

Covered so far

# 8/30/2021

# Basic Counting

Let  $\phi(m, n)$  be the number of mappings from  $[n]$  to  $[m]$ .

## Theorem

$$\phi(m, n) = m^n$$

**Proof** By induction on  $n$ .

$$\phi(m, 0) = 1 = m^0.$$

$$\begin{aligned}\phi(m, n+1) &= m\phi(m, n) \\ &= m \times m^n \\ &= m^{n+1}.\end{aligned}$$

$\phi(m, n)$  is also the number of sequences  $x_1 x_2 \cdots x_n$  where



Let  $\psi(n)$  be the number of subsets of  $[n]$ .

## Theorem

$$\psi(n) = 2^n.$$

**Proof** (1) By induction on  $n$ .

$$\psi(0) = 1 = 2^0.$$

$$\psi(n+1)$$

$$= \#\{\text{sets containing } n+1\} + \#\{\text{sets not containing } n+1\}$$

$$= \psi(n) + \psi(n)$$

$$= 2^n + 2^n$$

$$= 2^{n+1}.$$

There is a general principle that if there is a 1-1 correspondence between two finite sets  $A, B$  then  $|A| = |B|$ . Here is a use of this principle.

**Proof** (2).

For  $A \subseteq [n]$  define the map  $f_A : [n] \rightarrow \{0, 1\}$  by

$$f_A(x) = \begin{cases} 1 & x \in A \\ 0 & x \notin A \end{cases}.$$

$f_A$  is the characteristic function of  $A$ .

Distinct  $A$ 's give rise to distinct  $f_A$ 's and vice-versa.

Thus  $\psi(n)$  is the number of choices for  $f_A$ , which is  $2^n$  by Theorem 51. □

Let  $\psi_{\text{odd}}(n)$  be the number of odd subsets of  $[n]$  and let  $\psi_{\text{even}}(n)$  be the number of even subsets.

## Theorem

$$\psi_{\text{odd}}(n) = \psi_{\text{even}}(n) = 2^{n-1}.$$

**Proof** For  $A \subseteq [n-1]$  define

$$A' = \begin{cases} A & |A| \text{ is odd} \\ A \cup \{n\} & |A| \text{ is even} \end{cases}$$

The map  $A \rightarrow A'$  defines a bijection between  $[n-1]$  and the odd subsets of  $[n]$ . So  $2^{n-1} = \psi(n-1) = \psi_{\text{odd}}(n)$ . Furthermore,

$$\psi_{\text{even}}(n) = \psi(n) - \psi_{\text{odd}}(n) = 2^n - 2^{n-1} = 2^{n-1}.$$

Let  $\phi_{1-1}(m, n)$  be the number of 1-1 mappings from  $[n]$  to  $[m]$ .

## Theorem

$$\phi_{1-1}(m, n) = \prod_{i=0}^{n-1} (m - i). \quad (1)$$

**Proof** Denote the RHS of (1) by  $\pi(m, n)$ . If  $m < n$  then  $\phi_{1-1}(m, n) = \pi(m, n) = 0$ . So assume that  $m \geq n$ . Now we use induction on  $n$ .

If  $n = 0$  then we have  $\phi_{1-1}(m, 0) = \pi(m, 0) = 1$ .

In general, if  $n < m$  then

$$\begin{aligned} \phi_{1-1}(m, n+1) &= (m-n)\phi_{1-1}(m, n) \\ &= (m-n)\pi(m, n) \\ &= \pi(m, n+1). \end{aligned}$$

$\phi_{1-1}(m, n)$  also counts the number of length  $n$  **ordered** sequences **distinct** elements taken from a set of size  $m$ .

$$\phi_{1-1}(n, n) = n(n-1) \cdots 1 = n!$$

is the number of ordered sequences of  $[n]$  i.e. the number of **permutations** of  $[n]$ .



## Binomial Coefficients

$$\binom{n}{k} = \frac{n!}{(n-k)!k!} = \frac{n(n-1)\cdots(n-k+1)}{k(k-1)\cdots 1}$$

Let  $X$  be a finite set and let

$\binom{X}{k}$  denote the collection of  $k$ -subsets of  $X$ .

### Theorem

$$\left| \binom{X}{k} \right| = \binom{|X|}{k}.$$

**Proof** Let  $n = |X|$ ,

$$k! \left| \binom{X}{k} \right| = \phi_{1-1}(n, k) = n(n-1)\cdots(n-k+1).$$

Let  $m, n$  be non-negative integers. Let  $Z_+$  denote the non-negative integers. Let

$$S(m, n) = \{(i_1, i_2, \dots, i_n) \in Z_+^n : i_1 + i_2 + \dots + i_n = m\}.$$

### Theorem

$$|S(m, n)| = \binom{m+n-1}{n-1}.$$

**Proof** imagine  $m+n-1$  points in a line. Choose positions  $p_1 < p_2 < \dots < p_{n-1}$  and color these points red. Let  $p_0 = 0, p_n = m+1$ . The gap sizes between the red points

$$i_t = p_t - p_{t-1} - 1, t = 1, 2, \dots, n$$

form a sequence in  $S(m, n)$  and vice-versa. □

$|S(m, n)|$  is also the number of ways of coloring  $m$  indistinguishable balls using  $n$  colors.

Suppose that we want to count the number of ways of coloring these balls so that each color appears at least once i.e. to compute  $|S(m, n)^*|$  where, if  $N = \{1, 2, \dots, \}$

$$S(m, n)^* =$$

$$\{(i_1, i_2, \dots, i_n) \in N^n : i_1 + i_2 + \dots + i_n = m\}$$

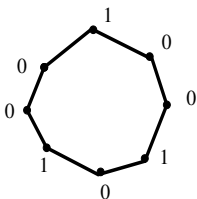
$$= \{(i_1 - 1, i_2 - 1, \dots, i_n - 1) \in Z_+^n :$$

$$(i_1 - 1) + (i_2 - 1) + \dots + (i_n - 1) = m - n\}$$

Thus,

$$|S(m, n)^*| = \binom{m - n + n - 1}{n - 1} = \binom{m - 1}{n - 1}.$$

## Separated 1's on a cycle



How many ways (patterns) are there of placing  $k$  1's and  $n - k$  0's at the vertices of a polygon with  $n$  vertices so that no two 1's are adjacent?

Choose a vertex  $v$  of the polygon in  $n$  ways and then place a 1 there. For the remainder we must choose  $a_1, \dots, a_k \geq 1$  such that  $a_1 + \dots + a_k = n - k$  and then go round the cycle (clockwise) putting  $a_1$  0's followed by a 1 and then  $a_2$  0's followed by a 1 etc..

Each pattern  $\pi$  arises  $k$  times in this way. There are  $k$  choices of  $v$  that correspond to a 1 of the pattern. Having chosen  $v$  there is a unique choice of  $a_1, a_2, \dots, a_k$  that will now give  $\pi$ .

There are  $\binom{n-k-1}{k-1}$  ways of choosing the  $a_i$  and so the answer to our question is

$$\frac{n}{k} \binom{n-k-1}{k-1}$$

## Theorem

### Symmetry

$$\binom{n}{r} = \binom{n}{n-r}$$

**Proof** Choosing  $r$  elements to include is equivalent to choosing  $n - r$  elements to exclude. □

## Theorem

### *Pascal's Triangle*

$$\binom{n}{k} + \binom{n}{k+1} = \binom{n+1}{k+1}$$

**Proof** A  $k+1$ -subset of  $[n+1]$  either  
(i) includes  $n+1$  —  $\binom{n}{k}$  choices or  
(ii) does not include  $n+1$  —  $\binom{n}{k+1}$  choices.

## Pascal's Triangle

The following array of binomial coefficients, constitutes the famous triangle:

```
1
1 1
1 2 1
1 3 3 1
1 4 6 4 1
1 5 10 10 5 1
1 6 15 20 15 6 1
1 7 21 35 35 21 7 1
...
```



## Theorem

$$\binom{k}{k} + \binom{k+1}{k} + \binom{k+2}{k} + \cdots + \binom{n}{k} = \binom{n+1}{k+1}. \quad (2)$$

**Proof** 1: Induction on  $n$  for arbitrary  $k$ .

*Base case:*  $n = k$ ;  $\binom{k}{k} = \binom{k+1}{k+1}$

*Inductive Step:* assume true for  $n \geq k$ .

$$\begin{aligned} \sum_{m=k}^{n+1} \binom{m}{k} &= \sum_{m=k}^n \binom{m}{k} + \binom{n+1}{k} \\ &= \binom{n+1}{k+1} + \binom{n+1}{k} \quad \text{Induction} \\ &= \binom{n+2}{k+1}. \quad \text{Pascal's triangle} \end{aligned}$$

**Proof 2:** Combinatorial argument.

If  $\mathcal{S}$  denotes the set of  $k + 1$ -subsets of  $[n + 1]$  and  $\mathcal{S}_m$  is the set of  $k + 1$ -subsets of  $[n + 1]$  which have largest element  $m + 1$  then

- $\mathcal{S}_k, \mathcal{S}_{k+1}, \dots, \mathcal{S}_n$  is a partition of  $\mathcal{S}$ .
- $|\mathcal{S}_k| + |\mathcal{S}_{k+1}| + \dots + |\mathcal{S}_n| = |\mathcal{S}|$ .
- $|\mathcal{S}_m| = \binom{m}{k}$ .



## Theorem

### *Vandermonde's Identity*

$$\sum_{r=0}^k \binom{m}{r} \binom{n}{k-r} = \binom{m+n}{k}.$$

**Proof** Split  $[m+n]$  into  $A = [m]$  and  $B = [m+n] \setminus [m]$ . Let  $\mathcal{S}$  denote the set of  $k$ -subsets of  $[m+n]$  and let  $\mathcal{S}_r = \{X \in \mathcal{S} : |X \cap A| = r\}$ . Then

- $\mathcal{S}_0, \mathcal{S}_1, \dots, \mathcal{S}_k$  is a partition of  $\mathcal{S}$ .
- $|\mathcal{S}_0| + |\mathcal{S}_1| + \dots + |\mathcal{S}_k| = |\mathcal{S}|$ .
- $|\mathcal{S}_r| = \binom{m}{r} \binom{n}{k-r}$ .
- $|\mathcal{S}| = \binom{m+n}{k}$ .



## Theorem

### *Binomial Theorem*

$$(1 + x)^n = \sum_{r=0}^n \binom{n}{r} x^r.$$

**Proof** Coefficient  $x^r$  in  $(1 + x)(1 + x) \cdots (1 + x)$ : choose  $x$  from  $r$  brackets and 1 from the rest.  $\square$

## Applications of Binomial Theorem

- $x = 1$ :

$$\binom{n}{0} + \binom{n}{1} + \cdots + \binom{n}{n} = (1 + 1)^n = 2^n.$$

LHS counts the number of subsets of all sizes in  $[n]$ .

- $x = -1$ :

$$\binom{n}{0} - \binom{n}{1} + \cdots + (-1)^n \binom{n}{n} = (1 - 1)^n = 0,$$

i.e.

$$\binom{n}{0} + \binom{n}{2} + \binom{n}{4} + \cdots = \binom{n}{1} + \binom{n}{3} + \binom{n}{5} + \cdots$$

and number of subsets of even cardinality = number of subsets of odd cardinality.

$$\sum_{k=0}^n k \binom{n}{k} = n2^{n-1}.$$

Differentiate both sides of the Binomial Theorem w.r.t.  $x$ .

$$n(1+x)^{n-1} = \sum_{k=0}^n k \binom{n}{k} x^{k-1}.$$

Now put  $x = 1$ .

# 9/1/2021

## Inclusion-Exclusion

2 sets:

$$|A_1 \cup A_2| = |A_1| + |A_2| - |A_1 \cap A_2|$$

So if  $A_1, A_2 \subseteq A$  and  $\bar{A}_i = A \setminus A_i$ ,  $i = 1, 2$  then

$$|\bar{A}_1 \cap \bar{A}_2| = |A| - |A_1| - |A_2| + |A_1 \cap A_2|$$

3 sets:

$$\begin{aligned} |\bar{A}_1 \cap \bar{A}_2 \cap \bar{A}_3| &= |A| - |A_1| - |A_2| - |A_3| \\ &\quad + |A_1 \cap A_2| + |A_1 \cap A_3| + |A_2 \cap A_3| \\ &\quad - |A_1 \cap A_2 \cap A_3|. \end{aligned}$$



## General Case

$A_1, A_2, \dots, A_N \subseteq A$  and each  $x \in A$  has a weight  $w_x$ . (In our examples  $w_x = 1$  for all  $x$  and so  $w(X) = |X|$ .)

For  $S \subseteq [N]$ ,  $A_S = \bigcap_{i \in S} A_i$  and  $w(S) = \sum_{x \in S} w_x$ .

E.g.  $A_{\{4,7,18\}} = A_4 \cap A_7 \cap A_{18}$ .

$A_\emptyset = A$ .

