{k}{2} \rceil \right)}{2} \right\rceil = \left\lceil \frac{\lfloor k/2 \rfloor n}{2} - \left\lceil \frac{k}{2} \rceil \right\rceil,$$

which for k = even directly matches the upper bound. For k = odd, we adjust the upper bound by recognizing that if $\ell =$ even, then $m\ell \leq k - 1$ and splitting into 2 cases:

1. $\sum_{j=m+1}^{k} jd_j = 0$. Then the depth of the call is

$$1 - \left\lceil \frac{\ell}{2} \right\rceil + \left\lceil \frac{\lfloor \ell/2 \rfloor \left(mn - 2k + 2m + 2\sum_{j=m+1}^{k} (j-m) d_j \right)}{2} \right\rceil$$
$$= 1 - \frac{\ell}{2} + \left\lceil \frac{m\ell(n+2) - 2k\ell}{4} \right\rceil$$
$$\leq 1 - \frac{\ell}{2} + \left\lceil \frac{(k-1)(n+2) - 2k\ell}{4} \right\rceil :$$

maximized for $\ell = 2$ and gives an upper bound $\left\lceil \frac{(k-1)(n+2)-4k}{4} \right\rceil = \left\lceil \frac{(k-1)n}{4} - \frac{k+1}{2} \right\rceil$.

2. $\sum_{j=m+1}^{k} jd_j > 0$. Then $m\ell \leq k - (m+1) \leq k - 2$ and the depth of the call is

$$1 - \left\lceil \frac{\ell}{2} \right\rceil + \left\lceil \frac{\lfloor \ell/2 \rfloor \left(mn - 2k + 2m + 2\sum_{j=m+1}^{k} (j-m) d_j \right)}{2} \right\rceil$$
$$\leq 1 - \frac{\ell}{2} + \left\lceil \frac{\ell (mn - 2k + 2m + 2(k-\ell m) - 2m)}{4} \right\rceil$$
$$= 1 - \frac{\ell}{2} + \left\lceil \frac{m\ell (n-2\ell)}{4} \right\rceil$$
$$\leq 1 - \frac{\ell}{2} + \left\lceil \frac{m\ell (n-2\ell)}{4} \right\rceil,$$

which again is maximized for $\ell = 2$ and gives an upper bound $\left\lceil \frac{(k-2)(n-4)}{4} \right\rceil = \left\lceil \frac{(k-1)n}{4} - \frac{k+1}{2} - \frac{2k+n-10}{4} \right\rceil \le \left\lceil \frac{(k-1)n}{4} - \frac{k+1}{2} \right\rceil$ since $k \ge 4$ and $n \ge 2$.¹³

The upper bound from either case matches the achievable bound for k = odd. If $\ell = \text{odd}$, then the earlier $\lfloor \frac{\ell}{2} \rfloor \leq \frac{\ell}{2}$ can be replaced by $= \frac{\ell-1}{2}$. Thus, the upper bound is

$$\leq 1 - \left\lceil \frac{\ell}{2} \right\rceil + \left\lceil \frac{m(\ell-1)\left(n+2-2\ell\right)}{4} \right\rceil \leq 1 - \left\lceil \frac{\ell}{2} \right\rceil + \left\lceil \frac{(k-1)\left(n+2-2\ell\right)}{4} \right\rceil$$

which is now maximized by $\ell = 3$ (since we assume $\ell = \text{odd})^{14}$ giving the upper bound: $\left\lceil \frac{(k-1)(n-4)}{4} \right\rceil - 1$, which is worse¹⁵ than the bound when $\ell = 2$ so we can ignore $\ell = \text{odd}$ when it comes to maximizing depth. Therefore our chosen λ , and i, achieve the greatest depth: $1 + \left\lceil \frac{\lfloor k/2 \rfloor n}{2} \right\rceil - \left\lceil \frac{k}{2} \right\rceil$.

To confirm that this recursive depth actually happens, we must also confirm that $G(n',k') \neq 0$ for any other *i*. I.e., we must show that $n' \geq 0$. For $i < \lceil \frac{k}{2} \rceil$

$$n' = (k-i)n - 2i + 2\sum_{j=0}^{i-1} (i-j)d_{k-j} = (k-i)n - 2i > \left(k - \left\lceil \frac{k}{2} \right\rceil\right)n - 2\left\lceil \frac{k}{2} \right\rceil = \left\lfloor \frac{k}{2} \right\rfloor n - 2\left\lceil \frac{k}{2} \right\rceil \ge 0$$

since $n \ge 3$ when k = even and since $n \ge 4$ when k = odd. For $i > \lfloor \frac{k}{2} \rfloor$,

$$n' = (k-i)n - 2i + 2\sum_{j=0}^{i-1} (i-j)d_{k-j} = (k-i)n - 2i + 2\left(i - \left\lceil \frac{k}{2} \right\rceil\right) 2$$
$$= (k-i)n + 2i - 4\left\lceil \frac{k}{2} \right\rceil = kn - i(n-2) - 4\left\lceil \frac{k}{2} \right\rceil \ge kn - (k-1)(n-2) - 4\left\lceil \frac{k}{2} \right\rceil$$
$$= kn - kn + n + 2k - 2 - 4\left\lceil \frac{k}{2} \right\rceil = n - 2 + 2k - 4\left\lceil \frac{k}{2} \right\rceil,$$

 $^{^{13}}$ Actually, we have k= odd so $k\geq 5$ and $n\geq 4;$ the bound improves further.

¹⁴Actually, $\ell = 5$ since we ignore $\ell = 3$.

¹⁵We only need $k \ge 1$.

which for k = even, is ≥ 0 since $n \geq 3$. And for $k = \text{odd } n' \geq 0$, since $n \geq 4$. By induction, the claim holds for all n, k.

We have glossed over a few details. We assumed that $k' = \ell \neq 3$. If $\ell = 3$, then this would lead to a different (shallower) recursive call than $\ell = 2$:

$$\left\lceil \frac{n'}{4} \right\rceil \le \left\lceil \frac{n'}{2} \right\rceil = \left\lceil \frac{\lfloor 2/2 \rfloor n'}{2} \right\rceil - \left\lceil \frac{2}{2} \right\rceil + 1.$$

We also need to ensure that if $k' \ge 4$, then $n' \ge 3$ or $n' \ge 4$ as needed. Actually, if $k' \ge 4$, and n' < 3 (n' < 4) then by Proposition 2.20 their depth is 1. And since the extra depth of using $\lambda = \left\lfloor \lfloor \frac{k}{2} \rfloor^2 \right\rfloor$ (with 0, 1 as needed), is¹⁶

$$\left\lceil \frac{\lfloor k/2 \rfloor n}{2} \right\rceil - \left\lceil \frac{k}{2} \right\rceil \ge 1,$$

 $\ell = 2$ produces a recursive call that is at least as deep.

Proposition 2.20. If k = even, then for n = 1, 2, the maximum depth of recursive calls is 1. If $k \ge 3$ is odd, then for n = 1, 2, 3, the maximum depth is 1.

Proof. Consider a partition that repeats part $m, \ell \ge 2$ times. First consider $n \le 2$. The recursive call with i = k - m is

$$n' = mn - 2k + 2m + 2\sum_{j=m+1}^{k} (j-m) d_j \le 4m - 2k + 2\sum_{j=m+1}^{k} (j-m) d_j$$
$$\le 4m - 2k + 2(k - m\ell) = m(4 - 2\ell) \le m(4 - 4) = 0.$$

Thus, the chain terminates after at most 1 step. For all other partitions (distinct), we already know the chain terminates after 1 step.

Now consider k = odd and n = 3. Note that $d_k = 0$ otherwise we have a distinct partition. The recursive call with i = k - 1 is

$$n' = (k - (k - 1))3 - 2(k - 1) + 2\sum_{j=0}^{(k-1)-1} ((k - 1) - j) d_{k-j}$$
$$= 5 - 2k + 2\sum_{j=0}^{k-1} (k - j) d_{k-j} - 2\sum_{j=0}^{k-1} d_{k-j} = 5 - 2k + 2k - 2\sum_{j=0}^{k-1} d_{k-j} = 5 - 2\sum_{j=1}^{k} d_j < 0$$

¹⁶Recall that $k \ge 4$ and $n \ge 3$ $(n \ge 4)$. Actually, this is the only spot that needs $n \ge 3$ when k = even. Otherwise $n \ge 2$ is sufficient.

if $|\lambda| \geq 3$. Thus, the only partitions that actually create recursive calls are those with $|\lambda| \leq 2$. And since k = odd, λ will be distinct and therefore terminate after at most 1 step.

The explicit depth formula is IMPLEMENTED AS **KOHdepthFAST**(n,k). It is somewhat odd that the depth is not symmetric in n, k. G(n, k) is symmetric so we can choose to calculate G(k, n) if n > k and have a shallower depth of recursive calls. The total number of calls may be the same; this was not analyzed.

2.5 Altered Recurrence

2.5.1 Restricted Partitions

We can alter the recurrence of Eqn. (2.3) in several ways to create G'(n,k) and maintain the property that G'(n,k) is symmetric and unimodal of darga nk.¹⁷ One simple way to do this is to restrict the partitions over which we sum. This simply reduces the use of Proposition 2.7. We can restrict the minimum/maximum size of a partition, min/max integer in a partition, distinct parts, modulo classes, etc.

Suppose we restrict to partitions with size $\leq s$ and denote this new function as G_s .

$$\begin{aligned} G_s(n,k) &= \sum_{\substack{(d_1,\dots,d_k); \sum_{i=1}^k id_i = k; \sum_{i=1}^k d_i \le s}} q^{k\left(\sum_{i=1}^k d_i\right) - k - \sum_{1 \le j < i \le k} (i-j)d_i d_j} \\ &= \prod_{i=0}^{k-1} G_s \left((k-i)n - 2i + 2\sum_{j=0}^{i-1} (i-j)d_{k-j}, d_{k-i} \right) \\ &= \sum_{\substack{(d_1,\dots,d_k); \sum_{i=1}^k id_i = k; \sum_{i=1}^k d_i \le s}} q^{k\left(\sum_{i=1}^k d_i\right) - k - \sum_{1 \le j < i \le k} (i-j)d_i d_j} \\ &\prod_{i=0}^{k-1} G\left((k-i)n - 2i + 2\sum_{j=0}^{i-1} (i-j)d_{k-j}, d_{k-i} \right). \end{aligned}$$

For $k \leq s$, $G_s(n,k) = G(n,k)$ since ALL partitions of k will have size $\leq s$. And since $d_{k-i} \leq s$ by definition of the sum, we can replace G_s by G (the actual q-binomial) in the product. I found conjectured recurrence relations of order 1 and degree 1 in both n, k for s = 1, 2, 3, 4. I then conjectured for general s:

¹⁷Dependent on restrictions, one may achieve G'(n,k) = 0.

$$q^{n}(q^{k+s}-1)G_{s}(n+1,k) - q^{k+1}(q^{n}-q^{s-2})G_{s}(n,k+1) + (q^{n}-q^{k+s-1})G_{s}(n+1,k+1) = 0.$$
(2.5)

Eqn. (2.5) was verified for $n, k \leq 20$ and $s \leq 10$. The bounds were chosen to make the verification run in a short time (≈ 2 minutes). In the algebra of formal power series, taking the limit as $s \to \infty$ (thus obtaining the original G(n, k) with no restriction), recovers

$$q^{n}(-1)G_{\infty}(n+1,k) - q^{k+1}(q^{n})G_{\infty}(n,k+1) + (q^{n})G_{\infty}(n+1,k+1) = 0,$$

$$G(n+1,k) + q^{k+1}G(n,k+1) = G(n+1,k+1).$$

which matches the recurrence relation in Eqn. (2.2). Recall that $G_s(n,k) = G(n,k)$ for $k \leq s$. It is somewhat surprising that G(n,k) follows many recurrence relations for bounded k.

We can try to enumerate all $G_s(n,k)$ using a translated Eqn. (2.5):

$$G_s(n,k) = \frac{1}{q^n - q^{k+s-1}} \left(q^n (q^{k+s-1} - 1) G_s(n,k-1) - q^k (q^n - q^{s-1}) G_s(n-1,k) \right).$$

IMPLEMENTED AS Gs(s,n,k,q). However, this recurrence by itself cannot enumerate all $G_s(n,k)$ as we will eventually hit the singularity-causing n = k + s - 1 if $n \ge s$. However, we do obtain an interesting relation along that line:

$$G_s(k+s,k) = \frac{1-q^{k+1}}{1-q^{k+s}}G_s(k+s-1,k+1) \quad \text{or} \quad G_{n-k+1}(n+1,k) = \frac{1-q^{k+1}}{1-q^{n+1}}G_{n-k+1}(n,k+1)$$

So far, Eqn. (2.5) has been verified for s = 1, 2, 3 using the explicit formulas in Eqns. (2.6), (2.9), and (2.10), respectively.

It is quite remarkable that the generalized recurrence in Eqn. (2.5) yields polynomials using the initial conditions of Eqn. (2.4),¹⁸ let alone unimodal polynomials. Are there other sets of initial conditions (potentially for chosen values of s) that will also yield unimodal polynomials when iterated in Eqn. (2.5)?¹⁹

 $^{{}^{18}}G_s(n,0) = G_s(0,k) = 1.$

¹⁹At least until the singularity.

There are a couple of avenues for attack of Conjecture 2.21. A good starting place may be to tackle the (easier?) question of whether the recurrence even produces symmetric polynomials.

Since we have base case verification, we could try to use induction to prove the conjecture. This would likely involve writing $G_{s+1} = G_s + g_{s+1}$, where g_{s+1} is the contribution from all of the partitions of size s + 1:

$$g_{s+1}(n,k) = \sum_{\substack{(d_1,\dots,d_k); \sum_{i=1}^k id_i = k; \sum_{i=1}^k d_i = s+1\\ \prod_{i=0}^{k-1} G\left((k-i)n - 2i + 2\sum_{j=0}^{i-1} (i-j)d_{k-j}, d_{k-i}\right).$$

Another possibility, since we have an explicit expression for G_s (and for G if needed), may be to simply plug in to Eqn. (2.5) and simplify cleverly.

A more general question would be to analyze recurrences of type

$$a(n,k)F(n+1,k+1) = b(n,k)F(n+1,k) + c(n,k)F(n,k+1)$$

that produce symmetric and unimodal polynomials. What are the requirements on a, b, c? If a_1, b_1, c_1 and a_2, b_2, c_2 both produce unimodal polynomials, what, if anything, can be said about $a_1 + a_2, b_1 + b_2, c_1 + c_2$? About a_1a_2, b_1b_2, c_1c_2 ? If one can build from previously known valid recurrences, one could potentially build up to Eqn. (2.5) from Eqn. (2.2).

"Natural Partitions"

Lemma 2.22. Suppose we restrict partitions to only use integers $\leq p$. If $p < \frac{2k}{n+2}$ then $G'(n,k) = 0.^{20}$

Proof. We have $d_i = 0$ for i > p. In the product of Eqn. (2.3), consider i = k - 1 for any

²⁰ The bound $n+2 < 2\left\lceil \frac{k}{p} \right\rceil$ is sufficient to make G'(n,k) = 0.

restricted partition λ . Then for the new n in the recursive call we obtain

$$n'(k-1) = (k - (k-1))n - 2(k-1) + 2 \sum_{j=0}^{(k-1)-1} ((k-1) - j)d_{k-j}$$

$$= n - 2(k-1) + 2 \sum_{j=k-p}^{k-2} (k-1-j)d_{k-j}$$

$$= n - 2(k-1) + 2 \sum_{j=2}^{p} (k-1 - (k-j))d_{k-(k-j)}$$

$$= n - 2(k-1) + 2 \sum_{j=1}^{p} (j-1)d_{j}$$

$$= n - 2(k-1) + 2k - 2|\lambda|$$

$$\leq n + 2 - 2 \left\lceil \frac{k}{p} \right\rceil$$

$$\leq n + 2 - 2 \frac{k}{p}$$

$$< n + 2 - 2 \frac{k}{2k/(n+2)} = 0.$$

Then $G'(n', d_1) = 0$ for any partition and therefore $G'(n, k) = \sum_{\lambda} 0 = 0$.

What do other natural partitions yield for a recursive call? Let us first consider $\lambda = [1^k]$. Then $d_1 = k$ and $d_{i \neq 1} = 0$.

$$q^{k\left(\sum_{i=1}^{k} d_{i}\right)-k-\sum_{1\leq j
$$=q^{k\cdot k-k}G\left(n-2(k-1),k\right)\prod_{i=0}^{k-2}G\left((k-i)n-2i+2\sum_{j=0}^{i-1}(i-j)\cdot 0,0\right)$$
$$=q^{k\cdot (k-1)}G\left(n-2(k-1),k\right)\prod_{i=0}^{k-2}G\left((k-i)n-2i,0\right)=q^{k(k-1)}G\left(n-2(k-1),k\right),$$$$

since $(k-i)n - 2i \ge 2n - 2(k-2)$ and if 2n - 2(k-2) < 0, then n - 2(k-1) < 0. If we define G'(n,k) as only being over partitions = $[1^k]$, then $G'(n,k) = q^{nk/2}$ if 2(k-1)|nand = 0 otherwise. One can confirm the monomial by either looking at how many times $q^{k(k-1)}$ is multiplied or by recognizing that G'(n,k) must be a monomial of darga nk.

Now consider $\lambda = [k]$. Then $d_k = 1$ and $d_{i \neq k} = 0$.

$$q^{k\left(\sum_{i=1}^{k} d_{i}\right)-k-\sum_{1\leq j< i\leq k}(i-j)d_{i}d_{j}}\prod_{i=0}^{k-1}G\left((k-i)n-2i+2\sum_{j=0}^{i-1}(i-j)d_{k-j},d_{k-i}\right)$$
$$=q^{k-k}G\left(kn,1\right)\prod_{i=1}^{k-1}G\left((k-i)n-2i+2\sum_{j=1}^{i-1}(i-j)d_{k-j}+2(i-0)d_{k},0\right)$$
$$=G\left(kn,1\right)\prod_{i=1}^{k-1}G\left((k-i)n,0\right)=G(nk,1)=\frac{1-q^{nk+1}}{1-q},$$

since $(k-i)n \ge n \ge 0$. So [k] provides the "base" of G(n,k): $1+q+\cdots+q^{nk}$.

If we restrict to partitions of size 1, then

$$G_1(n,k) = \frac{1 - q^{nk+1}}{1 - q}.$$
(2.6)

We can use this to verify the recurrence relation in Eqn. (2.5) for s = 1.

Lemma 2.23. Let us restrict partitions to those with distinct parts. Fix $k \in \mathbb{N}$. Let $\ell = \lfloor \frac{1}{2}(\sqrt{8k+1}-1) \rfloor$.²¹ Then G'(n,k) is a polynomial of degree k in q^n for $n \ge 2\ell - 2$.²²

Proof. Since $\lambda = [k]$ is distinct, G'(n,k) is of degree k in q^n . Let λ be a partition with distinct parts. Then $d_i \leq 1$ and $|\lambda| = \sum_{j=1}^k d_k \leq l$. And for each recursive call, $G'(n', d_i) = 0, 1$, or $\frac{1-q^{n'+1}}{1-q}$. Since k is fixed, it has a fixed number of partitions into distinct parts; thus, it remains to show $G'(n', d_i) \neq 0$ by showing $n'(i) \geq 0$ for all i.

$$n'(i) = (k-i)n - 2i + 2\sum_{j=0}^{i-1} (i-j)d_{k-j},$$

$$n'(i) - n'(i-1) = 2\sum_{j=0}^{i-1} (i-j)d_{k-j} - 2\sum_{j=0}^{i-1} (i-j)d_{k-j} + 2\sum_{j=0}^{i-2} d_{k-j} + 2d_{k-i+1} - n - 2$$

$$= 2\sum_{j=0}^{i-1} d_{k-j} - n - 2 \le 2\ell - n - 2 \le 0.$$

²¹I.e., the index of the largest triangular number $\leq k$ or the maximum size of a distinct partition of k. ²²It appears that $n = 2\ell - 3$ also fits the polynomial, though this was left unproven. Thus, the minimum occurs at i = k - 1 and has value

$$n'(k-1) = (k - (k - 1))n - 2(k - 1) + 2\sum_{j=0}^{k-1-1} (k - 1 - j)d_{k-j}$$
$$= n - 2(k - 1) + 2\sum_{j=0}^{k-1} (k - 1 - j)d_{k-j}$$
$$= n - 2(k - 1) + 2\sum_{j=0}^{k-1} (k - j)d_{k-j} - 2\sum_{j=0}^{k-1} d_{k-j}$$
$$= n - 2(k - 1) + 2k - 2\sum_{j=0}^{k-1} d_{k-j} = n + 2 - 2\ell \ge 0.$$

Let us return to the minimum of n' for general partitions. It occurs either at i = k - 1, or if $n \leq 2|\lambda| - 2 - 2(\# \text{ of 1s in } \lambda)$, then at the final i s.t. $n + 2 \geq 2 \sum_{j=0}^{i-1} d_{k-j}$: see the simplification of n'(i) - n'(i-1) above. The second difference is n'(i) - 2n'(i-1) + n'(i-2) = $2d_{k-i+1} \geq 0$. Thus, the extrema of n' found is in fact a minimum. And if we fix the i such that there exists a minimum (or equivalent to the minimum), we obtain a value of

$$(k-i)n - 2i + 2\sum_{j=0}^{i-1} (i-j)d_{k-j} \approx (k-i)\left(2\sum_{j=0}^{i-1} d_{k-j} - 2\right) - 2i + 2\sum_{j=0}^{i-1} (i-j)d_{k-j}$$
$$= 2\sum_{j=0}^{i-1} (k-j)d_{k-j} - 2k \le 2k - 2k = 0.$$

To avoid the approximation step, consider the next i:

$$(k - (i + 1))n - 2(i + 1) + 2\sum_{j=0}^{i} (i + 1 - j)d_{k-j}$$

$$< (k - (i + 1))\left(2\sum_{j=0}^{i} d_{k-j} - 2\right) - 2(i + 1) + 2\sum_{j=0}^{i} (i + 1 - j)d_{k-j}$$

$$= 2\sum_{j=0}^{i} (k - j)d_{k-j} - 2k \le 2k - 2k = 0.$$

The recursive call is then seen to be 0 for any partition that has $|\lambda| - |\{1s \text{ in } \lambda\}| \ge \frac{n+2}{2}$.

