

# Convergence rates in generalized Zeckendorf decomposition problems

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## Summary

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- Discuss new approaches to asymptotic behavior of variance
- Discuss new results on Gaussian behavior of gaps between summands

## Previous Results

## Definitions: Zeckendorf Decomposition

### Theorem (Zeckendorf)

*Let  $\{F_n\}_{n \in \mathbb{N}}$  denote the Fibonacci numbers with  $F_1 = 1$  and  $F_2 = 2$ . Every positive integer can be written uniquely as a sum of non-consecutive Fibonacci numbers.*

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### Example

$$101 = 89 + 8 + 3 + 1 = F_{10} + F_5 + F_3 + F_1$$

## Definitions: Positive Linear Recurrence Sequence

### Definition

A *Positive Linear Recurrence Sequence (PLRS)* is a sequence  $\{G_n\}$  satisfying

$$G_n = c_1 G_{n-1} + \cdots + c_L G_{n-L}$$

with nonnegative integer coefficients  $c_i$  with  $c_1, c_L, L \geq 1$  and initial conditions  $G_1 = 1$  and

$$G_n = c_1 G_{n-1} + c_2 G_{n-2} + \cdots + c_{n-1} G_1 + 1 \text{ for } 1 \leq n \leq L.$$



## Examples of PLRS

- 1 Fibonacci numbers:  $L = 2$ ,  $c_1 = c_2 = 1$ .  
 $G_1 = 1$ ,  $G_2 = 2$ ,  $G_3 = 3$ ,  $G_4 = 5, \dots$

## Examples of PLRS

- 1 Fibonacci numbers:  $L = 2, c_1 = c_2 = 1$ .  
 $G_1 = 1, G_2 = 2, G_3 = 3, G_4 = 5, \dots$
- 2 Powers of  $b$ :  $L = 2, c_1 = b - 1, c_2 = b$ .  
 $G_1 = 1, G_2 = b, G_3 = b^2, G_4 = b^3, \dots$

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 $G_1 = 1, G_2 = b, G_3 = b^2, G_4 = b^3, \dots$
- 3  $d$ -bonacci numbers:  $L = d, c_1 = c_2 = \dots = c_d = 1$ ,  
 $G_n = 2^{n-1}$  for  $n \leq d$ .

## Definition: Generalized Zeckendorf Decomposition

### Definition (Generalized Zeckendorf Decomposition)

Let  $\{G_n\}$  be a PLRS and  $m$  be a positive integer. Then

$$m = \sum_{i=1}^N a_i G_{N+1-i}$$

is a **legal** decomposition if  $a_1 > 0$  and the other  $a_i \geq 0$ , and one of the following conditions holds.

- 1 We have  $N < L$  and  $a_i = c_i$  for  $1 \leq i \leq N$ .
- 2 There exists an  $s \in \{1, \dots, L\}$  such that  $a_1 = c_1, a_2 = c_2, \dots, a_{s-1} = c_{s-1}, a_s < c_s$ , and  $\{b_i\}_{i=1}^{N-s}$  (with  $b_i = a_{s+i}$ ) is either legal or empty.

## Example

Consider the PLRS:

$$G_n = 3G_{n-1} + 2G_{n-2} + 2G_{n-4}.$$

Examples of legal decompositions:

- $m = 3G_9 + 2G_8 + G_6 + 3G_5 + G_4 + 2G_1.$
- $m = 3G_9 + 2G_8 + G_6 + 3G_5 + G_4 + 3G_1.$

Examples of NOT legal decompositions:

- $m = 4G_9.$
- $m = 3G_9 + 2G_8 + G_7.$
- $m = 3G_9 + 2G_8 + 2G_6.$

## Theorem: Generalized Zeckendorf Decomposition

### Theorem

*Let  $\{G_n\}$  be a PLRS. Then there is a unique legal decomposition for every positive integer  $m$ .*

## Definitions and Notations

### Definition

- Probability Space  $\Omega_n$ : The set of legal decompositions of integers in  $[G_n, G_{n+1})$ .
- Probability Measure: Let each of the  $G_{n+1} - G_n$  legal decompositions be weighted equally.
- Random Variables  $K_n$ : Set  $K_n(\omega)$  equal to the number of summands of  $\omega \in \Omega_n$ .

