

# Random Numerical Semigroups

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Thesis submitted in fulfilment of the requirements for the degree of  
*Bachelor of Science*  
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October 28, 2023

# Abstract

This is the text of the abstract, providing a brief summary of the context, research problem, main contributions and conclusions of this work.

# Dedication

For my mother. (name of person - this is optional!)

# Acknowledgements

I would like to thank (supervisor, family, research collaborators, anyone else who significantly helped you with this work).

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October 28, 2023  
Bogotá, Colombia

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# Chapter 1

## Introduction

The start of the introduction provides some context and brief background. This is a test for github.

### 1.1 Research Objectives and Overview

The research question which this Thesis aims to answer is...

The specific research objectives of this Thesis are:

1. Objective 1
2. Objective 2

Chapter 2 provides a comprehensive review of literature which is relevant to the overall aim. This includes ...

Chapter 4 aims to ...

This chapter resulted in the following publications:

- D. R. Franklin and K. J. Wilson, “A LaTeX Thesis Template for the School of Electrical and Data Engineering,” *IEEE Transactions on LaTeX Thesis Templates*, vol. 1, no. 1, Oct. 2021
- D. R. Franklin and K. J. Wilson, “A LaTeX Thesis Template for the School of Electrical and Data Engineering,” *IEEE Transactions on LaTeX Thesis Templates*, vol. 1, no. 1, Oct. 2021

Chapter 5 aims to ...

This chapter resulted in the following publications:

- D. R. Franklin and K. J. Wilson, “A LaTeX Thesis Template for the School of Electrical and Data Engineering,” *IEEE Transactions on LaTeX Thesis Templates*, vol. 1, no. 1, Oct. 2021

Chapter 6 aims to ...

This chapter resulted in the following publications:

- D. R. Franklin and K. J. Wilson, “A LaTeX Thesis Template for the School of Electrical and Data Engineering,” *IEEE Transactions on LaTeX Thesis Templates*, vol. 1, no. 1, Oct. 2021

Finally, Chapter 7 summarises the results and implications of this work, and provides recommended directions for continuation of this work in the future.

### 1.1.1 Additional Research Contributions

A number of additional research publications and presentations are listed below:

- xxx

Test [1]

# Chapter 2

## The Probabilistic Method

### 2.1 Introduction

The probabilistic method is a powerful tool, with applications in Combinatorics, Graph Theory, Number Theory and Computer Science. It is a nonconstructive method that proves the existence of an object with a certain property, by showing that the probability that a randomly chosen object has that property is greater than zero. The method requires an appropriate sample space and is best illustrated by an example:

**Definition 2.1.1.** A *tournament* is a directed graph  $T$  on  $n$  vertices such that for every pair of vertices  $i, j \in V(T)$ , exactly one of the edges  $(i, j)$  or  $(j, i)$  is in  $E(T)$ . [2]

The name of a tournament comes from the fact that it can be thought of as a sports tournament where each vertex represents a team and each team plays every other team exactly once. The edge  $(i, j)$  represents a win for team  $i$  over team  $j$ . A tournament  $T$  has property  $S_k$  if for every subset  $S \subseteq V(T)$  of size  $k$ , there is a vertex  $v \in V(T)$  such that  $(v, s) \in E(T)$  for all  $s \in S$  [2]. That is, for every set of  $k$  teams there is a team that beats all of them. For example, the tournament in Figure 2.1 has property  $S_1$  since every team is beaten by another team.

A natural question to ask is: is there a tournament with property  $S_k$  for every  $k$ ? The answer is yes. We will prove this using the probabilistic method. First we define a probability space over the set of tournaments on  $n$  vertices:

A *random* tournament on a set of  $n$  vertices is a tournament  $T$  such that for every pair of vertices  $i, j \in V(T)$ , the edge  $(i, j)$  is in  $E(T)$  with probability  $\frac{1}{2}$  and the edge  $(j, i)$  is in  $E(T)$  with probability  $\frac{1}{2}$ , independently of all other edges. Thus, every tournament on  $n$  vertices has the same probability, which means that this probability space is *symmetric*.

The main idea is to show that there is a large  $n$  as a function of  $k$ , such that the probability that a random tournament on  $n$  vertices has property  $S_k$  is greater than zero. This implies

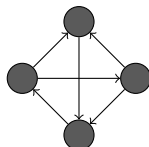


Figure 2.1: A tournament on 3 vertices with property  $S_1$ .

that there is at least one tournament with property  $S_k$ .

**Theorem 2.1.1.** *For every  $k \in \mathbb{N}$ , there is a tournament with property  $S_k$ . [2]*

**Proof.** Fix a subset  $S \subseteq V(T)$  of size  $k$ . Consider the event  $A_S$  that there is no vertex  $v \in V(T)$  such that  $(v, s) \in E(T)$  for all  $s \in S$ . For any vertex  $v \in V(T) \setminus S$ , the probability that  $(v, s) \notin E(T)$  for all  $s \in S$  is  $2^{-k}$ . Thus,

$$\Pr[A_S] = (1 - 2^{-k})^{n-k}.$$

Now, if we consider all subsets  $S \subseteq V(T)$  of size  $k$ , then the probability that  $T$  does not have property  $S_k$  is the probability that any of the events  $A_S$  occurs. Since there are  $\binom{n}{k}$  such subsets, by the union bound,

$$\Pr \left[ \bigvee_{S \subseteq V(T), |S|=k} A_S \right] \leq \sum_{S \subseteq V(T), |S|=k} \Pr[A_S] = \binom{n}{k} (1 - 2^{-k})^{n-k}.$$