Inclusion-Exclusion Formula:

$$w \left( \bigcap_{i=1}^N \bar{A}_i \right) = \sum_{S \subseteq [N]} (-1)^{|S|} w(A_S).$$

Simple example. How many integers in  $[1000]$  are not divisible by 5,6 or 8 i.e. what is the size of  $\bar{A}_1 \cap \bar{A}_2 \cap \bar{A}_3$  below? Here we take  $w_x = 1$  for all  $x$ .

$A = A_\emptyset$	$= \{1, 2, 3, \dots, \}$	$ A  = 1000$
$A_1$	$= \{5, 10, 15, \dots, \}$	$ A_1  = 200$
$A_2$	$= \{6, 12, 18, \dots, \}$	$ A_2  = 166$
$A_3$	$= \{8, 16, 24, \dots, \}$	$ A_3  = 125$
$A_{\{1,2\}}$	$= \{30, 60, 90, \dots, \}$	$ A_{\{1,2\}}  = 33$
$A_{\{1,3\}}$	$= \{40, 80, 120, \dots, \}$	$ A_{\{1,3\}}  = 25$
$A_{\{2,3\}}$	$= \{24, 48, 72, \dots, \}$	$ A_{\{2,3\}}  = 41$
$A_{\{1,2,3\}}$	$= \{120, 240, 360, \dots, \}$	$ A_{\{1,2,3\}}  = 8$

$$\begin{aligned}
 |\bar{A}_1 \cap \bar{A}_2 \cap \bar{A}_3| &= 1000 - (200 + 166 + 125) \\
 &\quad + (33 + 25 + 41) - 8 \\
 &= 600.
 \end{aligned}$$

## Derangements

A **derangement** of  $[n]$  is a permutation  $\pi$  such that

$$\pi(i) \neq i : i = 1, 2, \dots, n.$$

We must express the set of derangements  $D_n$  of  $[n]$  as the intersection of the complements of sets.

We let  $A_i = \{\text{permutations } \pi : \pi(i) = i\}$  and then

$$|D_n| = \left| \bigcap_{i=1}^n \bar{A}_i \right|.$$

We must now compute  $|A_S|$  for  $S \subseteq [n]$ .

$|A_1| = (n - 1)!$ : after fixing  $\pi(1) = 1$  there are  $(n - 1)!$  ways of permuting  $2, 3, \dots, n$ .

$|A_{\{1,2\}}| = (n - 2)!$ : after fixing  $\pi(1) = 1, \pi(2) = 2$  there are  $(n - 2)!$  ways of permuting  $3, 4, \dots, n$ .

In general

$$|A_S| = (n - |S|)!$$

$$\begin{aligned} |D_n| &= \sum_{S \subseteq [n]} (-1)^{|S|} (n - |S|)! \\ &= \sum_{k=0}^n (-1)^k \binom{n}{k} (n - k)! \\ &= \sum_{k=0}^n (-1)^k \frac{n!}{k!} \\ &= n! \sum_{k=0}^n (-1)^k \frac{1}{k!}. \end{aligned}$$

When  $n$  is large,

$$\sum_{k=0}^n (-1)^k \frac{1}{k!} \approx e^{-1}.$$

## Proof of inclusion-exclusion formula

$$\theta_{x,i} = \begin{cases} 1 & x \in A_i \\ 0 & x \notin A_i \end{cases}$$

$$(1 - \theta_{x,1})(1 - \theta_{x,2}) \cdots (1 - \theta_{x,N}) = \begin{cases} 1 & x \in \bigcap_{i=1}^N \bar{A}_i \\ 0 & \text{otherwise} \end{cases}$$

So

$$\begin{aligned} w \left( \bigcap_{i=1}^N \bar{A}_i \right) &= \sum_{x \in A} w_x (1 - \theta_{x,1})(1 - \theta_{x,2}) \cdots (1 - \theta_{x,N}) \\ &= \sum_{x \in A} w_x \sum_{S \subseteq [N]} (-1)^{|S|} \prod_{i \in S} \theta_{x,i} \\ &= \sum_{S \subseteq [N]} (-1)^{|S|} \sum_{x \in A} w_x \prod_{i \in S} \theta_{x,i} \\ &= \sum_{S \subseteq [N]} (-1)^{|S|} w(A_S). \end{aligned}$$

## Euler's Function $\phi(n)$ .

Let  $\phi(n)$  be the number of positive integers  $x \leq n$  which are mutually prime to  $n$  i.e. have no common factors with  $n$ , other than 1.

$$\phi(12) = 4.$$

Let  $n = p_1^{\alpha_1} p_2^{\alpha_2} p_1^{\alpha_2} \cdots p_k^{\alpha_k}$  be the prime factorisation of  $n$ .

$$A_i = \{x \in [n] : p_i \text{ divides } x\}, \quad 1 \leq i \leq k.$$

$$\phi(n) = \left| \bigcap_{i=1}^k \bar{A}_i \right|$$

$$|A_S| = \frac{n}{\prod_{i \in S} p_i} \quad S \subseteq [k].$$

$$\begin{aligned} \phi(n) &= \sum_{S \subseteq [k]} (-1)^{|S|} \frac{n}{\prod_{i \in S} p_i} \\ &= n \left(1 - \frac{1}{p_1}\right) \left(1 - \frac{1}{p_2}\right) \cdots \left(1 - \frac{1}{p_k}\right) \end{aligned}$$



## Surjections

Fix  $n, m$ . Let

$$A = \{f : [n] \rightarrow [m]\}$$

Thus  $|A| = m^n$ . Let

$$F(n, m) = \{f \in A : f \text{ is onto } [m]\}.$$

How big is  $F(n, m)$ ?

Let

$$A_i = \{f \in F : f(x) \neq i, \forall x \in [n]\}.$$

Then

$$F(n, m) = \bigcap_{i=1}^m \bar{A}_i.$$

For  $S \subseteq [m]$

$$\begin{aligned} A_S &= \{f \in A : f(x) \notin S, \forall x \in [n]\}. \\ &= \{f : [n] \rightarrow [m] \setminus S\}. \end{aligned}$$

So

$$|A_S| = (m - |S|)^n.$$

Hence

$$\begin{aligned} F(n, m) &= \sum_{S \subseteq [m]} (-1)^{|S|} (m - |S|)^n \\ &= \sum_{k=0}^m (-1)^k \binom{m}{k} (m - k)^n. \end{aligned}$$

## Scrambled Allocations

We have  $n$  boxes  $B_1, B_2, \dots, B_n$  and  $2n$  distinguishable balls  $b_1, b_2, \dots, b_{2n}$ .

An allocation of balls to boxes, **two balls to a box**, is said to be *scrambled* if there does **not** exist  $i$  such that box  $B_i$  contains balls  $b_{2i-1}, b_{2i}$ . Let  $\sigma_n$  be the number of scrambled allocations.

Let  $A_i$  be the set of allocations in which box  $B_i$  contains  $b_{2i-1}, b_{2i}$ . We show that

$$|A_S| = \frac{(2(n - |S|))!}{2^{n-|S|}}.$$

Inclusion-Exclusion then gives

$$\sigma_n = \sum_{k=0}^n (-1)^k \binom{n}{k} \frac{(2(n-k))!}{2^{n-k}}.$$

First consider  $A_\emptyset$ :

Each permutation  $\pi$  of  $[2n]$  yields an allocation of balls, placing  $b_{\pi(2i-1)}, b_{\pi(2i)}$  into box  $B_i$ , for  $i = 1, 2, \dots, n$ . The order of balls in the boxes is immaterial and so each allocation comes from exactly  $2^n$  distinct permutations, giving

$$|A_\emptyset| = \frac{(2n)!}{2^n}.$$

To get the formula for  $|A_S|$  observe that the contents of  $2|S|$  boxes are fixed and so we are in essence dealing with  $n - |S|$  boxes and  $2(n - |S|)$  balls.

# 9/3/2021

## Problème des Ménages

In how many ways  $M_n$  can  $n$  male-female couples be seated around a table, alternating male-female, so that no person is seated next to their partner?

Let  $A_i$  be the set of seatings in which couple  $i$  sit together.

If  $|S| = k$  then

$$|A_S| = 2k!(n - k)!^2 \times d_k.$$

$d_k$  is the number of ways of placing  $k$  1's on a cycle of length  $2n$  so that no two 1's are adjacent. (We place a person at each 1 and his/her partner on the succeeding 0).

2 choices for which seats are occupied by the men or women.

$k!$  ways of assigning the couples to the positions;  $(n - k)!^2$

ways of assigning the rest of the people.

$$d_k = \frac{2n}{k} \binom{2n-k-1}{k-1} = \frac{2n}{2n-k} \binom{2n-k}{k}.$$

(See slides 11 and 12).

$$\begin{aligned} M_n &= \sum_{k=0}^n (-1)^k \binom{n}{k} \times 2k!(n-k)!^2 \times \frac{2n}{2n-k} \binom{2n-k}{k} \\ &= 2n! \sum_{k=0}^n (-1)^k \frac{2n}{2n-k} \binom{2n-k}{k} (n-k)!. \end{aligned}$$

## The weight of elements in exactly $k$ sets:

Observe that

$$\prod_{i \in S} \theta_{x,i} \prod_{i \notin S} (1 - \theta_{x,i}) = 1 \text{ iff } x \in A_i, i \in S \text{ and } x \notin A_i, i \notin S.$$

$W_k$  is the total weight of elements in exactly  $k$  of the  $A_i$ :

$$\begin{aligned} N_k &= \sum_{x \in A} w_x \sum_{|S|=k} \prod_{i \in S} \theta_{x,i} \prod_{i \notin S} (1 - \theta_{x,i}) \\ &= \sum_{|S|=k} \sum_{x \in A} w_x \prod_{i \in S} \theta_{x,i} \prod_{i \notin S} (1 - \theta_{x,i}) \\ &= \sum_{|S|=k} \sum_{T \supseteq S} \sum_{x \in A} w_x (-1)^{|T \setminus S|} \prod_{i \in T} \theta_{x,i} \\ &= \sum_{|S|=k} \sum_{T \supseteq S} (-1)^{|T \setminus S|} w(A_T) \\ &= \sum_{\ell=k}^N \sum_{|T|=\ell} (-1)^{\ell-k} \binom{\ell}{k} w(A_T). \end{aligned}$$



As an example. Let  $D_{n,k}$  denote the number of permutations  $\pi$  of  $[n]$  for which there are exactly  $k$  indices  $i$  for which  $\pi(i) = i$ . Then

$$\begin{aligned} D_{n,k} &= \sum_{\ell=k}^n \binom{n}{\ell} (-1)^{\ell-k} \binom{\ell}{k} (n-\ell)! \\ &= \sum_{\ell=k}^n \frac{n!}{\ell!(n-\ell)!} (-1)^{\ell-k} \frac{\ell!}{k!(\ell-k)!} (n-\ell)! \\ &= \frac{n!}{k!} \sum_{\ell=k}^n \frac{(-1)^{\ell-k}}{(\ell-k)!} \\ &= \frac{n!}{k!} \sum_{r=0}^{n-k} \frac{(-1)^r}{r!} \\ &\approx \frac{n!}{ek!} \end{aligned}$$

when  $n$  is large and  $k$  is constant.

# 9/8/2021

## Recurrence Relations

Suppose  $a_0, a_1, a_2, \dots, a_n, \dots$  is an infinite sequence.  
A recurrence relation is a set of equations

$$a_n = f_n(a_{n-1}, a_{n-2}, \dots, a_{n-k}). \quad (3)$$

The whole sequence is determined by (3) and the values of  $a_0, a_1, \dots, a_{k-1}$ .

# Linear Recurrence

## Fibonacci Sequence

$$a_n = a_{n-1} + a_{n-2} \quad n \geq 2.$$

$$a_0 = a_1 = 1.$$

$b_n = |B_n| = |\{x \in \{a, b, c\}^n : aa \text{ does not occur in } x\}|.$

$b_1 = 3 : a b c$

$b_2 = 8 : ab ac ba bb bc ca cb cc$

$b_n = 2b_{n-1} + 2b_{n-2} \quad n \geq 2.$

$$b_n = 2b_{n-1} + 2b_{n-2} \quad n \geq 2.$$

Let

$$B_n = B_n^{(b)} \cup B_n^{(c)} \cup B_n^{(a)}$$

where  $B_n^{(\alpha)} = \{x \in B_n : x_1 = \alpha\}$  for  $\alpha = a, b, c$ .