Now we consider $\lambda = [k - \ell, \ell]$ for some $1 \le \ell < \frac{k}{2}$.

$$\begin{split} q^{k\left(\sum_{i=1}^{k}d_{i}\right)-k-\sum_{1\leq j< i\leq k}(i-j)d_{i}d_{j}}\prod_{i=0}^{k-1}G\left((k-i)n-2i+2\sum_{j=0}^{i-1}(i-j)d_{k-j},d_{k-i}\right)} \\ &=q^{2k-k-((k-\ell)-\ell)}\prod_{i=0}^{\ell-1}G\left((k-i)n-2i+2\sum_{j=0}^{i-1}(i-j)d_{k-j},0\right) \\ &\quad G\left((k-\ell)n-2\ell+2\sum_{j=0}^{\ell-1}(\ell-j)d_{k-j},d_{k-\ell}\right) \\ &\quad \prod_{i=\ell+1}^{k-\ell-1}G\left((k-i)n-2i+2\sum_{j=0}^{i-1}(i-j)d_{k-j},0\right) \\ &\quad G\left((k-(k-\ell))n-2(k-\ell)+2\sum_{j=0}^{k-\ell-1}(k-\ell-j)d_{k-j},d_{\ell}\right) \\ &\quad \prod_{i=k-\ell+1}^{k-1}G\left((k-i)n-2i+2\sum_{j=0}^{i-1}(i-j)d_{k-j},0\right) \\ &=q^{2\ell}\prod_{i=0}^{\ell-1}G\left((k-i)n-2i,0\right) \\ &\quad G\left((k-\ell)n-2\ell,1\right) \\ &\quad \prod_{i=\ell+1}^{k-\ell-1}G\left((k-i)n-2\ell,0\right) \\ &\quad G\left((k-2\ell,1) \\ &\quad \prod_{i=k-\ell+1}^{k-\ell-1}G\left((k-i)n+2i-2k,0\right), \end{split}$$

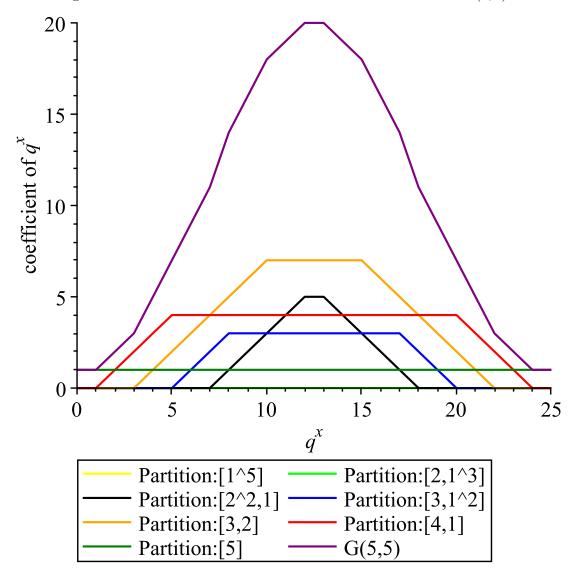
which to be nonzero requires that (assuming $n \ge 2$ otherwise we get 0 in the recursive call)

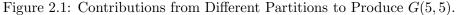
$$0 \le (k-i)n - 2i \ge (k-\ell+1)n - 2(\ell-1) \ge (2\ell-\ell+1)2 - 2(\ell-1) \ge 4,$$
 and
$$0 \le (k-i)n - 2\ell \ge (k-(k-\ell-1))n - 2\ell = (\ell+1)n - 2\ell \ge 2,$$
 and
$$0 \le (k-i)n + 2i - 2k \ge (k-(k-1))n + 2(k-1) - 2k = n - 2 \ge 0.$$

The leftmost inequality is what we need and the other inequalities deduce that to be true. Thus, the recursive call is

$$q^{2\ell} \frac{1 - q^{(k-\ell)n-2\ell+1}}{1-q} \cdot \frac{1 - q^{\ell n-2\ell+1}}{1-q}$$

For a visual of how the different partitions can contribute to the final polynomial, see Figure 2.1.





The partitions $[1^5]$ and $[2, 1^3]$ both contribute the 0 polynomial.

Lemma 2.24. If we restrict to partitions with consecutive difference $\geq k - d$, and $d < \frac{2}{3}k + 1$, then the coefficient of $q^{nk/2}$ in G'(n,k) is $n\frac{D(D+1)}{2} - (D^2 - 1)$ for $D = \lfloor \frac{d}{2} \rfloor$.

Proof. The bound on d implies that the maximum size of a partition is 2 by Proposition 2.25. D can be seen to be the relevant number because it is the number of possible size 2 partitions of k. $[k], [k-1,1], \ldots, [k-D,D]$ are valid since $k - D - D = k - 2 \lfloor \frac{d}{2} \rfloor \ge k - d$.

$$G'(n,k) = \frac{1-q^{nk+1}}{1-q} + \sum_{i=1}^{D} q^{2i} \frac{1-q^{(k-i)n-2i+1}}{1-q} \cdot \frac{1-q^{in-2i+1}}{1-q}.$$
 (2.7)

Note that the middle coefficient of $\left(\frac{1-q^a}{1-q}\right) \cdot \left(\frac{1-q^b}{1-q}\right)$ is going to be min(a, b). Also note that the darga of the summand in Eqn. (2.7) is

$$2 \cdot (2i) + [(k-i)n - 2i + 1 - 1] + [in - 2i + 1 - 1] = nk,$$

so $q^{nk/2}$ is the middle monomial.²³ Thus, the coefficient is

$$1 + \sum_{i=1}^{D} \min\left[(k-i)n - 2i + 1, in - 2i + 1\right] = 1 + \sum_{i=1}^{D} (in - 2i + 1) = \frac{1}{2} \left((n-2)D + 2\right) \left(D + 1\right).$$

Proposition 2.25. If we restrict to partitions with difference $\geq k - d$ and minimum part ℓ , then to have a partition of size s, it is required that

$$d \ge k \frac{(s+1)(s-2)}{s(s-1)} + \frac{2\ell}{s-1},$$

$$\ell \le \frac{ds^2 - ks^2 - ds + ks + 2k}{2s},$$

$$k \le \frac{(ds - d - 2\ell)s}{(s+1)(s-2)}.$$

Proof. A partition λ of size s, minimum part ℓ , and difference k - d will have total

$$k = \sum_{i=1}^{s} \lambda_i \ge \sum_{i=1}^{s} \left[\ell + (i-1)(k-d)\right] = \frac{s}{2}(2\ell - k + d + (k-d)s).$$

 $^{^{23}}$ As one should expect since the summand of Eqn. (2.7) is a specific case of the summand in Eqn. (2.3).

Size 2 Partitions

Another special partition to consider is $\lambda = \left[\frac{k}{2}^2\right]$ when k is even. Then $d_{k/2} = 2$ and $d_i = 0$ otherwise.

$$\begin{split} q^{k\left(\sum_{i=1}^{k}d_{i}\right)-k-\sum_{1\leq j< i\leq k}(i-j)d_{i}d_{j}}\prod_{i=0}^{k-1}G\left((k-i)n-2i+2\sum_{j=0}^{i-1}(i-j)d_{k-j},d_{k-i}\right) \\ &=q^{2k-k-0}\prod_{i=0}^{k/2-1}G\left((k-i)n-2i+2\sum_{j=0}^{i-1}(i-j)d_{k-j},d_{k-i}\right) \\ &\quad G\left(\left(k-\frac{k}{2}\right)n-2\frac{k}{2}+2\sum_{j=0}^{k/2-1}\left(\frac{k}{2}-j\right)d_{k-j},d_{k/2}\right) \\ &\quad \prod_{i=k/2+1}^{k-1}G\left((k-i)n-2i+2\sum_{j=0}^{i-1}(i-j)d_{k-j},d_{k-i}\right) \\ &=q^{k}G\left(\frac{k}{2}n-k,2\right)\prod_{i=0}^{k/2-1}G\left((k-i)n-2i,0\right)\prod_{i=k/2+1}^{k-1}G\left((k-i)n-2i+2(i-k/2)2,0\right) \\ &\quad =q^{k}G\left(\frac{k}{2}n-k,2\right)\prod_{i=0}^{k/2-1}G\left((k-i)n-2i,0\right)\prod_{i=k/2+1}^{k-1}G\left((k-i)n+2i-2k,0\right), \end{split}$$

which is nonzero only if (assuming $n \ge 2$ otherwise we get 0 in the recursive call)

$$0 \le (k-i)n - 2i \ge (k - (k/2 - 1))2 - 2(k/2 - 1) = (k/2 + 1)2 - k + 2 = 4,$$

$$0 \le (k-i)n + 2i - 2k \ge (k - (k-1))n + 2(k-1) - 2k = n - 2 \ge 0.$$

Thus, the recursive call is $q^k G\left(\frac{k}{2}n-k,2\right)$.

While proving Lemma 2.24, Maple produced this useful gem:

$$\sum_{i=1}^{D} q^{2i} \frac{1-q^{(k-i)n-2i+1}}{1-q} \cdot \frac{1-q^{in-2i+1}}{1-q} = q^2 \frac{(1+q^{nk-2D})(1-q^{2D})}{(1-q)^2(1-q^2)} - q^{n+1} \frac{(1+q^{nk-nD-n})(1-q^{nD})}{(1-q)^2(1-q^n)}.$$
 (2.8)

We can then look at using only partitions restricted to maximum size 2. The possible partitions are [k] and $[k - \ell, \ell]$ for $1 \le \ell \le \frac{k}{2}$. We can find the exact form of $G_2(n, k)$ by including all partitions with size ≤ 2 .

Lemma 2.26. For $n \ge 1$,

$$G_{2}(n,k) = \left(q^{nk+n+4} - q^{nk+n+3} + q^{nk+n+1} - q^{nk+4} - q^{(n-1)k+n+3} + q^{(n-1)k+3} + q^{k+n+1} - q^{k+1} - q^{k+1} - q^{n} + q^{3} - q + 1\right) / (1-q)^{2}(1-q^{2})(1-q^{n}).$$
(2.9)

If n = 0, then $G_2(0, k) = 1$.

Proof. If k = odd, then, utilizing Eqn. (2.8),

$$G_{2}(n,k) = \frac{1-q^{nk+1}}{1-q} + \sum_{\ell=1}^{(k-1)/2} q^{2\ell} \frac{1-q^{(k-\ell)n-2\ell+1}}{1-q} \cdot \frac{1-q^{\ell n-2\ell+1}}{1-q}$$
$$= \frac{1-q^{nk+1}}{1-q} + q^{2} \frac{(1+q^{nk-2(k-1)/2})(1-q^{2(k-1)/2})}{(1-q)^{2}(1-q^{2})}$$
$$- q^{n+1} \frac{(1+q^{nk-n(k-1)/2-n})(1-q^{n(k-1)/2})}{(1-q)^{2}(1-q^{n})}$$
$$= \frac{1-q^{nk+1}}{1-q} + q^{2} \frac{(1+q^{nk-k+1})(1-q^{k-1})}{(1-q)^{2}(1-q^{2})} - q^{n+1} \frac{1-q^{n(k-1)}}{(1-q)^{2}(1-q^{n})}.$$

If k = even, then

$$\begin{aligned} G_2(n,k) &= \frac{1-q^{nk+1}}{1-q} + \sum_{\ell=1}^{k/2-1} q^{2\ell} \frac{1-q^{(k-\ell)n-2\ell+1}}{1-q} \cdot \frac{1-q^{\ell n-2\ell+1}}{1-q} + q^k G_2\left(\frac{k}{2}n-k,2\right) \\ &= \frac{1-q^{nk+1}}{1-q} + q^2 \frac{(1+q^{nk-k+2})(1-q^{k-2})}{(1-q)^2(1-q^2)} - q^{n+1} \frac{(1+q^{nk/2})(1-q^{n(k-2)/2})}{(1-q)^2(1-q^n)} \\ &+ q^k \frac{(1-q^{nk/2-k+1})(1-q^{nk/2-k+2})}{(1-q)(1-q^2)}. \end{aligned}$$

We again utilized Eqn. (2.8) and noted that $G_2(n', 2) = G(n', 2)$ and used Eqn. (2.1). One can use Maple or any other mathematics software to verify that both cases "simplify" to the expression given above.

Notice also that the bound of $n \ge 2$ required from characterizing partition calls has been lowered to $n \ge 1$. If n = 1, then the size 2 partition calls will be 0, which is exactly what the extra summands reduce to.

We can use this to verify the recurrence relation in Eqn. (2.5) for s = 2. To compute $G_2(n,k)$ using the explicit expression, use $\mathbf{G2}(\mathbf{n},\mathbf{k},\mathbf{q})$.

The expression given in Eqn. (2.9) obfuscates that $G_2(n, k)$ is unimodal of darga nk. The expanded expressions in the proof allow for a heuristic argument by reasoning the darga of each summand. This would be a great way to prove unimodality of general functions. However, the useful decomposition is often difficult to ascertain.

Size 3 Partitions

By looking at $\lambda = [\ell_1, \ell_2, \ell_3] \vdash k$ for the separate cases $\ell_1 = \ell_2 = \ell_3$, $\ell_1 = \ell_2 \neq \ell_3$, $\ell_1 \neq \ell_2 = \ell_3$, and $\ell_1 \neq \ell_2 \neq \ell_3$, one can add this to Lemma 2.26, and find an explicit expression for $G_3(n, k)$.

We start with $[\ell_1 \neq \ell_2 \neq \ell_3]$. Then $d_{\ell_i} = 1$. We will need $n \ge 4$ for nonzero calls.

$$\begin{split} q^{k\left(\sum_{i=1}^{k} d_{i}\right)-k-\sum_{1\leq j< i\leq k}(i-j)d_{i}d_{j}} \prod_{i=0}^{k-1} G\left((k-i)n-2i+2\sum_{j=0}^{i-1}(i-j)d_{k-j},d_{k-i}\right) \\ &= q^{2\left(\ell_{2}+2\ell_{3}\right)} \prod_{i=0}^{k-\ell_{1}-1} G\left((k-i)n-2i,0\right) \\ G\left(\ell_{1}n-2\left(\ell_{2}+\ell_{3}\right),1\right) \\ &\prod_{i=k-\ell_{1}+1}^{k-\ell_{2}-1} G\left((k-i)n-2\left(k-\ell_{1}\right),0\right) \\ G\left(\ell_{2}n-2\left(\ell_{2}+\ell_{3}\right),1\right) \\ &\prod_{i=k-\ell_{2}+1}^{k-\ell_{3}-1} G\left((k-i)(n-2)-2\ell_{3},0\right) \\ G\left(\ell_{3}(n-4),1\right) \\ &\prod_{i=k-\ell_{3}+1}^{k-1} G\left((k-i)(n-4),0\right) \\ &= q^{2\left(\ell_{2}+2\ell_{3}\right)} \cdot \frac{1-q^{\ell_{1}n-2\left(\ell_{2}+\ell_{3}\right)+1}}{1-q} \cdot \frac{1-q^{\ell_{2}n-2\left(\ell_{2}+\ell_{3}\right)+1}}{1-q} \cdot \frac{1-q^{\ell_{3}(n-4)+1}}{1-q}. \end{split}$$

Now $[k - 2\ell, \ell^2]$, for $\ell < \frac{k}{3}$. Then $d_{k-2\ell} = 1$, $d_\ell = 2$. Again, we need $n \ge 4$.

$$\begin{split} q^{k\left(\sum_{i=1}^{k} d_{i}\right)-k-\sum_{1\leq j< i\leq k}(i-j)d_{i}d_{j}} \prod_{i=0}^{k-1} G\left((k-i)n-2i+2\sum_{j=0}^{i-1}(i-j)d_{k-j}, d_{k-i}\right) \\ &= q^{6\ell} \prod_{i=0}^{2\ell-1} G\left((k-i)n-2i, 0\right) \\ G\left((k-2\ell)n-4\ell, 1\right) \\ &\prod_{i=2\ell+1}^{k-\ell-1} G\left((k-i)n-4\ell, 0\right) \\ G\left(\ell n-4\ell, 2\right) \\ &\prod_{i=k-\ell+1}^{k-1} G\left((k-i)(n-4), 0\right) \\ &= q^{6\ell} \cdot \frac{1-q^{(k-2\ell)n-4\ell+1}}{1-q} \cdot \frac{(1-q^{\ell n-4\ell+1})(1-q^{\ell n-4\ell+2})}{(1-q)(1-q^{2})}. \end{split}$$

We continue by analyzing $[\ell^2, k - 2\ell]$, for $\frac{k}{2} > \ell > \frac{k}{3}$. Then $d_\ell = 2$, $d_{k-2\ell} = 1$. $n \ge 4$ is needed as usual.

$$\begin{split} q^{k\left(\sum_{i=1}^{k}d_{i}\right)-k-\sum_{1\leq j< i\leq k}(i-j)d_{i}d_{j}}\prod_{i=0}^{k-1}G\left((k-i)n-2i+2\sum_{j=0}^{i-1}(i-j)d_{k-j},d_{k-i}\right)\\ &=q^{4k-6\ell}\prod_{i=0}^{k-\ell-1}G\left((k-i)n-2i,0\right)\\ G\left(\ell n-2k+2\ell,2\right)\\ &\prod_{i=k-\ell+1}^{2\ell-1}G\left((k-i)n+2i-4(k-\ell),0\right)\\ G\left((k-2\ell)n+8\ell-4k,1\right)\\ &\prod_{i=2\ell+1}^{k-1}G\left((k-i)(n-4),0\right)\\ &=q^{4k-6\ell}\cdot\frac{\left(1-q^{\ell n-2k+2\ell+1}\right)\left(1-q^{\ell n-2k+2\ell+2}\right)}{\left(1-q\right)\left(1-q^{2}\right)}\cdot\frac{1-q^{(k-2\ell)n+8\ell-4k+1}}{1-q}. \end{split}$$

Finally, consider $\left[\frac{k}{3}^3\right]$ for $k \cong 0 \pmod{3}$. Then $d_{k/3} = 3$. Again, we need $n \ge 4$.

$$q^{k(\sum_{i=1}^{k} d_{i})-k-\sum_{1\leq j< i\leq k}(i-j)d_{i}d_{j}}\prod_{i=0}^{k-1}G\left((k-i)n-2i+2\sum_{j=0}^{i-1}(i-j)d_{k-j},d_{k-i}\right)$$

$$=q^{2k}\prod_{i=0}^{2k/3-1}G\left((k-i)n-2i,0\right)$$

$$G\left((n-4)k/3,3\right)$$

$$\prod_{i=2k/3+1}^{k-1}G\left((k-i)(n-4),0\right)$$

$$=q^{2k}\frac{(1-q^{(n-4)k/3+1})(1-q^{(n-4)k/3+2})(1-q^{(n-4)k/3+3})}{(1-q)(1-q^{2})(1-q^{3})}.$$

Lemma 2.27. *For* $n \ge 2$ *,*

$$G_{3}(n,k) = \begin{bmatrix} q^{4k} \cdot q^{3} \left(1-q^{n}\right) \left(q-q^{n}\right) & - q^{3k} \cdot q \left(1+q\right) \left(1-q^{n}\right) \left(q-q^{2}+q^{5}-q^{n}\right) \\ - q^{2k} \left(q^{nk} \left(q^{2n} \left(q^{9}-q^{8}-q^{7}+q^{6}+q^{5}-q^{3}+q\right)-q^{n} \left(q^{10}-q^{8}+q^{6}+q^{5}\right)+q^{10}\right) \\ - q^{2n}+q^{n} \left(q^{5}+q^{4}-q^{2}+1\right)-q^{9}+q^{7}-q^{5}-q^{4}+q^{3}+q^{2}-q \end{bmatrix} \\ + q^{k} \cdot q^{nk} q^{3} \left(1+q\right) \left(1-q^{n}\right) \left(q^{5}-q^{n}+q^{n+3}-q^{n+4}\right)-q^{nk} q^{6} \left(1-q^{n}\right) \left(q-q^{n}\right) \end{bmatrix} \\ \left/ \left(1-q\right)^{2} \left(1-q^{2}\right)^{2} \left(1-q^{3}\right) \left(1-q^{n-1}\right) \left(1-q^{n}\right) q^{2k+1}. \end{aligned}$$

$$(2.10)$$

For n < 2, $G_3(n, k) = G_2(n, k)$.