We want to show that, for some  $n$ , this probability of this event is less than 1. Using A.0.1, we have that

$$\Pr \left[ \bigvee_{S \subseteq V(T), |S|=k} A_S \right] \leq \binom{n}{k} (1 - 2^{-k})^{n-k} \leq \left( \frac{en}{k} \right)^k \left( e^{-2^{-k}} \right)^{n-k} = e^{k \log \left( \frac{n}{k} \right) - \frac{n-k}{2^k}}. \quad (2.1)$$

Thus, 2.1 is less than one if

$$k \log \left( \frac{n}{k} \right) - \frac{n-k}{2^k} = k \log n - k \log k + \frac{k}{2^k} - \frac{n}{2^k} < k \log n - \frac{n}{2^k} < 0,$$

which holds if and only if

$$\frac{n}{\log n} > k2^k.$$

Now, if  $n = (\log 2)k^2 2^k$ , then

$$\lim_{n \rightarrow \infty} \frac{n}{k2^k \log n} = \lim_{k \rightarrow \infty} \frac{k \log 2}{2 \log k + k \log 2 + \log \log 2} = 1.$$

Therefore, we conclude for some  $n = k^2 2^k (1 + o(1))$ , the probability that a random tournament on  $n$  vertices does not have property  $S_k$  is less than one. Thus, the probability that there exists a tournament on  $n$  vertices with property  $S_k$  is greater than zero, which means that there exists at least one tournament with property  $S_k$ .

We make two observations:

1. In the proof, we used the *union bound*. The union bound is a common technique in the probabilistic method. It states that for any events  $A_1, \dots, A_n$ ,

$$\Pr[A_1 \cup \dots \cup A_n] \leq \Pr[A_1] + \dots + \Pr[A_n].$$

We will extensively use this technique in this thesis. In a measure space, the union bound is the same property as *subadditivity*.

2. The proof is nonconstructive. It does not give us a way to find a tournament with property  $S_k$ . It only shows that there is at least one. This is a common feature of the probabilistic method.

Section 2.2 discusses Theme 1. Section 2.3 discusses Theme 2....

## 2.2 Union Bound

### 2.2.1 Example

### 2.2.2 Subtopic B

### 2.2.3 Subtopic C

## 2.3 Linearity of Expectation

Etc. etc. 2.1.1

## 2.4 Second Moment Method

## 2.5 Threshold Functions

Let  $n \in \mathbb{N}$  and  $0 \leq p \leq 1$ . The random graph  $G(n, p)$  is a probability space over the set of graphs on  $n$  labeled vertices determined by

$$\Pr[\{i, j\} \in G] = p$$

with these events mutually independent [2]. Given a graph theoretic property  $A$ , there is a probability that  $G(n, p)$  satisfies  $A$ , which we write as  $\Pr[G(n, p) \models A]$ .

**Definition 2.5.1.**  $r(n)$  is a threshold function for a graph theoretic property  $A$  if

1. When  $p(n) \in o(r(n))$ ,  $\lim_{n \rightarrow \infty} \Pr[G(n, p(n)) \models A] = 0$ ,
2. When  $r(n) \in o(p(n))$ ,  $\lim_{n \rightarrow \infty} \Pr[G(n, p(n)) \models A] = 1$ ,

or vice versa. [2]

We give an example of a threshold function which illustrates a common method for proving that a function is a threshold.

### Threshold function for having isolated vertices

Let  $G$  be a graph on  $n$  labeled vertices. An isolated vertex of  $G$  is a vertex which does not belong to any of the edges of  $G$ . Let  $A$  be the property that  $G$  contains an isolated vertex. We will prove that  $r(n) = \frac{\ln n}{n}$  is a threshold for  $A$ .

For each vertex  $i$  in  $G$  define the variable

$$X_i = \begin{cases} 1 & \text{if } i \text{ is an isolated vertex,} \\ 0 & \text{if } i \text{ is not an isolated vertex.} \end{cases}$$

Now, the probability that a vertex  $i$  is isolated is  $(1-p)^{n-1}$  since it is the probability that none of the other  $n-1$  vertices is connected to  $i$ . Let  $X = \sum_{i=1}^n X_i$ , then the expected number of isolated vertices is

$$E[X] = \sum_{i=1}^n E[X_i] = \sum_{i=1}^n \Pr[X_i] = n(1-p)^{n-1}.$$

Let  $p = k \frac{\ln n}{n}$  for  $k \in \mathbb{R}_{>0}$ . Then

$$\begin{aligned} \lim_{n \rightarrow \infty} E[X] &= \lim_{n \rightarrow \infty} n \left( 1 - k \frac{\ln n}{n} \right)^{n-1} \\ &= n e^{-k \ln n} = n^{1-k}. \end{aligned}$$

Therefore,  $\lim_{n \rightarrow \infty} E[X] = 0$  if  $k > 1$ . Since  $E[X] \geq \Pr[X > 0]$ , we conclude that

$$\lim_{n \rightarrow \infty} \Pr[G(n, p) \models A] = \lim_{n \rightarrow \infty} \Pr[X > 0] = 0.$$

Now, for  $k < 1$ , the fact that  $\lim_{n \rightarrow \infty} E[X] = \infty$  is not enough to conclude that  $\lim_{n \rightarrow \infty} \Pr[G(n, p) \models A] = 1$ . We have to use the second moment method.