Now  $|B_n^{(b)}| = |B_n^{(c)}| = |B_{n-1}|$ . The map  $f : B_n^{(b)} \rightarrow B_{n-1}$ ,

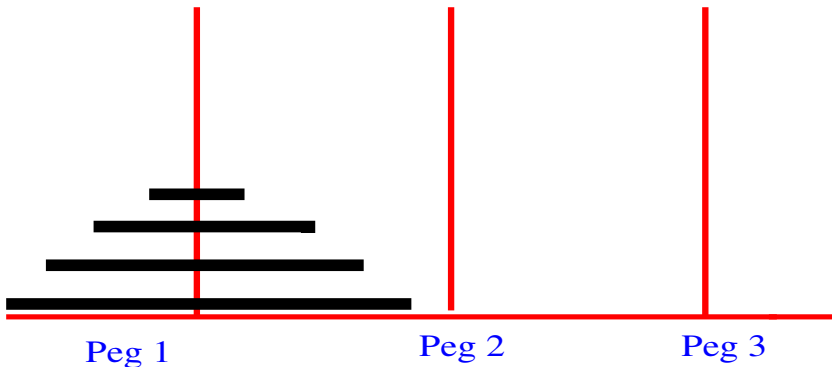
$$f(bx_2x_3 \dots x_n) = x_2x_3 \dots x_n \text{ is a bijection.}$$

$B_n^{(a)} = \{x \in B_n : x_1 = a \text{ and } x_2 = b \text{ or } c\}$ . The map  $g : B_n^{(a)} \rightarrow B_{n-1}^{(b)} \cup B_{n-1}^{(c)}$ ,

$$g(ax_2x_3 \dots x_n) = x_2x_3 \dots x_n \text{ is a bijection.}$$

Hence,  $|B_n^{(a)}| = 2|B_{n-1}|$ .

## Towers of Hanoi



$H_n$  is the minimum number of moves needed to shift  $n$  rings from Peg 1 to Peg 2. One is not allowed to place a larger ring on top of a smaller ring.

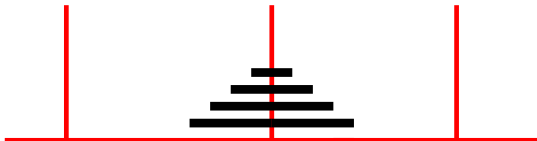
XXX



$H_{n-1}$  moves



1 move



$H_{n-1}$  moves



We see that  $H_1 = 1$  and  $H_n = 2H_{n-1} + 1$  for  $n \geq 2$ .

So,

$$\frac{H_n}{2^n} - \frac{H_{n-1}}{2^{n-1}} = \frac{1}{2^n}.$$

Summing these equations give

$$\frac{H_n}{2^n} - \frac{H_1}{2} = \frac{1}{2^n} + \frac{1}{2^{n-1}} + \cdots + \frac{1}{4} = \frac{1}{2} - \frac{1}{2^n}.$$

So

$$H_n = 2^n - 1.$$

$A$  has  $n$  dollars. Everyday  $A$  buys one of a Bun (1 dollar), an Ice-Cream (2 dollars) or a Pastry (2 dollars). How many ways are there (sequences) for  $A$  to spend his money?  
Ex. **BBPIIPBI** represents “Day 1, buy Bun. Day 2, buy Bun etc.”.

$$\begin{aligned}u_n &= \text{number of ways} \\ &= u_{n,B} + u_{n,I} + u_{n,P}\end{aligned}$$

where  $u_{n,B}$  is the number of ways where  $A$  buys a Bun on day 1 etc.

$$u_{n,B} = u_{n-1}, \quad u_{n,I} = u_{n,P} = u_{n-2}.$$

So

$$u_n = u_{n-1} + 2u_{n-2},$$

and

$$u_0 = u_1 = 1.$$

If  $a_0, a_1, \dots, a_n$  is a sequence of real numbers then its **(ordinary) generating function**  $a(x)$  is given by

$$a(x) = a_0 + a_1x + a_2x^2 + \cdots a_nx^n + \cdots$$

and we write

$$a_n = [x^n]a(x).$$

For more on this subject see **Generatingfunctionology** by the late Herbert S. Wilf. The book is available from <https://www.math.upenn.edu//wilf/DownldGF.html>

$$a_n = 1$$

$$a(x) = \frac{1}{1-x} = 1 + x + x^2 + \cdots + x^n + \cdots$$

$$a_n = n + 1.$$

$$a(x) = \frac{1}{(1-x)^2} = 1 + 2x + 3x^2 + \cdots + (n+1)x^n + \cdots$$

$$a_n = n.$$

$$a(x) = \frac{x}{(1-x)^2} = x + 2x^2 + 3x^3 + \cdots + nx^n + \cdots$$

## Generalised binomial theorem:

$$a_n = \binom{\alpha}{n}$$

$$a(x) = (1+x)^\alpha = \sum_{n=0}^{\infty} \binom{\alpha}{n} x^n.$$

where

$$\binom{\alpha}{n} = \frac{\alpha(\alpha-1)(\alpha-2)\cdots(\alpha-n+1)}{n!}.$$

$$a_n = \binom{m+n-1}{n}$$

$$a(x) = \frac{1}{(1-x)^m} = \sum_{n=0}^{\infty} \binom{-m}{n} (-x)^n = \sum_{n=0}^{\infty} \binom{m+n-1}{n} x^n.$$

General view.

Given a recurrence relation for the sequence  $(a_n)$ , we

(a) Deduce from it, an equation satisfied by the generating function  $a(x) = \sum_n a_n x^n$ .

(b) Solve this equation to get an explicit expression for the generating function.

(c) Extract the coefficient  $a_n$  of  $x^n$  from  $a(x)$ , by expanding  $a(x)$  as a power series.

## Solution of linear recurrences

$$a_n - 6a_{n-1} + 9a_{n-2} = 0 \quad n \geq 2.$$

$$a_0 = 1, a_1 = 9.$$

$$\sum_{n=2}^{\infty} (a_n - 6a_{n-1} + 9a_{n-2})x^n = 0. \quad (4)$$

$$\begin{aligned}\sum_{n=2}^{\infty} a_n x^n &= a(x) - a_0 - a_1 x \\ &= a(x) - 1 - 9x.\end{aligned}$$

$$\begin{aligned}\sum_{n=2}^{\infty} 6a_{n-1} x^n &= 6x \sum_{n=2}^{\infty} a_{n-1} x^{n-1} \\ &= 6x(a(x) - a_0) \\ &= 6x(a(x) - 1).\end{aligned}$$

$$\begin{aligned}\sum_{n=2}^{\infty} 9a_{n-2} x^n &= 9x^2 \sum_{n=2}^{\infty} a_{n-2} x^{n-2} \\ &= 9x^2 a(x).\end{aligned}$$



$$a(x) - 1 - 9x - 6x(a(x) - 1) + 9x^2 a(x) = 0$$

or

$$a(x)(1 - 6x + 9x^2) - (1 + 3x) = 0.$$

$$\begin{aligned} a(x) &= \frac{1 + 3x}{1 - 6x + 9x^2} = \frac{1 + 3x}{(1 - 3x)^2} \\ &= \sum_{n=0}^{\infty} (n+1)3^n x^n + 3x \sum_{n=0}^{\infty} (n+1)3^n x^n \\ &= \sum_{n=0}^{\infty} (n+1)3^n x^n + \sum_{n=0}^{\infty} n3^n x^n \\ &= \sum_{n=0}^{\infty} (2n+1)3^n x^n. \end{aligned}$$

$$a_n = (2n+1)3^n.$$

Fibonacci sequence:

$$\sum_{n=2}^{\infty} (a_n - a_{n-1} - a_{n-2})x^n = 0.$$

$$\sum_{n=2}^{\infty} a_n x^n - \sum_{n=2}^{\infty} a_{n-1} x^n - \sum_{n=2}^{\infty} a_{n-2} x^n = 0.$$

$$(a(x) - a_0 - a_1 x) - (x(a(x) - a_0)) - x^2 a(x) = 0.$$

$$a(x) = \frac{1}{1 - x - x^2}.$$

$$\begin{aligned} a(x) &= -\frac{1}{(\xi_1 - x)(\xi_2 - x)} \\ &= \frac{1}{\xi_1 - \xi_2} \left( \frac{1}{\xi_1 - x} - \frac{1}{\xi_2 - x} \right) \\ &= \frac{1}{\xi_1 - \xi_2} \left( \frac{\xi_1^{-1}}{1 - x/\xi_1} - \frac{\xi_2^{-1}}{1 - x/\xi_2} \right) \end{aligned}$$

where

$$\xi_1 = -\frac{\sqrt{5} + 1}{2} \text{ and } \xi_2 = \frac{\sqrt{5} - 1}{2}$$

are the 2 roots of

$$x^2 + x - 1 = 0.$$

Therefore,

$$\begin{aligned} a(x) &= \frac{\xi_1^{-1}}{\xi_1 - \xi_2} \sum_{n=0}^{\infty} \xi_1^{-n} x^n - \frac{\xi_2^{-1}}{\xi_1 - \xi_2} \sum_{n=0}^{\infty} \xi_2^{-n} x^n \\ &= \sum_{n=0}^{\infty} \frac{\xi_1^{-n-1} - \xi_2^{-n-1}}{\xi_1 - \xi_2} x^n \end{aligned}$$

and so

$$\begin{aligned} a_n &= \frac{\xi_1^{-n-1} - \xi_2^{-n-1}}{\xi_1 - \xi_2} \\ &= \frac{1}{\sqrt{5}} \left( \left( \frac{\sqrt{5}+1}{2} \right)^{n+1} - \left( \frac{1-\sqrt{5}}{2} \right)^{n+1} \right). \end{aligned}$$

# 9/10/2021

## Inhomogeneous problem

$$a_n - 3a_{n-1} = n^2 \quad n \geq 1.$$

$$a_0 = 1.$$

$$\begin{aligned} \sum_{n=1}^{\infty} (a_n - 3a_{n-1})x^n &= \sum_{n=1}^{\infty} n^2 x^n \\ \sum_{n=1}^{\infty} n^2 x^n &= \sum_{n=2}^{\infty} n(n-1)x^n + \sum_{n=1}^{\infty} nx^n \\ &= \frac{2x^2}{(1-x)^3} + \frac{x}{(1-x)^2} \\ &= \frac{x + x^2}{(1-x)^3} \\ \sum_{n=1}^{\infty} (a_n - 3a_{n-1})x^n &= a(x) - 1 - 3xa(x) \\ &= a(x)(1 - 3x) - 1. \end{aligned}$$

$$\begin{aligned} a(x) &= \frac{x + x^2}{(1-x)^3(1-3x)} + \frac{1}{1-3x} \\ &= \frac{A}{1-x} + \frac{B}{(1-x)^2} + \frac{C}{(1-x)^3} + \frac{D+1}{1-3x} \end{aligned}$$

where

$$\begin{aligned} x + x^2 &\cong A(1-x)^2(1-3x) + B(1-x)(1-3x) \\ &\quad + C(1-3x) + D(1-x)^3. \end{aligned}$$

Then

$$A = -1/2, B = 0, C = -1, D = 3/2.$$

So

$$\begin{aligned}a(x) &= \frac{-1/2}{1-x} - \frac{1}{(1-x)^3} + \frac{5/2}{1-3x} \\ &= -\frac{1}{2} \sum_{n=0}^{\infty} x^n - \sum_{n=0}^{\infty} \binom{n+2}{2} x^n + \frac{5}{2} \sum_{n=0}^{\infty} 3^n x^n\end{aligned}$$

So

$$\begin{aligned}a_n &= -\frac{1}{2} - \binom{n+2}{2} + \frac{5}{2} 3^n \\ &= -\frac{3}{2} - \frac{3n}{2} - \frac{n^2}{2} + \frac{5}{2} 3^n.\end{aligned}$$



## Products of generating functions

$$a(x) = \sum_{n=0}^{\infty} a_n x^n, \quad b(x) = \sum_{n=0}^{\infty} b_n x^n.$$

$$\begin{aligned} a(x)b(x) &= (a_0 + a_1x + a_2x^2 + \cdots) \times \\ &\quad (b_0 + b_1x + b_2x^2 + \cdots) \\ &= a_0b_0 + (a_0b_1 + a_1b_0)x + \\ &\quad (a_0b_2 + a_1b_1 + a_2b_0)x^2 + \cdots \\ &= \sum_{n=0}^{\infty} c_n x^n \end{aligned}$$

where

$$c_n = \sum_{k=0}^n a_k b_{n-k}.$$

## Derangements

$$n! = \sum_{k=0}^n \binom{n}{k} d_{n-k}.$$

**Explanation:**  $\binom{n}{k} d_{n-k}$  is the number of permutations with exactly  $k$  cycles of length 1. Choose  $k$  elements ( $\binom{n}{k}$  ways) for which  $\pi(i) = i$  and then choose a derangement of the remaining  $n - k$  elements.