## Asymptotic Behavior of Variance



## Old Result

### Theorem (Miller and Wang, 2012)

When  $\{G_n\}$  is a PLRS, there exists constants  $A, B, C, D, \gamma_1 \in (0, 1), \gamma_2 \in (0, 1)$  such that

$$\mathbf{E}[K_n] = An + B + o(\gamma_1^n) \quad (1)$$

$$\mathbf{Var}[K_n] = Cn + D + o(\gamma_2^n) \quad (2)$$

### Remark

- Proof of (1) is **easy**.
- Proof of (2) is **hard**. Key difficulty: bound of  $C$ .

## Main New Result

### Theorem

*When  $\{G_n\}$  is a PLRS with length  $L$ , there exists  $C_{min} > 0$  such that*

$$\mathbf{Var}[K_n] > C_{min} \cdot n$$

*for all  $n > L$ .*

## Definition of Blocks

### Definition

- We define a Type 1 block as an integer sequence corresponding to Condition 1.
- We define a Type 2 block as an integer sequence corresponding to Condition 2.

### Example

$$G_n = 2G_{n-1} + 2G_{n-2} + 2G_{n-4}$$

Suppose  $m = G_5 + 2G_3 + G_2 + 2G_1$ , we write the decomposition into an integer sequence:  $[1, 0, 2, 1, 2]$ .

- The Type 2 Blocks:  $[1], [0], [2, 1]$ .
- The Type 1 Block:  $[2]$ .

## Definition of Blocks

### Definition

- We define the size of a block as the total number of summands in it.
- We define the length of a block as the total number of indices in it.

### Example

$$G_n = 2G_{n-1} + 3G_{n-3} + 2G_{n-4}$$

A possible Type 2 Block is [2, 0, 3, 1]. It has size 6 and length 4.

## Key Observations of Type 1 Blocks

Properties:

- It appears at most once in any legal decomposition.
- It has to be the last block if it does exist.

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So,

- Type 1 block matters little when  $n$  is large.
- Second to last block has to be a Type 2 block.

## Key Observations of Type 2 Blocks

- The length of a Type 2 block is fully determined by its size. So we can define a function  $\ell(t)$  such that a Type 2 block with size  $t$  has length  $\ell(t)$ .

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- The length of a Type 2 block is fully determined by its size. So we can define a function  $\ell(t)$  such that a Type 2 block with size  $t$  has length  $\ell(t)$ .
- A Type 2 block always has nonnegative size and strictly positive length.
- A legal decomposition will stay legal if we add a Type 2 block to it or remove a Type 2 block from it.

## Key Idea

### Lemma

For a fixed  $t$ , there is a bijection  $h_t$  between:

- all legal decompositions with total length  $n$  and the second to last block with size  $t$ , and
- all legal decompositions with total length  $n - \ell(t)$ .

### Example

$$G_n = 2G_{n-1} + 2G_{n-2} + 0 + 2G_{n-4}.$$

$$m = G_6 + G_5 + 2G_4 + G_1.$$

Its block representation:  $[1], [1], [2, 0], [0], [1]$ .

After removing  $[0]$ :  $[1], [1], [2, 0], [1]$ .

The resulting legal decomposition:  $G_5 + G_4 + 2G_3 + G_1$ .