Proof. For $n \ge 4$,

$$\begin{split} G_{3}(n,k) &= G_{2}(n,k) + \sum_{\ell_{3}=1}^{\lfloor k/3 \rfloor - 1} \sum_{\ell_{2}=\ell_{3}+1}^{\lfloor (k-\ell_{3}-1)/2 \rfloor} q^{2(\ell_{2}+2\ell_{3})} \cdot \frac{1 - q^{(k-\ell_{3}-\ell_{2})n-2(\ell_{2}+\ell_{3})+1}}{1 - q} \\ &\quad \cdot \frac{1 - q^{\ell_{2}n-2(\ell_{2}+\ell_{3})+1}}{1 - q} \cdot \frac{1 - q^{\ell_{3}(n-4)+1}}{1 - q} \\ &\quad + \sum_{\ell=1}^{\lfloor (k-1)/3 \rfloor} q^{6\ell} \cdot \frac{1 - q^{(k-2\ell)n-4\ell+1}}{1 - q} \cdot \frac{(1 - q^{\ell n-4\ell+1})(1 - q^{\ell n-4\ell+2})}{(1 - q)(1 - q^{2})} \\ &\quad + \sum_{\ell=\lceil (k+1)/3 \rceil}^{\lfloor (k-1)/2 \rfloor} q^{4k-6\ell} \cdot \frac{(1 - q^{\ell n-2k+2\ell+1})(1 - q^{\ell n-2k+2\ell+2})}{(1 - q)(1 - q^{2})} \cdot \frac{1 - q^{(k-2\ell)n+8\ell-4k+1}}{1 - q}. \end{split}$$

The first nested summation is for the distinct size 3 partitions. The other 2 sums are for partitions of type $[k - 2\ell, \ell, \ell]$ and $[\ell, \ell, k - 2\ell]$, respectively. If $k \cong 0 \pmod{3}$ then, for the

 $\left[\frac{k}{3}^3\right]$ partition, we need to add the extra term:

$$q^{2k} \frac{(1-q^{(n-4)k/3+1})(1-q^{(n-4)k/3+2})(1-q^{(n-4)k/3+3})}{(1-q)(1-q^2)(1-q^3)}$$

Amazingly, Maple is able to simplify the above expression into closed form once one makes assumptions about $k \pmod{6}$. It is also necessary to divide the first sum into 2 parts based on the parity of l_3 . In the divided sum, we are replacing ℓ_3 by $2\ell_3 - 1$ and $2\ell_3$, respectively.

$$\begin{split} \sum_{\ell_3=1}^{\lfloor (\lfloor k/3 \rfloor -1)/2 \rfloor} \Bigg[\sum_{\ell_2=2\ell_3}^{\lfloor k/2 \rfloor -\ell_3} q^{2(\ell_2+2(2\ell_3-1))} \cdot \frac{1-q^{(k-(2\ell_3-1)-\ell_2)n-2(\ell_2+2\ell_3)+3}}{1-q} \\ & \cdot \frac{1-q^{\ell_2n-2(\ell_2+2\ell_3)+3}}{1-q} \cdot \frac{1-q^{(2\ell_3-1)(n-4)+1}}{1-q} \\ & + \sum_{\ell_2=2\ell_3+1}^{\lfloor (k-1)/2 \rfloor -\ell_3} q^{2(\ell_2+4\ell_3)} \cdot \frac{1-q^{(k-2\ell_3-\ell_2)n-2(\ell_2+2\ell_3)+1}}{1-q} \\ & \cdot \frac{1-q^{\ell_2n-2(\ell_2+2\ell_3)+1}}{1-q} \cdot \frac{1-q^{2\ell_3(n-4)+1}}{1-q} \Bigg]. \end{split}$$

If $k \cong 0, 1, 2 \pmod{6}$ then $\lfloor \frac{k}{3} \rfloor - 1 \cong 1 \pmod{2}$ and so we need to add the final inner sum when $\ell_3 = \lfloor \frac{k}{3} \rfloor - 1$:

$$\sum_{\ell_{2}=\lfloor k/3 \rfloor}^{\lfloor (k-\lfloor k/3 \rfloor)/2 \rfloor} q^{2(\ell_{2}+2(\lfloor k/3 \rfloor-1))} \cdot \frac{1-q^{(k-\ell_{2}-\lfloor k/3 \rfloor+1)n-2(\ell_{2}+\lfloor k/3 \rfloor)+3}}{1-q} \cdot \frac{1-q^{\ell_{2}n-2(\ell_{2}+\lfloor k/3 \rfloor)+3}}{1-q} \cdot \frac{1-q^{(\lfloor k/3 \rfloor-1)(n-4)+1}}{1-q}.$$

For each $k \cong 0, \ldots, 5 \pmod{6}$, the decomposition simplifies to the same result: Eqn. (2.10).²⁴

If n < 4, then the additions from size 3 partition calls will all be 0 (see the characterization of size 3 partition calls above) and so $G_3(n,k) = G_2(n,k)$. It is a simple exercise to check that the difference evaluates to 0 when n = 2, 3. The difference between the two

²⁴This was only fully done for the $k \cong 5 \pmod{6}$ case. However, the other cases were partly simplified and found to be in agreement empirically; after removing the assumption on $k \pmod{6}$, all cases still gave the same result. And the explicit formula in Eqn. (2.10) matches the polynomials found by recursion for all $n = 2, \ldots, 30$ and $k = 0, \ldots 30$. If the reader does not wish to take our word for it, they are encouraged to show the other cases for themselves. We highly recommend using Maple or another computer assistant if you do not have time to waste.

polynomials is given by

$$G_{3}(n,k) - G_{2}(n,k) = \left[q^{4k} \cdot q^{2} \left(1 - q^{n}\right) \left(q - q^{n}\right) - q^{3k} \cdot q \left(1 + q\right) \left(1 - q^{n}\right) \left(q^{n+3} - q^{n+2} - q^{n} + q^{3}\right) + q^{2k} \cdot q \left(q^{3} - q^{n}\right) \left(q^{nk} \left(q^{n+1} - q^{3} - q^{2} + 1\right) + q^{n} (q^{3} - q - 1) + q^{2}\right) - q^{k} \cdot q^{nk} q^{3} \left(1 + q\right) \left(1 - q^{n}\right) \left(1 - q - q^{3} + q^{n}\right) - q^{nk} q^{5} \left(1 - q^{n}\right) \left(q - q^{n}\right)\right] / (1 - q)^{2} (1 - q^{2})^{2} (1 - q^{3}) (1 - q^{n-1}) (1 - q^{n}) q^{2k}.$$

We can use this result to verify the recurrence relation in Eqn. (2.5) for s = 3. To compute $G_3(n,k)$ for numeric n, k using the explicit expression, use **G3(n,k,q)**.

Because we will always have $k' \leq s$, it is possible to automate the process of finding $G_s(n,k)$ for any s and have it terminate in finite time. Characterize all of the new size s partitions and add their contributions to $G_{s-1}(n,k)$. However, the number of new types of partitions to consider grows exponentially as 2^{s-1} ; each new part is either the same as the previous part, or smaller. Also, the partitions become more and more complex to describe. One general partition that we can tackle is $\left[\frac{k^{\ell}}{\ell}\right]$. Then $d_{k/\ell} = \ell$ and the recursive call is

$$\begin{split} q^{k\left(\sum_{i=1}^{k} d_{i}\right)-k-\sum_{1\leq j< i\leq k}(i-j)d_{i}d_{j}} \prod_{i=0}^{k-1} G\left((k-i)n-2i+2\sum_{j=0}^{i-1}(i-j)d_{k-j}, d_{k-i}\right) \\ &= q^{k(\ell-1)} \prod_{i=0}^{k(\ell-1)/\ell-1} G\left((k-i)n-2i,0\right) \\ G\left(\frac{k}{\ell}n-2k\frac{\ell-1}{\ell},\ell\right) \\ &= \prod_{i=k(\ell-1)/\ell+1}^{k-1} G\left((k-i)n-2i+2(i-k(\ell-1)/\ell)\ell,0\right) \\ &= q^{k(\ell-1)} G\left(\frac{k}{\ell}(n-2(\ell-1)),\ell\right), \end{split}$$

which to be nonzero requires that $n \ge 2(\ell - 1)$.

The final functions found in Lemmas 2.26 and 2.27 are by no means obviously unimodal. If one instead writes them in decomposed form, they can be reasoned to be unimodal by looking at the darga of each term. Knowing the proper decomposition beforehand allows for easy proof of unimodality, but finding any decomposition is typically extremely difficult. This is the motivation behind reverse engineering.

2.5.2 Adjusted Initial Conditions

We can also multiply each recursive call by constant factors or multiply initial conditions while still maintaining base case darga = nk.

$$G(n < 0, k) = 0, \quad G(n, k < 0) = 0, \quad G(0, k) = \nu, \quad G(n, 0) = \nu, \quad G(n, 1) = \nu \frac{1 - q^{n+1}}{1 - q}.$$
(2.11)

We can also adjust the recursive call on each $\lambda \vdash k$ by multiplying it by ρ . ρ need not be constant; it can be a function of $|\lambda|$, the largest part of λ , or even more generally, each λ gets its own specific value. We choose to "normalize" the final answer by dividing by ν^k (and ρ when it is constant) so that we do not have arbitrary differences. Even after normalizing, there will still be differences from G(n, k).

Though we have not yet found meaningful results using the techniques of this section, we are hopeful that in the future we can find a pattern for the difference between the largest and second-largest coefficients.

2.5.3 Adjusted Recursive Call

A more complex method is to adjust the recursive call. We return to using the initial conditions in Eqn. (2.4).

$$G(n,k;a,b,c) = \sum_{\substack{(d_1,\dots,d_k);\sum_{i=1}^k id_i = k}} q^{k\left(\sum_{i=1}^k d_i\right) - k - \sum_{1 \le j < i \le k} (i-j)d_id_j} \cdot q^{ka + \left(\sum_{i=1}^k d_i\right)b + c[k(k+1)(n/2-a) - k(k+2b) + \sum_{i=1}^k d_ii^2]}} \prod_{i=0}^{k-1} G\left((k-i)n - 2i + 2\sum_{j=0}^{i-1} (i-j)d_{k-j} - 2a(k-i) - 2b, d_{k-i} - 2c; a, b, c\right).$$

The reduction in darga in the recursive call is exactly balanced by the extra factors of q. However, if c > 0, since there is some i s.t. $d_{k-i} = 0$ for any partition, the product will be 0 by initial conditions, and thus G(n, k; a, b, c > 0) = 0. And if c < 0, we obtain infinite recursion for nontrivial choices of n, k. Therefore, we should ignore this parameter.

$$G(n,k;a,b) = \sum_{\substack{(d_1,\dots,d_k);\sum_{i=1}^k id_i = k}} q^{(k+b)\left(\sum_{i=1}^k d_i\right) + k(a-1) - \sum_{1 \le j < i \le k} (i-j)d_id_j}$$
$$\prod_{i=0}^{k-1} G\left((k-i)(n-2a) - 2(i+b) + 2\sum_{j=0}^{i-1} (i-j)d_{k-j}, d_{k-i};a,b\right). \quad (2.12)$$

IMPLEMENTED AS KOHgeneral.

Unless otherwise stated, we are now assuming the type of partition is unrestricted.

Proposition 2.28. 1. G(n,k;a,b) will have smallest degree ka + b.

2. If $a + b > \frac{n}{2}$, then G(n, k; a, b) = 0. 3. If $a + b = \frac{n}{2}$, then $G(n, k; a, b) = q^{ka+b} \frac{1-q^{nk+1-2(ka+b)}}{1-q}$.

Proof. 1. This follows from inspecting factors of q. Note that $k\left(\sum_{i=1}^{k} d_i\right) - k - \sum_{1 \le j < i \le k} (i-j) d_i d_j \ge 0$ since G(n,k) has smallest degree 1. Recall it = 0 for $\lambda = [k]$.

2. Let $\lambda \vdash k$. Consider n'(k-1) in the recursive call:

$$n'(k-1) = (k - (k - 1))(n - 2a) - 2((k - 1) + b) + 2\sum_{j=0}^{(k-1)-1} ((k - 1) - j)d_{k-j}$$

= (1)(n - 2a) - 2(k - 1 + b) + 2 $\sum_{j=0}^{k-1} (k - 1 - j)d_{k-j}$
= n - 2a - 2k + 2 - 2b + 2 $\sum_{j=1}^{k} (k - 1 - (k - j))d_j$
= n - 2a - 2b + 2 + $\left(-2k + 2\sum_{j=1}^{k} jd_j\right) - 2\sum_{j=1}^{k} d_j$
= n - 2(a + b) + 2 - 2 $\sum_{j=1}^{k} d_j$
 $\leq n - 2(a + b) + 2 - 2(1) < 0.$

Thus, $G(n'(k-1), d_1; a, b) = 0$ and since λ was arbitrary, G(n, k; a, b) = 0.

3. The above has equality only if $|\lambda|=1,$ i.e., $\lambda=[k].$ Then

$$\begin{split} G(n,k;a,b) &= q^{(k+b)\left(\sum_{i=1}^{k} d_i\right) + k(a-1) - \sum_{1 \le j < i \le k} (i-j)d_i d_j} \\ &\prod_{i=0}^{k-1} G\left((k-i)(n-2a) - 2(i+b) + 2\sum_{j=0}^{i-1} (i-j)d_{k-j}, d_{k-i}; a, b \right) \\ &= q^{ka+b} G\left((k)(n-2a) - 2(b), 1; a, b \right) \\ &\prod_{i=1}^{k-1} G\left((k-i)(n-2a) - 2(i+b) + 2\sum_{j=0}^{i-1} (i-j)d_{k-j}, 0; a, b \right) \\ &= q^{ka+b} G\left(k(n-2a) - 2b, 1; a, b \right) \\ &\prod_{i=1}^{k-1} G\left((k-i)(n-2a) - 2(i+b) + 2(i-0), 0; a, b \right). \end{split}$$

It only remains to confirm $(k - i)(n - 2a) - 2b \ge 0$, which follows from $a = \frac{n}{2} - b$.

2.6 OEIS

When we take $q \to 1^-$ for the normal q-binomials, we obtain the common binomial coefficients. What happens if we do that to the modified G_s polynomials? There is no direct

combinatorial interpretation (yet), but we can derive some cases to start the search. Recall the explicit forms of $G_1(n,k)$, $G_2(n,k)$, and $G_3(n,k)$ in Eqns. (2.6), (2.9), and (2.10), respectively.

$$\begin{split} \lim_{q \to 1^{-}} G_1(n,k) &= nk+1, \\ \lim_{q \to 1^{-}} G_2(n,k) &= \frac{1}{12}(k+1)(k^2n^2 - 3k^2n - kn^2 + 2k^2 + 9kn - 8k + 12), \\ \lim_{q \to 1^{-}} G_3(n,k) &= \frac{1}{720}(k+2)(k+1)\bigg(k^3n^3 - 9k^3n^2 - 3k^2n^3 + 26k^3n + 42k^2n^2 + 2kn^3 - 24k^3 - 153k^2n - 33kn^2 + 162k^2 + 247kn - 378k + 360\bigg). \end{split}$$

We can also conjecture the form of $G_4(n,n)$ and $G_5(n,n)$ from experimental data:

$$\lim_{q \to 1^{-}} G_4(n,n) = \frac{1}{120960} n(n+2)(n+1) \left(n^8 - 32n^7 + 462n^6 - 3836n^5 + 20013n^4 - 66836n^3 + 140804n^2 - 171216n + 100800 \right),$$
$$\lim_{q \to 1^{-}} G_5(n,n) = \frac{n(n+3)(n+2)(n+1)}{43545600} \left(n^{10} - 50n^9 + 1140n^8 - 15420n^7 + 136533n^6 - 824370n^5 + 3436190n^4 - 9762880n^3 + 18198936n^2 - 20242080n + 10886400 \right).$$

The sequences for $G_1(n, n), \ldots, G_5(n, n)$ are now in the OEIS [OEIa, OEIb, OEIc, OEId, OEIe]. Only s = 1 was already in the OEIS.

Table 2.1: Unimodal Sequences from the OEIS

<u>A002522</u> :	s = 1	1, 2, 5, 10, 17, 26, 37, 50, 65, 82, 101, 122, 145, 170, 197, 226, 257, 290,
<u>A302612</u> :	s = 2	1, 2, 6, 20, 65, 186, 462, 1016, 2025, 3730, 6446, 10572, 16601, 25130,
<u>A302644</u> :	s = 3	1, 2, 6, 20, 70, 252, 896, 2976, 8955, 24310, 60038, 136500, 289016,
A302645:	s = 4	1, 2, 6, 20, 70, 252, 924, 3432, 12705, 45430, 152438, 472836, 1352078,
<u>A302646</u> :	s = 5	1, 2, 6, 20, 70, 252, 924, 3432, 12870, 48620, 183755, 683046, 2443168.

[H] The sequences will approach the central binomial coefficients <u>A000984</u> [OEIf] since $G_s \to G_\infty = G$.

2.7 Conclusion and Future Work

In this work we found a new way to generate unimodal polynomials for which unimodality is, *a priori*, very difficult to prove. Many of the recurrences found produce surprisingly unimodal polynomials given the correct initial conditions. The proof is "trivial" because of our method of reverse engineering. There are many promising avenues for future work.

- We were not able to make interesting use of the techniques in Section 5.2. With much more care, it may be possible to glean some interesting results by adjusting initial conditions.
- 2. An ultimate goal would be to find a restriction on the partitions and other parameters that yields non-trivial polynomials that can be expressed as a function in 2 variables. We have a few for the restricted partition size, but more is always better!
- 3. Find other initial conditions that still yield unimodal polynomials when iterated in the recurrence of Eqn. (2.5).
- 4. Try to fully automate the process of finding $G_s(n,k)$ in Section 2.5.1. Use them to verify Eqn. (2.5) for larger values of s. The ideal result would be to prove Eqn. (2.5) for ALL s.
- 5. I also tried looking at how restricting to odd partitions compares to restricting to distinct partitions. As these are conjugate sets, I thought there may be a relation in the resulting polynomials. Alas, I could see no connection. One could reexamine these pairs or look at other partition sets that are known to be equinumerous, e.g., odd and distinct partitions compared to self-congruent partitions.
- An appealing result would be to find combinatorial interpretations for any of the restricted polynomials. Particularly the G_s polynomials and the OEIS sequences <u>A302612</u> [OEIb], <u>A302644</u> [OEIc], <u>A302645</u> [OEId], and <u>A302646</u> [OEIe].
- 7. Instead of only using an adjusted Eqn. (2.12), one might be able to "arbitrarily" combine polynomials of known darga to create another polynomial with known darga. Do this in a similar recursive manner. Instead of using partitions, are there other

combinatorial objects that can be applied? What other unimodal and symmetric polynomials are out there to use as starting blocks? What recurrences do they satisfy that can be tweaked to reverse-engineer unimodal polynomials?

- 8. Unimodal probability distributions have special properties. One could use these polynomials for that purpose after normalizing by G'(n, k; q = 1). Various properties can be discovered using Gauss's inequality [Gau23] or others.
- 9. Unimodal polynomials can be constructed from multiplication very easily. Is the inverse process of factoring them easier than for general polynomials? If so, there could be applications in cryptography and coding theory where factoring is a common theme.

Chapter 3 Lattice Walk Enumeration

3.1 Introduction

This chapter is available as a stand-alone paper on arXiv.org, number 1803.10920. It is being submitted to the *Journal of Integer Sequences*.

Trying to enumerate all of the walks in a 2D lattice is a fun combinatorial problem and there are numerous applications, from polymers to sports. Computers provide a wonderful tool for analyzing these walks; we provide a Maple package for automatically describing generating functions of walks restricted to any step set in a 2D lattice. We always obtain a closed system of relations for generating functions of walks that are bounded, semibounded, or unbounded. For bounded walks, this leads to explicit rational solutions! For semi-bounded or unbounded walks, we may get lucky and obtain algebraic solutions; if not, we still have a short self-referential description of the generating function.

We consider walks in the two-dimensional square lattice with an "arbitrary" set of integral steps (x, y) subject to $x \ge 0$. In addition to unbounded walks, we also separately constrain the walks to lie in regions bounded above and below as well as bounded only below.

One would then like to count all possible walks of a certain length possibly with a specific total change in y value. Rather than a brute-force search of the entire space, looking for 1 value, one could use generating function relations. As a bonus, one would obtain not only the initial generating function of desire, but also many related ones that may be of interest.

Studies of this kind have been done in the literature with simple step sets such as Dyck paths ($\{[1, 1], [1, -1]\}$) [DR71, BORW05] and old-time basketball games [AZ07]. Philippe Duchon analyzed the case of nonnegative bridges with step set $\{[1, -2], [1, 3]\}$; see OEIS¹ sequence <u>A060941</u> [OEIg]. For further developments on the subject, see [vR00] and the references therein.

The accompanying Maple package is able to extend and inform on old sequences and create many new sequences. Much of the analysis, thus far, has been on steps with x-value exactly equal to 1. One of the aspects of this paper that sets it apart is the ease with which it can analyze more generic cases.

3.1.1 Motivation

We want to enumerate walking paths constrained to specific allowable steps. Most of the time we are looking for paths that begin and end on the x-axis.

An earlier motivation for bounded walks came from Physics: analyzing polymers constrained between plates [BORW05]. Ayyer and Zeilberger gave one solution in an earlier paper [AZ07] that provided the main motivation for this research.

The kernel method has received attention lately for analyzing specific cases of walks [BKK⁺17]. There are several advantages to describing walks using the method of this paper. The main idea is the same: writing functional equations to describe possible steps in a walk. The difference is that this method then describes "new" components of the functional equation in an iterative manner. The kernel method uses analytical number theory on the roots and can be reliant on very case-specific techniques. Compared to the kernel method, we believe our method is a lot easier to understand combinatorially, is more insightful, faster, and easier to produce.

Trying to picture an entire walk at once can be difficult. This is where the awesome powers of dynamical programming come into play. Instead of trying to think about the entirety of a walk, think about a part: either the beginning step, end step, or the middle step

¹Online Encyclopedia of Integer Sequences [Nei].

across the x-axis. Break the walk down into different parts (irreducible versus reducible). This is what the generating function equations accomplish.

Originally, we wrote out a single equation to describe the initial walk of interest, then the next, then the next, until it eventually became a closed system. (And the wonderful part is that our descriptions ALWAYS lead to closed systems.) Finally, we solved the system we created. Since this is a very algorithmic approach to answering the question, why not have a computer work for us?

3.1.2 Definitions

I will generically use the term walk to indicate any sequence of points $\{(x_0, y_0), \ldots, (x_s, y_s)\}$ in the xy-plane. Though the walk will not necessarily start at the origin, if the starting point is not given, it is assumed to begin at (0,0). The steps of a walk are $\{(x_1 - x_0, y_1 - y_0), \ldots, (x_s - x_{s-1}, y_s - y_{s-1})\}$ and are built from some step set S, a set of ordered pairs. I exclusively consider walks that are monotonically weakly increasing: $x_{i+1} - x_i \ge 0$. A walk is nonnegative (nonpositive) if the walk never crosses below (above) the x-axis. I will sometimes refer to the y-value as the altitude of the walk.

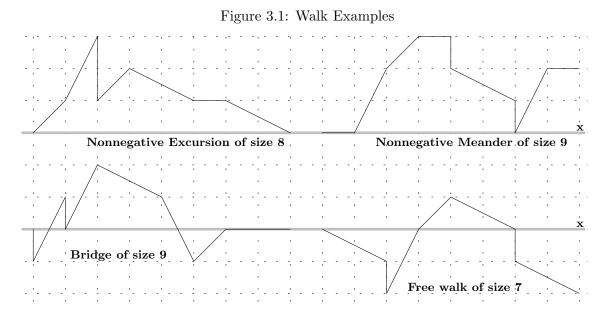
Definition 3.1 (Walks). A bridge is an unbounded walk that begins at the origin and ends on the x-axis. I say bounded bridge for a bounded walk that begins at the origin and ends on the x-axis.

An excursion is a semi-bounded (not necessarily nonnegative) walk that begins at the origin and ends on the x-axis.

A free walk can end anywhere (not ending at any specific altitude). A meander is a semibounded free walk.

The length of a walk is $n = \sum_{i=0}^{s} x_i$. The size of a walk is s.

See Figure 3.1 for examples of the different types of walks.



All of the example walks are of length 8 and are considered to have begun at the origin. They are all built from the same step set $S = \{[0, -1], [1, 0], [1, 1], [1, 2], [2, -1]\}$. Note the vertical line in the excursion is actually 2 steps. The excursion happens to be irreducible while the rest are not.

Banderier, *et al.* [BKK⁺17] uses *walk/path* to reference any sequence of steps that start at the origin. The only difference in definitions is that they always count excursions and meanders as nonnegative. I will mostly consider them as nonnegative and explicitly say when I am not.

Definition 3.2. The interior of a walk consists of every point other than the endpoints: $\{(x_1, y_1), \ldots, (x_{s-1}, y_{s-1})\}.$

An irreducible walk is one in which the interior has a strictly higher altitude than the lower endpoint: $\min\{y_1, \ldots, y_{s-1}\} > \min\{y_0, y_s\}$. For purposes of this paper and Maple package, the stationary walk² and walks that are direct steps to the right are NOT considered to be irreducible walks.³

Irreducible is also used to refer to walks that do not exactly hit the final altitude until the final step.

 $^{^{2}}A$ single point.

 $^{^{3}}As$ the interior (the edge between points) is not strictly higher than the endpoints.