**Theorem.** *If  $E[X] \rightarrow \infty$  and  $\text{Var}[X] = o(E[X]^2)$ , then  $\lim_{n \rightarrow \infty} \Pr[X > 0] = 1$ . [2]*

**Proof.** We will prove that, in this case,  $\text{Var}[X] = o(E[X]^2)$ . First,

$$\begin{aligned} \sum_{i \neq j} E[X_i X_j] &= \sum_{i \neq j} \Pr[X_i = X_j = 1] \\ &= n(n-1)(1-p)^{n-1}(1-p)^{n-2} \\ &= n(n-1)(1-p)^{2n-3}, \end{aligned}$$

for if  $i$  is an isolated vertex, then there is no edge between  $i$  and  $j$  so we only have to account for the remaining  $n-2$  edges that contain  $j$ .

Thus, since  $\sum_{i=1}^n E[X_i^2] = \sum_{i=1}^n E[X_i] = E[X]$  and  $\lim_{n \rightarrow \infty} p = 0$ ,

$$\begin{aligned} \lim_{n \rightarrow \infty} \frac{\text{Var}[X]}{E[X]^2} &= \lim_{n \rightarrow \infty} \frac{E[X^2] - E[X]^2}{E[X]^2} = \lim_{n \rightarrow \infty} \frac{\sum_{i=1}^n E[X_i^2] + \sum_{i \neq j} E[X_i X_j]}{E[X]^2} - 1 \\ &= 0 + \lim_{n \rightarrow \infty} \frac{n(n-1)(1-p)^{2n-3}}{n^2(1-p)^{2n-2}} - 1 = \lim_{n \rightarrow \infty} \frac{1}{1-p} - 1 = 0. \end{aligned}$$

We conclude that  $\text{Var}[X] \in o(E[X]^2)$  and so, if  $k < 1$ ,

$$\lim_{n \rightarrow \infty} \Pr[G(n, p) \models A] = \lim_{n \rightarrow \infty} \Pr[X > 0] = 1.$$

Therefore,  $r(n) = \frac{\ln n}{n}$  is a threshold function for property  $A$ .

# Chapter 3

## Numerical Semigroups

### 3.1 Introduction

Some contextual overview of what we are going to discuss. Include the first example in the book.

Section 2.2 discusses Theme 1. Section 2.3 discusses Theme 2....

A *numerical semigroup* is a subset  $S \subseteq \mathbb{N}_0$  which is closed under addition, i.e.  $a, b \in S$  implies  $a + b \in S$ . For instance,  $\mathbb{N}_0$ ,  $\mathbb{N}_0 \setminus \{0\}$ ,  $2\mathbb{N}_0$  are all numerical semigroups, but  $\mathbb{N}_0 \setminus \{2\}$  is not. Some literature requires that a semigroup has a finite complement in  $\mathbb{Z}_{\geq 0}$  [3], but we prefer the more general definition.

### 3.2 Invariants

#### 3.2.1 Subtopic A

#### 3.2.2 Subtopic B

#### 3.2.3 Subtopic C

### 3.3 Wilf's Conjecture

Etc. etc.

# Chapter 4

## Random Numerical Semigroups

### 4.1 Introduction

In this Chapter, XXX is presented. Note: to cross-reference to other parts of the document you do so like this - see Section 5.2.2.

Section 4.2 discusses Theme 1. Section 4.4 discusses Theme 2....

### 4.2 Box Model

Figure 4.1 illustrates the key elements of Thesis synthesis.  $\text{\LaTeX}$  will tend to place figures where it wants (they ‘float’ - generally they should be at the top or bottom of a page); you can override the default behaviour if you want, but you probably don’t want to bother doing that until after your content is pretty much done. Instead, keep the figures as close as possible to the text; you can tweak this afterwards if you want by adding an option:

```
\begin{figure}[htb]
```

This says put it here, if you can, otherwise at the top, or otherwise at the bottom. BUT I strongly suggest not using **h** as it looks terrible.

Maybe you need an inline URL at this point: Here’s one! <https://en.wikibooks.org/wiki/LaTeX>.

Mathematics can be inline, for example  $x = \int_0^\infty y^2 dy$ , or can be in display mode, as shown in (4.1):

$$F(x, y, z) = \sqrt{x^{y^z}} \tag{4.1}$$

You probably want to show some of your results in a table, such as Table 4.1.

The basic creative process is shown in Figure 4.2. Specifically, Figure 4.2a shows the first step, while Figure 4.2b and 4.2c show the remaining key steps of the procedure.

#### 4.2.1 Subtopic A

Algorithm 1 shows a classical algorithm typeset in  $\text{\LaTeX}$ . This is also a float.