## Key Idea

- Set  $Z_n(\omega)$  equal to the size of the second to last block for the legal decomposition  $\omega \in \Omega_n$ . Then if  $Z_n(\omega) = t$ , we have

$$K_n(\omega) = K_{n-\ell(t)}(h_t(\omega)) + t.$$

## Key Idea

- Set  $Z_n(\omega)$  equal to the size of the second to last block for the legal decomposition  $\omega \in \Omega_n$ . Then if  $Z_n(\omega) = t$ , we have

$$K_n(\omega) = K_{n-l(t)}(h_t(\omega)) + t.$$

- In other words, we get the conditional expectations:

$$\mathbf{E}[K_n | Z_n = t] = \mathbf{E}[K_{n-l(t)}] + t$$

$$\begin{aligned} \mathbf{E}[K_n^2 | Z_n = t] &= \mathbf{E}[(K_{n-l(t)} + t)^2] \\ &= \mathbf{E}[K_{n-l(t)}^2] + 2t \mathbf{E}[K_{n-l(t)}] + t^2. \end{aligned}$$

## Proof of Main Theorem

- We first explicitly choose  $C_{min}$ . Then, we use strong induction on  $n$  to prove  $\mathbf{Var}[K_n] > C_{min} \cdot n$ .

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- From induction hypothesis, we can bound  $\mathbf{E}[K_{n-\ell(t)}^2]$  by estimating  $\mathbf{Var}[K_{n-\ell(t)}] + (\mathbf{E}[K_{n-\ell(t)}])^2$ .

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- In inductive step, we will be able to bound  $\mathbf{Var}[K_n]$  by estimating  $\mathbf{E}[K_n^2] - (\mathbf{E}[K_n])^2$ .



## Further Works

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- Can we generalize to more general sequences? Even without a recurrence relation?
- Can we define legal decompositions and blocks for other sequences?

## Gaussian Behavior of Gaps

## Technical Note

For the rest of the talk, assume PLRS refers to recurrences

$$G_n = c_1 G_{n-1} + \cdots + c_L G_{n-L}$$

with  $c_i > 0$  for  $1 \leq i \leq L$ .

(Previously, we only needed  $c_1, c_L > 0$  and  $c_i \geq 0$  for  $1 < i < L$ .)

## Gaps in Decompositions

### Definition

Let  $m$  be a positive integer with decomposition

$$m = \sum_{i=1}^N a_i G_{N+1-i} = G_{i_1} + G_{i_2} + \cdots + G_{i_k}$$

and  $i_1 \geq i_2 \geq \cdots \geq i_k$ . Then the *gaps* in the decomposition of  $m$  are the numbers  $i_1 - i_2, i_2 - i_3, \dots, i_{k-1} - i_k$ .

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and  $i_1 \geq i_2 \geq \cdots \geq i_k$ . Then the *gaps* in the decomposition of  $m$  are the numbers  $i_1 - i_2, i_2 - i_3, \dots, i_{k-1} - i_k$ .

### Example

The gaps for the decomposition of  $m = 101$  are 5, 2, 2.

$$101 = F_{10} + F_5 + F_3 + F_1.$$

## Gaps in Decompositions

### Recall:

- Let  $k(m)$  be the number of summands in the decomposition of  $m$ .
- Let  $K_n$  be  $k(m)$  for an  $m$  chosen uniformly in  $[G_n, G_{n+1})$ .



## Gaps in Decompositions

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- Let  $K_n$  be  $k(m)$  for an  $m$  chosen uniformly in  $[G_n, G_{n+1})$ .
- Let  $k_g(m)$  be the number of size- $g$  gaps in the decomposition of  $m$ .
- Let  $K_{g,n}$  be  $k_g(m)$  for an  $m$  chosen uniformly in  $[G_n, G_{n+1})$ .