3.1.3 Chapter Organization

I will exclusively use t as the variable in generating functions. I also abbreviate generating function(s) as g.f.(s). This chapter is organized in the following sections:

- Section 2: Bounded: This is the section in which a computer does the best; it can give an exact g.f. solution since we have a "linear" system of equations. Computers are very good at solving these quickly and efficiently.
- Section 3: Semi-bounded: An algebraic expression for the g.f. is no longer guaranteed. However, we can always find a polynomial in $\mathbb{Z}[t]$ for which the g.f. is a root.⁴
- Section 4: Guess-and-check: Finding a minimal polynomial is guaranteed so why not just guess? This section gives a time and memory comparison showing why that would be a poor decision.
- Section 5: Algebraic-to-Recursive: Using the minimal polynomial of the g.f. is not always the fastest for enumerating. This section introduces another Maple package that converts the polynomial into a 1D recurrence.
- Section 6: Unbounded: We find minimal polynomials for g.f.s of unbounded walks and show an alternative method of producing recurrences in specific cases.
- Section 7: Asymptotics: We discover asymptotic results for several step sets to show relationships between the number of excursions, bridges, and, to a lesser degree, meanders.
- Section 8: Applications: We use the Maple package to find some extended results and talk about probabilistic behavior. We also express minimal polynomials for excursions of small step sets.
- Section 9: Conclusions and Future Work: A brief recap of what we can do with the ScoringPaths Maple package and how we can extend the work.

⁴In the formal power series sense.

This paper is produced in conjunction with the Maple package **ScoringPaths**. It is down-loadable from

$math.rutgers.edu/{\sim}bte14/Code/ScoringPaths/ScoringPaths.txt.$

I will mention various functions of the package in **bold**. All functions mentioned in this paper are included in the accompanying Maple package. Some functions have been borrowed from other packages with credit and included in **ScoringPaths** for completeness. All comparisons of time and memory are done with Maple 17's **CodeTools[Usage]** on Linux version 3.10.0-514.el7.x86_64 with 8GB of RAM. All values are averages of at least 30 trials unless otherwise specified. Pay careful attention to the units in some examples.

We are considering discrete walks. As such, we can consider steps with only integral values. If the x-steps are fractional, there will likely not be any issues. If the y-steps are fractional, many functions will not work as intended. If all of the steps share a common factor in the y-value, it should be factored out leaving an equivalent problem. The bounds will need to be factored and truncated as well. Leaving the common factor may cause problems with some functions.

The g.f. may produce non-zero coefficients only every m^{th} value. It may be desirable to make the substitution $t \to t^{1/m}$ for ease of reading. I commonly use B(n), and occasionally C(n) as the number of walks of length n. The step set and walk restrictions will be obvious from context or given explicitly.

A few functions to create sets of random steps have been included for quick demonstrations; **RandomStepSet** produces a generic set, **RandomZeroStepSet** parses for walks that begin and end on the x-axis, **RandomSemiBoundedStepSet** parses for walks bounded below, and **RandomUnBoundedStepSet** produces a step set without any [0, y]steps.

There are several "paper" functions that automatically produce an article with information about a given step set. See the "Paper-Producing Functions" section of the **Help** function.

The package gfun by Salvy and Zimmermann [SZ94], a staple included with recent

Maple versions, is also very useful for manipulating g.f.s. It contains many functions for translating between algebraic expressions, recurrences, and differential equations satisfied by g.f.s.

3.2 Bounded

The first case is walks that are bounded above and below. Consider an arbitrary set of steps $S = \{(x_1, y_1), \ldots, (x_n, y_n)\}$. The goal is to find the g.f., denoted $f_{a,b}$, for walks with step set S, starting at (0,0), and bounded above and below by $a \ge 0$ and $b \le 0$, respectively. All walks/paths in this section are constrained to a step set S and bounded above and below.

3.2.1 Walking "Anywhere"

First assume that the walk can end anywhere between the lines y = a and y = b. A walk either never goes anywhere, +1, or it takes a step (t^x) and continues as if it is a walk starting at a new point: $f_{a-y,b-y}$. The following relation accomplishes that.

$$f_{a,b} = 1 + \sum_{(x,y)\in\mathcal{S}} t^x f_{a-y,b-y}.$$

This by itself gets us nowhere. But if a - y < 0 or b - y > 0, then $f_{a-y,b-y} = 0$ since it is already starting in a prohibited region. Now write out all (a - b) equations:

$$f_{0,b-a} = 1 + \sum_{(x,y)\in\mathcal{S}} t^x f_{0-y,b-a-y},$$

$$f_{1,b-a+1} = 1 + \sum_{(x,y)\in\mathcal{S}} t^x f_{1-y,b-a+1-y},$$

$$\vdots$$

$$f_{a-b,0} = 1 + \sum_{(x,y)\in\mathcal{S}} t^x f_{a-b-y,0-y}.$$

This is a (linear!) system of a - b equations with a - b variables, after discarding the constant 0 g.f.s, for which Maple's solve function can easily find the solution. As a bonus, we not only have $f_{a,b}$, but also every $f_{m,n}$ such that m - n = a - b (and $m \ge 0, n \le 0$).

The function **BoundedScoringPathsEqsVars** will produce all of the equations and variables for this system extremely quickly. **BoundedScoringPaths** will output only the g.f. $f_{a,b}$. Again, this is very fast since it is solving a linear system of only (a - b) equations. To verify the values, one can use **BoundedScoringPathsNumber**, which computes the number of walks by recursion.

Allowable steps are any as long as there is not a (0, -) AND a (0, +) such that they can sum to 0 within the bounds. Specifically, if (0, m) and (0, -n) are steps, then it is allowable (will not produce infinite values) as long as the width is small enough: a - b < n + m - gcd(m, n).

Proof. Let $S = \{[0,m], [0,-n]\}$ and assume m > n. Also assume gcd(m,n) = 1. If $a-b \ge n+m-1$, then we can always take at least one of the S steps. Since there is a finite number of altitudes, then must be acollision at some point: a loop.

If gcd(m,n) = g > 1, then reconsider the same problem after factoring g out of everything: $a' = \left\lfloor \frac{a}{g} \right\rfloor, b' = \left\lceil \frac{b}{g} \right\rceil, S' = \{[0, m/g], [0, -n/g]\}.$

This bound only works for two 0-steps. For more than two 0-steps, the width must be even shorter. For some combinations (e.g., $\{[0, u - v], [0, -u], [0, u + v]\}$), there is no allowable width that includes all steps.⁵

Examples

Example 3.3 (Close (American) Football Games). Consider trying to enumerate the number of American football games in which the teams are never separated by more than 1 score. On a given play,⁶ it is possible for one team to score 2, 3, 6, 7, or 8 points. It is also possible for one team to score 6 points and the opposing team to score 1 or 2 points, though these are pretty rare occurrences. What if we want to enumerate the number of games with n

⁵Assuming that one can get to the edges of the boundary. (u+v) + (-u) + (u-v) + (-u) = 0: a loop.

scoring plays that end separated by no more than 1 score? We can use the step set

$$\begin{split} \mathcal{S} &= \{ [1,2], [1,3], [1,6], [1,7], [1,8], [1,5], [1,4], \\ & \quad [1,-2], [1,-3], [1,-6], [1,-7], [1,-8], [1,-5], [1,-4] \} \end{split}$$

Each step represents 1 scoring play and how many points the home team gained relative to the away team. Since the maximum scoring play is 8 points, we make the bounds y = -8, 8. The g.f. is then found with **BoundedScoringPaths**(S, 8, -8, t):

$$\frac{1+10t+13t^2-37t^3-40t^4+28t^5+26t^6-2t^7}{1-4t-59t^2-77t^3+170t^4+234t^5-92t^6-142t^7-4t^8+6t^9}$$

This sequence is new in the OEIS: A301379 [OEIh].

Example 3.4 (Speed Enumeration - Bounded 1). Consider finding the first 1000 coefficients of the g.f. found in Example 3.3 above. We could use brute-force recursion in **BoundedScoringPathsNumber** or take a taylor series expansion since we have an explicit form for the g.f. from **BoundedScoringPaths**.

Method	Memory Used	Memory Allocation	CPU Time	Real Time
Brute-Force Recursion	$60.23 \mathrm{MiB}$	24MiB	$309.53 \mathrm{ms}$	$308.90 \mathrm{ms}$
G.F. Construction	3.27MiB	0 bytes	$30.17 \mathrm{ms}$	$33.50 \mathrm{ms}$
Taylor Enumeration	4.86MiB	0 bytes	$4.93 \mathrm{ms}$	$5.07 \mathrm{ms}$
Total	8.13MiB	0 bytes	35.10ms	38.57ms

Table 3.1: Bounded Walk Enumeration

Using the g.f. method of enumeration is about 8 times as fast and uses a much smaller amount of memory: (1/8th).

3.2.2 Returning to the *x*-axis

The above equations are for g.f.s of walks that are able to end ANYWHERE. What if we want a g.f. for walks that must end back at the x-axis? Ayyer and Zeilberger solved the bounded bridge case (but not the bounded free walk case) using different relations but

the same general method [AZ08]; they also use the equivalent problem of walks between parallel lines of positive slope instead of between horizontal lines.

Let $f_{a,b}$ now denote the g.f. for walks that begin at (0,0) and end on the x-axis. Again, we either never get moving, +1, or we take a step and then must take a path back to the x-axis. Let $e_{a,b,c}$ denote the g.f. for paths that start at (0,c), end on the x-axis and never touch the x-axis beforehand. Now that $e_{a,b,c}$ is introduced, we can write the relation for $f_{a,b}$:

$$f_{a,b} = 1 + \left(\sum_{(x,0)\in\mathcal{S}} t^x\right) f_{a,b} + \left(\sum_{(x,y)\in\mathcal{S}; y\neq 0} t^x e_{a,b,y}\right) f_{a,b}.$$
(3.1)

There is also the possibility of just moving along the x-axis to start. After returning to the x-axis, we can now take any walk as before. Hence the multiplication by $f_{a,b}$. We now need the equations for $e_{a,b,c}$ for $a \ge c \ge b$. If any step would return to the x-axis, then we are done. Otherwise, it must continue as a new $e_{a,b,c'}$ path.

$$e_{a,b,c} = \sum_{(x,-c)\in\mathcal{S}} t^x + \sum_{(x,y)\in\mathcal{S}; y\neq -c} t^x e_{a,b,c+y}.$$

We now have a set of linear equations to solve for the $e_{a,b,c}$. To produce all of the equations and variables, use **EqualBoundedScoringPathsEqsVars**. Once the $e_{a,b,c}$ system is solved, Eqn. (3.1) is linear in $f_{a,b}$ so we (or rather a computer) will find a rational g.f. solution! Use **EqualBoundedScoringPaths** for the single solution. To verify the result, one can check using the enumeration in **EqualBoundedScoringPathsNumber**. The irreducible g.f.s can be checked with **SpecificIrreducibleBoundedScoringPathsNumber**.

Old-Time Basketball

The methods of relating generating functions were inspired by Ayyer and Zeilberger's work [AZ07]. They found the following relation.

Theorem 3.5. Let F_w denote the generating function for the number of walks subject to step set $S = \{[\frac{1}{2}, 1], [\frac{1}{2}, -1], [1, 2], [1, -2]\}$, that start at (0, 0), end at (n, 0), and never go below the x-axis or above the line y = w. Then F_w satisfies the following recurrence relation:

$$F_w = 1 - tF_w + 2tF_wF_{w-1} + 2t^2F_wF_{w-1}F_{w-2}$$
$$- (t^3 + t^4)F_wF_{w-1}F_{w-2}F_{w-3} + t^5F_wF_{w-1}F_{w-2}F_{w-3}F_{w-4}$$

Proposition 3.6. The initial conditions are given as follows:

$$F_0 = 1, \qquad F_1 = \frac{1}{1-t}, \qquad F_2 = \frac{1-t}{1-2t+3t^2},$$

$$F_3 = \frac{1-2t+3t^2}{1-3t-5t^2-2t^3+t^4}, \qquad F_4 = \frac{1-3t-5t^2-2t^3+t^4}{1-4t-6t^2+2t^3}.$$

If we compute EqualBoundedScoringPaths($\{[1/2,1],[1/2,-1],[1,2],[1,-2]\},w,0,t$) for w = 0, ..., 4, then the initial conditions match. And we can verify Theorem 3.5 empirically for any fixed w.

Examples

Example 3.7 (Tied (American) Football Games). Consider the similar Example 3.3, but this time we want the teams to be tied at the end of the game. Our step set is the same. The only difference is in which method we use: EqualBoundedScoringPaths(S,8,-8,t):

$$\frac{1-4t-45t^2-43t^3+98t^4+108t^5-24t^6-30t^7}{1-4t-59t^2-77t^3+170t^4+234t^5-92t^6-142t^7-4t^8+6t^9}.$$

This sequence is new in the OEIS: <u>A301380</u> [OEIi].

Example 3.8 (Speed Enumeration - Bounded 2). Once again let us find the first 1000 coefficients of a g.f.: this time the one found in Example 3.7 above.

Method	Memory Used	Memory Allocation	CPU Time	Real Time
Brute-Force Recursion	$55.98 \mathrm{MiB}$	24MiB	$278.93 \mathrm{ms}$	$279.03 \mathrm{ms}$
G.F. Construction	10.13MiB	8MiB	$102.43 \mathrm{ms}$	$103.20 \mathrm{ms}$
Taylor Enumeration	4.85MiB	0 bytes	$5.20\mathrm{ms}$	$5.10\mathrm{ms}$
Total	14.98MiB	8MiB	$107.63 \mathrm{ms}$	108.30ms

Table 3.2: Bounded Bridge Enumeration 1000 terms

Again, the g.f. method is faster: this time, about 2.5 times as fast and 1/4th the memory.

3.3 Semi-bounded

We now remove the restriction of bounding walks from above.⁷ Ayyer and Zeilberger also provided relations for describing excursions that are similar to those included here [AZ08]. Duchon had previously tackled the case of excursions using much different language, but the same overall method [Duc00].

Let S denote the set of steps as in the previous section. Let f now denote the g.f. for nonnegative excursions with step set S that begin at (0,0) and end on the x-axis. How will we describe f in an equation? We could try to use the same method as Section 3.2.2 and write

$$f = 1 + \left(\sum_{(x,0)\in\mathcal{S}} t^x\right) f + \left(\sum_{(x,y)\in\mathcal{S}; y\neq 0} t^x e_y\right) f.$$

where e_y is the g.f. for walks that start at (0, y), end at (n, 0) and never hit the x-axis beforehand. However, to describe e_y as in Section 3.2.2 (by looking at the first step) would require writing equations for all $e_{i>0}$ since a walk could get arbitrarily far from the x-axis before returning.⁸ At some point, all of the e_i equations would look essentially the same and it may be possible to take limits of their form to find a solution for e_y . However, there is an easier method.

Ayyer and Zeilberger [AZ07] used a standard idea in combinatorics of "irreducible" walks to describe F_w in Section 3.2.2. We will use them as well. Let $f_{a,b}$ denote the g.f. for nonnegative walks that start at (0, a) and end at (n, b). Note that $f_{0,0} = f$ is what we are typically looking for. Let $g_{a,b}$ denote the g.f. for walks that start at (0, a), end at (n, b), and stay above the line $y = \min(a, b)$ except at the respective endpoint. Note that

⁷To look at walks solely bounded above, simply change the sign of the y-value of every step. Or note that every nonnegative excursion is in bijection with a nonpositive excursion by reversing the order of steps.

⁸We could instead look at the final step, but that is essentially the method described in the next paragraph.

 $g_{a,b} = g_{a-b,0}$ (or $= g_{0,b-a}$ if b > a).⁹ Then

$$f_{0,0} = 1 + \left(g_{0,0} + \sum_{(x,0)\in\mathcal{S}} t^x\right) f_{0,0}.$$
(3.2)

Either the walk is stationary (+1) or it returns to the x-axis $(g_{0,0} + \sum_{(x,0)\in\mathcal{S}} t^x)$ and then continues as if it were new (multiplication by $f_{0,0}$).

$$g_{0,0} = \sum_{(x_1,y_1)\in\mathcal{S}; y_1>0} \sum_{(x_2,y_2)\in\mathcal{S}; y_2<0} t^{x_1} f_{y_1-1,-y_2-1} t^{x_2}.$$
(3.3)

An irreducible (0,0)-walk can be characterized by how it departs from the x-axis (t^{x_1}) and how it returns (t^{x_2}) . There must be an intermediate walk between these two steps $(f_{y_1-1,-y_2-1})$ that does not touch the x-axis: hence the shift. Now we need to describe each new $f_{a,b}$.

$$a > b$$
 $f_{a,b} = \sum_{i=0}^{b} g_{a-i,0} f_{0,b-i},$ (3.4)

$$a = b$$
 $f_{a,a} = \sum_{i=0}^{b-1} g_{a-i,0} f_{0,b-i} + f_{0,0},$ (3.5)

$$a < b$$
 $f_{a,b} = \sum_{i=0}^{a-1} g_{a-i,0} f_{0,b-i} + f_{0,0} g_{0,b-a}.$ (3.6)

We can characterize a walk by how close it gets to the x-axis. An $f_{a,b}$ walk can go *i* levels below the level of $y = \min(a, b)$. In this case, we have an irreducible walk down to the lowest point $(g_{a-i,0})$, and then an arbitrary walk to the final level that does not go any lower $(f_{0,b-i})$. If a = b, then a walk that does not go below level y = a is equivalent to $f_{0,0}$. If a < b, then a walk that does not go below level y = a consists of an arbitrary walk back to the same level without going lower $(f_{0,0})$, followed by an irreducible walk to the final level of y = b $(g_{0,b-a})$. Now we must describe the irreducible walks that have not been covered.

$$g_{a,0} = \sum_{(x,y)\in\mathcal{S}; y<0} f_{a-1,-y-1}t^x, \qquad (3.7)$$

$$g_{0,a} = \sum_{(x,y)\in\mathcal{S}; y>0} t^x f_{y-1,a-1}.$$
(3.8)

⁹In this definition, $g_{0,0}$ does not include stationary walks or walks that are solely a step directly to the right. If this is what one desires, then use $g = g_{0,0} + 1 + \left(\sum_{(x,0)\in S} t^x\right)$.

An irreducible walk can be characterized by how it reaches (t^x) the lowest point. The rest of the walk is arbitrary as long as it does not hit the x-axis: again, hence the shift.

We need to show that this iterated process does eventually terminate. This can easily be justified because the largest index of either g or f will be $\langle \max_{(x,y)\in S} |y|$: the maximum step size. Thus, there is a finite number of g, f that we need to describe in order to have a closed system. Use **EqualSemiBoundedScoringPathsEqsVars** to generate these variables and their equations.

The equations are no longer linear and so Maple's built-in solve function does not work as nicely. Instead, we use the **Basis** procedure in the **Groebner** package to find

Definition 3.9 (Minimal Polynomial). the (minimal) algebraic equation satisfied by the generating function: p such that p(f) = 0 in terms of formal power series. We refer to p as the minimal polynomial. When discussing the degree of the minimal polynomial, we are considering the degree of f in p unless explicitly stated that we are considering t.

To produce this polynomial, use **EqualSemiBoundedScoringPaths**. To produce an "ideal" polynomial for each g, f, use **AllEqualSemiBoundedScoringPaths**.¹⁰ If you want the minimal polynomial for only a single variable, use **SpecificEqualSemiBoundedScoringPaths**. The minimal polynomial is typically found when the Groebner basis receives the desired variable as the lowest order lexicographically. Occasionally the Groebner basis will not produce the minimal polynomial, but instead a product of it and another polynomial. One can recover the minimal polynomial by factoring the output and testing which factor satisfies p(f) = 0 in terms of formal power series.¹¹ Every function that outputs an algebraic expression has had **FindProperRoot** appended to the end to properly parse the minimal polynomial.

Example 3.10 (Semi-bounded Example). Let $S = \{[0, -3], [1, -2], [2, 0], [3, 1]\}$. The g.f.,

¹⁰This can be slow as it requires finding a Groebner basis for *every* variable.

¹¹By checking a truncated version of f. In most cases f = 1 is sufficient to see which factor works. In general, enumerate m terms first using the proper enumerating function and see if the corresponding polynomial is $O(t^m)$, signifying a root.

F, for the number of excursions that do not go below y = -1 and have step set S is found to satisfy, using EqualSemiBoundedScoringPaths(S,-1,t,F),

$$t^{18}F^4 + t^{14}(t^2 - 1)F^3 + 2t^7(t^2 - 1)^2F^2 + (t^2 - 1)^5F + (t^2 - 1)^4 = 0.$$

This is actually a fairly simple answer. If we change $[0, -3] \rightarrow [0, 3]$, then the minimal polynomial is degree 10 in F and takes 4 lines to write.

This polynomial can then be used to discover the enumeration hidden in the coefficients of the taylor series expansion by setting $f_0 = 1$ and then iterating $f_0 \rightarrow p(f_0) + f_0$ to find a fixed point solution.¹² Finding the coefficients in this manner requires finding p: use EqualSemiBoundedScoringPathsCoefficients.

Finding the minimal polynomial can be a time-consuming process. An alternative method is to iterate a fixed-point solution of a vector of all g, f and then pick out $f_{0,0}$. Initialize $\{f_{a,a} = 1\}$ and every other g.f. to $0.^{13}$ Now iterate all of the g, f into their respective equations, truncating to the desired coefficient. Eventually we will reach a fixed-point for the vector of solutions. To do this, use **EqualSemiBoundedScoringPathsSeries**. However, for most reasonable calculations of the coefficients (< 1000 terms), finding the values via brute-force recursion (with **SpecificEqualBoundedScoringPathsNumber**) is faster than either iterating technique.

Example 3.11 (Speed Enumeration - Excursions). We want to obtain the number of nonnegative excursions with step set $S = \{[1, -2], [1, -1], [1, 0], [1, 1], [1, 2]\}$. Let F denote the corresponding g.f.; then¹⁴

$$t^{4}F^{4} - t^{2}(t+1)F^{3} + t(t+2)F^{2} - (t+1)F + 1 = 0.$$
(3.9)

A truncated solution in formal power series, and the one that makes sense in terms of our problem, is

$$F = 1 + t + 3t^{2} + 9t^{3} + 32t^{4} + 120t^{5} + 473t^{6} + 1925t^{7} + 8034t^{8} + 34188t^{9} + 147787t^{10} + 14778t^{10} + 1478t^{10} + 14778t^{10} + 14778t^{10$$

¹²This glosses over why the convergence works. It becomes an issue in the unbounded case.

¹³Staying stationary is only valid for general walks that begin and end at the same level.

¹⁴ Using EqualSemiBoundedScoringPaths(S, 0, t, F).

We compare and contrast the amount of time and memory to enumerate the first 500 and 1000 coefficients of F in the following tables.