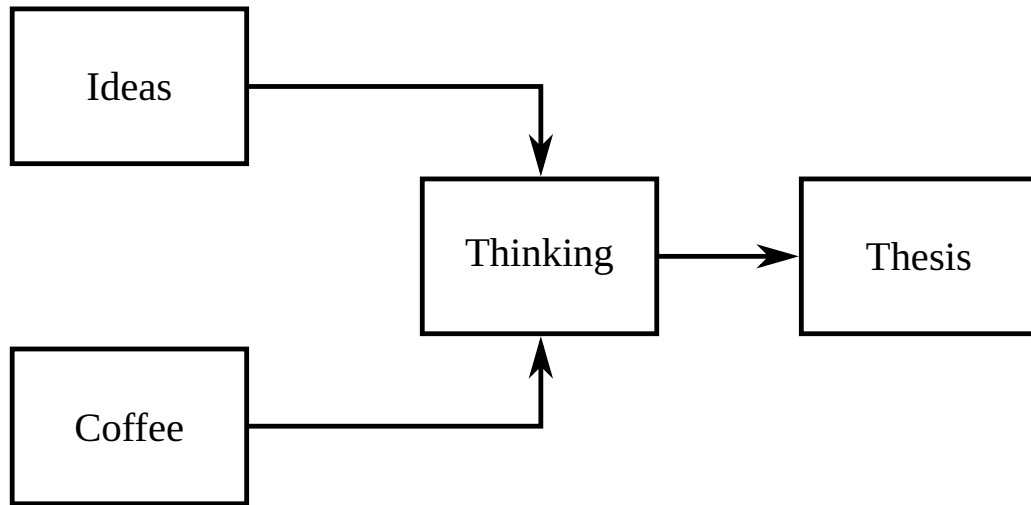


Figure 4.1: This the approximate process for producing a Thesis. It typically takes 3-4 years.

Table 4.1: Table captions normally go at the top.

Left column	$\mu$	$\sigma$
Item 1	0.3	0.5
Item 2	0.9	0.4

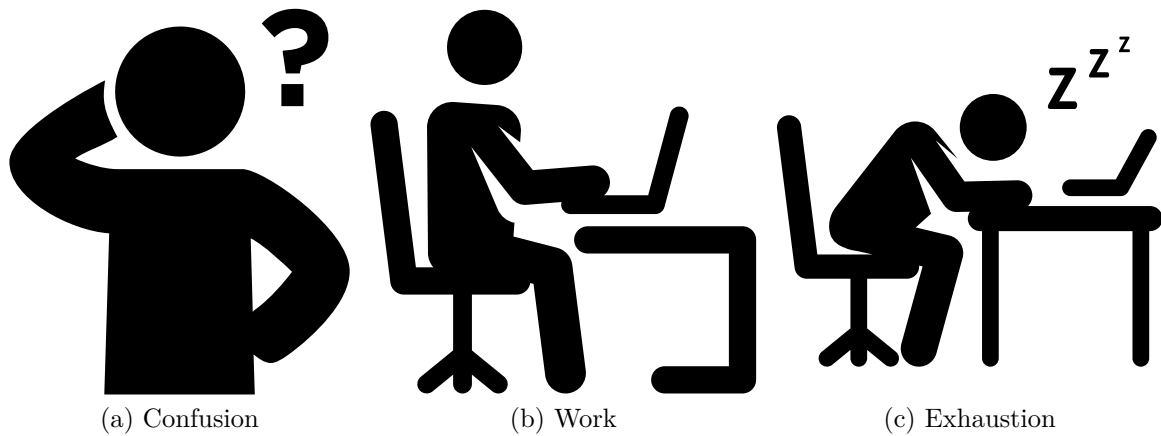


Figure 4.2: The stages of the creative process.

---

**Algorithm 1** An algorithm with caption

---

**Require:**  $n \geq 0$ **Ensure:**  $y = x^n$  $y \leftarrow 1$  $X \leftarrow x$  $N \leftarrow n$ **while**  $N \neq 0$  **do**    **if**  $N$  is even **then**         $X \leftarrow X \times X$          $N \leftarrow \frac{N}{2}$     **else if**  $N$  is odd **then**         $y \leftarrow y \times X$          $N \leftarrow N - 1$     **end if****end while**

---

 $\triangleright$  This is a comment

### 4.2.2 Results

### 4.2.3 Subtopic C

## 4.3 ER-type model

We generate a random numerical semigroup with a model similar to the Erdos-Renyi model for random graphs.

**Definition 4.3.1.** For  $p \in [0, 1]$  and  $M \in \mathbb{N}$ , a random numerical semigroup  $S(M, p)$  is a probability space over the set of semigroups  $S = \langle \mathcal{A} \rangle$  with  $\mathcal{A} \subseteq \{1, \dots, M\}$ , determined by

$$\Pr[n \in \mathcal{A}] = p,$$

with these events mutually independent.

## 4.4 Downward model

Etc. etc.

## 4.5 Conclusion

# Chapter 5

## Experiments

### 5.1 ER-type model experiments

In this Chapter, XXX is presented. Include pseudocode.

Section 5.2 discusses Theme 1. Section 5.3 discusses Theme 2....

### 5.2 Downward model experiments

#### 5.2.1 Subtopic A

#### 5.2.2 Subtopic B

#### 5.2.3 Subtopic C

### 5.3 Theme 2

Etc. etc.

# Chapter 6

## Results

### 6.1 Introduction

In this Chapter, XXX is presented.

Section 6.2 discusses Theme 1. Section 6.4 discusses Theme 2....

### 6.2 Lower Bound

### 6.3 Expected Frobenius Number

We prove a Theorem found in [4] without the use of the simplicial complex.

