

# MATH 7018: Probabilistic Combinatorics

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# 1 Introduction

We begin with three theorems.

## Theorem 1.1: Erdős, from Graph Theory

The Ramsey number  $r(t, t)$  satisfies  $r(t, t) \geq 2^{t/2}$ .

## Theorem 1.2: Erdős, from Additive Combinatorics

Every set  $B$  of nonzero integers contains a sum-free subset  $A \subseteq B$  of size  $|A| \geq \frac{1}{3}|B|$ .

## Theorem 1.3: Spencer, from Extremal Set Theory

Any set family  $\mathcal{F} \subseteq 2^{[n]}$  such that no  $S, T \in \mathcal{F}$  satisfy  $S \subsetneq T$  satisfies  $|\mathcal{F}| \leq \binom{n}{\lfloor n/2 \rfloor}$ .

## Theorem 1.4: from Coding Theory

Any binary prefix code  $C \subseteq \{0, 1\}^*$  satisfies  $\sum_{s \in C} \frac{1}{2^{|s|}} \leq 1$ .

## Definition 1.5

Ramsey number  $r(t, t)$  is the smallest  $n$  such that any graph  $G$  on  $n$  vertices has either a clique of size  $t$  or an independent set of size  $t$ . Equivalently, any coloring of  $E(K_n)$  in red, blue has either a monochromatic red or blue clique of size  $t$ . The equivalence can be seen by color all edges of  $G$  as red and  $E(K_n) \setminus E(G)$  as blue. For example,  $r(3, 3) = 6$ , it is not 5 because  $C_5$  and its complement do not have a triangle.

*Proof of Theorem 1.1.* Let  $G = G(n, 1/2)$  which is the Erdős-Renyi graph with  $n$  vertices and each edge appears independently with probability  $1/2$ .

Look at  $S \subseteq V(G)$  of size  $t$ . Then

$$\Pr[S \text{ is a clique}] = 2^{-\binom{t}{2}}$$

$$\Pr[S \text{ is an independent set}] = 2^{-\binom{t}{2}}.$$

Hence,

$$\begin{aligned} & \Pr[G \text{ has a } t\text{-clique or } t\text{-independent set}] \\ & \leq \sum_{S \subseteq V(G)} \Pr[S \text{ is a } t\text{-clique or } t\text{-independent set}] \\ & = \binom{n}{t} \frac{2}{2^{\binom{t}{2}}} < \frac{n^t}{2^{\binom{t}{2}}} = 1 \text{ by picking } n = 2^{(t-1)/2}. \end{aligned}$$

Notice the strict inequality is always true when  $t \geq 2$ . Thus, with positive probability,  $G$  has no  $t$ -clique nor  $t$ -independent set.  $\square$

## Definition 1.6

$S \subseteq \mathbb{Z}$  is sum-free if  $\nexists a, b, c \in S$  such that  $a + b = c$ .

Ex:  $B = [n] = \{1, \dots, n\}$ .  $A = \text{odd numbers in } B$  is sum-free with  $|A| \geq \lfloor \frac{1}{2}|B| \rfloor$ ;  $A = \text{largest } n/2 \text{ numbers in } B$  is sum-free with  $|A| \geq \lfloor \frac{1}{2}|B| \rfloor$ .

*Proof of Theorem 1.2.* Pick a big prime number  $p > 2 \max_{b \in B} |b|$ , say  $p = 3k + 2$  (which exists by the prime number theorem). Define

$$A := \{b \in B \mid (xb \pmod p) \in [k + 1, 2k + 1]\}$$

where  $x$  is a uniformly random element of  $[p - 1]$ .

$$\begin{aligned} \mathbb{E}[|A|] &= \sum_{b \in B} \Pr[xb \in [k + 1, 2k + 1]] \\ &= |B| \frac{k + 1}{3k + 1} > \frac{1}{3}|B| \end{aligned}$$

with positive probability  $|A| > \frac{1}{3}|B|$ . □

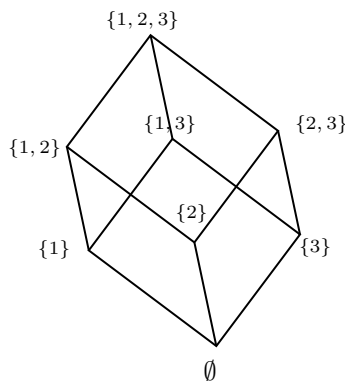


Figure 1: Subsets Diagram

*Example 1.1.* For  $n = 3$ , consider  $2^{[n]}$ . Look for biggest antichain in  $2^{[n]}$ , that is, choice of sets where no set lies above another. For example,  $\{1\}, \{2\}, \{3\}$  and  $\{1, 2\}, \{1, 3\}, \{2, 3\}$  are biggest antichains.

*Proof of Theorem 1.3.* Let  $\pi$  be a random element of  $S_n$ , which is a symmetric group. That is,  $\pi = \pi_1 \pi_2 \dots \pi_n$  is a random permutation of  $[n]$ . We consider all prefixes of  $\pi$  and let  $\mathcal{F}$  be an antichain in  $2^{[n]}$ . Define  $X$  to be the number of elements in  $\mathcal{F}$  which appear among prefix of  $\pi$ . Note that here we say appear among prefix, that is, if  $\pi = 312$ , then the  $s$  appear among its prefix are  $\emptyset, \{3\}, \{1, 3\}, \{1, 2, 3\}$ , where  $\{1\}$  is not because any permutation of 1 does not make a prefix for 312.

First, notice that  $X \leq 1$  by the fact that  $\mathcal{F}$  is an antichain, so  $\mathbb{E}[X] \leq 1$ . Then

$$\begin{aligned} \mathbb{E}[X] &= \sum_{s \in \mathcal{F}} \Pr[s \text{ is a prefix of } \pi] \\ &= \sum_{s \in \mathcal{F}} \frac{1}{\binom{n}{|s|}} \geq \frac{|\mathcal{F}|}{\binom{n}{\lfloor n/2 \rfloor}} \end{aligned}$$

where the probability of  $s$  being a prefix of  $\pi$  is the probability of  $\pi = (\text{permutations of } s)(\text{permutations of } [n] \setminus s)$ . Thus, there are  $s!(n - |s|)!$  of such  $\pi$  and  $n!$  possibly  $\pi$  in total. The probability is as above. □

*Proof of Theorem 1.4.*  $\{0, 1\}^* := \cup_{n \geq 0} \{0, 1\}^n$ . A set  $C \subseteq \{0, 1\}^*$  is a prefix code if no  $s, t \in C$  satisfy that  $s$  is a prefix of  $t$ . For example,  $C = \{00, 01, 10, 11\}$ ,  $C = \{10, 110, 1110, 111110\}$  are prefix codes but  $C = \{110, 1101\}$  is not. The theorem states that if we want  $|C|$  to be large, then  $|S|$  needs to be large in general.

Sample an infinite binary string  $S$  uniformly at random. Let  $X$  be the number of elements of  $C$  that appear as a prefix of  $S$  (not among like in the previous proof). Similarly,  $X \leq 1$ , or  $C$  is not a prefix code. Hence,  $\mathbb{E}[X] \leq 1$  and

$$\mathbb{E}[X] = \sum_{s \in C} \frac{1}{2^{|s|}}$$

where  $\frac{1}{2^{|s|}}$  is the probability of  $s \in C$  being a prefix of the binary string  $S$ . □