- 1. Set up the g.f. equations and then iterate a vector of solutions using EqualSemi-BoundedScoringPathsSeries.
- 2. Solve for the minimal polynomial (Eqn. (3.9)) and iterate a single solution with EqualSemiBoundedScoringPaths.
- 3. Use Maple's taylor function on the minimal polynomial.¹⁵
- 4. Use brute-force recursion and Maple's option remember: SpecificEqualSemiBoundedScoringPathsNumber.

Method	Memory Used	Memory Allocation	CPU Time	Real Time
Vector Set-Up	$16.58 \mathrm{KiB}$	0 bytes	$300 \mu s$	$766 \mu s$
Iterating	100.04GiB	55.64MiB	4.45m	4.09m
Total	100.04GiB	55.64MiB	4.45m	4.09m
Polynomial	2.63MiB	0 bytes	$19.23 \mathrm{ms}$	$18.63 \mathrm{ms}$
Iterating Total	23.61GiB	20.84MiB	64.52s	58.56s
	23.61GiB	20.84MiB	64.54s	58.58s
Polynomial	$2.63 \mathrm{MiB}$	0 bytes	$19.23 \mathrm{ms}$	$18.63 \mathrm{ms}$
taylor	328.97MiB	53.85MiB	3.20s	3.18s
Total	331.60MiB	53.85MiB	3.22s	3.20s
Brute-Force Recursion	$186.60 \mathrm{MiB}$	328 MiB	1.569s	1.472s

Table 3.3: 500 term Excursion Enumeration

Finding the polynomials takes negligible time and memory in comparison to actually enumerating the coefficients beyond the first few terms.

¹⁵Requires replacing F by Z and using **RootOf(p)**.

Method	Memory Used	Memory Allocation	CPU Time	Real Time
Vector Iterating	1.31TiB	159.26MiB	$63.84\mathrm{m}$	$51.86\mathrm{m}$
Single Iterating	300.84 GiB	$15.32 \mathrm{MiB}$	$14.50\mathrm{m}$	$11.93\mathrm{m}$
taylor	$1.82 \mathrm{GiB}$	489.68MiB	12.20s	11.12s
Brute-Force Recursion	$0.85 \mathrm{GiB}$	508 MiB	7.53s	7.13s

Table 3.4: 1000 term Excursion Enumeration

So using recursion is the fastest, but also must reserve the most memory. The other methods use more memory in total but can recycle much of what they used. We can best all of the methods with another recursion that is faster and uses less memory; see Example 3.16.

Before implementing **FindProperRoot**, the Groebner basis output a degree 5 polynomial that took over 5 times as long to iterate as the degree 4 minimal polynomial. Whether iterating a single polynomial or the entire vector of solutions is faster depends on the degree of the minimal polynomial and the number of variables in the closed system. And the number of terms to enumerate. An interesting question would be to look at the time-complexity of this method of enumeration. The vector and single polynomial iteration could potentially be optimized to pick out coefficients from each monomial instead of expanding everything and then picking only relatively few terms. This would help the minimal polynomial iteration much more as the vector iteration only relies on degree 2 expressions.

3.3.1 Arbitrary Lower Limit

What if we want to consider walks that stay above an arbitrary lower bound y = -c as in Example 3.10? This is actually very easy. By shifting the walk to be nonnegative, we are now looking for $f_{c,c}$. Describe $f_{c,c}$ using Eqn. (3.5). Again, iterating on all new g, f will eventually yield a closed system since the indices are bounded by $\max(\max_{(x,y)\in S} |y|-1, c)$.

This shifting technique is what the Maple package utilizes when it is given a lower limit other than 0 for semi-bounded walks. The Maple functions are always focused on obtaining g.f.s for walks that begin at the origin: other produced g.f. are purely bonus. To obtain a g.f. that starts at (0, c), input a lower limit of -c, and take the g.f. that corresponds to starting at the origin.

3.3.2 Meanders

What if we do not care where the walk ends, as long as it stays above y = -c? Neither Ayyer and Zeilberger [AZ08] nor Duchon [Duc00] investigated semi-bounded free walks. In this case, we can actually utilize the irreducible walks we have just created. Recall the irreducible g.f.s in Eqns. (3.7) and (3.8).

Let k_a denote the g.f. for nonnegative meanders that begin at (0, a), restricted to step set S. Then

$$k_0 = 1 + \left(g_{0,0} + \sum_{(x,0)\in\mathcal{S}} t^x\right) k_0 + \sum_{(x,y)\in\mathcal{S}; y>0} t^x k_{y-1}$$

The walk can be stationary (+1), it can return to the x-axis $(g_{0,0} + \sum_{(x,0) \in S} t^x)$ and continue as a new meander (k_0) , or we take the first step (t^x) and continue as a new meander that never returns to the x-axis (k_{y-1}) . $g_{0,0}$, the g.f. for irreducible walks that return to the x-axis, already has a description from our previous work. We need only describe the new k_i .¹⁶

$$k_a = \sum_{i=0}^{a-1} g_{a,i} k_0 + k_0 = \left(\sum_{i=1}^{a} g_{i,0} + 1\right) k_0.$$

The meander can drop down to any lower altitude $(g_{a,i} = g_{a-i,0})$ and then continue as a new meander (k_0) , never dropping further. Or the meander will never go below y = a so it is equivalent to a meander from the origin (k_0) .

And since we have already described the irreducible walks earlier, we now have a closed system that we can use to solve for k_0 . To produce the entire system of equations, use **SemiBoundedScoringPathsEqsVars**. Again, we use **Groebner**[**Basis**] to find a minimal polynomial: **SemiBoundedScoringPaths**. To find the minimal polynomials for ALL¹⁷ of the variables, use **AllSemiBoundedScoringPaths**, though this will likely take

¹⁶If they exist.

¹⁷Including all of the irreducible and specific altitude walks from the previous section.

a while. For a specific k_a , use **SpecificSemiBoundedScoringPaths**. For any $g_{a,b}, f_{a,b}$, it is best¹⁸ to use **SpecificEqualSemiBoundedScoringPaths**.

Example 3.12. For comparison, we use the same step set as in Example 3.10: $S = \{[0, -3], [1, -2], [2, 0], [3, 1]\}$. The g.f., K, for the number of meanders that do not go below y = -1 and have step set S is found to satisfy, using SemiBoundedScoringPaths(S,-1,t,K),

$$t^{19} (t^{2} + t + 1) K^{4} + t^{12} (4t^{6} + t^{4} + 3t^{3} + 6t^{2} + t - 3) K^{3} + 3t^{6} (2t^{9} - 2t^{8} + t^{7} + 4t^{6} + 2t^{5} - 2t^{4} - t^{3} + 3t^{2} - 1) K^{2} + (4t^{12} - 8t^{11} + 7t^{10} + 11t^{9} - 7t^{8} - 3t^{7} + 7t^{6} + 6t^{5} - 6t^{4} - 2t^{3} + 4t^{2} - 1) K + t^{9} - 3t^{8} + 4t^{7} + 2t^{6} - 6t^{5} + 4t^{4} + 3t^{3} - 3t^{2} + 1 = 0.$$

Sadly, the coefficients do not factor as nicely as in Example 3.10.

Now that we have a method of describing the g.f. for meanders, let us compare how fast it is for enumeration.

Example 3.13 (Speed Enumeration - Meanders). We use the simpler step set $S = \{[1, -2], [1, -1], [1, 0], [1, 1], [1, 2]\}$ and let K be the g.f. for the number of nonnegative meanders with step set S. Then¹⁹

$$t^{2}(5t-1)^{2}K^{4} + t(5t-1)^{2}K^{3} + 3t(5t-1)K^{2} + (5t-1)K + 1 = 0.$$
(3.10)

A truncated solution is

 $K = 1 + 3t + 12t^2 + 51t^3 + 226t^4 + 1025t^5 + 4724t^6 + 22022t^7 + 103550t^8 + 490191t^9 + 2333057t^{10}.$

We repeat our analysis of differing methods of enumeration from Example 3.11.

- 1. Iterating a vector of solutions using SemiBoundedScoringPathsSeries.
- 2. Iterating a fixed point solution after solving for the polynomial (Eqn. (3.10)) with SemiBoundedScoringPaths.²⁰

¹⁸And necessary. Compatibility was removed for ease of use.

¹⁹ Using **SemiBoundedScoringPaths**(S,0,t,K).

²⁰All of this can be accomplished with the one function SemiBoundedScoringPathsCoefficients.

- 3. Using Maple's taylor function on the minimal polynomial.²¹
- 4. Enumerating using brute-force recursion and Maple's option remember: SpecificSemiBoundedScoringPathsNumber.

Method	Memory Used	Memory Allocation	CPU Time	Real Time
Vector Set-Up	21.07KiB	0 bytes	$233 \mu \mathrm{s}$	$233 \mu s$
Vector Iterating	$2.18 \mathrm{TiB}$	$15.29 \mathrm{MiB}$	$102.64\mathrm{m}$	84.48m
Total	2.18TiB	15.29MiB	102.64m	84.48m
Polynomial	6.37MiB	0 bytes	43.00ms	43.03ms
Single Iterating	304.99 GiB	$30.58 \mathrm{MiB}$	$13.88\mathrm{m}$	$11.64 \mathrm{m}$
Total	305.00GiB	30.58MiB	13.88m	11.64m
Polynomial	$6.37 \mathrm{MiB}$	0 bytes	43.00ms	$43.03 \mathrm{ms}$
taylor	1.85 GiB	490MiB	11.33	10.62s
Total	1.86GiB	490MiB	11.37s	10.66s
Brute-Force Recursion	1.08GiB	0.52 GiB	7.93s	7.465s

Table 3.5: 1000 term Meander Enumeration

Again, there is an even faster recursive formula. See Example 3.17.

3.4 Guess-and-Check Method

Zeilberger provided a guess-and-check method and Maple package, **W1D** [EZ15], for finding algebraic expressions. In the semi-bounded case where all steps have x-step 1, Phillippe Duchon guaranteed that the results are algebraic. Thus, the guess-and-check method WILL work, eventually, if you set the search parameter high enough. For semi-bounded cases with differing x-steps and unbounded cases, there is no guarantee that guessing will eventually work [MR09]. Though you may get lucky and produce an algebraic equation that has the minimal polynomial as a root.

²¹Requires replacing K by Z and using **RootOf(p**).

Another advantage of this paper's method over guess-and-check is that this method is typically much much faster. Setting up the equations takes a set amount of time that is linear in the maximum step size, $|\mathcal{S}|$, and the lower limit. The time sink comes in finding a Groebner basis, but this is still typically faster.

Empir will take a list of the first few coefficients of the g.f. F and attempt to find an algebraic equation that F satisfies by guessing the degree and trying to solve for the coefficients. **EmpirF** will do this faster²² by utilizing the **gfun** package. Both **Empir** and **EmpirF** require enumerating the first few terms.

Example 3.14 (Excursion Time Comparison). Consider looking for an algebraic equation for the g.f. for nonnegative excursions with step set $S = \{[1, -2], [1, -1], [1, 0], [1, 1], [1, 2]\}$. It takes²³

Method	Memory Used	Memory Allocation	CPU Time	Real Time
ESBSP	2.63MiB	0 bytes	$19.23 \mathrm{ms}$	$18.63 \mathrm{ms}$
Empir	91.61MiB	$5.60 \mathrm{MiB}$	$734 \mathrm{ms}$	$735 \mathrm{ms}$
EmpirF	4.33MiB	0 bytes	$33.87 \mathrm{ms}$	$35.90\mathrm{ms}$

Table 3.6: Finding Excursion Minimal Polynomial $\mathcal{S} = \{[1, -2], [1, -1], [1, 0], [1, 1], [1, 2]\}$

EqualSemiBoundedScoringPaths (ESBSP) takes about 3% of the time and memory that Empir requires and 60% of the time and memory as EmpirF.

Example 3.15 (Meander Time Comparison). Now try to find an algebraic equation satisfied by the g.f. for nonnegative meanders with step set $S = \{[1, -2], [1, -1], [1, 0], [1, 1], [1, 2]\}$.²⁴

²²Almost always.

²³ The actual minimal polynomial is shown in Example 3.11.

²⁴ The minimal polynomial is shown in Example 3.13.

IE.	$c = 5.7$. Finding Meander Minimar Folynonial $\mathcal{O} = \{[1, -2], [1, -1], [1, 0], [1, 1], [1, 0], [1, 1], [1, 0], [1, 1], [1, 0], [1, 1], [1, 0], [1, 1], [1, 0], [1, 1], [1, 0], [1, 1], [1, 0], [1, $					
	Method	Memory Used	Memory Allocation	CPU Time	Real Time	
	SBSP	$6.37 \mathrm{MiB}$	0 bytes	43.00ms	$43.03 \mathrm{ms}$	
	Empir	$91.68 \mathrm{MiB}$	$5.45 \mathrm{MiB}$	$749 \mathrm{ms}$	$770 \mathrm{ms}$	
	EmpirF	4.74MiB	0 bytes	$36.53 \mathrm{ms}$	$38.63 \mathrm{ms}$	

Table 3.7: Finding Meander Minimal Polynomial $S = \{[1, -2], [1, -1], [1, 0], [1, 1], [1, 2]\}$

SBSP=SemiBoundedScoringPaths.

As expected, this package is still much faster than **Empir**. Interestingly, **EmpirF** appears to be as fast, if not a little faster. **EmpirF** is still handicapped in its range of applications and so the slight speed-up is sacrificed for versatility.

In fact, adding [1,3] to the step set already makes **EmpirF** fail for the preset search bound. The minimal polynomial has degree and order 10 in that case: easily found by **SemiBoundedScoringPaths** in 1/4s.

The other bonus is that this paper's method allows for any size x-step. **W2D** could potentially be used to solve this problem once it has been restricted to look for the coefficient of $x^n \cdot y^0$. However, this would make the already slow guessing method even slower.

3.5 Algebraic to Recursive

3.5.1 Conversion

There is a classical method for deducing from the algebraic function satisfying the g.f. a linear recurrence with polynomial coefficients satisfied by the coefficients of the g.f. in question. See Chapter 6 of "The Concrete Tetrahedron" by Kauers and Paule [KP11].

The method is implemented in the Maple package gfun [SZ94] and also in Zeilberger's Maple package **SCHUTZENBERGER**, that is used here. The **SCHUTZENBERGER** package also contains **EmpirF** for obtaining the minimal polynomial, but we now have a much better method of producing the minimal polynomial. To convert an algebraic formula to a recurrence formula, use **algtorec**. Let B(n) denote the number of walks of length n. Interestingly, sometimes a larger (non-minimal) polynomial produces a better (lower-order) recurrence.

Example 3.16 (Better Excursion Recursion). Let $S = \{[1, -2], [1, -1], [1, 0], [1, 1], [1, 2]\}$. Let F denote the g.f. for nonnegative excursions.²⁵ Then its coefficients satisfy

$$0 = 3125(n+1)(n+2)(n+3)(n+4)B(n)$$

- 250(n+4)(n+3)(n+2)(27n+122)B(n+1)
+ 25(n+4)(n+3)(107n² + 1457n + 4316)B(n+2)
+ 10(n+4)(304n³ + 3233n² + 9864n + 6513)B(n+3)
- (2821n⁴ + 56794n³ + 425771n² + 1407974n - 1731540)B(n+4)
+ 2(n+7)(413n³ + 6986n² + 39356n + 73830)B(n+5)
- (n+8)(n+7)(99n² + 1241n + 3900)B(n+6)
+ 2(2n+15)(n+9)(n+8)(n+7)B(n+7). (3.11)

The following are the time and memory requirements for various stages of enumerating. We need to create the minimal polynomial with **EqualSemiBoundedScoringPaths**, convert it to an improved recursive formula, Eqn. (3.11), for the coefficients of F with **algtorec**, and then enumerate with **SeqFromRec**.

Table 5.6. Enumerating Excursions more Enclently					
Method	Memory Used	Memory Allocation	CPU Time	Real Time	
ESBSP	$2.63 \mathrm{MiB}$	0 bytes	$19.23 \mathrm{ms}$	$18.63 \mathrm{ms}$	
algtorec	$35.84 \mathrm{MiB}$	$3.12 \mathrm{MiB}$	$267.93\mathrm{ms}$	$262.40\mathrm{ms}$	
SeqFromRec	123.05MiB	24MiB	375.27ms	$375.37\mathrm{ms}$	
Total	$161.52 \mathrm{MiB}$	27.12MiB	662.43ms	$656.40\mathrm{ms}$	

 Table 3.8: Enumerating Excursions More Efficiently

${\bf ESBSP}{=}{\bf EqualSemiBoundedScoringPaths}$

As expected, this streamlined recurrence is much faster and less memory-intensive than the basic recurrence in Example 3.11. There is an up-front cost for creating the improved

²⁵ The minimal polynomial is given in Example 3.11.

recurrence, but if the goal is to enumerate enough terms, it can be worth it. In this case, enough is less than 500.

Before **FindProperRoot** was implemented to automatically parse the minimal polynomial, we converted a larger polynomial into a sixth-order recurrence in about 2.4 seconds.

Example 3.17 (Better Meander Recursion). Let $S = \{[1, -2], [1, -1], [1, 0], [1, 1], [1, 2]\}$ and K denote the g.f. for nonnegative meanders.²⁶ The coefficients now satisfy

$$\begin{split} 0 &= 625(n+1)(n+2)(n+3)(n+4)B(n) \\ &- 250(5n+21)(n+4)(n+3)(n+2)B(n+1) \\ &+ 50(n+4)(n+3)(7n^2+95n+270)B(n+2) \\ &+ 20(n+4)(32n^3+367n^2+1365n+1620)B(n+3) \\ &- (n+5)(463n^3+6691n^2+32442n+52704)B(n+4) \\ &+ 2(n+5)(n+6)(53n^2+593n+1674)B(n+5) \\ &- 4(n+5)(2n+13)(n+7)(n+6)B(n+6). \end{split}$$

Oddly it is a lower order recurrence despite the slightly higher complexity of describing meanders. The following are the time and memory requirements for various stages of enumerating.

Method	Memory Used	Memory Allocation	CPU Time	Real Time
SBSP	6.37MiB	0 bytes	43.00ms	43.03ms
algtorec	$110.77 \mathrm{MiB}$	29.39MiB	$659.73\mathrm{ms}$	$633.10\mathrm{ms}$
SeqFromRec	$89.58 \mathrm{MiB}$	4MiB	284.57ms	$257.23 \mathrm{ms}$
Total	206.72MiB	33.39MiB	987.30ms	$933.36\mathrm{ms}$

Table 3.9: Enumerating Meanders More Efficiently

${\bf SBSP}{=}{\bf SemiBoundedScoringPaths}$

This improved recurrence is about 8 times as fast as the basic recurrence in Example

²⁶ The minimal polynomial is given in Example 3.13.

3.13. It is not quite as much of a savings as the case of excursions, but it is still extremely good.

3.5.2 Searching

There is an alternative to converting the minimal polynomial into a linear recurrence. Since we know that this will be possible, we could simply guess at the form of the recurrence and use a suitable number of starting values to determine the coefficients. **Findrec** will accomplish this guessing method. The problem is that we do not know an upper bound for the order and degree. Setting the bound extremely high (or searching until a recurrence is found) would suffice. However, this is not ideal.

Example 3.18. Let $S = \{[1,2], [1,-3]\}$. The minimal polynomial is given later in Section 3.8.3. Let B(n) denote the number of nonnegative excursions of length 5n. B(n) was found to satisfy a 4th order, degree 11 recurrence relation using **Findrec** and a 7th order, degree 9 recurrence relation using **algtorec**.²⁷ Both relations²⁸ are fairly large and so are not included here.²⁹ We compare and contrast the methods in the table below.

Table 3.10: Alternative Recurrence Step Set $\{[1,2], [1,-3]\}$

Method	Degree	Order	Memory Used	Allocated	CPU Time	Real Time
algtorec	9	7	$165.84 \mathrm{GiB}$	2.46 GiB	46.99h	12.06h
Findrec	11	4	0.50 GiB	388.01MiB	4.47s	4.31s

Simply guessing at the form appears to be the MUCH better choice here.

Example 3.19. Let $S = \{[1, -2], [1, -1], [1, 0], [1, 1], [1, 2]\}$. See Example 3.11 for the minimal polynomial. Let B(n) denote the number of nonnegative excursions of length n.

 $^{^{27}2}$ trials.

²⁸ The **Findrec** recurrence matches with Andrew Lohr's calculation as expected since they are computing the same result.

²⁹ They are available at math.rutgers.edu/~bte14/Articles/ScoringPaths/23Recurrence.txt.

Then

$$\begin{split} 0 &= 2 \left(n+4 \right) \left(2n+11 \right) \left(n+7 \right) \left(n+6 \right) B(n+5) \\ &- \left(n+6 \right) \left(43n^3+597n^2+2738n+4142 \right) B(n+4) \\ &+ \left(124n^4+2110n^3+13305n^2+36815n+37686 \right) B(n+3) \\ &- 5 \left(n+3 \right) \left(2n^3-22n^2-305n-726 \right) B(n+2) \\ &- 25 \left(n+3 \right) \left(n+2 \right) \left(8n^2+72n+159 \right) B(n+1) \\ &+ 125 \left(n+5 \right) \left(n+3 \right) \left(n+2 \right) \left(n+1 \right) B(n). \end{split}$$

This was found with Findrec while algeorec found

$$\begin{split} 0 &= 2 \left(2n + 15 \right) (n + 9) \left(n + 8 \right) (n + 7) B(n + 7) \\ &- \left(n + 8 \right) (n + 7) \left(99n^2 + 1241n + 3900 \right) B(n + 6) \\ &+ 2 \left(n + 7 \right) \left(413n^3 + 6986n^2 + 39356n + 73830 \right) B(n + 5) \\ &- \left(2821n^4 + 56794n^3 + 425771n^2 + 1407974n + 1731540 \right) B(n + 4) \\ &+ 10 \left(n + 4 \right) \left(304n^3 + 3233n^2 + 9864n + 6513 \right) B(n + 3) \\ &+ 25 \left(n + 4 \right) \left(n + 3 \right) \left(107n^2 + 1457n + 4316 \right) B(n + 2) \\ &- 250 \left(n + 4 \right) \left(n + 3 \right) \left(n + 2 \right) \left(27n + 122 \right) B(n + 1) \\ &+ 3125 \left(n + 1 \right) \left(n + 2 \right) \left(n + 3 \right) \left(n + 4 \right) B(n). \end{split}$$

The comparison in methods is given below.

Table 3.11: Alternative Recurrence Step Set $\{[1, -2], [1, -1], [1, 0], [1, 1], [1, 2]\}$

Method	Degree	Order	Memory Used	Allocated	CPU Time	Real Time
algtorec	4	7	38.32MiB	42.61MiB	$355.60\mathrm{ms}$	$354.44\mathrm{ms}$
Findrec	4	5	62.38MiB	28.00MiB	$618.07\mathrm{ms}$	$630.87\mathrm{ms}$

Conversion is actually more efficient, though not ideal, in this case. Though the difference is not as extreme as that shown in Table 3.18.