So

$$\begin{aligned} 1 &= \sum_{k=0}^n \frac{1}{k!} \frac{d_{n-k}}{(n-k)!} \\ \sum_{n=0}^{\infty} x^n &= \sum_{n=0}^{\infty} \left( \sum_{k=0}^n \frac{1}{k!} \frac{d_{n-k}}{(n-k)!} \right) x^n. \end{aligned} \tag{5}$$

Let

$$d(x) = \sum_{m=0}^{\infty} \frac{d_m}{m!} x^m.$$

From (5) we have

$$\begin{aligned} \frac{1}{1-x} &= e^x d(x) \\ d(x) &= \frac{e^{-x}}{1-x} \\ &= \sum_{n=0}^{\infty} \sum_{k=0}^n \left( \frac{(-1)^k}{k!} \right) x^n. \end{aligned}$$

So

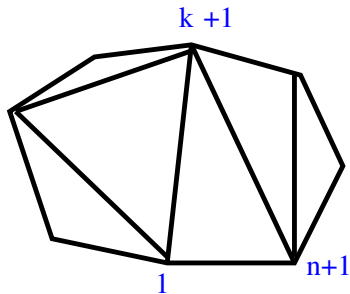
$$\frac{d_n}{n!} = \sum_{k=0}^n \frac{(-1)^k}{k!}.$$

## Triangulation of $n$ -gon

Let

$$\begin{aligned} a_n &= \text{number of triangulations of } P_{n+1} \\ &= \sum_{k=0}^n a_k a_{n-k} \quad n \geq 2 \end{aligned} \tag{6}$$

$$a_0 = 0, a_1 = a_2 = 1.$$



Explanation of (6):

$a_k a_{n-k}$  counts the number of triangulations in which edge  $1, n+1$  is contained in triangle  $1, k+1, n+1$ .

There are  $a_k$  ways of triangulating  $1, 2, \dots, k+1, 1$  and for each such there are  $a_{n-k}$  ways of triangulating  $k+1, k+2, \dots, n+1, k+1$ .

$$x + \sum_{n=2}^{\infty} a_n x^n = x + \sum_{n=2}^{\infty} \left( \sum_{k=0}^n a_k a_{n-k} \right) x^n.$$

But,

$$x + \sum_{n=2}^{\infty} a_n x^n = a(x)$$

since  $a_0 = 0, a_1 = 1$ .

$$\begin{aligned} \sum_{n=2}^{\infty} \left( \sum_{k=0}^n a_k a_{n-k} \right) x^n &= \sum_{n=0}^{\infty} \left( \sum_{k=0}^n a_k a_{n-k} \right) x^n \\ &= a(x)^2. \end{aligned}$$

So

$$a(x) = x + a(x)^2$$

and hence

$$a(x) = \frac{1 + \sqrt{1 - 4x}}{2} \text{ or } \frac{1 - \sqrt{1 - 4x}}{2}.$$

But  $a(0) = 0$  and so

$$\begin{aligned} a(x) &= \frac{1 - \sqrt{1 - 4x}}{2} \\ &= \frac{1}{2} - \frac{1}{2} \left( 1 + \sum_{n=1}^{\infty} \frac{(-1)^{n-1}}{n2^{2n-1}} \binom{2n-2}{n-1} (-4x)^n \right) \\ &= \sum_{n=1}^{\infty} \frac{1}{n} \binom{2n-2}{n-1} x^n. \end{aligned}$$

So

$$a_n = \frac{1}{n} \binom{2n-2}{n-1}.$$

# 9/13/2021



## Colouring Problem

### Theorem

Let  $A_1, A_2, \dots, A_n$  be subsets of  $A$  and  $|A_i| = k$  for  $1 \leq i \leq n$ . If  $n < 2^{k-1}$  then there exists a partition  $A = R \cup B$  such that

$$A_i \cap R \neq \emptyset \text{ and } A_i \cap B \neq \emptyset \quad 1 \leq i \leq n.$$

[ $R$  = Red elements and  $B$  = Blue elements.]

**Proof** Randomly colour  $A$ .

$\Omega = \{R, B\}^A = \{f : A \rightarrow \{R, B\}\}$ , uniform distribution.

$$BAD = \{\exists i : A_i \subseteq R \text{ or } A_i \subseteq B\}.$$

**Claim:**  $\Pr(BAD) < 1$ .

Thus  $\Omega \setminus BAD \neq \emptyset$  and this proves the theorem.

$$BAD(i) = \{A_i \subseteq R \text{ or } A_i \subseteq B\} \text{ and } BAD = \bigcup_{i=1}^n BAD(i).$$

**Boole's Inequality:** if  $\mathcal{A}_1, \mathcal{A}_2, \dots, \mathcal{A}_N$  are a collection of events, then

$$\Pr \left( \bigcup_{i=1}^N \mathcal{A}_i \right) \leq \sum_{i=1}^N \Pr(\mathcal{A}_i).$$

This easily proved by induction on  $N$ . When  $N = 2$  we use

$$\Pr(\mathcal{A}_1 \cup \mathcal{A}_2) = \Pr(\mathcal{A}_1) + \Pr(\mathcal{A}_2) - \Pr(\mathcal{A}_1 \cap \mathcal{A}_2) \leq \Pr(\mathcal{A}_1 \cup \mathcal{A}_2).$$

In general,

$$\Pr \left( \bigcup_{i=1}^N \mathcal{A}_i \right) \leq \Pr \left( \bigcup_{i=1}^{N-1} \mathcal{A}_i \right) + \Pr(\mathcal{A}_N) \leq \sum_{i=1}^{N-1} \Pr(\mathcal{A}_i) + \Pr(\mathcal{A}_N).$$

The first inequality is the two event case and the second is by induction on  $N$ .

So,

$$\begin{aligned}\Pr(\text{BAD}) &\leq \sum_{i=1}^n \Pr(\text{BAD}(i)) \\ &= \sum_{i=1}^n \left(\frac{1}{2}\right)^{k-1} \\ &= n/2^{k-1} \\ &< 1.\end{aligned}$$

Example of system which is not 2-colorable.

Let  $n = \binom{2k-1}{k}$  and  $A = [2k - 1]$  and

$$\{A_1, A_2, \dots, A_n\} = \binom{[2k - 1]}{k}.$$

Then in any 2-coloring of  $A_1, A_2, \dots, A_n$  there is a set  $A_i$  all of whose elements are of one color.

Suppose  $A$  is partitioned into 2 sets  $R, B$ . At least one of these two sets is of size at least  $k$  (since  $(k - 1) + (k - 1) < 2k - 1$ ). Suppose then that  $R \geq k$  and let  $S$  be any  $k$ -subset of  $R$ . Then there exists  $i$  such that  $A_i = S \subseteq R$ .

## Tournaments

$n$  players in a tournament each play each other i.e. there are  $\binom{n}{2}$  games.

Fix some  $k$ . Is it possible that for every set  $S$  of  $k$  players there is a person  $w_S$  who beats everyone in  $S$ ?

Suppose that the results of the tournament are decided by a random coin toss.

Fix  $S$ ,  $|S| = k$  and let  $\mathcal{E}_S$  be the event that nobody beats everyone in  $S$ .

The event

$$\mathcal{E} = \bigcup_S \mathcal{E}_S$$

is that there is a set  $S$  for which  $w_S$  does not exist.

We only have to show that  $\Pr(\mathcal{E}) < 1$ .

$$\begin{aligned}\Pr(\mathcal{E}) &\leq \sum_{|S|=k} \Pr(\mathcal{E}_S) \\ &= \binom{n}{k} (1 - 2^{-k})^{n-k} \\ &< n^k e^{-(n-k)2^{-k}} \\ &= \exp\{k \ln n - (n-k)2^{-k}\} \\ &\rightarrow 0\end{aligned}$$

since we are assuming here that  $k$  is fixed independent of  $n$ .

## Graph Crossing Number

The crossing number of a graph  $G$  is the minimum number of edge crossings of a drawing of  $G$  in the plane.

Euler's formula implies that a planar graph with  $n$  vertices has at most  $3n$  edges.

This implies that a graph  $G = (V, E)$  requires at least  $|E| - 3|V|$  crossings.

### Theorem

If  $|E| > 4|V|$  then  $G$  has crossing number  $\Omega(|E|^3/|V|^2)$ .

If  $|E| \approx |V|^{3/2}$  then this gives  $\Omega(|V|^{5/2})$  whereas  $|E| - 3|V| = O(|V|^{3/2})$ .



# Proof

Suppose that  $G$  has a drawing with  $k$  crossings and let  $0 < p < 1$ .

Let  $G_p = (V_p, E_p)$  denote the subgraph of  $G$  obtained by including each vertex in  $V_p$  independently with probability  $p$ .

$E_p$  is then the set of edges  $\{x, y\}$  such that  $x, y \in V_p$ .

$$\mathbf{E}(|V_p|) = p|V| \text{ and } \mathbf{E}(|E_p|) = p^2|E|.$$

Also,

$$\mathbf{E}(\text{number of crossings in the drawing of } G_p) = p^4 k.$$

So,

$$p^4 k \geq \mathbf{E}(|E_p| - 3|V_p|) = p^2|E| - 3p|V|.$$

So

$$k \geq \frac{p^2|E| - 3p|V|}{p^4}.$$

Maximising the RHS over  $p \leq 1$  gives  $p = 4|V|/|E|$  and

$$k \geq \frac{|E|^3}{64|V|^2}.$$



# 9/15/2021

## Random Binary Search Trees



A binary tree consists of a set of *nodes*, one of which is the *root*. Each node is connected to 0,1 or 2 nodes below it and every node other than the root is connected to exactly one node above it. The root is the highest node.

The depth of a node is the number of edges in its path to the root.

The depth of a tree is the maximum over the depths of its nodes.

Starting with a tree  $T_0$  consisting of a single root  $r$ , we grow a tree  $T_n$  as follows:

The  $n$ 'th *particle* starts at  $r$  and flips a fair coin. It goes left (L) with probability  $1/2$  and right (R) with probability  $1/2$ .

It tries to move along the tree in the chosen direction. If there is a node below it in this direction then it goes there and continues its random moves. Otherwise it creates a new node where it wanted to move and stops.

Let  $D_n$  be the depth of this tree.

**Claim:** for any  $t \geq 0$ ,

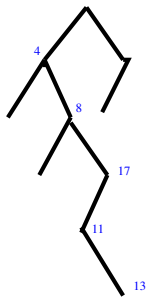
$$\Pr(D_n \geq t) \leq (n2^{-(t-1)/2})^t.$$

**Proof** The process requires at most  $n^2$  coin flips and so we let  $\Omega = \{L, R\}^{n^2}$  – most coin flips will not be needed most of the time.

$$DEEP = \{D_n \geq t\}.$$

For  $P \in \{L, R\}^t$  and  $S \subseteq [n]$ ,  $|S| = t$  let  $DEEP(P, S) = \{\text{the particles } S = \{s_1, s_2, \dots, s_t\} \text{ follow } P \text{ in the tree i.e. the first } i \text{ moves of } s_i \text{ are along } P, 1 \leq i \leq t\}$ .

$$DEEP = \bigcup_P \bigcup_S DEEP(P, S).$$



$$S=\{4,8,11,13,17\}$$

$t=5$  and DEEP(P,S) occurs if

4 goes L...

8 goes LR...

17 goes LRR...

11 goes LRRL...

13 goes LRRLR...

$$\begin{aligned}
\Pr(\text{DEEP}) &\leq \sum_P \sum_S \Pr(\text{DEEP}(P, S)) \\
&= \sum_P \sum_S 2^{-(1+2+\dots+t)} \\
&= \sum_P \sum_S 2^{-t(t+1)/2} \\
&= 2^t \binom{n}{t} 2^{-t(t+1)/2} \\
&\leq 2^t n^t 2^{-t(t+1)/2} \\
&= (n 2^{-(t-1)/2})^t.
\end{aligned}$$

So if we put  $t = A \log_2 n$  then

$$\Pr(D_n \geq A \log_2 n) \leq (2n^{1-A/2})^{A \log_2 n}$$

which is very small, for  $A > 2$ .



## A problem with hats

There are  $n$  people standing a circle. They are blind-folded and someone places a hat on each person's head. The hat has been randomly colored Red or Blue.

They take off their blind-folds and everyone can see everyone else's hat. Each person then simultaneously declares (i) my hat is red or (ii) my hat is blue or (iii) or I pass.

They win a big prize if the people who opt for (i) or (ii) are all correct. They pay a big penalty if there is a person who incorrectly guesses the color of their hat.

Is there a strategy which means they will win with probability better than  $1/2$ ?

Suppose that we partition  $Q_n = \{0, 1\}^n$  into 2 sets  $W, L$  which have the property that  $L$  is a **cover** i.e. if

$x = x_1x_2 \cdots x_n \in W = Q_n \setminus L$  then there is  $y_1y_2 \cdots y_n \in L$  such that  $h(x, y) = 1$  where

$$h(x, y) = |\{j : x_j \neq y_j\}|.$$

Hamming distance between  $x$  and  $y$ .

Assume that  $0 \equiv \text{Red}$  and  $1 \equiv \text{Blue}$ . Person  $i$  knows  $x_j$  for  $j \neq i$  (color of hat  $j$ ) and if there is a **unique** value  $\xi$  of  $x_i$  which places  $x$  in  $W$  then person  $i$  will declare that their hat has color  $\xi$ .

The people assume that  $x \in W$  and if indeed  $x \in W$  then there is at least one person who will be in this situation and any such person will guess correctly.

Is there a small cover  $L$ ?

Let  $p = \frac{\ln n}{n}$ . Choose  $L_1$  randomly by placing  $y \in Q_n$  into  $L_1$  with probability  $p$ .

Then let  $L_2$  be those  $z \in Q_n$  which are not at Hamming distance  $\leq 1$  from some member of  $L_1$ .