**Theorem 6.3.1.** *Let  $S \sim S(M, p)$ , where  $p = p(M)$  is a monotone decreasing function of  $M$ . If  $\frac{1}{M} \ll p \ll 1$ , then  $S$  is cofinite, i.e., the set of gaps is finite, a.a.s and*

$$\lim_{M \rightarrow \infty} E[e(S)] = \lim_{M \rightarrow \infty} E[g(S)] = \lim_{M \rightarrow \infty} E[F(S)] = \infty.$$

**Proof.** Let  $X := \min(S \setminus \{0\})$  be a random variable. Then, for  $0 < n \leq M$ ,

$$\Pr[X = n] = p(1 - p)^{n-1},$$

and so

$$\begin{aligned} E[X] &= \sum_{n=0}^{\infty} n \Pr[X = n] = \sum_{n=0}^M np(1 - p)^{n-1} = p \frac{d}{dp} \left[ - \sum_{n=0}^M (1 - p)^n \right] \\ &= p \frac{d}{dp} \frac{(1 - p)^{M+1} - 1}{p} = p \frac{1 - (1 - p)^{M+1} - (M + 1)(1 - p)^M p}{p^2} \\ &= \frac{1 - (1 - p)^M - M(1 - p)^M p}{p} \geq \frac{1 - e^{-Mp} - Mpe^{-Mp}}{p}. \end{aligned}$$

Thus, since  $\lim_{M \rightarrow \infty} Mp = \infty$ , then  $\lim_{M \rightarrow \infty} Mpe^{-Mp} = \lim_{M \rightarrow \infty} e^{-Mp} = 0$ , which implies that

$$\lim_{M \rightarrow \infty} E[X] = \lim_{M \rightarrow \infty} \frac{1 - e^{-Mp} - Mpe^{-Mp}}{p} = \infty.$$

Also, note that if  $p = \frac{c}{M}$ ,  $c \in \mathbb{R}_+$  ( $0 < e^{-c} + ce^{-c} < 1$ ),

$$\lim_{M \rightarrow \infty} E[X] = \lim_{M \rightarrow \infty} \frac{1 - e^{-c} - ce^{-c}}{p} = \infty.$$

**Proof.** Fix  $a \in \mathbb{N}$  such that  $a > 11$  and let  $A = \{1, \dots, \lfloor \frac{a}{p} \rfloor\}$ . Since  $\frac{1}{M} \ll p$ , we have that  $\lfloor \frac{a}{p} \rfloor \leq M$  for large enough  $M$ . Consider the following events:

- $E_1$ : No generator selected is less than  $\frac{1}{ap}$ .

Let  $X_1$  be the number of generators selected from  $\{1, \dots, \lfloor \frac{1}{ap} \rfloor\}$ . Then

$$\Pr[\overline{E_1}] = \Pr[X_1 > 0] \leq E[X_1] \leq p \cdot \frac{1}{ap} = \frac{1}{a}.$$

- $E_2$ : At most  $\frac{3a}{2}$  generators are selected from  $A$ .

Let  $X_2$  be the number of generators selected in  $A$ , then  $X_2$  is a binomial random variable with  $n = \frac{a}{p}$  and we can use the bound (Feller [I can add this to the appendix](#))

$$\Pr[\overline{E_2}] = \Pr\left[X_2 > \frac{3a}{2}\right] \leq \frac{\frac{3a}{2}(1-p)}{(\frac{3a}{2} - a)^2} \leq \frac{6}{a}.$$

Also, note that by the union bound

$$\Pr[E_1 \wedge E_2] \leq 1 - \frac{1}{a} - \frac{6}{a} = 1 - \frac{7}{a}.$$

- $E_3$ : At least  $\frac{a}{2}$  generators are selected from  $A$ .

Similarly, we can use the bound for the other tail of the distribution so that

$$\Pr[\overline{E_3}] = \Pr\left[X_2 < \frac{a}{2}\right] \leq \frac{(n - \frac{a}{2})p}{(np - \frac{a}{2})^2} = \frac{a - (\frac{a}{2})p}{(\frac{a}{2})^2} \leq \frac{4}{a}.$$

- $E_4$ : The generators selected from  $A$  are minimal.

Let  $Y_{(1)}, Y_{(2)}, \dots, Y_{(k)}$  denote the first  $k$  generators selected in  $A$ . Assume  $E_1$  and  $E_2$ . We have that  $E_1$  implies  $Y_{(1)} \geq \frac{1}{ap}$  and  $E_2$  implies  $k \leq \frac{3a}{2}$ .

First we bound for the probability that, given  $E_1$  and  $E_2$ ,  $b \in A$  is selected as a generator. By conditional probability

$$\begin{aligned} \Pr[b \text{ is selected}] &= \Pr[b \text{ is selected} | E_1 \wedge E_2] \Pr[E_1 \wedge E_2] \\ &\quad + \Pr[b \text{ is selected} | \overline{E_1 \wedge E_2}] \Pr[\overline{E_1 \wedge E_2}], \end{aligned}$$

and so

$$\Pr[b \text{ is selected} | E_1 \wedge E_2] \leq \frac{\Pr[b \text{ is selected}]}{\Pr[E_1 \wedge E_2]} \leq \frac{p}{1 - \frac{7}{a}}.$$