With smaller step sizes, conversion appears to be "better". This is likely due to the

intensive task of converting higher degree polynomials, while guessing avoids that hurdle. For small enough examples, guessing is a larger search space than conversion.

Because of the way **Findrec** is programmed, it is guaranteed that a found recurrence has order less than or equal to that of **algtorec**. Depending on the goal when enumerating, lower degree (save computation) or lower order (save memory) is optimal.

One can use **algotrec**'s order and degree as a proper bound for use in **Findrec**. But instead of directly converting the minimal polynomial, it may be possible to find a bound on the order and degree of a recurrence based on the degrees of the g.f. and t in the minimal polynomial. We may also be able to derive an upper bound based on the number of variables in the closed system we created.

I was unable to prove any useful bounds, and these examples are evidence against bounding the order and degree directly by the degrees of the g.f. and t. All three examples have minimal polynomials with smaller degree (and total degree) than the order of the found recurrences. In fact, we know that the sum of the degree and order of any recurrence has to be greater than the total degree of the minimal polynomial in these cases since we searched everything lower.

The first example showed that **Findrec** can be an *EXTREME* improvement over **algtorec** and the second example showed when **algtorec** can just edge out **Findrec**. The following example demonstrates that **algtorec** can be the significantly better choice.

Example 3.20. Let $S = \{[1, -1], [3, -1], [1, 0], [3, 0], [2, 1], [1, 2], [2, 2]\}$. The minimal polynomial, letting K denote the g.f. of meanders with step set S, is

 $1 + (t^{3} + 5t^{2} + 4t - 1)K + t(4t + 3)(2t^{3} + 2t^{2} + 3t - 1)K^{2} + t(t + 1)(2t^{3} + 2t^{2} + 3t - 1)K^{3}.$

The recurrences found by conversions and searching are much too large to include here. For their information, see

 $math.rutgers.edu/\sim bte14/Articles/ScoringPaths/RecurrenceOutputFile.$

		[1], [0, 1], [1, 0], [0, 0], [2, 1], [1, 2], [2, 2]				
Method	Degree	Order	Memory Used	Allocated	CPU Time	Real Time
Polynomial			$2.55 \mathrm{MiB}$	24.00MiB	$20.80 \mathrm{ms}$	$24.50\mathrm{ms}$
algtorec	3	31	73.92MiB	11.39MiB	491.03ms	475.13ms
Total			76.47MiB	35.39MiB	511.83ms	499.63ms
161 terms			24.11MiB	0 bytes	$190.00 \mathrm{ms}$	$193.00\mathrm{ms}$
Findrec	5	15	2.46GiB	504MiB	23.30s	21.20s
Total			2.48GiB	504MiB	23.49s	21.39s

Table 3.12: Alternative Recurrence Step Set $\{[1, -1], [3, -1], [1, 0], [3, 0], [2, 1], [1, 2], [2, 2]\}$

Conversion is surprisingly 43 times faster and has roughly $1/32^{nd}$ of the memory requirements!

We chose to enumerate 161 terms of the sequence because that was sufficient, once we knew the degree and order of the converted recurrence, to guarantee **Findrec** would encounter a solution. Typically, we do not use both **Findrec** and **algtorec**; we should simply find a set large bound of numbers. But enumerating was a small portion of **Findrec**'s time so it does not matter too much to this example.

We have used meanders instead of excursions for our 3^{rd} example, but they are similar enough to compare. The important facets of the problem are the degree of the minimal polynomial and the number of steps.

It would appear that increasing the number of steps does not affect the runtime of **algtorec** as much as the runtime of **Findrec**. The runtime of **algtorec** is highly dependent on the degree of the minimal polynomial.³⁰ The degree of the minimal polynomial is generally correlated with the number of variables in the system, though not necessarily equal or bounded one way or the other. Example 3.11 has 7 variables in the closed system and a minimal polynomial of degree only 4. Section 3.8.3 shows that a step set of $\{[1,3], [1,-5]\}$ for nonnegative excursions yields a closed system with 18 variables yet the minimal polynomial has degree 56!

³⁰The degree of t is much less relevant.

The most important aspects of the step set to **algtorec** runtime are then the maximum and minimum³¹ y-steps since these dictate how many other walks we must consider.

Findrec is much weaker with larger step sets. Inherently, the recurrences will become more complicated, which means **Findrec** must consider many more terms in its guesses. The worst case (most number of terms to address) will occur when the order and degree of our search are the same.

Since **algtorec** works better with minimal polynomials of lower degree, and finding the minimal polynomial is generally very fast (see Example 3.13), I have chosen to have programs such as **PaperSemiBounded** and **BookEqualSemiBounded** use **Findrec** when the minimal polynomial has degree ≥ 8 and **algtorec** otherwise. The cutoff was arbitrarily chosen based on a few random examples. A rigorous cutoff would be useful for more widespread implementation.

3.6 Unbounded

We remove the lower bound in this section and transition from analyzing excursions to analyzing bridges. I was unable to find any current literature that analyzes the unbounded case with generating functions; thus, I am led to believe that this section is novel work. Again, all walks are assumed to be constructed from a step set S.

3.6.1 Walking "Anywhere"

Enumerating walks that go anywhere *can* be a trivial task. First off, the *y*-values do not matter except for describing multiple steps with the same *x*-step. If all of the steps have *x*-value m > 0, then the number of walks of length $m \cdot n$ is simply $|\mathcal{S}|^n$. If we have steps with varying *x*-values, then the problem becomes finding combinations of the *x*-values that sum to *n*. In general, the g.f. will be

$$\frac{1}{1 - \sum_{(x,y) \in \mathcal{S}} t^x}.$$

³¹Most negative.

This g.f. is produced by **UnBoundedScoringPaths**. The terms are generated by (using Taylor series) **UnBoundedScoringPathsCoefficients** or (via recursion) **UnBounded-ScoringPathsNumber**. Using Maple's taylor series expansion is typically faster, but both are extremely quick at enumerating the first 1000 terms.³²

3.6.2 Returning to the *x*-axis

A lot of the work we have done in the semi-bounded case will prove useful here. We cannot use the exact same method as in semi-bounded Section 3.3. Suppose we tried describing a walk with a negative change in altitude. The first and last steps could both be positive. Then we would need to describe a walk that has a larger negative change in altitude. And so on.

We try a slightly different method. Let G denote the g.f. for walks with step set S that begin at (0,0) and end on the x-axis: a bridge. We choose to introduce another g.f., which we denote h_i , for the number of walks from (0,i) to (n,0) that do not touch the x-axis beforehand, with the exception of h_0 starting on the x-axis.³³ Recall the definitions of $g_{a,b}$ and $f_{a,b}$. The first equation is similar to $f_{0,0}$:

$$G = 1 + \left(h_0 + \sum_{(x,0)\in\mathcal{S}} t^x\right)G.$$

A walk can be stationary, or it returns to the x-axis after some number of steps, at which point it can take another G-walk.

$$h_0 = 2g_{0,0} + \sum_{(x,y)\in\mathcal{S}; y\leq -2} \sum_{i=1}^{-y-1} g_{0,i} t^x h_{i+y} + \sum_{(x,y)\in\mathcal{S}; y\geq 2} \sum_{i=1}^{y-1} g_{y-i,0} t^x h_i$$

An h_0 -walk can be purely positive $(g_{0,0})$ or purely negative $(g_{0,0})$.³⁴ Note that walks below the x-axis are in bijection with walks above the x-axis by reversing the order of steps.³⁵ We can then use $g_{0,-a} = g_{a,0}$ and $f_{-a,-b} = f_{b,a}$. It is also possible for the h-walk to cross

³²By utilizing Maple's **option remember** in the recursive case.

 $^{^{33}}h_0$ does NOT include stationary walks nor walks that are steps directly to the right. For that, one will want to use $h = h_0 + 1 + \sum_{(x,0) \in S} t^x$.

³⁴This is also partially why irreducible walks do not include steps to the right. If $g_{0,0}$ did, then I would have to write $h_0 = 2g_{0,0} - \sum_{(x,0)\in S} t^x$, which is less elegant.

 $^{^{35}}$ Equivalent to rotating the walk 180° .

the x-axis without touching it. This would involve an irreducible walk $(g_{0,i})$, a step across the x-axis (t^x) , and another walk (h_{i+y}) to return to the x-axis. The previous sentence described crossing the x-axis from above to below. We could instead cross from below to above. This would consist of an irreducible walk $(g_{0,i-y} = g_{y-i,0})$, a step across the x-axis (t^x) , and another walk (h_i) to return to the x-axis.

We already have equations to describe the irreducible walks. We now need to describe the other h-walks.

$$j > 0 \qquad h_j = g_{j,0} + \sum_{(x,y) \in \mathcal{S}; y \le -2} \sum_{i=1}^{-y-1} f_{j-1,i-1} t^x h_{i+y},$$
$$j < 0 \qquad h_j = g_{0,-j} + \sum_{(x,y) \in \mathcal{S}; y \ge 2} \sum_{i=1}^{y-1} f_{y-i-1,-j-1} t^x h_i.$$

An $h_{j>0}$ -walk can either be a walk "directly" to the x-axis $(g_{j,0})$ or it consists of an arbitrary walk that does not touch the x-axis $(f_{j-1,i-1})$, followed by a step across (t^x) , and another walk (h_{i+y}) to return to the x-axis. Similarly an $h_{j<0}$ -walk can either be a walk "directly" to the x-axis $(g_{j,0} = g_{0,-j})$ or it consists of an arbitrary walk that does not touch the x-axis $(f_{j+1,i-y+1} = f_{y-i-1,-j-1})$, followed by a step across (t^x) , and another walk (h_i) to return to the x-axis.

A brief lapse in concentration allowed me to write the alternative (and much simpler) relation:

$$h_0 = 2g_{0,0} + \sum_{(x,y)\in\mathcal{S}; y\neq 0} t^x h_y.$$

The reason for shunning this "simpler" description is that the h_y may be irreducible walks directly back to the x-axis, and thus double-count a walk from $g_{0,0}$.

The system of equations will be closed since the index of h_i -walks is bounded by $\max_{(x,y)\in S} |y| - 1$. All of the equations and variables are produced in **EqualUnBound**edScoringPathsEqsVars. Again, we use the **Basis** procedure in the **Groebner** package to find a polynomial p such that p(G) = 0 in terms of formal power series. To produce this polynomial, use **EqualUnBoundedScoringPaths**. To produce the minimal polynomial for h_j , use **SpecificUnBoundedScoringPaths**. We can try to produce the coefficients of the taylor series using the iteration $G_0 \rightarrow p(G_0) + G_0$. However, because of convergence issues, this will typically not work. It is also potentially not ideal because it requires finding a Groebner basis, which can be time-consuming. However, iterating a vector of solutions à la the semi-bounded Section 3.3 does work. To check these values, use **SpecificUnBoundedScoringPathsNumber** to verify G and **SpecificIrreducibleUnBoundedScoringPathsNumber** to verify h_j .

3.6.3 Alternative Method

There is an alternative method for enumerating unbounded walks in the specific case that all of the steps have x-step 1 [AZ90]. The functions themselves are taken from the Maple package **EKHAD** by Zeilberger. I will start with an example.

Example 3.21. Consider the function $G(t,n) = (t^{-2} + t^{-1} + t + t^2)^n$. This represents a possible step set of $S = \{[1, -2], [1, -1], [1, 1], [1, 2]\}$. A specific monomial in the expansion of G(t, n) comes from all the ways of picking a power of t from each term; this is equivalent to picking which step to take and in which order.

- 1. G(t,0) = 1. There is only 1 walk: the stationary walk.
- 2. $G(t,1) = t^{-2} + t^{-1} + t + t^2$. There are 4 possible steps each leading to a different final altitude.
- 3. $G(t,2) = t^{-4} + 2t^{-3} + t^{-2} + 2t^{-1} + 4 + 2t + t^2 + 2t^3 + t^4$. There are 4 ways to return to the x-axis after 2 steps.

G(t,n) enumerates all the ways to change altitude by c in the coefficient of the t^c monomial.

In general, for a step set S, let $G(t, n) = \left(\sum_{(x,y)\in S} t^y\right)^n$. The procedure $\mathbf{AZd}(\mathbf{A}, \mathbf{t}, \mathbf{n}, \mathbf{N})$ from package **EKHAD** will give a recurrence for the (contour around 0) integral of A (with respect to t) under hypergeometric assumptions, i.e., \mathbf{AZd} returns the residue. And if we represent A as a power series, that means we extract the t^{-1} term. So to find a recurrence for the number of walks that return to the x-axis, we will use A = G(t, n)/t. Example 3.22. The input $AZd((t^{-2} + t^{-1} + t + t^2)^n, t, n, N)$ yields as output:

$$18(2n+1)(5n+8)(n+1) + (n+1)(35n^2 + 91n + 54)N - 2(n+2)(2n+3)(5n+3)N^2.$$

This translates as, if we let B(n) denote the number of walks that return to the x-axis after n steps,

$$18(2n+1)(5n+8)(n+1)B(n) + (n+1)(35n^2 + 91n + 54)B(n+1) - 2(n+2)(2n+3)(5n+3)B(n+2) = 0.$$
(3.12)

We can find a recursion for any change in altitude, not just bridges. If we are interested in walks that change in elevation by c, then we should use $A = G(t, n)/t^{c+1}$.

3.6.4 Selected Step Sets

The following examples are all produced by the one call:

EqualUnBoundedScoringPaths(\mathcal{S}, t, G).

Example 3.23 (Old-Time Basketball). We will follow up the previous example by using this paper's g.f. method of dynamical programming with the same step set. Let $S = \{[1, -2], [1, -1], [1, 1], [1, 2]\}$. Let G denote the g.f. for the number of unbounded bridges subject to step set S. Then G satisfies

$$(9t+4)(4t-1)^2G^4 - 2(3t-2)(4t-1)G^2 + t = 0.$$

And the coefficients satisfy (using algebra)

$$108(n+1)(2n+1)B(n) + (78n^2 + 246n + 216)B(n+1) - (n+2)(17n-9)B(n+2) - 2(n+3)(2n+5)B(n+3) = 0.$$

Interestingly, we have obtained a different recurrence than what was found with the AZd method: Eqn. (3.12). Both recurrences are correct. Typically, AZd will produce a lower order recurrence but with higher degree polynomial coefficients.

The simple alteration of the step set to $S' = \{[2, -2], [1, -1], [1, 1], [2, 2]\}$ makes the alternative method in Section 3.6.3 ineffectual. However, the method presented in this

paper is not bothered in the slightest. Let G' denote the g.f. for the number of unbounded bridges subject to step set S'. Then G' satisfies

$$(8t^{2}+5) (4t^{4}-8t^{2}+1)^{2} G'^{4}+2 (4t^{2}-3) (4t^{4}-8t^{2}+1) G'^{2}+1=0.$$

Now we only have the option of conversion or guessing to find a recurrence for the coefficients of G'. Using algeorec took only 69ms and produced:

$$32(n+3)(n+2)B(n) + 4(5n^2 + 110n + 356)B(n+2) - 8(15n^2 + 160n + 439)B(n+4) - (59n + 438)(n+6)B(n+6) + 10(n+8)(n+6)B(n+8) = 0.$$

The reason to have x-step as all 1s versus having x-step equal to y-step is that the choice changes what G counts. If all x-steps are 1, G will count the number of ways to be tied after n total baskets. If x-step equals |y|-step, then G' counts the number of ways to be tied at a score of $\frac{n}{2}$ to $\frac{n}{2}$. Note that this interpretation means G' has coefficient 0 for all odd powers of t; teams cannot be tied after an odd number of points have been scored. We could make the substitution $t \to \sqrt{t}$ in G' and then G' would count the number of ways to be tied at a score of n - n.

Example 3.24 (Current Basketball). Let $S = \{[1, -3], [1, -2], [1, -1], [1, 1], [1, 2], [1, 3]\}$. Let G denote the g.f. for the number of unbounded bridges subject to step set S. Then G satisfies

$$(8t^{2} - 68t - 27)^{2} (6t - 1)^{4} (2t + 1)^{4} G^{8} + 4 (68t^{2} + 10t - 9) (8t^{2} - 68t - 27) (6t - 1)^{3} (2t + 1)^{3} G^{6} + 2 (9120t^{4} + 3744t^{3} - 1264t^{2} - 212t + 135) (6t - 1)^{2} (2t + 1)^{2} G^{4} + 4 (1216t^{4} + 832t^{3} + 4t^{2} - 46t + 7) (6t - 1) (2t + 1) G^{2} + (16t^{2} + 8t - 1)^{2} = 0.$$

EqualUnBoundedScoringPaths(S,t,G) originally gave a much larger polynomial that G satisfies. After FindProperRoot was added to parse the output, the smaller polynomial above was the result. The coefficients satisfy (using algebra)

$$\begin{aligned} 36864 \left(n+1\right) \left(n+2\right) \left(n+3\right) B(n) \\ &- 3072 \left(n+3\right) \left(n+2\right) \left(97n+142\right) B(n+1) \\ &- 64 \left(n+3\right) \left(4031n^2+17601n+19504\right) B(n+2) \\ &- 16 \left(1684n^3+13491n^2+31178n+15240\right) B(n+3) \\ &+ 24 \left(663n^3+7222n^2+28628n+41563\right) B(n+4) \\ &+ 4 \left(467n^3+7011n^2+35842n+62490\right) B(n+5) \\ &- 2 \left(n+6\right) \left(115n^2+1080n+2273\right) B(n+6) \\ &- 3 \left(3n+20\right) \left(3n+19\right) \left(n+7\right) B(n+7) = 0. \end{aligned}$$

or, using AZd,

$$96 (n + 1) (n + 2) (n + 3) (2058n^{3} + 20335n^{2} + 66857n + 73300) B(n) -8 (n + 2) (n + 3) (201684n^{4} + 2295356n^{3} + 9540055n^{2} + 16998380n + 10742400) B(n + 1) -4 (n + 3) (310758n^{5} + 4313617n^{4} + 23611469n^{3} + 63712598n^{2} + 84804508n + 44608800) B(n + 2) -2 (n + 3) (41160n^{5} + 591920n^{4} + 3361281n^{3} + 9453847n^{2} + 13262292n + 7512000) B(n + 3)$$

 $+3(3n+11)(3n+10)(n+4)(2058n^3+14161n^2+32361n+24720)B(n+4) = 0.$

Let $S' = \{[3, -3], [2, -2], [1, -1], [1, 1], [2, 2], [3, 3]\}$. Let G' denote the g.f. for the number of unbounded bridges subject to step set S'. Then G' satisfies

$$(108t^{6} - 99t^{4} - 52t^{2} - 44)^{2} (4t^{6} + 4t^{4} + 8t^{2} - 1)^{4} G'^{8} + 4 (36t^{6} + 29t^{4} + 24t^{2} - 20) (108t^{6} - 99t^{4} - 52t^{2} - 44) (4t^{6} + 4t^{4} + 8t^{2} - 1)^{3} G'^{6} + 2 (2160t^{12} + 4248t^{10} + 4347t^{8} + 992t^{6} - 16t^{4} - 1184t^{2} + 976) (4t^{6} + 4t^{4} + 8t^{2} - 1)^{2} G'^{4} + 4 (112t^{12} + 288t^{10} + 665t^{8} + 412t^{6} + 552t^{4} - 112t^{2} + 96) (4t^{6} + 4t^{4} + 8t^{2} - 1) G'^{2} + (4t^{6} + 3t^{4} + 12t^{2} + 4)^{2} = 0.$$

Again, EqualUnBoundedScoringPaths(S',t,G') originally gave a more complicated polynomial before FindProperRoot was added. We cannot use AZd for this case. However, we can still obtain a recurrence using algeorec. The conversion only took about 13 seconds but produced a recurrence of order 66 and degree 3 and as such is not included here.

3.6.5 Time Comparison

To reiterate the benefits of this paper and Maple package over guessing for the polynomial, the following examples illustrate the time savings.

Example 3.25 (Finding a Polynomial). Let $S = \{[1, -2], [1, -1], [1, 0], [1, 1]\}$. Let G be the g.f. for bridges subject to step set S. Then

$$(16t3 + 8t2 + 11t - 4)G3 + (3 - 2t)G + 1 = 0.$$

This takes

10010 0.10	$\frac{1}{1000} \frac{1}{1000} \frac{1}{1000$				
Method	Memory Used	Memory Allocation	CPU Time	Real Time	
EUSP	2.10MiB	0 bytes	$18.27\mathrm{ms}$	$21.53 \mathrm{ms}$	
Empir	34.94MiB	$0.73 \mathrm{MiB}$	$294.93 \mathrm{ms}$	$295.63\mathrm{ms}$	
EmpirF	1.44MiB	0 bytes	$14.27\mathrm{ms}$	$16.23 \mathrm{ms}$	

Table 3.13: Finding Minimal Polynomial $S = \{[1, -2], [1, -1], [1, 0], [1, 1]\}$

In this case, **EmpirF** appears to be (25%) faster than **EqualUnBoundedScoring-Paths** (**EUSP**) but the time is so small the difference is likely to go unnoticed. The next example shows why the **ScoringPaths** package is superior to using **EmpirF**.

Example 3.26 (Finding a Polynomial 2). Let $S = \{[1, -2], [1, -1], [1, 0], [1, 1], [1, 2]\}$. Let *G* be the g.f. for bridges subject to step set *S*. Then

$$(5t+4)(5t-1)^2(t-1)^2G^4 + 2(t-1)(5t-2)(5t-1)G^2 + t = 0.$$

This takes

Table 3.	Table 3.14: Finding Minimal Polynomial $S = \{[1, -2], [1, -1], [1, 0], [1, 1], [1, 2]\}$						
Method	Memory Used	Memory Allocation Change	CPU Time	Real Time			
EUSP	7.70MiB	0 bytes	$58.33 \mathrm{ms}$	$61.43 \mathrm{ms}$			
Empir	377.91MiB	31.29MiB	2.89s	2.75s			
EmpirF		Failed					

m 11 9 14 **٦** 614 0] [1

EmpirF happened to fail for this S while EqualUnBoundedScoringPaths (EUSP) only tripled in time and memory usage. The only difference between this example and Example 3.25 is the second step with positive y-value.

Eventually guessing should work, but we have no way of knowing what the upper bound of search space is before blindly investigating. Expressing the g.f. equations is a set amount of time that is guaranteed to produce the solution. The time complexity to find the minimal polynomial with Groebner bases is an interesting question we leave to the reader.

We have many different ways to enumerate the number of bridges with step set \mathcal{S} . We will compare their speed and memory usage here.