Clearly  $L = L_1 \cup L_2$  is a cover and

$$\mathbf{E}(|L|) = 2^n p + 2^n (1 - p)^{n+1} \leq 2^n (p + e^{-np}) \leq 2^n \frac{2 \ln n}{n}.$$

So there must exist a cover of size at most  $2^n \frac{2 \ln n}{n}$  and the players can win with probability at least  $1 - \frac{2 \ln n}{n}$ .

# 9/22/2021

## Hoeffding's Inequality – I

Let  $X_1, X_2, \dots, X_n$  be independent random variables taking values such that  $\Pr(X_i = 1) = 1/2 = \Pr(X_i = -1)$  for  $i = 1, 2, \dots, n$ . Let  $X = X_1 + X_2 + \dots + X_n$ . Then for any  $t \geq 0$

$$\Pr(|X| \geq t) < 2e^{-t^2/2n}.$$

**Proof:** For any  $\lambda > 0$  we have

$$\begin{aligned}\Pr(X \geq t) &= \Pr(e^{\lambda X} \geq e^{\lambda t}) \\ &\leq e^{-\lambda t} \mathbf{E}(e^{\lambda X}).\end{aligned}$$

Now for  $i = 1, 2, \dots, n$  we have

$$\mathbf{E}(e^{\lambda X_i}) = \frac{e^{-\lambda} + e^{\lambda}}{2} = 1 + \frac{\lambda^2}{2!} + \frac{\lambda^4}{4!} + \dots < e^{\lambda^2/2}.$$

So, by independence,

$$\mathbf{E}(e^{\lambda X}) = \mathbf{E}\left(\prod_{i=1}^n e^{\lambda X_i}\right) = \prod_{i=1}^n \mathbf{E}(e^{\lambda X_i}) \leq e^{\lambda^2 n/2}.$$

Hence,

$$\Pr(X \geq t) \leq e^{-\lambda t + \lambda^2 n/2}.$$

We choose  $\lambda = t/n$  to minimise  $-\lambda t + \lambda^2 n/2$ . This yields

$$\Pr(X \geq t) \leq e^{-t^2/2n}.$$

Similarly,

$$\begin{aligned} \Pr(X \leq -t) &= \Pr(e^{-\lambda X} \geq e^{\lambda t}) \\ &\leq e^{-\lambda t} \mathbf{E}(e^{-\lambda X}) \\ &\leq e^{-\lambda t + \lambda^2 n/2}. \end{aligned}$$

# Discrepancy

Suppose that  $|X| = n$  and  $\mathcal{F} \subseteq \mathcal{P}(X)$ . If we color the elements of  $X$  with Red and Blue i.e. partition  $X$  in  $R \cup B$  then the **discrepancy**  $disc(\mathcal{F}, R, B)$  of this coloring is defined

$$disc(\mathcal{F}, R, B) = \max_{F \in \mathcal{F}} disc(F, R, B)$$

where  $disc(F, R, B) = ||R \cap F| - |B \cap F||$  i.e. the absolute difference between the number of elements of  $F$  that are colored Red and the number that are colored Blue.

# Claim:

If  $|\mathcal{F}| = m$  then there exists a coloring  $R, B$  such that  $\text{disc}(\mathcal{F}, R, B) \leq (2n \log_e(2m))^{1/2}$ .

**Proof** Fix  $F \in \mathcal{F}$  and let  $s = |F|$ . If we color  $X$  randomly and let  $Z = |R \cap F| - |B \cap F|$  then  $Z$  is the sum of  $s$  independent  $\pm 1$  random variables.

So, by the Hoeffding inequality,

$$\Pr(|Z| \geq (2n \log_e(2m))^{1/2}) < 2e^{-n \log_e(2m)/s} \leq \frac{1}{m}.$$



# 9/24/2021

## The Local Lemma

We go back to the coloring problem at the beginning of these slides. We now place a different restriction on the sets involved.

### Theorem

*Let  $A_1, A_2, \dots, A_n$  be subsets of  $A$  where  $|A_i| \geq k$  for  $1 \leq i \leq n$ . If each  $A_i$  intersects at most  $2^{k-3}$  other sets then there exists a partition  $A = R \cup B$  such that*

$$A_i \cap R \neq \emptyset \text{ and } A_i \cap B \neq \emptyset \quad 1 \leq i \leq n.$$

**Symmetric Local Lemma:** We consider the following situation.

$X = \{x_1, x_2, \dots, x_N\}$  is a collection of independent random variables. Suppose that we have events  $\mathcal{E}_i, i = 1, 2, \dots, m$  where  $\mathcal{E}_i$  depends only on the set  $X_i \subseteq X$ . Thus if  $X_i \cap X_j = \emptyset$  then  $\mathcal{E}_i$  and  $\mathcal{E}_j$  are independent.

The **dependency graph**  $\Gamma$  has vertex set  $[m]$  and an edge  $(i, j)$  iff  $X_i \cap X_j \neq \emptyset$ .

## Theorem

Let

$p = \max_i \Pr(\mathcal{E}_i)$  and let  $d$  be the maximum degree of  $\Gamma$ .

$4dp \leq 1$  implies that  $\Pr\left(\bigcap_{i=1}^m \bar{\mathcal{E}}_i\right) \geq (1 - 2p)^m > 0$ .

**Proof of Theorem 14:** We randomly color the elements of  $A$  Red and Blue. Let  $\mathcal{E}_j$  be the event that  $A_j$  is mono-colored. Clearly,  $\Pr(\mathcal{E}_j) \leq 2^{-(k-1)}$ . Thus,

$$p \leq 2^{-(k-1)}.$$

The degree of vertex  $i$  of  $\Gamma$  is the number of  $j$  such that  $A_i \cap A_j \neq \emptyset$ . So, by assumption,

$$d \leq 2^{k-3}.$$

Theorem 15 implies that  $\Pr(\bigcap_{i=1}^n \bar{\mathcal{E}}_i) > 0$  and so the required coloring exists.

## Theorem

Let  $G = (V, E)$  be an  $r$ -regular graph. If  $r$  is sufficiently large, then  $E$  can be partitioned into  $E_1, E_2$  so that if  $G_i = (V, E_i), i = 1, 2$  then

$$\frac{r}{2} - (20r \log r)^{1/2} \leq \delta(G_i) \leq \Delta(G_i) \leq \frac{r}{2} + (20r \log r)^{1/2}.$$

**Proof:** We randomly partition the edges of  $G$  by independently placing  $e$  into  $E_1$  or  $E_2$  with probability  $1/2$ . For  $v \in V$ , we let  $\mathcal{E}_v$  be the event that the degree  $d_1(v)$  in  $G_1$  satisfies

$$d_1(v) \notin \left[ \frac{r}{2} - (3r \log r)^{1/2}, \frac{r}{2} + (3r \log r)^{1/2} \right].$$

It follows from Hoeffding's Inequality - I with  $t = (3r \log r)^{1/2}$  that

$$\Pr(\mathcal{E}_v) \leq 2e^{-t^2/2r} = 2r^{-3/2}. \quad (7)$$

Furthermore,  $\mathcal{E}_v$  is independent of the events  $\mathcal{E}_w$  for vertices  $w$  at distance 2 or more from  $v$  in  $G$ . Thus,

$$d \leq r.$$

Clearly,  $4 \cdot 2r^{-3/2} \cdot r \leq 1$  for  $r$  large and the result follows from Theorem 15. I.e.  $\Pr(\bigcap_{v \in V} \bar{\mathcal{E}}_v) > 0$  which implies that there exists a partition where none of the events  $\mathcal{E}_v, v \in V$  occur.

# 9/27/2021

For the next application, let  $D = (V, E)$  be a  $k$ -regular digraph. By this we mean that each vertex has exactly  $k$  in-neighbors and  $k$  out-neighbors.

## Theorem

*Every  $k$ -regular digraph has a collection of  $\lfloor k/(4 \log k) \rfloor$  vertex disjoint cycles.*

**Proof:** Let  $r = \lfloor k/(4 \log k) \rfloor$  and color the vertices of  $D$  with colors  $[r]$ . For  $v \in V$ , let  $\mathcal{E}_v$  be the event that there is a color missing at the out-neighbors of  $v$ . We will show that

$\Pr(\bigcap_{v \in V} \bar{\mathcal{E}}_v) > 0$ .

Suppose then that none of the events  $\mathcal{E}_v, v \in V$  occur. Consider the graph  $D_j$  induced by a single color  $j \in [r]$ . Note that  $D_j$  is not the empty graph. Let  $P_j = (v_1, v_2, \dots, v_m)$  be a longest directed path in  $D_j$ . Let  $w$  be an out-neighbor of  $v_m$  of color  $j$ . We must have  $w \in \{v_1, \dots, v_m\}$ , else  $P_j$  is not a longest path in  $D_j$ . Thus each  $D_j, j \in [r]$  contains a cycle and these cycles are vertex disjoint.



We first estimate

$$\Pr(\mathcal{E}_v) \leq k \left(1 - \frac{1}{r}\right)^k \leq ke^{-k/r} \leq ke^{-4 \log k} = k^{-3}.$$

On the other hand, if  $N^+(v)$  denotes the out-neighbors of  $v$  plus  $v$  then  $\mathcal{E}_v$  is independent of all events  $\mathcal{E}_w$  for which  $N^+(v) \cap N^+(w) = \emptyset$ . It follows that

$$d \leq k^2.$$

To apply Theorem 15 we need to have  $4k^{-3}k^2 \leq 1$ . This is true for  $k \geq 4$ . For  $k \leq 3$  we have  $r = 1$  and the local lemma is not needed.

# Extremal Problems

Let  $\mathcal{P}_n = \{A : A \subseteq [n]\}$  denote the *power set* of  $[n]$ .

$\mathcal{A} \subseteq \mathcal{P}_n$  is a *Sperner family* if  $A, B \in \mathcal{A}$  implies that  $A \not\subseteq B$  and  $B \not\subseteq A$

## Theorem

If  $\mathcal{A} \subseteq \mathcal{P}_n$  is a Sperner family  $|\mathcal{A}| \leq \binom{n}{\lfloor n/2 \rfloor}$ .

**Proof** We will show that

$$\sum_{A \in \mathcal{A}} \frac{1}{\binom{n}{|A|}} \leq 1. \quad (8)$$

Now  $\binom{n}{k} \leq \binom{n}{\lfloor n/2 \rfloor}$  for all  $k$  and so

$$1 \geq \sum_{A \in \mathcal{A}} \frac{1}{\binom{n}{\lfloor n/2 \rfloor}} = \frac{|\mathcal{A}|}{\binom{n}{\lfloor n/2 \rfloor}}.$$

**Proof of (8):** Let  $\pi$  be a random permutation of  $[n]$ .

For a set  $A \in \mathcal{A}$  let  $\mathcal{E}_A$  be the event

$$\{\pi(1), \pi(2), \dots, \pi(|A|)\} = A.$$

If  $A, B \in \mathcal{A}$  then the events  $\mathcal{E}_A, \mathcal{E}_B$  are disjoint.

So

$$\sum_{A \in \mathcal{A}} \Pr(\mathcal{E}_A) \leq 1.$$

On the other hand, if  $A \in \mathcal{A}$  then

$$\Pr(\mathcal{E}_A) = \frac{|A|!(n - |A|)!}{n!} = \frac{1}{\binom{n}{|A|}}$$

and (18) follows.

The set of all sets of size  $\lfloor n/2 \rfloor$  is a Sperner family and so the bound in the above theorem is best possible.

Inequality (8) can be generalised as follows: Let  $s \geq 1$  be fixed. Let  $\mathcal{A}$  be a family of subsets of  $[n]$  such that **there do not exist** distinct  $A_1, A_2, \dots, A_{s+1} \in \mathcal{A}$  such that  $A_1 \subseteq A_2 \subseteq \dots \subseteq A_{s+1}$ .

### Theorem

$$\sum_{A \in \mathcal{A}} \frac{1}{\binom{n}{|A|}} \leq s.$$

**Proof** Let  $\pi$  be a random permutation of  $[n]$ .

Let  $\mathcal{E}(A)$  be the event  $\{\pi(1), \pi(2), \dots, \pi(|A|) = A\}$ .

Let

$$Z_i = \begin{cases} 1 & \mathcal{E}(A_i) \text{ occurs.} \\ 0 & \text{otherwise.} \end{cases}$$

and let  $Z = \sum_i Z_i$  be the number of events  $\mathcal{E}(A_i)$  that occur.

Now our family is such that  $Z \leq s$  for all  $\pi$  and so

$$E(Z) = \sum_i E(Z_i) = \sum_i \Pr(\mathcal{E}(A_i)) \leq s.$$

On the other hand,  $A \in \mathcal{A}$  implies that  $\Pr(\mathcal{E}(A)) = \frac{1}{\binom{n}{|A|}}$  and the required inequality follows.  $\square$

# 9/29/2021

# Extremal Problems

**Intersecting Families** A family  $\mathcal{A} \subseteq \mathcal{P}_n$  is an *intersecting* family if  $A, B \in \mathcal{A}$  implies  $A \cap B \neq \emptyset$ .