## Gaps in Decompositions

- $k(m)$  : number of summands in  $m$
- $k_g(m)$  : number of size- $g$  gaps in  $m$

### Example

$$101 = F_{10} + F_5 + F_3 + F_1$$

$$k(101) = 4, k_2(101) = 2, k_3(101) = k_4(101) = 0, \\ k_5(101) = 1$$

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### Example

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$$k(101) = 4, k_2(101) = 2, k_3(101) = k_4(101) = 0, \\ k_5(101) = 1$$

### Fact

$$k(m) = 1 + \sum_{g=0}^{\infty} k_g(m)$$

## Results

Understand: Distribution of number of summands:  $K_n$

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Don't Understand: Distribution of number of size- $g$  gaps:  
 $K_{g,n}$

## Results: Mean and Variance

$K_n$ : number of summands in random  $m \in [G_n, G_{n+1})$

### Theorem (Lekkerkerker, 1951)

When  $\{G_n\}$  is Fibonacci,  $\mathbf{E}[K_n] = C_{Lek}n + O(1)$  for  $C_{Lek} = \frac{1}{\varphi^2+1}$ . ( $C_{Lek} \approx 0.276$ )

### Theorem (Miller and Wang, 2012)

When  $\{G_n\}$  is a PLRS, there exists  $A > 0$ ,  $B$ , and  $\gamma_1 \in (0, 1)$  such that  $\mathbf{E}[K_n] = An + B + O(\gamma_1^n)$ .

### Theorem (Miller and Wang, 2012)

When  $\{G_n\}$  is a PLRS, there exists  $C > 0$ ,  $D$ , and  $\gamma_2 \in (0, 1)$  such that  $\mathbf{Var}[K_n] = Cn + D + O(\gamma_2^n)$ .

## Results: Mean and Variance

$K_{g,n}$ : number of size- $g$  gaps in random  $m \in [G_n, G_{n+1})$

### Theorem (Main Result 1: Lekkerkerker for gaps)

When  $\{G_n\}$  is a PLRS, for every integer  $g \geq 0$ , there exists  $A_g, B_g$  and  $\gamma_{g,1} \in (0, 1)$  such that  $\mathbf{E}[K_{g,n}] = A_g n + B_g + O(\gamma_{g,1}^n)$ . By Bower et al. 2013,  $A_g$  is known for all PLRS  $\{G_n\}$  and  $g$ .

### Theorem (Main Result 2: Variance is linear for gaps)

When  $\{G_n\}$  is a PLRS, for every integer  $g \geq 0$ , there exists  $C_g, D_g$  and  $\gamma_{g,2} \in (0, 1)$  such that  $\mathbf{Var}[K_{g,n}] = C_g n + D_g + O(\gamma_{g,2}^n)$ .

## Results: Asymptotic Gaussianity

$K_n$ : number of summands in random  $m \in [G_n, G_{n+1})$

$K_{g,n}$ : number of size- $g$  gaps in random  $m \in [G_n, G_{n+1})$

### Theorem (Kopp, Koloğlu, Miller, and Wang, 2011)

*When  $\{G_n\}$  is Fibonacci, as  $n \rightarrow \infty$ ,  $K_n$  approaches Gaussian.*

### Theorem (Miller and Wang, 2012)

*When  $\{G_n\}$  is PLRS, as  $n \rightarrow \infty$ ,  $K_n$  approaches Gaussian.*

### Theorem (Main Result 3: Gaussian Behavior for Gaps)

*When  $\{G_n\}$  is PLRS, as  $n \rightarrow \infty$ ,  $K_{g,n}$  approaches Gaussian.*

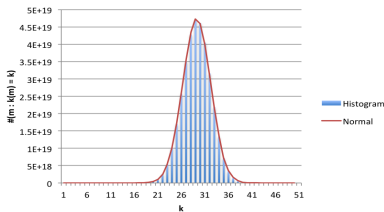


## Results: Summary

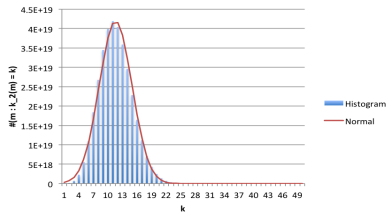
### Theorem (Main Results, Summary)

The mean  $\mu_{g,n}$  and variance  $\sigma_{g,n}^2$  of  $K_{g,n}$  grow linearly in  $n$ , and  $(K_{g,n} - \mu_{g,n})/\sigma_{g,n}$  converges to the standard normal  $N(0, 1)$  as  $n \rightarrow \infty$ .