Now, note that  $Y_{(2)}$  is not minimal if a multiple of  $Y_{(1)}$  is selected in  $A$ . Thus, if we fix  $Y_{(1)} = y_1 \geq \frac{1}{ap}$ ,  $Y_{(1)}$  is not minimal if  $b \in \{2y_1, 3y_1, \dots, c_1 y_1\}$  is selected, where  $c_1 y_1$  is

the largest multiple of  $y_1$  which does not exceed  $\frac{a}{p}$ . Since  $y_1 \geq \frac{1}{ap}$ , we have that  $c_1 \leq a^2$ . Then, using the union bound,

$$\Pr[Y_{(2)} \text{ is not minimal} | E_1 \wedge E_2 \wedge Y_{(1)} = y_1] \leq \frac{pa^2}{1 - \frac{7}{a}}.$$

If we sum over all possible  $y_1$ , we get that

$$\Pr[Y_{(2)} \text{ is not minimal} | E_1 \wedge E_2] \leq \frac{pa^2}{1 - \frac{7}{a}}.$$

Similarly, for  $2 \leq t \leq k$  and fixed  $Y_{(1)} = y_1, \dots, Y_{(t-1)} = y_{t-1}$ ,  $Y_{(t)}$  is not minimal if the first  $t-1$  numbers selected from  $A$  can generate  $Y_{(t)}$ . For the possible numbers generated by the first  $t$  numbers selected, there are at most  $a^2$  choices for each coefficient, so there are at most  $a^{2t}$  such linear combinations. Then

$$\Pr[Y_{(t)} \text{ is not minimal} | E_1 \wedge E_2] \leq \frac{pa^{2t}}{1 - \frac{7}{a}}.$$

Therefore, since  $Y_{(1)}$  is always minimal, we can use the union bound and  $k \leq \frac{3a}{2}$  to conclude that

$$\Pr[E_4 | E_1 \wedge E_2] \geq 1 - \frac{p}{1 - \frac{7}{a}} \sum_{t=1}^{\frac{3a}{2}-1} a^{2t} = 1 - o(1).$$

Thus,

$$\Pr[E_4] = \Pr[E_4 | E_1 \wedge E_2] \Pr[E_1 \wedge E_2] \geq 1 - \frac{7}{a} - o(1).$$

Finally, note that by union bound,

$$\Pr[E_4 \wedge E_3] \geq 1 - \frac{11}{a} - o(1).$$

Therefore, for every  $N \in \mathbb{N}$  and  $\varepsilon > 0$ , there exists  $K$  such that  $M \geq K$  implies

$$\Pr[f(S) > N], \Pr[g(S) > N], \Pr[e(S) > N] > 1 - \varepsilon.$$

.

### 6.3.1 Subtopic A

### 6.3.2 Subtopic B

### 6.3.3 Subtopic C

## 6.4 Lower Bound result

I conjecture that the hypothesis that  $q$  is prime can be relaxed.

### 6.4.1 Lemma

- Let  $q$  be a prime number and  $S$  be a random subset of  $\mathbb{Z}_q$  of size  $4\lfloor 3\log_2 q \rfloor$ . As  $q$  tends to infinity,  $2\lfloor 3\log_2 q \rfloor S$  covers  $\mathbb{Z}_q$  almost always.

Let  $q$  be a prime number and let  $s \in \mathbb{N}$  such that  $s \leq q$ . Let  $S$  be a uniformly random subset of  $\mathbb{Z}_q$  of size  $s$ , that is,

$$\Pr(S) = \frac{1}{\binom{q}{s}}.$$

For a given  $z \in \mathbb{Z}_q$  and  $k \in \mathbb{N}$  for which  $k \leq s/2$ , let

$$N_z^k := \left\{ K \subseteq \mathbb{Z}_q : |K| = k, \sum_{t \in K} t = z \right\}.$$

Note that  $|N_z^k| = \frac{1}{q} \binom{q}{k}$ , since  $K \in N_0^k$  if and only if  $K + k^{-1}z \in N_z^k$  for every  $z \in \mathbb{Z}_q$ .

For  $K \in N_z^k$ , let  $A_K$  be the event that  $K \subset S$ . Let  $X_K$  be the indicator variable of  $A_K$ . We define the random variable

$$X_z = \sum_{K \in N_z^k} X_K.$$

Note that  $X_z$  counts the number of sets of size  $k$  which add up to  $z$ . We provide two ways of finding  $E[X_z]$ . The first one uses that, for any  $K \in N_z^k$ ,

$$E[X_K] = \Pr[A_K] = \frac{\binom{q-k}{s-k}}{\binom{q}{s}},$$

and so we get that

$$E[X_z] = \sum_{K \in N_z^k} E[X_K] = |N_z^k| E[X_K] = \frac{1}{q} \binom{q}{k} \frac{\binom{q-k}{s-k}}{\binom{q}{s}} = \frac{1}{q} \binom{s}{k}.$$

This motivates the second way, for we know that

$$\sum_{z \in \mathbb{Z}_q} X_z = \binom{s}{k} = \sum_{z \in \mathbb{Z}_q} E[X_z].$$

As in the argument for finding  $|N_z^k|$ , for every  $z \in \mathbb{Z}_q$ ,

$$E[X_0] = \sum_{K \in N_0^k} E[X_K] = \sum_{K \in N_0^k} E[X_{K+k^{-1}z}] = \sum_{K \in N_z^k} E[X_K] = E[X_z].$$