Example 3.27 (Speed Enumeration - Unbounded). As with most previous examples, we use the benchmark step set $S = \{[1, -2], [1, -1], [1, 0], [1, 1], [1, 2]\}$. The following are time and memory requirements for enumerating 1000 terms of the sequence of bridges with step set S. The different methods are

- 1. Iterating the single polynomial found with EqualUnBoundedScoringPaths does not work due to convergence issues beyond the scope of this paper.
- 2. taylor cannot be used. It gives the error: "does not have series solution". taylor tends to work only if the minimal polynomial contains a G^1 term. See Example 3.25.
- 3. Iterating a vector solution with EqualUnBoundedScoringPathsSeries.
- 4. Brute force recursion using **SpecificUnBoundedScoringPathsNumber**.
- 5. Converting minimal polynomial to recurrence:

- (a) Obtain the polynomial with EqualUnBoundedScoringPaths.
- (b) Convert to recurrence with algebraic. The coefficients satisfy

$$0 = 125(n+1)(n+2)B(n) - 25(n+2)(n-1)B(n+1)$$

- 5(21n² + 89n + 96)B(n+2) + (n² - 43n - 140)B(n+3)
+ 2(n+4)(n+7)B(n+4).

(c) Enumerate using SeqFromRec.

6. Using AZd and then enumerating with SeqFromRec.

$$0 = 25(3n+8)(n+2)(n+1)B(n) - 5(3n+5)(2n+5)(n+2)B(n+1) - (3n+8)(19n^2 + 76n + 75)B(n+2) + 2(3n+5)(2n+5)(n+3)B(n+3).$$

Method	Memory Used	<u>m Unbounded Enume</u> Memory Allocation		Real Time
Vector Set-Up	20.47KiB	0 bytes	$200 \mu s$	$200 \mu s$
Vector Iterating	$2.79 \mathrm{TiB}$	$30.54 \mathrm{MiB}$	$2.27\mathrm{h}$	$110.54 \mathrm{m}$
Polynomial	7.76MiB	0 bytes	$59.40\mathrm{ms}$	$63.03 \mathrm{ms}$
Conversion	$5.97 \mathrm{MiB}$	8KiB	$49.03 \mathrm{ms}$	$51.97 \mathrm{ms}$
Enumeration	$20.64 \mathrm{MiB}$	2.16MiB	84.87ms	$81.50 \mathrm{ms}$
Total	34.37MiB	2.17MiB	193.30ms	196.50ms
AZd	2.45MiB	0 bytes	$25.77\mathrm{ms}$	$30.53 \mathrm{ms}$
SeqFromRec	19.34MiB	3.59MiB	74.40ms	74.43ms
Total	21.79MiB	3.59MiB	100.17ms	$104.96 \mathrm{ms}$
Brute-Force Recursion	1.41GiB	4.29MiB	12.69s	11.91s

Table 3.15: 1000 term Unbounded Enumeration

The alternate method is the fastest in this case. It is about half of the time as converting, but the actual enumeration after the preliminary set-up is very similar; the memory usage is similarly low. However, **AZd** is only appropriate for specific cases with constant x-step. The key take-away is that enumerating by combining **ScoringPaths** and **algtorec** is much faster and leaner than brute-force recursion and applicable in a wide-range of cases. If we change the step set back to $S = \{[1, -2], [1, -1], [1, 0], [1, 1]\}$, then we can use **taylor** on the minimal polynomial, as well as iterating the single solution (after an appropriate change of iteration). The difference is that Example 3.25 had a G^1 term in its minimal polynomial while Example 3.26 did not. This allows one to write G as copies of itself, which is what the iterative method is effectively accomplishing.

3.7 Asymptotics

For determining asymptotic behavior, I avoided analyzing the bounded cases as those resulted in explicit rational solutions, which can already be handled very easily. The asymptotic behavior of unbounded general walks is simply finding the number of combinations of x-steps to yield a length of n; the actual walk altitude does not matter.

For semi-bounded and unbounded cases, once we have recurrences for the coefficients, we can derive asymptotic expressions. Because of the nature of these quantities, the number of walks of length n will always follow b^n for some base b [FS09]. Wimp and Zeilberger [WZ85] created a method for automatically determining asymptotics for a linear recurrence, including the constant of the leading term! Their package **AsyRec** provides **Asy**, which attempts to determine the exponential power with which a recurrence grows. **AsyC**, along with suitable starting values, will also give the correct constant.

One task we can use this for is finding what proportion of walks are nonnegative. Let $S = \{[1/2, -1], [1/2, 1]\}$.³⁶ Let F denote the g.f. for nonnegative excursions, and B(n) denote the number of nonnegative excursions of length n. Then³⁷

$$\begin{split} 0 &= 1 - F + tF^2, \\ 0 &= (4n+2)B(n) - (n+2)B(n+1), \\ B(n) &\sim \frac{4^n}{\sqrt{\pi}n^{3/2}} \cdot \left(1 - \frac{9}{8n} + \frac{145}{128n^2} - \frac{1155}{1024n^3} + \frac{36939}{32768n^4} - \frac{295911}{262144n^5}.\right) \end{split}$$

 $^{^{36}}x\math{-}step$ is 1/2 so that excursions and bridges do not all have even length, producing a bunch of extraneous 0s.

³⁷All of this can be produced by the one command **EqualSemiBoundedPaper**(S,20,true,true).

Because the minimal polynomial is simple, we could use the Lagrange Inversion Formula to compute exact values of B(n).³⁸ But for most step sets with $|\mathcal{S}| > 2$, this will not work, so the LIF is not considered here. Let G denote the g.f. for bridges and C(n) the number of bridges of length n. Then³⁹

$$0 = 1 + (4t - 1)G^{2},$$

$$0 = (4n + 2)C(n) - (n + 1)C(n + 1),$$

$$C(n) \sim \frac{4^{n}}{\sqrt{\pi}\sqrt{n}} \cdot \left(1 - \frac{1}{8n} + \frac{1}{128n^{2}} + \frac{5}{1024n^{3}} - \frac{21}{32768n^{4}} - \frac{399}{262144n^{5}}\right).$$

Then the proportion of binary bridges that are Dyck paths is asymptotically

$$\frac{B(n)}{C(n)} \sim \frac{1}{n}.$$

This matches with the known exact proportion of $\frac{1}{n+1}$.

The above example was not exactly revolutionary, but the method allows for "quick" analysis of walks with any step set.

Dyck paths have been studied quite extensively. The typical result is for walks with step set $\{[1,0], [0,1]\}$ from (0,0) to (n,n) that stay below the line y = x. This is equivalent to enumerating nonnegative excursions with step set $\{[1,-1], [1,1]\}$. Changing the slope of the upper bound to rational $\frac{a}{b}$ is equivalent to using a step set $\{[1,-a], [1,b]\}$; this is called Rational Catalan Combinatorics. For a general conversion from walks below a line with slope m, reflect the walk across the line through the origin with slope $\frac{m}{2}$. This produces an equivalent nonnegative walk in the right hand plane. Andrew Lohr studied the field of paths below rational slope with the goal of obtaining asymptotic constants [LZ17]. For his results, go to http://sites.math.rutgers.edu/~ajl213/DrZ/RSP.pdf.

Lohr states a result by Duchon in 2000 [Duc00] that for any slope $\frac{a}{b}$, the number of paths below a line of that slope is asymptotically $\Theta\left(\frac{1}{n}\binom{(a+b)n}{an}\right)$. Are we able to say anything about step sets of size > 2? Intuition would say the growth rate should be larger, but how much larger?

³⁸This is what Duchon used for several simple cases [Duc00].

³⁹This can be produced by **UnBoundedPaper**(S,20,true,true).

First we need to tackle the scenario if we have steps with x-step 0. Recall we can only have steps directly up OR down. The maximum number of up-steps we could have in an excursion or bridge of length n is

$$n\frac{-\min\{\frac{y}{x}:(x,y)\in\mathcal{S},x\neq0\}}{\min\{y:(0,y)\in\mathcal{S}\}}$$

Replace the mins with maxs for down-steps.⁴⁰ We can only take so many steps in one direction before we have to start returning in order to make it to the *x*-axis before length n. Let c denote the fraction above (c = 0 for step sets without vertical steps). The "worst case" scenario⁴¹ is that the remaining steps all have *x*-step 1. Then we have $\binom{n+cn}{n}$ ways to place the vertical steps. Let S denote the size of S with the 0-steps removed. Then the number of walks is upper bounded (usually very poorly) by

$$\binom{n(1+c)}{n} \cdot S^n \approx \sqrt{\frac{1+c}{2\pi cn}} \left[S(1+c)(1+1/c)^c \right]^n.$$

So the number of every type of walk is $O\left(\frac{b^n}{\sqrt{n}}\right)$ for some $b.^{42}$ The bound may seem to contradict the Dyck path example, but remember that we cut the steps in half to avoid extraneous 0s. A good goal would be to improve this bound to something more meaningful.

Example 3.28. Let $S = \{[1, -1], [1, 0], [1, 2]\}$. Then, assuming F, B, G, C are as before,

$$\begin{split} 0 &= 1 + (t-1)F + t^3 F^3, \\ 0 &= 31(n+1)(n+2)B(n) - 6(2n+5)(n+2)B(n+1) \\ &+ 2(6n^2 + 36n + 53)B(n+2) - 2(2n+9)(n+3)B(n+3), \\ B(n) &\sim 0.8001188640 \cdot \frac{2.889881575^n}{n^{3/2}} \cdot \left(1 - \frac{1.7475722}{n} - \frac{2.6532889}{n^2} + \frac{4.0131981}{n^3}\right), \end{split}$$

⁴⁰We technically do not need the $x \neq 0$ since we assumed there were no steps directly down.

⁴¹Largest number of walks.

⁴²Which matches with the earlier mentioned asymptotic.

and

$$0 = 1 - 3(t - 1)G + (31t^3 - 12t^2 + 12t - 4)G^3,$$

$$0 = 31(n + 1)(n + 2)C(n) - 6(n + 2)(2n + 3)C(n + 1)$$

$$+ 2(6n^2 + 24n + 23)C(n + 2) - 2(n + 3)(2n + 3)C(n + 3),$$

$$C(n) \sim 0.3488331868 \cdot \frac{2.889881575^n}{\sqrt{n}} \cdot \left(1 - \frac{0.24757219}{n} - \frac{0.03549572}{n^2} + \frac{0.046925761}{n^3}\right).$$

The exponential bound is simply $|S|^n = 3^n$: fairly close to the actual asymptotic base. And the proportion of excursions to bridges is

$$\frac{B(n)}{C(n)} \sim \frac{2.293700526}{n} :$$

also just a constant times $\frac{1}{n}$.

Finding the asymptotic behavior does not work in every case, e.g., meanders with step set $\{[0, -1], [1, -1], [2, -1], [2, 0], [2, 1]\}$. If the asymptotic does not match with empirical data, then **AsyC** will notify the user of this.

Is the ratio of nonnegative excursions to bridges always some constant times $\frac{1}{n}$? The following example is more experimental evidence for this conjecture.

Example 3.29. Let $S = \{[1, -2], [3, 0], [0, 1], [2, 1], [2, -2]\}$ and F, B, G, C as before. Then

$$0 = 1 + (t-1)(t^{2} + t + 1)F + t(1+t)(t^{2} + 1)^{2}F^{3},$$

$$B(n) \sim \frac{4}{15} \cdot \frac{7.898354145^{n}}{n^{3/2}},$$

$$0 = 1 - 3(t-1)(t^{2} + t + 1)G + (4t^{9} + 15t^{6} + 27t^{5} + 54t^{4} + 66t^{3} + 27t^{2} + 27t - 4)G^{3},$$

$$C(n) \sim \frac{15\sqrt{3}}{58} \cdot \frac{7.898354145^{n}}{\sqrt{n}}.$$

The actual recurrences obtained from conversion (algtorec) are not listed here due to size. They are degree 2 order 18 and degree 2 order 16, respectively. The base of the exponential is exactly the same for B and C; they are the same root of the same polynomial.⁴³ And once again,

$$\frac{B(n)}{C(n)} \sim \frac{232\sqrt{3}}{675} \cdot \frac{1}{n}.$$

 $^{^{43}}$ The polynomial in question is $4x^9 - 27x^8 - 27x^7 - 66x^6 - 54x^5 - 27x^4 - 15x^3 - 4$ and the root has index=1 in Maple notation.

To compute the ratio for any step set, use the shorthand RatioOfWalks(S).

We can also try to reason the ratio heuristically. Suppose we start with an excursion, E. We can reorder the steps cyclically (or in any order) and still have a bridge. Consider the set of such walks; there are size of E such walks. The reason for reordering cyclically would be to maintain the "unique" excursion in the set, while the rest are bridges. Actually, Eis only unique if it was an irreducible excursion but potentially there is not much overlap.

Similarly, from a bridge B we could reorder the steps cyclically and we must obtain at least 1 excursion: starting from the step after the minimum altitude of B. We actually have the same number of excursions in this cyclic set as points of minimum altitude.

We have a relation between excursions and bridges that appears to be roughly linear. Some issues may be that an excursion of length n does not necessarily have size n. We still have a linear relationship between the two bounded by $\max\{x, (x, y) \in S\}$. And even steps with x-step 0 do not mess this up too horribly because we are considering excursions and bridges; the walk returns to the x-axis so can only go so far away before it must start returning since the return rate $\frac{y}{x}$ is finite (we do not have steps directly up AND down). This again leads to a linear number of x-step 0 allowed steps.

3.7.1 Discriminant

Another interesting result from analyzing asymptotic behavior is what the base of the exponent appears to be. For all three of our examples, the base can be found by taking the reciprocal of the smallest modulus of the roots of the coefficient of the leading term in the minimal polynomial of the bridges g.f., i.e.,

$$\left(\min\{|z|:4z-1=0\}\right)^{-1} = 4,$$

$$\left(\min\{|z|:31z^3 - 12z^2 + 12z - 4 = 0\}\right)^{-1} = \frac{3}{2^{2/3}} + 1 \approx 2.89,$$

$$\left(\min\{|z|:4z^9 + 15z^6 + 27z^5 + 54z^4 + 66z^3 + 27z^2 + 27z - 4 = 0\}\right)^{-1} = 7.90.$$

This is the first estimate of the asymptotics from the inverse of the radius of convergence. Actually, the polynomial whose root modulus we need can be taken to be the discriminant of the minimal polynomial, which in the step sets given is the same for excursions and bridges.

In general, one method that appears to find the base b for the exponential asymptotic behavior is

- 1. Find the discriminant of the minimal polynomial.
- 2. Take the smallest positive real root.
- 3. Take the reciprocal of that root.

This will miss the possible sub-exponential factors of $n^{-3/2}$, etc. The shorthand for this computation is implemented as **AsymptoticBase**. A more detailed analysis of singularities and the associated asymptotics is available in *Analytic Combinatorics* by Flajolet and Sedgewick [FS09].

3.7.2 Meanders

We have yet to analyze the asymptotic behavior of meanders. Let K denote the g.f. and D(n) the number of nonnegative meanders of length n. For the case of Dyck paths with a step set of $\{[1,1],[1,-1]\}$ (since we can have odd length walks now), meanders satisfy

$$1 + (2t - 1)K + t(2t - 1)K^{2} = 0,$$

$$4(n+1)D(n) + 2D(n+1) - (n+3)D(n+2) = 0,$$

$$D(n) \sim \frac{0.797}{\sqrt{n}}2^{n}$$

This says there are roughly 2.283 nonnegative meanders for each bridge of the same length. All of the asymptotic behavior in this section was found using the conversion to recurrence and extracting the asymptotics from there.

It turns out that meanders can follow very different asymptotic behavior depending on the step set. **Example 3.30.** For $S = \{[1, -1], [1, 0], [1, 2]\},\$

$$\begin{split} 0 &= 1 + (4t-1)K + 3t(3t-1)K^2 + t(3t-1)^2K^3, \\ 0 &= 93(n+1)(n+2)(n+3)D(n) - 2(80n+293)(n+3)(n+2)D(n+1) \\ &+ (n+3)(115n^2+793n+1318)D(n+2) \\ &- 4(18n^3+210n^2+820n+1071)D(n+3) \\ &+ 4(n+4)(7n^2+65n+152)D(n+4) - 2(n+5)(n+4)(2n+11)D(n+5), \\ D(n) &\sim \frac{3-\sqrt{5}}{2}3^n. \end{split}$$

The base of the exponent happens to match the inverse of the smallest modulus of the roots of the discriminant of the minimal polynomial.

So meanders can, by virtue of their endpoint flexibility, greatly outnumber excursions and bridges.

Finally, meanders do not always follow $|S|^n$. Let $S = \{[1, -1], [1, 0], [1, 1], [2, 2]\}$. Then

$$0 = 1 + (3t^2 + 3t - 1)K + t(3t + 1)(t^2 + 3t - 1)K^2 + t^2(t^2 + 3t - 1)^2K^3,$$

$$D(n) \sim 0.307 \left(\frac{3 + \sqrt{13}}{2}\right)^n \approx 0.307 \cdot 3.303^n.$$

The recurrence found was 11^{th} order and 3^{rd} degree so is not included here.⁴⁴ It was produced in less than one second using **algtorec**. And $S = \{[1, -2], [2, -1], [1, 0], [1, 2], [2, 1]\},$ yields relations of

$$0 = 1 + (2t^2 + 3t - 1)K + t(t + 2)(2t^2 + 3t - 1)K^2 + t(2t^2 + 3t - 1)^2K^3 + (2t^2 + 3t - 1)^2t^2K^4,$$
$$D(n) \sim \frac{7\sqrt{2}}{13\sqrt{n}} \left(\frac{3 + \sqrt{17}}{2}\right)^n \approx \frac{0.7615}{\sqrt{n}} \cdot 3.562^n.$$

The recurrence for this step set was 25^{th} order and of 4^{th} degree.

 $^{^{44}}$ The 0.307 is actually a root of $169x^4 - 1014x^3 + 507x^2 - 78x + 4$. Maple's identify command was used on experimental data.

The behavior of meanders appears to be a lot harder to peg down than the anticipated $\frac{b^n}{\sqrt{n}}$ for bridges and $\frac{b^n}{n^{3/2}}$ for excursions. Though it appears that meanders are always at least as large as bridges asymptotically.

3.8 Applications

3.8.1 Combining Solutions

After obtaining all of these g.f.s, we can produce much more. Sports are always a subject of interest for a good portion of the population. Sports statistics are an integral part of many a fan base. So a question of interest beyond the number of ways to be tied, may be the number of ways to win by at least X. The task at hand then is how to describe a walk such as that. We have spent the majority of this paper breaking down walks into smaller components. We can now use those components to build other types of walks. The important step is making sure that we count all of our walks, and do not double count any walks.

Suppose we want to win by ≥ 2 and never trail by more than 3. $f_{3,5}$ will count walks that drop by no more than 3, and we will have a 2 point lead at the end. Then any walk that stays above that line (k_0) will produce what we want. So does $f_{3,5} \cdot k_0$ count what we want? Not necessarily. $f_{3,5}$ ensures that at some point we are exactly 2 points ahead of where we began. But depending on the step set, we may skip over this lead and never actually hit the altitude 2 steps higher than our beginning. In addition, $f_{3,5}$ may finish with a step up and then down, while k_0 could start that way; effectively tracing the same walk in "different" ways. Thus, $f_{3,5} \cdot k_0$ double counts some walks and misses others.

There is at least one way, though not as elegant, of describing these walks. Let L denote the g.f. of interest. Then $L = k_3 - f_{3,0} - f_{3,1} - f_{3,2} - f_{3,3} - f_{3,4}$. We count all meanders that never drop more than 3 points, and then remove those that change in altitude by exactly -3, -2, -1, 0, 1. **Example 3.31.** Let $S = \{[0, -1], [1, 0], [1, 1], [1, 2]\}$. Then L satisfies

$$\begin{split} 0 &= t \bigg(3t^{12} - 63t^{11} + 555t^{10} - 2673t^9 + 7671t^8 - 13371t^7 \\ &\quad + 13745t^6 - 7554t^5 + 1615t^4 + 179t^3 - 138t^2 + 21t - 1 \bigg) \\ &\quad + \bigg(9t^{14} - 216t^{13} + 2286t^{12} - 13986t^{11} + 54558t^{10} - 141402t^9 + 246687t^8 \\ &\quad - 288270t^7 + 221709t^6 - 109548t^5 + 33981t^4 - 6448t^3 + 715t^2 - 42t + 1 \bigg) L \\ &\quad - 3t^3 \big(3t^7 - 36t^6 + 153t^5 - 261t^4 + 126t^3 + 50t^2 - 26t + 2 \big) L^2 + 9t^6 L^3, \end{split}$$

and has truncated expansion

$$L = t + 21t^{2} + 305t^{3} + 4064t^{4} + 52431t^{5} + 666657t^{6} + 8420130t^{7} + 106070229t^{8} + 1335635352t^{9} + 16832212452t^{10}.$$

The minimal polynomial was derived by finding a Groebner basis. We had to describe all of $\{k_3, f_{3,0}, f_{3,1}, f_{3,2}, f_{3,3}, f_{3,4}\}$ as well, which included a chain of describing further walks. But those methods are already set and have been shown to be closed.

3.8.2 Weighted Walks

One extension that would be fairly easy to implement is adding a weight to each step; [x, y] has an associated weight w. Then, for example, we would write⁴⁵

$$f_{a,b} = 1 + \sum_{(x,y,w) \in \mathcal{S}} w \cdot t^x f_{a-y,b-y}.$$

Simply multiply by the weight whenever we take a certain step. Now we can accomplish more with the weights in place. If $\sum_{(x,y,w)\in S} w = 1$ (and all x = 1), then w represents the probability of taking a specific step. And then $f_{a,b}$ is the g.f. for the probability that a given walk of length n maintains altitude $a \ge y \ge b$. If we want to know the probability of a bounded walk being a bounded bridge, use weights to describe $f'_{a,b}$, the g.f. for probability of a general walk being a bounded bridge, and divide its coefficients by the appropriate coefficient of $f_{a,b}$, the probability of a general walk being bounded. This is an explicit

⁴⁵Taken from general walks in Bounded Section 3.2.

description of using $\Pr[A|B] = \frac{\Pr[A \cap B]}{\Pr[B]}$. If all of the weights are the same, then simply enumerate each one to get the probabilities; it is much easier.

Example 3.32. Let us try finding the probability of a general walk with step set $S = \{[1, 2, 1/3], [1, -1, 1/6], [1, -2, 1/2]\}$ being bounded above by y = 3 and below by y = -2. Then the g.f. is explicitly

$$3\frac{3888+3888t-756t^2-972t^3-12t^4-t^5}{11664-7776t^2-432t^3+1296t^4+t^6}$$

which has taylor expansion

$$1 + t + \frac{17}{36}t^2 + \frac{49}{108}t^3 + \frac{77}{324}t^4 + \frac{811}{3888}t^5 + \frac{53}{432}t^6 + \frac{3407}{34992}t^7 + \frac{26483}{419904}t^8 + \frac{58247}{1259712}t^9 + O\left(t^{10}\right)$$

We need the steps to all have the same x-value, otherwise we aren't representing the probability that a given walk of length n has some property. We would somehow be combining the probability that a walk has length n and the probability that it satisfies our desired property(ies) in a way I cannot currently describe.