Histogram:  $k(m)$  for  $m$  in  $[F_{100}, F_{101}]$



Histogram:  $k_2(m)$  for  $m$  in  $[F_{100}, F_{101}]$



## Proof Sketch

### Theorem (Main Result 3: Fibonacci Numbers, $g = 2$ )

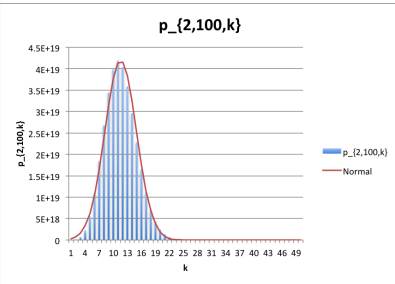
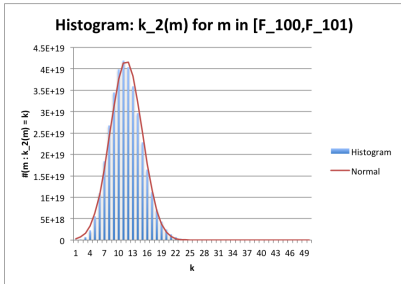
*When  $\{G_n\}$  is Fibonacci, as  $n \rightarrow \infty$ ,  $(K_{2,n} - \mu_{2,n})/\sigma_{2,n}$  converges to standard normal.*

# Proof Sketch

## Theorem (Main Result 3: Fibonacci Numbers, $g = 2$ )

When  $\{G_n\}$  is Fibonacci, as  $n \rightarrow \infty$ ,  $(K_{2,n} - \mu_{2,n})/\sigma_{2,n}$  converges to standard normal.

Let  $p_{2,n,k} = \#\{m \in [F_n, F_{n+1}) \text{ with } k \text{ size-2 gaps}\} |$ .  
 $\Pr[K_{2,n} = k] \propto p_{2,n,k}$ .



## Proof Sketch

$$p_{2,n,k} = \#\{m \in [F_n, F_{n+1}) \text{ with } k \text{ size-2 gaps}\}.$$

$n \backslash k$	0	1	2	3	4	5	6
0	0	0	0	0	0	0	0
1	1	0	0	0	0	0	0
2	1	0	0	0	0	0	0
3	1	1	0	0	0	0	0
4	2	1	0	0	0	0	0
5	3	1	1	0	0	0	0
6	4	3	1	0	0	0	0
7	6	5	1	1	0	0	0
8	9	7	4	1	0	0	0

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6	4	3	1	0	0	0	0
7	6	5	1	1	0	0	0
8	9	7	4	1	0	0	0

### Lemma

$$p_{2,n,k} = p_{2,n-1,k} + p_{2,n-2,k-1} + p_{2,n-3,k-1} - p_{2,n-3,k}$$

## Proof Sketch

### Lemma (Key Lemma)

If  $a_{n,k}$  is a “nice” two dimensional homogenous recurrence

$$a_{n,k} = \sum_{i=1}^{i_0} \sum_{j=0}^{j_0} t_{i,j} a_{n-i,k-j}$$

for constants  $t_{i,j}$ , then the random variable  $X_n$  given by  $\Pr[X_n = k] \propto a_{n,k}$  approaches Gaussian as  $n \rightarrow \infty$ .

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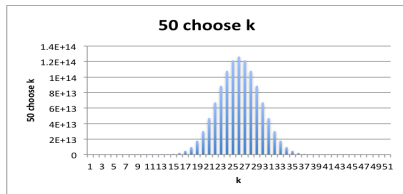
**Famous example:**  $a_{n,k} = a_{n-1,k} + a_{n-1,k-1}$  gives...