Therefore, we also find that

$$E[X_z] = \frac{1}{p} \binom{s}{k}. \tag{6.1}$$

Now, for  $K, L \in N_z^k$ , let  $j \in \mathbb{N}$  and define

$$\Delta_j := \sum_{|K \cap L|=j} \Pr[A_K \wedge A_L].$$

Fix  $j \leq k$ , then

$$\Pr[A_K \wedge A_L] = \frac{\binom{q-2k+j}{s-2k+j}}{\binom{q}{s}}.$$

We can bound the number of events for which  $|K \cap L| = j$ . First we choose  $K$  as any set in  $N_z^k$  and then we choose the remaining  $k - j$  elements as any subset of  $\mathbb{Z}_q \setminus K$  with size  $k - j$ . Thus,

$$\Delta_j \leq \frac{\binom{p}{k} \binom{q-k}{k-j} \binom{q-2k+j}{s-2k+j}}{q \binom{q}{s}}.$$

This implies that, using (1),

$$\begin{aligned} \frac{\Delta_j}{E[X_z]^2} &\leq \frac{\binom{q}{k} \binom{q-k}{k-j} \binom{q-2k+j}{s-2k+j}}{\frac{1}{q} \binom{s}{k} \frac{1}{q} \binom{s}{k} q \binom{q}{s}} \\ &= \frac{\frac{q!}{(q-k)!k!} \frac{(p-k)!}{(k-j)!(q-2k+k)!} \frac{(q-2k+j)!}{(s-2k+j)!(q-s)!}}{\frac{1}{q} \binom{s}{k} \frac{s!}{(s-k)!k!} \frac{q!}{(q-s)!s!}} \\ &= \frac{q \binom{s-k}{k-j}}{\binom{s}{k}}. \end{aligned}$$

Let  $s = 4\lfloor 3\log_2 q \rfloor$  and  $k = 2\lfloor 3\log_2 q \rfloor$ , where  $\alpha \in (0, 1)$ . Using that  $\binom{s-k}{t}$  is maximized at  $t = \lfloor (s-k)/2 \rfloor$ ,

$$\frac{\Delta_j}{E[X_z]^2} \leq \frac{q \binom{2\lfloor 3\log_2 q \rfloor}{\lfloor 3\log_2 q \rfloor}}{\binom{4\lfloor 3\log_2 q \rfloor}{2\lfloor 3\log_2 q \rfloor}} \leq \frac{q}{\binom{2\lfloor 3\log_2 q \rfloor}{\lfloor 3\log_2 q \rfloor}} \leq \frac{q}{2^{\lfloor 3\log_2 q \rfloor}} \sim \frac{1}{q^2},$$

since  $\binom{2\lfloor q^\alpha \rfloor}{\lfloor 3\log_2 q \rfloor}^2 \leq \binom{4\lfloor 3\log_2 q \rfloor}{2\lfloor 3\log_2 q \rfloor}$  (I can prove this in a lemma or in the appendix).

This proves that

$$\Pr[X_z = 0] \leq \frac{\Delta}{E[X_z]^2} = \sum_{j=0}^k \frac{\Delta_j}{E[X_z]^2} \leq \frac{(k+1)}{q^2}.$$

Therefore, by the union bound,

$$\Pr\left[\bigvee_{z \in \mathbb{Z}_q} X_z = 0\right] \leq \frac{(k+1)}{q}.$$

## 6.5 Theorem

- Let  $g(x)$  be a function for which  $x(\log x)^2 \in o(g(x))$ . Then

$$\lim_{p \rightarrow 0} \Pr\left[F(S) \leq g\left(\frac{1}{p}\right)\right] = 1.$$

The proof of this Theorem consists of several parts. The strategy is to prove that the Ápery set of a subsemigroup of  $S$  is completed before step  $g\left(\frac{1}{p}\right)$  with high probability, since  $F(S)$  is less than the maximum element of this Ápery set. The proof has the following structure:



1. First, we will find a step for which a prime  $q$  is chosen with high probability (E1).
2. Then, in the spirit of the [Lemma](#), we will find a step such that  $s$  elements, which are different modulo  $q$ , are chosen with high probability (E2).
3. Finally, we will apply the [Lemma](#) to the Ápery set of a subsemigroup of  $S$  generated by the subset in part 2.

*Proof.*

### Part 1

Let  $h(x)$  be a function such that  $h(x) \in o(x(\log x)^2)$  and  $x \log x \in o(h(x))$ . Let  $t(x) = 20x \log x$ . Consider the event  $E_1$  that there exists a prime  $q \in S$ , such that

$$t\left(\frac{1}{p}\right) \leq q \leq h\left(\frac{1}{p}\right).$$

Let  $q_n$  be the  $n$ -th prime number and let  $k_x$  be the number of primes between  $20x \log x$  and  $h(x)$ . For  $n \geq 6$ , by the [Prime Number Theorem](#),

$$n(\log n + \log \log n - 1) < q_n < n(\log n + \log \log n) = o(h(n)).$$

Thus,  $n = o(k_n)$  ([I can prove this if it is not clear](#)) and, for every  $c > 0$ ,

$$\lim_{p \rightarrow 0} \Pr[\neg E_1] \geq \lim_{p \rightarrow 0} (1 - p)^{\frac{k_1}{p}} \geq \lim_{p \rightarrow 0} (1 - p)^{\frac{c}{p}} = e^{-c}.$$

Therefore,

$$\lim_{p \rightarrow 0} \Pr[E_1] = 1.$$

### Part 2

Now, assume  $E_1$ . Then  $S$  contains a prime number  $q$  for which

$$t\left(\frac{1}{p}\right) \leq q \leq h\left(\frac{1}{p}\right).$$

Let  $s = 4\lfloor 3 \log_2 q \rfloor$ , as in the [Lemma](#).