3.8.3 2-step Examples

With all of the tools at our disposal, let's now use them to produce more information about many sequences.

Bridges with $S = \{[1, 1], [1, -k]\}$

It is well known⁴⁶ that the g.f. for nonnegative excursions, denoted f, satisfies

$$f = 1 + t^{k+1} f^{k+1}.$$

 $^{^{46}\}mathrm{Try}$ proving it for yourself. Or prove it for any finite k with this package.

We would like to show something about bridges with this step set. The g.f. for the first k = 0, ..., 7 step sets have minimal polynomials

$$\begin{split} (t-1)G+1, \\ (4t^2-1)G^2+1, \\ (27t^3-4)G^3+3G+1, \\ (256t^4-27)G^4+18G^2+8G+1, \\ (3125t^5-256)G^5+160G^3+80G^2+15G+1, \\ (46656t^6-3125)G^6+1875G^4+1000G^3+225G^2+24G+1, \\ (823543t^7-46656)G^7+27216G^5+15120G^4+3780G^3+504G^2+35G+1. \end{split}$$

The general form appears to have leading term $([(k+1)t]^{k+1} - k^k)G^{k+1}$. The G^1 term is fairly easy to see have coefficient (k+1)(k-1). The G^2 term, after some manipulations, has coefficients that follow $\frac{k^2}{2}(k+1)(k-2)$. The computer could only obtain up to k = 9before memory requirements became too large (> 3GB allocated). From that limited data, I found that the G^3 term has coefficient that follows $\frac{k^3}{6}(k+1)(k-1)(k-3)$. Does this generalize? And if so, how?

These results are not groundbreaking as we can already enumerate the number of bridges of length (a + b)n with step set $\{[1, a], [1, -b]\}$ with the simple $\binom{(a+b)n}{an}$. They do allow a different view of the walks by considering how they can be built from copies of themselves.

Duchon Numbers

The Duchon numbers⁴⁷ are one of the cases that Lohr analyzed for asymptotic behavior [LZ17]. We can derive more information for the g.f., which we will denote f, of the Duchon numbers. The sequence is defined as the number of paths of length 5n from (0,0) to the line y = 2x/3 with unit North and East steps that stay below the line or touch it. It is equivalent

⁴⁷OEIS A060941 [OEIg].

to enumerating excursions with step set $\{[1/5, 2], [1/5, -3]\}$. To find the minimal polynomial for its g.f., all we have to do is type **EqualSemiBoundedScoringPaths**(S, 0, t, f) and hit return.

$$0 = 1 - f + 2tf^5 - tf^6 + tf^7 + t^2f^{10}.$$

We can also produce the minimal polynomial for the related g.f. of irreducible walks (those that only touch at the endpoints). We almost accomplish this with the call **SpecificEqualSemiBoundedScoringPaths**($\mathcal{S}, 0, t, f, g, g[0, 0]$). But to match OEIS <u>A293946</u> [OEIj], we must allow the stationary walk. Simply substitute g[0, 0] = g - 1and we are done.

$$0 = g^{10} - 19g^9 + 162g^8 - 816g^7 + 2688g^6 - (2t + 6048)g^5 + (19t + 9408)g^4 - (73t + 9984)g^3 + (142t + 6912)g^2 - (140t + 2816)g + t^2 + 56t + 512.$$

Excursions with $S = \{[1,2], [1,-k]\}$

We have some information for this family in the Duchon numbers: k = 3. But what do they look like in general? We will assume gcd(2, k) = 1 otherwise we could reduce the steps for an equivalent problem. The first few g.f. have minimal polynomial (we have made the transformation $t \to t^{1/(2+k)}$ for ease of reading)

$$\begin{array}{rcl} \underline{A001764}:&k=1&f^{3}t-f+1,\\\\ \underline{A060941}:&k=3&f^{10}t^{2}+f^{7}t-f^{6}t+2f^{5}t-f+1,\\\\ \underline{A300386}:&k=5&f^{21}t^{3}+2f^{16}t^{2}-f^{15}t^{2}+3f^{14}t^{2}\\&+f^{11}t-f^{10}t+2f^{9}t-2f^{8}t+3f^{7}t-f+1,\\\\ \underline{A300387}:&k=7&f^{36}t^{4}+3f^{29}t^{3}-f^{28}t^{3}+4f^{27}t^{3}+3f^{22}t^{2}-2f^{21}t^{2}+6f^{20}t^{2}\\&-3f^{19}t^{2}+6f^{18}t^{2}+f^{15}t-f^{14}t+2f^{13}t-2f^{12}t\\&+3f^{11}t-3f^{10}t+4f^{9}t-f+1,\\\\ \underline{A300388}:&k=9&f^{55}t^{5}+4f^{46}t^{4}-f^{45}t^{4}+5f^{44}t^{4}+6f^{37}t^{3}-3f^{36}t^{3}+12f^{35}t^{3}\\&-4f^{34}t^{3}+10f^{33}t^{3}+4f^{28}t^{2}-3f^{27}t^{2}+9f^{26}t^{2}-6f^{25}t^{2}\\&+12f^{24}t^{2}-6f^{23}t^{2}+10f^{22}t^{2}+f^{19}t-f^{18}t+2f^{17}t\\&-2f^{16}t+3f^{15}t-3f^{14}t+4f^{13}t-4f^{12}t+5f^{11}t-f+1,\\\\ \underline{A300389}:&k=11&f^{78}t^{6}+5f^{67}t^{5}-f^{66}t^{5}+6f^{65}t^{5}+10f^{56}t^{4}-4f^{55}t^{4}+20f^{54}t^{4}\\&-5f^{53}t^{4}+15f^{52}t^{4}+10f^{45}t^{3}-6f^{44}t^{3}+24f^{43}t^{3}-12f^{42}t^{3}\\&+30f^{41}t^{3}-10f^{40}t^{3}+20f^{39}t^{3}+5f^{34}t^{2}-4f^{33}t^{2}+12f^{32}t^{2}\\&-9f^{31}t^{2}+18f^{30}t^{2}-12f^{29}t^{2}+20f^{28}t^{2}-10f^{27}t^{2}\\&+15f^{26}t^{2}+f^{23}t-f^{22}t+2f^{21}t-2f^{20}t+3f^{19}t-3f^{18}t\\&+4f^{17}t-4f^{16}t+5f^{15}t-5f^{14}t+6f^{13}t-f+1.\\ \end{array}$$

The degree is simply $\frac{1}{2}(k+1)(k+2)$. If one looks closer they may recognize that the degrees then decrease at a consistent rate. The degree drops by k and then by 1 twice to create a "group" of 3 terms. The degree then drops again by k-2 and then by 1 to create a group of 5 terms. It can be seen that each polynomial follows this $1, 3, 5, 7, \ldots$ pattern, with the power of t decreasing by 1 in each successive group. There is always a -f + 1 included.

We have empirically shown the general form of the minimal polynomial, but were unable to describe what the coefficients themselves are.

<u>A001764</u> [OEIk] and <u>A060941</u> [OEIg] are sequences currently in the OEIS. <u>A300386</u> [OEII], <u>A300387</u> [OEIm], <u>A300388</u> [OEIn], and <u>A300389</u> [OEIo] are new and have been recently submitted and accepted.

Excursions with $S = \{[1,3], [1,-k]\}$

Let us push our computers further. What does this family look like? We cannot derive a lot of information empirically as the $\{[1,3], [1,-5]\}$ case already takes 28 seconds to run. The case k = 7 ran for over one day and had used 3GiB of allocated memory before it was terminated. Again, we made the transformation $t \to t^{1/(3+k)}$ for compact reading.

$$\begin{array}{rll} \underline{A002293}:&k=1&f^4t-f+1,\\\\ \underline{A060941}:&k=2&f^{10}t^2+f^7t-f^6t+2f^5t-f+1,\\\\ \underline{A300390}:&k=4&f^{35}t^5-f^{31}t^4+f^{30}t^4-f^{29}t^4+5f^{28}t^4-f^{25}t^3+f^{24}t^3+3f^{23}t^3\\&-4f^{22}t^3+10f^{21}t^3+f^{19}t^2-f^{18}t^2+5f^{17}t^2+3f^{16}t^2-6f^{15}t^2\\&+10f^{14}t^2+f^{13}t-f^{12}t+3f^{10}t+f^9t-4f^8t+5f^7t-f+1,\\\\ \underline{A300391}:&k=5&f^{56}t^7-2f^{51}t^6+f^{50}t^6-f^{49}t^6+7f^{48}t^6+f^{46}t^5-f^{45}t^5-3f^{43}t^5\\&+5f^{42}t^5-6f^{41}t^5+21f^{40}t^5-3f^{37}t^4-3f^{36}t^4+8f^{35}t^4+10f^{34}t^4\\&-15f^{33}t^4+35f^{32}t^4-2f^{31}t^3+2f^{30}t^3-9f^{28}t^3+22f^{27}t^3\\&+10f^{26}t^3-20f^{25}t^3+35f^{24}t^3+3f^{22}t^2+5f^{21}t^2-9f^{20}t^2\\&+18f^{19}t^2+5f^{18}t^2-15f^{17}t^2+f^{16}\left(21t+1\right)t-f^{15}t+3f^{13}t\\&-3f^{12}t+5f^{11}t+f^{10}t-6f^9t+7f^8t-f+1.\end{array}$$

<u>A002293</u> [OEIp] and <u>A060941</u> [OEIg] are already in the OEIS while <u>A300390</u> [OEIq] and <u>A300391</u> [OEIr] are new. There appears to be some pattern again in the degree of f, though it is not discernible at first with so few datum. The degree happens to follow $\binom{3+k}{3}$.

Conjecture 3.33. Let $S = \{[1, a], [1, -b]\}$ with gcd(a, b) = 1. Let f denote the g.f. for excursions (or meanders or bridges) with step set S. Then the minimal polynomial of f has degree $\binom{a+b}{a}$.

The above conjecture is supported by all of the examples in this Section 3.8.3, as well as several other quick tests of accuracy. Though what does this mean in terms of deconstructing a walk with step set S?

This is somewhat intuitive for the unbounded case since there are $\binom{a+b}{a}$ ways to have a bridge of smallest length: a+b. And we have provided ample evidence that the asymptotic behavior is only off by $\frac{c}{n}$ for these excursions.

3.9 Conclusion and Future Work

To analyze walks, we began by examining the first step (Bounded Section 3.2), the last step (Semi-bounded Section 3.3), and finally a middle step that crosses the x-axis (Unbounded Section 3.6).

The way we dissected the g.f. equations is not unique. You could describe them in slightly different ways that may be more optimal. However, the minimal polynomial is called that for a reason; there is no "better" way to describe the g.f. except for an exact solution in special cases. It is important to make sure you do not double count or miss any walks in your descriptions. Define your various types of walks in a very particular manner.

For bounded cases we produced the new OEIS sequences <u>A301379</u> [OEIh] and <u>A301380</u> [OEIi] as well as <u>A301381</u> [OEIs] and <u>A300998</u> [OEIt], which are not included in this thesis, though they were produced in conjunction. Enumerating some semi-bounded and unbounded examples may be much faster by actually solving the polynomial for the g.f. in cases where deg(p) ≤ 4 or p has "nice" roots.

One interesting note that appears in the semi-bounded case is that, no matter what we have chosen as our step set, the minimal polynomial has had terms -F + 1. This seems to indicate that there is always a way to write semi-bounded excursions and meanders as some combination (and deductions) of copies of ONLY ITSELF. Though it may be that the self-description is extremely complicated seemingly without (but it must be there) any combinatorial interpretation: see Example 3.10. We cannot necessarily do that with unbounded cases because there is not always a -F term.

Another trick we can use the minimal polynomial for is a bijective proof. All of the walks counted by terms with positive coefficients are also equally counted by those terms with negative coefficients. The trouble with this is that the two sides are very artificial and rarely something of interest by themselves.

In Section 3.7, we tried looking at the asymptotics of excursions and bridges. One note that came up was the relationship between the number of excursions and bridges. It is known that the number of Dyck paths is $\frac{1}{n+1}$ of the total number of bridges. With a different looking step set, we still obtained a $\frac{c}{n}$ asymptotic relationship (for a constant c). Is this always the case? We provided some heuristics and examples in support of the relationship but no found definitive proof. The relationship to meanders of the same step set appears to be much harder to state exactly.

A little further in Section 3.8.3, we examined many 2-step cases. I contributed the new sequences <u>A300386</u> [OEII], <u>A300387</u> [OEIm], <u>A300388</u> [OEIn], <u>A300389</u> [OEIo], <u>A300390</u> [OEIq], and <u>A300391</u> [OEIr] to the OEIS. These are equivalent to walks that stay below certain lines of rational slope. A further question: how does one translate walks with general step sets bounded by lines into bridges, excursions, or meanders? What about bounded by something other than straight lines?

An extension that would be fairly easy, though laborious to implement, would be generalizing from 2D walks to 3D walks. Or to n-dimensional walks.

Beyond simply enumerating walks, we might want to know more about them: how many peaks or valleys do they contain? what is the area beneath the curve? how many times does the walk hit its maximal/minimal altitude? We can try to answer these questions by adding in a catalytic variable to count this new measurement. The generating function relations are very similar, but the g.f.s themselves are now functions of 2 (or more) variables. This can lead to systems that are not closed (under current descriptions). However, the system can still be iterated to enumerate terms.

Ayyer and Zeilberger analyzed how many times a bounded bridge or excursion hits its boundaries [AZ08] in the bounded and semi-bounded cases. The extension to measuring the area under the curve has been started. The bounded case still yields explicit solutions but the semi-bounded and unbounded cases are now left as a system of equations rather than a minimal polynomial.

Thank you for reading this chapter. I hope you have enjoyed it and can make use of this package.

Chapter 4

Conclusion

Let us look back at what we have accomplished.

In Chapter 2 we adapted a method of proving the unimodal nature of q-binomials to build more families of unimodal polynomials. The special facet of the constructed families is that the unimodal nature is not immediately obvious. Either an explicit formula or a recurrence with the correct initial conditions produce surprisingly unimodal polynomials.

In order to enumerate lattice walks in Chapter 3 we generalized the technique of describing generating function relations with irreducible walks. This allowed us to give exact generating functions in the bounded case and minimal polynomials in semi-bounded and unbounded cases. Connected to these minimal polynomials are recurrences for enumeration. Once we take a step back, we realize that they produce surprisingly integral sequences given the correct initial conditions.

In both major chapters, we produced new sequences for the OEIS. We have revised results of mathematicians before us to work in more ways than originally intended. We created two Maple packages to accompany the work and provide avenues for quick and easy investigation into more general questions. We contributed to the field of experimental mathematics by devising methods for exploration and using them to gain insights into unimodal polynomial construction and lattice walk generating functions that we otherwise could not have gleaned.

References

- [AAR00] Jenny Alvarez, Miguel Amadis, and Leobardo Rosales, Unimodality and logconcavity of polynomials, Tech. report, The Summer Institute in Mathematics for Undergraduates (SIMU). (Cited on page 161.), 2000.
- [And98] George E. Andrews, *The Theory of Partitions*, Cambridge, England: Cambridge University Press, July 1998.
- [AZ90] Gert Almkvist and Doron Zeilberger, *The method of differentiating under the integral sign*, Journal of Symbolic Computation **10** (1990), 571–591.
- [AZ07] Arvind Ayyer and Doron Zeilberger, The Number of [Old-Time] Basketball games with Final Score n:n where the Home Team was never losing but also never ahead by more than w Points, The Electronic Journal of Combinatorics 14 (2007).
- [AZ08] _____, Two dimensional directed lattice walks with boundaries, Tapas in experimental mathematics, vol. 457, Contemporary Mathematics, Amer. Math. Soc., Providence, RI, 2008, pp. 1–19.
- [BB90] Francesco Brenti and Thomas H. Brylawski, Unimodal Polynomials Arising from Symmetric Functions, Proceedings of the American Mathematical Soceity, vol. 108, April 1990, pp. 1133–1141.
- [BKK⁺17] Cyril Banderier, Christian Krattenthaler, Alan Krinik, Dmitry Kruchinin, Vladimir Kruchinin, David Tuan Nguyen, and Michael Wallner, Explicit formulas for enumeration of lattice paths: basketball and the kernel method, arXiv:1609.06473v2 (2017).
- [Bón06] Miklós Bóna, A Walk Through Combinatorics, Hackensack, New Jersey: World Scientific Publishing Co, 2006.
- [BORW05] R. Brak, A.L. Owczarek, A. Rechnitzer, and S.G. Whittington, A directed walk model of a long chain polymer in a slit with attractive walls, J. Phys. A 38 (2005), 4309–4325.
- [Brä15] Petter Brändén, Unimodality, log-concavity, real-rootedness and beyond, Handbook of Enumerative Combinatorics 87 (2015), 437.
- [DFV80] Y. I. Petunin D. F. Vysochanskij, Justification of the 3\u0333 rule for unimodal distributions, Theory of Probability and Mathematical Statistics 21 (1980), 25-36.
- [DR71] E. A. DiMarzio and R. J. Rubin, Adsorption of a Chain Polymer between Two Plates, J. Chem. Phys. 55 (1971), 4318–4336.

- [Duc00] Philippe Duchon, On the enumeration and generation of generalized Dyck words, Discrete Mathematics **225** (2000), 121–135.
- [EZ15] Shalosh B. Ekhad and Doron Zeilberger, *The Method(!) of "Guess and Check"*, arXiv:1502.04377v1 (2015).
- [FS09] Philippe Flajolet and Robert Sedgewick, *Analytic Combinatorics*, Cambridge University Press, June 2009.
- [Gau23] Carl Friedrich Gauss, Theoria Combinationis Observationum Erroribus Minimis Obnoxiae, Pars Prior, Commentationes Societatis Regiae Scientiarum Gottingensis Recentiores 5 (1823).
- [KP11] Manuel Kauers and Peter Paule, *The Concrete Tetrahedron*, Springer, 2011.
- [LZ17] Andrew Lohr and Doron Zeilberger, Asymptotics of Rational Catalan Combinatorics, RSP, April 2017.
- [MR09] Marni Mishna and Andrew Rechnitzer, *Two non-holonomic lattice walks in the quarter plane*, Theor. Computer Science **410** (2009), 3616–3630.
- [Nei] Neil J. A. Sloane, OEIS Foundation Inc. (2018), The On-Line Encyclopedia of Integer Sequences, http://oeis.org/.
- [OEIa] OEIS Foundation Inc. (2018), The On-Line Encyclopedia of Integer Sequences, http://oeis.org/A002522.
- [OEIb] OEIS Foundation Inc. (2018), The On-Line Encyclopedia of Integer Sequences, http://oeis.org/A302612.
- [OEIc] OEIS Foundation Inc. (2018), The On-Line Encyclopedia of Integer Sequences, http://oeis.org/A302644.
- [OEId] OEIS Foundation Inc. (2018), The On-Line Encyclopedia of Integer Sequences, http://oeis.org/A302645.
- [OEIe] OEIS Foundation Inc. (2018), The On-Line Encyclopedia of Integer Sequences, http://oeis.org/A302646.
- [OEIf] OEIS Foundation Inc. (2018), The On-Line Encyclopedia of Integer Sequences, http://oeis.org/A000984.
- [OEIg] OEIS Foundation Inc. (2018), The On-Line Encyclopedia of Integer Sequences, http://oeis.org/A060941.
- [OEIh] OEIS Foundation Inc. (2018), The On-Line Encyclopedia of Integer Sequences, http://oeis.org/A301379.
- [OEIi] OEIS Foundation Inc. (2018), The On-Line Encyclopedia of Integer Sequences, http://oeis.org/A301380.
- [OEIj] OEIS Foundation Inc. (2018), The On-Line Encyclopedia of Integer Sequences, http://oeis.org/A293946.

- [OEIk] OEIS Foundation Inc. (2018), The On-Line Encyclopedia of Integer Sequences, http://oeis.org/A001764.
- [OEII] OEIS Foundation Inc. (2018), The On-Line Encyclopedia of Integer Sequences, http://oeis.org/A300386.
- [OEIm] OEIS Foundation Inc. (2018), The On-Line Encyclopedia of Integer Sequences, http://oeis.org/A300387.
- [OEIn] OEIS Foundation Inc. (2018), The On-Line Encyclopedia of Integer Sequences, http://oeis.org/A300388.
- [OEIo] OEIS Foundation Inc. (2018), The On-Line Encyclopedia of Integer Sequences, http://oeis.org/A300389.
- [OEIp] OEIS Foundation Inc. (2018), The On-Line Encyclopedia of Integer Sequences, http://oeis.org/A002293.
- [OEIq] OEIS Foundation Inc. (2018), The On-Line Encyclopedia of Integer Sequences, http://oeis.org/A300390.
- [OEIr] OEIS Foundation Inc. (2018), The On-Line Encyclopedia of Integer Sequences, http://oeis.org/A300391.
- [OEIs] OEIS Foundation Inc. (2018), The On-Line Encyclopedia of Integer Sequences, http://oeis.org/A301381.
- [OEIt] OEIS Foundation Inc. (2018), The On-Line Encyclopedia of Integer Sequences, http://oeis.org/A300998.
- [O'H90] Kathleen O'Hara, Unimodality of Gaussian Coefficients: A Constructive Proof, Journal of Combinatorial Theory, Series A 53 (1990), 29–52.
- [Pro82] Robert A. Proctor, Solution of Two Difficult Combinatorial Problems with Linear Algebra, The American Mathematical Monthly 89 (1982), no. 10, 721– 734.
- [Sta89] Richard P Stanley, Log-Concave and Unimodal Sequences in Algebra, Combinatorics, and Geometry, Annals of the New York Academy of Sciences 576 (1989), no. 1, 500–535.
- [SZ94] Bruno Salvy and Paul Zimmermann, *Gfun: a Maple package for the manipulation of generating and holonomic functions in one variable*, ACM Transactions on Mathematical Software (TOMS) **20** (1994), no. 2, 163–177.
- [Vat09] Vince Vatter, *Binomial coefficients, unimodality, log-concavity*, Math 68, Algebraic Combinatorics, Dartmouth College, September 2009.
- [vR00] E. J. Janse van Rensburg, *The statistical mechanics of interacting walks, poly*gons, animals and vesicles, vol. 18, Oxford University Press, Oxford, 2000.
- [WZ85] Jet Wimp and Doron Zeilberger, *Resurrecting the asymptotics of linear recurrences*, Journal of mathematical analysis and applications **111** (1985), no. 1, 162–176.

[Zei89b] _____, Kathy O'Hara's constructive proof of the unimodality of the Gaussian polynomials, Amer. Math. Monthly **96** (1989), no. 7, 590–602.