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**Famous example:**  $a_{n,k} = a_{n-1,k} + a_{n-1,k-1}$  gives...  
the binomials:  $a_{n,k} = \binom{n}{k}$ !





## Proof Sketch

### Lemma (Key Lemma)

If  $a_{n,k}$  is a “nice” two dimensional homogenous recurrence then the random variable  $X_n$  given by  $\Pr[X_n = k] \propto a_{n,k}$  approaches Gaussian as  $n \rightarrow \infty$ .

### Applications

- $a_{n,k} = a_{n-1,k} + a_{n-1,k-1}$   
 $X_n = \# \text{ heads after } n \text{ coin flips } (\Pr[X_n = k] = \frac{\binom{n}{k}}{2^n})$
- $a_{n,k} = a_{n-1,k} + a_{n-2,k-1} + a_{n-3,k-1} - a_{n-3,k}$   
 $X_n = \# \text{ size-2 gaps of random } m \in [F_n, F_{n+1})$   
 $(a_{n,k} = p_{2,n,k}, X_n = K_{2,n})$
- $a_{n,k} = a_{n-1,k} + a_{n-2,k-1}$   
 $X_n = \# \text{ summands of random } m \in [F_n, F_{n+1})$   
 $(X_n = K_n)$

## Proof Sketch

### Lemma (Key Lemma)

If  $a_{n,k}$  is a “nice” two dimensional homogenous recurrence then the random variable  $X_n$  given by  $\Pr[X_n = k] \propto a_{n,k}$  approaches Gaussian as  $n \rightarrow \infty$ .

Let  $\tilde{\mu}_n(m) = \mathbf{E}[(X_n - \mu_n)^m]$ .

### Lemma (Method of Moments)

Suffices to prove

$$\lim_{n \rightarrow \infty} \frac{\tilde{\mu}_n(2m)}{\tilde{\mu}_n(2)^m} = (2m - 1)!! \quad \lim_{n \rightarrow \infty} \frac{\tilde{\mu}_n(2m + 1)}{\tilde{\mu}_n(2)^{m + \frac{1}{2}}} = 0.$$

## Proof Sketch

Let  $\tilde{\mu}_n(m) = \mathbf{E}[(X_n - \mu_n)^m]$ . We have

$$\tilde{\mu}_n(m) = \sum_{\ell=0}^m \binom{m}{\ell} \sum_{t_{i,j} \neq 0} \frac{F_{n-i} t_{i,j}}{F_n} \cdot (j + \mu_{n-i} - \mu_n)^\ell \cdot \tilde{\mu}_{n-i}(m - \ell).$$

## Proof Sketch

Let  $\tilde{\mu}_n(m) = \mathbf{E}[(X_n - \mu_n)^m]$ . We have

$$\tilde{\mu}_n(m) = \sum_{\ell=0}^m \binom{m}{\ell} \sum_{t_{i,j} \neq 0} \frac{F_{n-i} t_{i,j}}{F_n} \cdot (j + \mu_{n-i} - \mu_n)^\ell \cdot \tilde{\mu}_{n-i}(m - \ell).$$

### Lemma

For each integer  $m \geq 0$ , there exist degree  $m$  polynomials  $Q_{2m}, Q_{2m+1}$  and  $\gamma_{2m}, \gamma_{2m+1} \in (0, 1)$  such that

$$\begin{aligned} \tilde{\mu}_n(2m) &= Q_{2m}(n) + O(\gamma_{2m}^n) \\ \tilde{\mu}_n(2m+1) &= Q_{2m+1}(n) + O(\gamma_{2m+1}^n). \end{aligned}$$

Furthermore, if  $C_{2m}$  is the leading coefficient of  $Q_{2m}$ , then for all  $m$ ,  $C_{2m} = (2m - 1)!! \cdot C_2^m$ .

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- What is the rate at which our  $K_{g,n}$  converges to a normal distribution?