Let  $A := \{q + 1, q + 2, \dots, 2q\}$ . Consider the event **E2** that at least  $s$  generators are selected in  $A$ . Let  $X_1$  be the number of generators selected in  $A$ , then  $X_1$  is a binomial random variable with parameters  $n = q$  and  $p$ . Then, in a similar way to  $E2$  in [Theorem 1](#), we use the [Binomial Distribution Tail Bound](#) to show that, assuming that  $p$  is small enough so that  $qp > s$  for all possible  $q$ ,

$$\Pr[\overline{E_2} | E_1] = \Pr[X_1 < s] \leq \Pr[X_2 < s] \leq \frac{(n - s)p}{(np - s)^2} = \frac{(q - s)p}{(qp - s)^2}.$$

Thus, bounding by the worst case asymptotically, ([needs to be explained better](#))

$$\lim_{p \rightarrow 0} P[\overline{E_2} | E_1] = \lim_{p \rightarrow 0} \frac{\left(h\left(\frac{1}{p}\right) - 4\left\lfloor 3 \log_2 h\left(\frac{1}{p}\right) \right\rfloor\right)p}{\left(20 \log \frac{1}{p} - 4\left\lfloor 3 \log_2 t\left(\frac{1}{p}\right) \right\rfloor\right)^2} = 0.$$

We conclude that

$$\lim_{p \rightarrow 0} \Pr[E_2|E_1] = 1,$$

and so

$$\lim_{p \rightarrow 0} \Pr[E_1 \wedge E_2] = \lim_{p \rightarrow 0} \Pr[E_2|E_1] \Pr[E_1] = 1.$$

### Part 3

Finally, assume  $E_1$  and  $E_2$ . Let  $B = \{Y_1, \dots, Y_s\}$  be a randomly selected subset of size  $s$  of the generators selected in  $E_2$ . Since the generators are chosen randomly and  $|A| = q$ , we can apply the [Lemma](#) to the Ápery set of the subsemigroup generated by  $B$ , denoted by  $G(B)$ , and conclude that the Ápery set of  $G(B)$  will be completed before step  $h\left(\frac{1}{p}\right) 2 \left\lfloor 3 \log_2 h\left(\frac{1}{p}\right) \right\rfloor$  with high probability as  $p \rightarrow 0$ .

Thus, if  $g(x)$  be a function for which  $x(\log x)^2 \in o(g(x))$  ([Probably needs to be explained better](#)),

$$\lim_{p \rightarrow 0} \Pr \left[ F(G(B)) \leq g\left(\frac{1}{p}\right) \right] = 1.$$

Since  $F(S) \leq F(G(B))$ , we conclude that

$$\lim_{p \rightarrow 0} \Pr \left[ F(S) \leq g\left(\frac{1}{p}\right) \right] = 1.$$

## 6.6 Conclusion

# Chapter 7

## Conclusions and Future Work

### 7.1 Summary of Outcomes

### 7.2 Recommendations & Future Work

### 7.3 Concluding Remarks

In summary, ...

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# Appendix A

## Example Appendix

Here you might present some additional results, derivations, proofs etc. that were not included in the main text.

**Proposition A.0.1.**  $\binom{n}{k} \leq \left(\frac{en}{k}\right)^k$  for  $1 \leq k \leq n$ .

# Appendix B

## Software Documentation

Here's an example source code listing, where the code is read in from an external file:

```
1 % Function to create a nice rotating animated GIF of 3D volumetric data V
2
3 function animation (V)
4
5 h = volshow (V, 'BackgroundColor', [0 0 0], 'Renderer', 'MaximumIntensityProjection', 'CameraPosition', [2 2 0], 'CameraUpVector', ←
    [1 0 0], 'ColorMap', jet);
6
7 camproj ('perspective');
8
9 N = 500;
10
11 filename = 'animation.gif';
12 vec = linspace(0, 4 * pi(), N)';
13 myPosition = 2 * [zeros(size(vec)) cos(vec) sin(vec)];
14
15 for idx = 1:N
16 % Update current view.
17     h.CameraPosition = myPosition(idx, :);
18 % Use getframe to capture image.
19     I = getframe(gcf);
20
21     [indI, cm] = rgb2ind (I.cdata,256);
22 % Write frame to the GIF File.
23     if idx == 1
24         imwrite(indI, cm, filename, 'gif', 'Loopcount', inf, 'DelayTime', 0.05);
25     else
26         imwrite(indI, cm, filename, 'gif', 'WriteMode', 'append', 'DelayTime', 0.05);
27     end
28 end
```

### B.1 Code Availability

All scripts and source code used for simulation and analysis of the ... are available here:

<https://bitbucket.org/username/gitrepo.git>

### B.2 Software Requirements

- MATLAB code is confirmed working with version XXXX;
- Simulations require the use of gcc version XXX or llvm/clang version YYYY

### B.3 Simulation Code - How to Run