## Theorem

If  $\mathcal{A}$  is an intersecting family then  $|\mathcal{A}| \leq 2^{n-1}$ .

**Proof** Pair up each  $A \in \mathcal{P}_n$  with its complement  $A^c = [n] \setminus A$ . This gives us  $2^{n-1}$  pairs altogether. Since  $\mathcal{A}$  is intersecting it can contain at most one member of each pair. □

If  $\mathcal{A} = \{A \subseteq [n] : 1 \in A\}$  then  $\mathcal{A}$  is intersecting and  $|\mathcal{A}| = 2^{n-1}$  and so the above theorem is best possible.

## Theorem

If  $\mathcal{A}$  is an intersecting family and  $A \in \mathcal{A}$  implies that  $|A| = k \leq \lfloor n/2 \rfloor$  then

$$|\mathcal{A}| \leq \binom{n-1}{k-1}$$

**Proof** If  $\pi$  is a permutation of  $[n]$  and  $A \subseteq [n]$  let

$$\theta(\pi, A) = \begin{cases} 1 & \exists s : \{\pi(s), \pi(s+1), \dots, \pi(s+k-1)\} = A \\ 0 & \text{otherwise} \end{cases}$$

where  $\pi(i) = \pi(i-n)$  if  $i > n$ .

We will show that for any permutation  $\pi$ ,

$$\sum_{A \in \mathcal{A}} \theta(\pi, A) \leq k. \tag{9}$$



Assume (9). We first observe that if  $\pi$  is a random permutation then

$$\mathbf{E}(\theta(\pi, A)) = n \frac{k!(n-k)!}{n!} = \frac{k}{\binom{n-1}{k-1}}$$

and so, from (9),

$$k \geq \mathbf{E}\left(\sum_{A \in \mathcal{A}} \theta(\pi, A)\right) = \sum_{A \in \mathcal{A}} \frac{k}{\binom{n-1}{|A|-1}}$$

Hence

$$|\mathcal{A}| \leq \binom{n-1}{k-1}$$

Assume w.l.o.g. that  $\pi$  is the identity permutation.

Let  $A_t = \{t, t + 1, \dots, t + k - 1\}$  and suppose that  $A_s \in \mathcal{A}$ .

All of the other sets  $A_t$  that intersect  $A_s$  can be partitioned into pairs  $A_{s-i}, A_{s+k-i}$ ,  $1 \leq i \leq k - 1$  and the members of each pair are disjoint. Thus  $\mathcal{A}$  can contain at most one from each pair. This verifies (9).

## Kraft's Inequality

Let  $x_1, x_2, \dots, x_m$  be a collection of sequences over an alphabet  $\Sigma$  of size  $s$ . Let  $x_j$  have length  $n_j$  and let  $n = \max\{n_1, n_2, \dots, n_m\}$ .

Assume next that no sequence is a prefix of any other sequence: Sequence  $x_j = a_1 a_2 \cdots a_{n_j}$  is a prefix of  $x_i = b_1 b_2 \cdots b_{n_i}$  if  $a_i = b_i$  for  $i = 1, 2, \dots, n_j$ .

### Theorem

$$\sum_{i=1}^m r^{-n_i} \leq 1.$$

**Proof:** Let  $x$  be a random sequence of length  $n$ . Let  $\mathcal{E}_i$  be the event  $x_i$  is a prefix of  $x$ . Then

(a)  $\Pr(\mathcal{E}_i) = r^{-n_i}$ .

(b) The event  $\mathcal{E}_i, i = 1, 2, \dots, m$  are disjoint.

(If  $\mathcal{E}_i$  and  $\mathcal{E}_j$  both occur and  $n_i \leq n_j$  then  $x_i$  is a prefix of  $x_j$ .)

Property (b) implies that

$$\Pr\left(\bigcup_{i=1}^m \mathcal{E}_i\right) = \Pr(\mathcal{E}_1) + \Pr(\mathcal{E}_2) + \dots + \Pr(\mathcal{E}_m) \leq 1.$$

The theorem now follows from Property (a). □

## Sunflowers

A **sunflower** of size  $r$  is a family of sets  $A_1, A_2, \dots, A_r$  such that every element that belongs to more than one of the sets belongs to all of them.

Let  $f(k, r)$  be the maximum size of a family of  $k$ -sets without a sunflower of size  $r$ .

### Theorem

$$f(k, r) \leq (r-1)^k k!.$$

**Proof** Let  $\mathcal{F}$  be a family of  $k$ -sets without a sunflower of size  $r$ . Let  $A_1, A_2, \dots, A_t$  be a maximum subfamily of pairwise disjoint subsets in  $\mathcal{F}$ .

Since a family of pairwise disjoint is a sunflower, we must have  $t < r$ .

Now let  $A = \bigcup_{i=1}^t A_i$ . For every  $a \in A$  consider the family  $\mathcal{F}_a = \{S \setminus \{a\} : S \in \mathcal{F}, a \in S\}$ .

Now the size of  $A$  is at most  $(r-1)k$ .

The size of each  $\mathcal{F}_a$  is at most  $f(k-1, r)$ . This is because a sunflower in  $\mathcal{F}_a$  is a sunflower in  $\mathcal{F}$ .

So,

$$f(k, r) \leq (r-1)k \times f(k-1, r) \leq (r-1)k \times (r-1)^{k-1} (k-1)!,$$

by induction. □

# 10/1/2021

# Linear Algebraic Methods

**Odd Town** In order to cut down the number of committees a town of  $n$  people has instituted the following rules:

- (a) Each club shall have an odd number of members.
- (b) Each pair of clubs shall share an even number of members.

## Theorem

*With these rules, there are at most  $n$  clubs.*



# Linear Algebraic Methods

**Proof** Suppose that the clubs are  $C_1, C_2, \dots, C_m \subseteq [n]$ .

Let  $\bar{v}_i = (v_{i,1}, v_{i,2}, \dots, v_{i,n})$  denote the incidence vector of  $C_i$  for  $1 \leq i \leq m$  i.e.  $v_{i,j} = 1$  iff  $j \in C_i$ . We treat these vectors as being over the two element field  $\mathbb{F}_2$ .

We claim that  $\bar{v}_1, \bar{v}_2, \dots, \bar{v}_m$  are linearly independent and the theorem will follow.

The rules imply that (i)  $\bar{v}_i \cdot \bar{v}_i = 1$  and (ii)  $\bar{v}_i \cdot \bar{v}_j = 0$  for  $1 \leq i \neq j \leq m$ .

(Remember that we are working over  $\mathbb{F}_2$ .)

Suppose then that

$$c_1 \bar{v}_1 + c_2 \bar{v}_2 + \cdots + c_m \bar{v}_m = 0.$$

We show that  $c_1 = c_2 = \cdots = c_m = 0$ .

Indeed, we have

$$\begin{aligned} 0 &= \bar{v}_j \cdot (c_1 \bar{v}_1 + c_2 \bar{v}_2 + \cdots + c_m \bar{v}_m) \\ &= c_1 \bar{v}_1 \cdot \bar{v}_j + c_2 \bar{v}_2 \cdot \bar{v}_j + \cdots + c_m \bar{v}_m \cdot \bar{v}_j \\ &= c_j, \end{aligned}$$

for  $j = 1, 2, \dots, m$ . □

**Decomposing  $K_n$  into bipartite subgraphs:** here we show

## Theorem

If  $G_k, k = 1, 2, \dots, m$  is a collection of complete bipartite graphs with vertex partitions  $A_k, B_k$ , such that every edge of  $K_n$  is in exactly one subgraph, then  $m \geq n - 1$ . (Note that  $A_k \cap B_k = \emptyset$  here.)

**Proof** This is tight since we can take  $A_k = \{k\}, B_k = \{k + 1, \dots, n\}$  for  $k = 1, 2, \dots, n - 1$ .

Define  $n \times n$  matrices  $M_k$  where  $M_k(i, j) = 1$  if  $i \in A_k, j \in B_k$  and  $M_k(i, j) = 0$  otherwise.

Let  $S = M_1 + M_2 + \dots + M_m$ . Then  $S + S^T = J_n - I_n$  where  $I_n$  is the identity matrix and  $J_n$  is the all ones matrix.

# Linear Algebraic Methods

We show next that  $\text{rank}(\mathbf{S}) \geq n - 1$  and then the theorem follows from

$$\text{rank}(\mathbf{S}) \leq \text{rank}(\mathbf{M}_1) + \text{rank}(\mathbf{M}_2) + \cdots + \text{rank}(\mathbf{M}_m) \leq m.$$

Suppose then that  $\text{rank}(\mathbf{S}) \leq n - 2$  so that there exists a non-zero solution  $\mathbf{x} = (x_1, x_2, \dots, x_n)^T$  to the system of equations

$$\mathbf{S}\mathbf{x} = \mathbf{0}, \quad \sum_{i=1}^n x_i = 0.$$

But then,  $\mathbf{J}_n \mathbf{x} = \mathbf{0}$  and  $\mathbf{S}^T \mathbf{x} = -\mathbf{x}$  and  $-\|\mathbf{x}\|^2 = -\mathbf{x}^T \mathbf{S}^T \mathbf{x} = 0$ , contradiction. □

## Nonuniform Fisher Inequality;

### Theorem

Let  $C_1, C_2, \dots, C_m$  be distinct subsets of  $[n]$  such that for every  $i \neq j$  we have  $|C_i \cap C_j| = s$  where  $1 \leq s < n$ . Then  $m \leq n$ .

**Proof** If  $|C_1| = s$  then  $C_i \supset C_1, i = 2, 3, \dots, m$  and the sets  $C_i \setminus C_1$  are pairwise disjoint for  $i \geq 2$ .

It follows in this case that  $m \leq 1 + n - s \leq n$ .

Assume from now on that  $c_i = |C_i| - s > 0$  for  $i \in [m]$ .

# Linear Algebraic Methods

Let  $M$  be the  $m \times n$  0/1 matrix where  $M(i, j) = 1$  iff  $j \in C_i$ .

Let

$$A = MM^T = sJ + D$$

where  $J$  is the  $m \times m$  all 1's matrix and  $D$  is the diagonal matrix, where  $D(i, i) = c_i$ .

We show that  $A$  and hence  $M$  has rank  $m$ , implying that  $m \leq n$  as claimed.

We will in fact show that  $\mathbf{x}^T A \mathbf{x} > 0$  for all  $0 \neq \mathbf{x} \in \mathbb{R}^m$ . This means that  $A \mathbf{x} \neq 0$  when  $\mathbf{x} \neq 0$ .

# Linear Algebraic Methods

If  $\mathbf{x} = (x_1, x_2, \dots, x_m)^T$  then

$$\mathbf{x}^T \mathbf{A} \mathbf{x} = s(x_1 + x_2 + \dots + x_m)^2 + \sum_{i=1}^m c_i x_i^2 > 0.$$



# 10/4/2021



# Pigeon Hole Principle

We have two disks, each partitioned into 200 sectors of the same size. 100 of the sectors of Disk 1 are coloured Red and 100 are colored Blue. The 200 sectors of Disk 2 are arbitrarily coloured Red and Blue.

It is always possible to place Disk 2 on top of Disk 1 so that the centres coincide, the sectors line up and at least 100 sectors of Disk 2 have the same colour as the sector underneath them.

Fix the position of Disk 1. There are 200 positions for Disk 2 and let  $q_i$  denote the number of matches if Disk 2 is placed in position  $i$ . Now for each sector of Disk 2 there are 100 positions  $i$  in which the colour of the sector underneath it coincides with its own.

# Pigeon Hole Principle

Therefore

$$q_1 + q_2 + \cdots + q_{200} = 200 \times 100 \quad (10)$$

and so there is an  $i$  such that  $q_i \geq 100$ .

Explanation of (19).

Consider 0-1  $200 \times 200$  matrix  $A(i, j)$  where  $A(i, j) = 1$  iff sector  $j$  lies on top of a sector with the same colour when in position  $i$ . Row  $i$  of  $A$  has  $q_i$  1's and column  $j$  of  $A$  has 100 1's. The LHS of (19) counts the number of 1's by adding rows and the RHS counts the number of 1's by adding columns.

# Pigeon Hole Principle

Alternative solution: Place Disk 2 randomly on Disk 1 so that the sectors align. For  $i = 1, 2, \dots, 200$  let

$$X_i = \begin{cases} 1 & \text{sector } i \text{ of disk 2 is on sector of disk 1 of same color} \\ 0 & \text{otherwise} \end{cases}$$

We have

$$\mathbf{E}(X_i) = 1/2 \quad \text{for } i = 1, 2, \dots, 200.$$

So if  $X = X_1 + \dots + X_{200}$  is the number of sectors sitting above sectors of the same color, then  $\mathbf{E}(X) = 100$  and there must exist at least one way to achieve 100.

# Pigeon Hole Principle

## Theorem

(Erdős-Szekeres) *An arbitrary sequence of integers  $(a_1, a_2, \dots, a_{k^2+1})$  contains a monotone subsequence of length  $k + 1$ .*

**Proof.** Let  $(a_i, a_i^1, a_i^2, \dots, a_i^{\ell-1})$  be the longest *monotone increasing* subsequence of  $(a_1, \dots, a_{k^2+1})$  that starts with  $a_i$ ,  $(1 \leq i \leq k^2 + 1)$ , and let  $\ell(a_i)$  be its length.

If for some  $1 \leq i \leq k^2 + 1$ ,  $\ell(a_i) \geq k + 1$ , then  $(a_i, a_i^1, a_i^2, \dots, a_i^{\ell-1})$  is a monotone increasing subsequence of length  $\geq k + 1$ .

So assume that  $\ell(a_i) \leq k$  holds for every  $1 \leq i \leq k^2 + 1$ .

# Pigeon Hole Principle

Consider  $k$  holes  $1, 2, \dots, k$  and place  $i$  into hole  $\ell(a_i)$ .

There are  $k^2 + 1$  subsequences and  $\leq k$  non-empty holes (different lengths), so by the pigeon-hole principle there will exist  $\ell^*$  such that there are (at least)  $k + 1$  indices  $i_1 < i_2 < \dots < i_{k+1}$  such that  $\ell(a_{i_t}) = \ell^*$  for  $1 \leq t \leq k + 1$ .

Then we must have  $a_{i_1} \geq a_{i_2} \geq \dots \geq a_{i_{k+1}}$ .

Indeed, assume to the contrary that  $a_{i_m} < a_{i_n}$  for some  $1 \leq m < n \leq k + 1$ . Then  $a_{i_m} \leq a_{i_n} \leq a_{i_n}^1 \leq a_{i_n}^2 \leq \dots \leq a_{i_n}^{\ell^* - 1}$ , i.e.,  $\ell(a_{i_m}) \geq \ell^* + 1$ , a contradiction.  $\square$

# Pigeon Hole Principle

The sequence

$$n, n-1, \dots, 1, 2n, 2n-1, \dots, n+1, \dots, n^2, n^2-1, \dots, n^2-n+1$$

has no monotone subsequence of length  $n+1$  and so the Erdős-Szekerés result is best possible.

# Pigeon Hole Principle

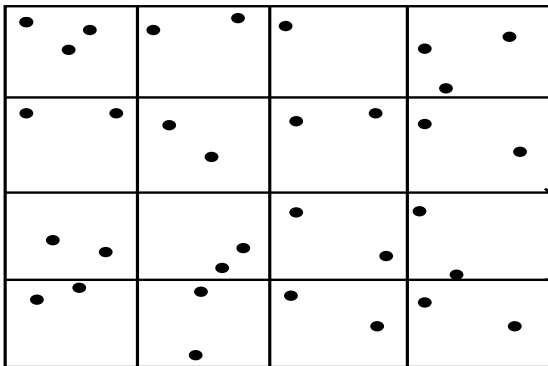
Let  $P_1, P_2, \dots, P_n$  be  $n$  points in the unit square  $[0, 1]^2$ . We will show that there exist  $i, j, k \in [n]$  such that the triangle  $P_i P_j P_k$  has area

$$\leq \frac{1}{2(\lfloor \sqrt{(n-1)/2} \rfloor)^2} \sim \frac{1}{n}$$

for large  $n$ .

# Pigeon Hole Principle

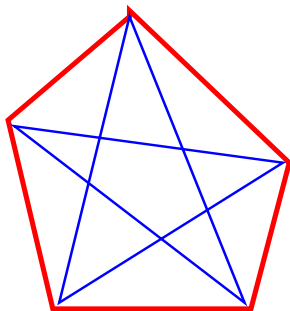
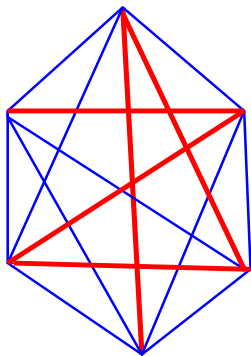
Let  $m = \lfloor \sqrt{(n-1)/2} \rfloor$  and divide the square up into  $m^2 < \frac{n}{2}$  subsquares. By the pigeonhole principle, there must be a square containing  $\geq 3$  points. Let 3 of these points be  $P_i P_j P_k$ . The area of the corresponding triangle is at most one half of the area of an individual square.





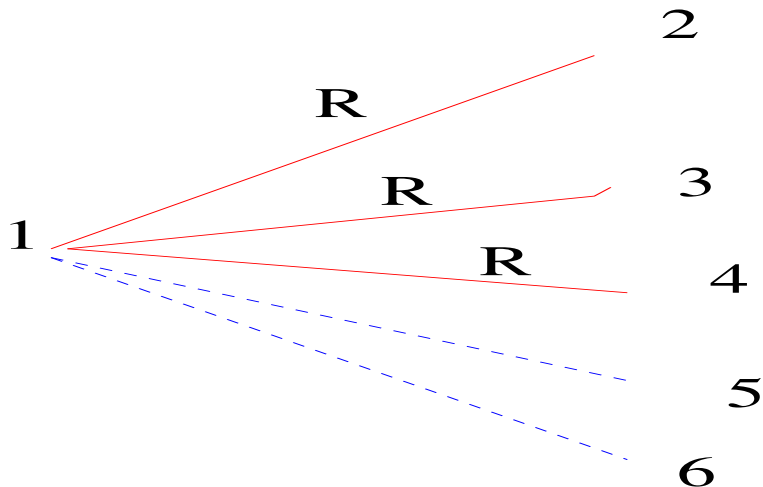
# Ramsey Theory

Suppose we 2-colour the edges of  $K_6$  of Red and Blue. There *must* be either a Red triangle or a Blue triangle.



This is not true for  $K_5$ .

# Ramsey Theory



There are 3 edges of the same colour incident with vertex 1, say  $(1,2)$ ,  $(1,3)$ ,  $(1,4)$  are Red. Either  $(2,3,4)$  is a blue triangle or one of the edges of  $(2,3,4)$  is Red, say  $(2,3)$ . But the latter

## Ramsey's Theorem

For all positive integers  $k, \ell$  there exists  $R(k, \ell)$  such that if  $N \geq R(k, \ell)$  and the edges of  $K_N$  are coloured Red or Blue then then either there is a “Red  $k$ -clique” or there is a “Blue  $\ell$ -clique. A clique is a complete subgraph and it is Red if all of its edges are coloured red etc.

$$\begin{aligned}R(1, k) &= R(k, 1) = 1 \\R(2, k) &= R(k, 2) = k\end{aligned}$$

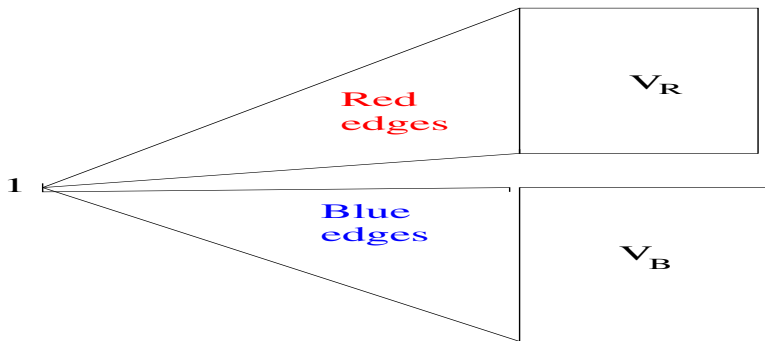
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# Ramsey Theory

## Theorem

$$R(k, l) \leq R(k, l - 1) + R(k - 1, l).$$

**Proof** Let  $N = R(k, l - 1) + R(k - 1, l)$ .



$V_R = \{x : (1, x) \text{ is coloured Red}\}$  and  $V_B = \{x : (1, x) \text{ is coloured Blue}\}$

# Ramsey Theory

$$|V_R| \geq R(k-1, \ell) \text{ or } |V_B| \geq R(k, \ell-1).$$

Since

$$\begin{aligned} |V_R| + |V_B| &= N - 1 \\ &= R(k, \ell - 1) + R(k - 1, \ell) - 1. \end{aligned}$$

Suppose for example that  $|V_R| \geq R(k-1, \ell)$ . Then either  $V_R$  contains a Blue  $\ell$ -clique – done, or it contains a Red  $k-1$ -clique  $K$ . But then  $K \cup \{1\}$  is a Red  $k$ -clique. Similarly, if  $|V_B| \geq R(k, \ell-1)$  then either  $V_B$  contains a Red  $k$ -clique – done, or it contains a Blue  $\ell-1$ -clique  $L$  and then  $L \cup \{1\}$  is a Blue  $\ell$ -clique. □

## Theorem

$$R(k, l) \leq \binom{k+l-2}{k-1}.$$

**Proof** Induction on  $k+l$ . True for  $k+l \leq 5$  say. Then

$$\begin{aligned} R(k, l) &\leq R(k, l-1) + R(k-1, l) \\ &\leq \binom{k+l-3}{k-1} + \binom{k+l-3}{k-2} \\ &= \binom{k+l-2}{k-1}. \end{aligned}$$



So, for example,

$$R(k, k) \leq \binom{2k-2}{k-1}$$

## Theorem

$$R(k, k) > 2^{k/2}$$

**Proof** We must prove that if  $n \leq 2^{k/2}$  then there exists a Red-Blue colouring of the edges of  $K_n$  which contains no Red  $k$ -clique and no Blue  $k$ -clique. We can assume  $k \geq 4$  since we know  $R(3, 3) = 6$ .

We show that this is true with positive probability in a *random* Red-Blue colouring. So let  $\Omega$  be the set of all Red-Blue edge colourings of  $K_n$  with uniform distribution. Equivalently we independently colour each edge Red with probability  $1/2$  and Blue with probability  $1/2$ .



# Ramsey Theory

Let

$\mathcal{E}_R$  be the event: {There is a Red  $k$ -clique} and

$\mathcal{E}_B$  be the event: {There is a Blue  $k$ -clique}.

We show

$$\Pr(\mathcal{E}_R \cup \mathcal{E}_B) < 1.$$

Let  $C_1, C_2, \dots, C_N$ ,  $N = \binom{n}{k}$  be the vertices of the  $N$   $k$ -cliques of  $K_n$ .

Let  $\mathcal{E}_{R,j}$  be the event:  $\{C_j \text{ is Red}\}$  and let  $\mathcal{E}_{B,j}$  be the event:  $\{C_j \text{ is Blue}\}$ .

$$\begin{aligned}\Pr(\mathcal{E}_R \cup \mathcal{E}_B) &\leq \Pr(\mathcal{E}_R) + \Pr(\mathcal{E}_B) = 2\Pr(\mathcal{E}_R) \\ &= 2\Pr\left(\bigcup_{j=1}^N \mathcal{E}_{R,j}\right) \leq 2\sum_{j=1}^N \Pr(\mathcal{E}_{R,j}) \\ &= 2\sum_{j=1}^N \left(\frac{1}{2}\right)^{\binom{k}{2}} = 2\binom{n}{k} \left(\frac{1}{2}\right)^{\binom{k}{2}} \\ &\leq 2\frac{n^k}{k!} \left(\frac{1}{2}\right)^{\binom{k}{2}} \\ &\leq 2\frac{2^{k^2/2}}{k!} \left(\frac{1}{2}\right)^{\binom{k}{2}} \\ &= \frac{2^{1+k/2}}{k!} \\ &< 1.\end{aligned}$$

# Ramsey Theory

Very few of the Ramsey numbers are known exactly. Here are a few known values.

$$R(3, 3) = 6$$

$$R(3, 4) = 9$$

$$R(4, 4) = 18$$

$$R(4, 5) = 25$$

$$43 \leq R(5, 5) \leq 49$$

# Ramsey Theory

## Schur's Theorem

Let  $r_k = N(3, 3, \dots, 3; 2)$  be the smallest  $n$  such that if we  $k$ -color the edges of  $K_n$  then there is a mono-chromatic triangle.

### Theorem

*For all partitions  $S_1, S_2, \dots, S_k$  of  $[r_k]$ , there exist  $i$  and  $x, y, z \in S_i$  such that  $x + y = z$ .*

**Proof** Given a partition  $S_1, S_2, \dots, S_k$  of  $[n]$  where  $n \geq r_k$  we define a coloring of the edges of  $K_n$  by coloring  $(u, v)$  with color  $j$  where  $|u - v| \in S_j$ .

There will be a mono-chromatic triangle i.e. there exist  $j$  and  $x < y < z$  such that  $u = y - x, v = z - x, w = z - y \in S_j$ .  
But  $u + v = w$ .



# Ramsey Theory

A set of points  $X$  in the plane is in **general position** if no 3 points of  $X$  are collinear.

## Theorem

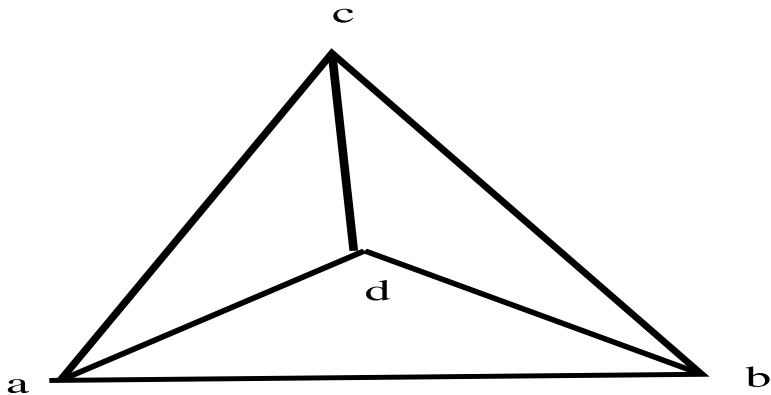
*If  $n \geq N(k, k; 3)$  and  $X$  is a set of  $n$  points in the plane which are in general position then  $X$  contains a  $k$ -subset  $Y$  which form the vertices of a convex polygon.*

**Proof** We first observe that if **every** 4-subset of  $Y \subseteq X$  forms a convex quadrilateral then  $Y$  itself induces a convex polygon.

Now label the points in  $S$  from  $X_1$  to  $X_n$  and then color each triangle  $T = \{X_i, X_j, X_k\}$ ,  $i < j < k$  as follows: If traversing triangle  $X_i X_j X_k$  in this order goes round it clockwise, color  $T$  Red, otherwise color  $T$  Blue.

# Ramsey Theory

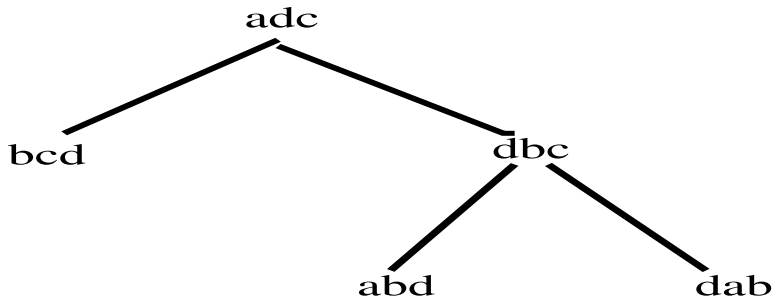
Now there must exist a  $k$ -set  $T$  such that all triangles formed from  $T$  have the same color. All we have to show is that  $T$  does not contain the following configuration:



Assume w.l.o.g. that  $a < b < c$  which implies that  $X_i X_j X_k$  is colored Blue.

All triangles in the previous picture are colored Blue.

So the possibilities are



and all are impossible.

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# Ramsey Theory

We define  $r(H_1, H_2)$  to be the minimum  $n$  such that in in Red-Blue coloring of the edges of  $K_n$  there is either (i) a Red copy of  $H_1$  or (ii) a Blue copy of  $H_2$ .

As an example, consider  $r(P_3, P_3)$  where  $P_t$  denotes a path with  $t$  edges.

We show that

$$r(P_3, P_3) = 5.$$

$R(P_3, P_3) > 4$ : We color edges incident with 1 Red and the remaining edges  $\{(2, 3), (3, 4), (4, 1)\}$  Blue. There is no mono-chromatic  $P_3$ .

# Ramsey Theory

$R(P_3, P_3) \leq 5$ : There must be two edges of the same color incident with 1.

Assume then that  $(1, 2), (1, 3)$  are both Red.

If any of  $(2, 4), (2, 5), (3, 4), (3, 5)$  are Red then we have a Red  $P_3$ .

If all four of these edges are Blue then  $(4, 2, 5, 3)$  is Blue.

# Ramsey Theory

We show next that  $r(K_{1,s}, P_t) \leq s + t$ . Here  $K_{1,s}$  is a **star**: i.e. a vertex  $v$  and  $t$  incident edges.

Let  $n = s + t$ . If there is no vertex of Red degree  $s$  then the minimum degree in the graph induced by the Blue edges is at least  $t$ .

We then note that a graph of minimum degree  $\delta$  contains a path of length  $\delta$ .

# Partially Ordered Sets

A **partially ordered set** or **poset** is a set  $P$  and a binary relation  $\preceq$  such that for all  $a, b, c \in P$

- 1  $a \preceq a$  (reflexivity).
- 2  $a \preceq b$  and  $b \preceq c$  implies  $a \preceq c$  (transitivity).
- 3  $a \preceq b$  and  $b \preceq a$  implies  $a = b$ . (anti-symmetry).

## Examples

- 1  $P = \{1, 2, \dots\}$  and  $a \leq b$  has the usual meaning.
- 2  $P = \{1, 2, \dots\}$  and  $a \preceq b$  if  $a$  divides  $b$ .
- 3  $P = \{A_1, A_2, \dots, A_m\}$  where the  $A_i$  are sets and  $\preceq = \subseteq$ .

# Partially Ordered Sets

A pair of elements  $a, b$  are **comparable** if  $a \preceq b$  or  $b \preceq a$ .  
Otherwise they are **incomparable**.

A poset without incomparable elements (Example 1) is a linear or total order.

We write  $a < b$  if  $a \preceq b$  and  $a \neq b$ .

A **chain** is a sequence  $a_1 < a_2 < \dots < a_s$ .

A set  $A$  is an **anti-chain** if every pair of elements in  $A$  are incomparable.

Thus a Sperner family is an anti-chain in our third example.

# Partially Ordered Sets

## Theorem

Let  $P$  be a finite poset, then

$$\min\{m : \exists \text{ anti-chains } A_1, A_2, \dots, A_m \text{ with } P = \bigcup_{i=1}^m A_i\} = \max\{|C| : C \text{ is a chain}\}.$$

The minimum number of anti-chains needed to cover  $P$  is at least the size of any chain, since a chain can contain at most one element from each anti-chain.

# Partially Ordered Sets

We prove the converse by induction on the maximum length  $\mu$  of a chain. We have to show that  $P$  can be partitioned into  $\mu$  anti-chains.

If  $\mu = 1$  then  $P$  itself is an anti-chain and this provides the basis of the induction.

So now suppose that  $C = x_1 < x_2 < \dots < x_\mu$  is a maximum length chain and let  $A$  be the set of maximal elements of  $P$ .

(An element is  $x$  maximal if  $\nexists y$  such that  $y > x$ .)

$A$  is an anti-chain.

Now consider  $P' = P \setminus A$ .  $P'$  contains no chain of length  $\mu$ . If it contained  $y_1 < y_2 < \dots < y_\mu$  then since  $y_\mu \notin A$ , there exists  $a \in A$  such that  $P$  contains the chain  $y_1 < y_2 < \dots < y_\mu < a$ , contradiction.

Thus the maximum length of a chain in  $P'$  is  $\mu - 1$  and so it can be partitioned into anti-chains  $A_1 \cup A_2 \cup \dots \cup A_{\mu-1}$ . Putting  $A_\mu = A$  completes the proof.  $\square$



# Partially Ordered Sets

Suppose that  $C_1, C_2, \dots, C_m$  are a collection of chains such that  $P = \bigcup_{i=1}^m C_i$ .

Suppose that  $A$  is an anti-chain. Then  $m \geq |A|$  because if  $m < |A|$  then by the pigeon-hole principle there will be two elements of  $A$  in some chain.

## Theorem

*(Dilworth)* Let  $P$  be a finite poset, then

$$\min\{m : \exists \text{ chains } C_1, C_2, \dots, C_m \text{ with } P = \bigcup_{i=1}^m C_i\} = \max\{|A| : A \text{ is an anti-chain}\}.$$

## Intervals Problem

$I_1, I_2, \dots, I_{mn+1}$  are closed intervals on the real line i.e.  
 $I_j = [a_j, b_j]$  where  $a_j \leq b_j$  for  $1 \leq j \leq mn + 1$ .

### Theorem

*Either (i) there are  $m + 1$  intervals that are pair-wise disjoint or  
(ii) there are  $n + 1$  intervals with a non-empty intersection*

Define a partial ordering  $\preceq$  on the intervals by  $I_r \preceq I_s$  iff  $b_r \leq a_s$ .  
Suppose that  $I_{i_1}, I_{i_2}, \dots, I_{i_t}$  is a collection of pair-wise disjoint intervals. Assume that  $a_{i_1} < a_{i_2} < \dots < a_{i_t}$ . Then  $I_{i_1} < I_{i_2} < \dots < I_{i_t}$  form a chain and conversely a chain of length  $t$  comes from a set of  $t$  pair-wise disjoint intervals.

So if (i) does not hold, then the maximum length of a chain is  $m$ .

# Partially Ordered Sets

This means that the minimum number of chains needed to cover the poset is at least  $\lceil \frac{mn+1}{m} \rceil = n + 1$ .

Dilworth's theorem implies that there must exist an anti-chain  $\{I_{j_1}, I_{j_2}, \dots, I_{j_{n+1}}\}$ .

Let  $a = \max\{a_{j_1}, a_{j_2}, \dots, a_{j_{n+1}}\}$  and  $b = \min\{b_{j_1}, b_{j_2}, \dots, b_{j_{n+1}}\}$ .

We must have  $a \leq b$  else the two intervals giving  $a, b$  are disjoint.

But then every interval of the anti-chain contains  $[a, b]$ .

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Suppose that  $C_1, C_2, \dots, C_m$  are a collection of chains such that  $P = \bigcup_{i=1}^m C_i$ .

Suppose that  $A$  is an anti-chain. Then  $m \geq |A|$  because if  $m < |A|$  then by the pigeon-hole principle there will be two elements of  $A$  in some chain.

### Theorem

*(Dilworth)* Let  $P$  be a finite poset, then  
 $\min\{m : \exists \text{ chains } C_1, C_2, \dots, C_m \text{ with } P = \bigcup_{i=1}^m C_i\} =$   
 $\max\{|A| : A \text{ is an anti-chain}\}.$

We have already argued that  $\max\{|A|\} \leq \min\{m\}$ .

We will prove there is equality here by induction on  $|P|$ .

The result is trivial if  $|P| = 0$ .

Now assume that  $|P| > 0$  and that  $\mu$  is the maximum size of an anti-chain in  $P$ . We show that  $P$  can be partitioned into  $\mu$  chains.

Let  $C = x_1 < x_2 < \dots < x_p$  be a *maximal* chain in  $P$  i.e. we cannot add elements to it and keep it a chain.

**Case 1** Every anti-chain in  $P \setminus C$  has  $\leq \mu - 1$  elements. Then by induction  $P \setminus C = \bigcup_{i=1}^{\mu-1} C_i$  and then  $P = C \cup \bigcup_{i=1}^{\mu-1} C_i$  and we are done.

## Case 2

There exists an anti-chain  $A = \{a_1, a_2, \dots, a_\mu\}$  in  $P \setminus C$ . Let

- $P^- = \{x \in P : x \preceq a_i \text{ for some } i\}$ .
- $P^+ = \{x \in P : x \succeq a_i \text{ for some } i\}$ .

Note that

- 1  $P = P^- \cup P^+$ . Otherwise there is an element  $x$  of  $P$  which is incomparable with every element of  $A$  and so  $\mu$  is not the maximum size of an anti-chain.
- 2  $P^- \cap P^+ = A$ . Otherwise there exists  $x, i, j$  such that  $a_i < x < a_j$  and so  $A$  is not an anti-chain.
- 3  $x_p \notin P^-$ . Otherwise  $x_p < a_i$  for some  $i$  and the chain  $C$  is not maximal.

Applying the inductive hypothesis to  $P^-$  ( $|P^-| < |P|$  follows from 3) we see that  $P^-$  can be partitioned into  $\mu$  chains  $C_1^-, C_2^-, \dots, C_\mu^-$ .

Now the elements of  $A$  must be distributed one to a chain and so we can assume that  $a_i \in C_i^-$  for  $i = 1, 2, \dots, \mu$ .

$a_i$  must be the maximum element of chain  $C_i^-$ , else the maximum of  $C_i^-$  is in  $(P^- \cap P^+) \setminus A$ , which contradicts 2.

Applying the same argument to  $P^+$  we get chains  $C_1^+, C_2^+, \dots, C_\mu^+$  with  $a_i$  as the minimum element of  $C_i^+$  for  $i = 1, 2, \dots, \mu$ .

Then from 2 we see that  $P = C_1 \cup C_2 \cup \dots \cup C_\mu$  where  $C_i = C_i^- \cup C_i^+$  is a chain for  $i = 1, 2, \dots, \mu$ . □



# Three applications of Dilworth's Theorem

(i) Another proof of

## Theorem

*Erdős and Szekeres*

$a_1, a_2, \dots, a_{n^2+1}$  contains a monotone subsequence of length  $n + 1$ .

Let  $P = \{(i, a_i) : 1 \leq i \leq n^2 + 1\}$  and let say  $(i, a_i) \preceq (j, a_j)$  if  $i < j$  and  $a_i \leq a_j$ .

A chain in  $P$  corresponds to a monotone increasing subsequence. So, suppose that there are no monotone increasing sequences of length  $n + 1$ . Then any cover of  $P$  by chains requires at least  $n + 1$  chains and so, by Dilworth's theorem, there exists an anti-chain  $A$  of size  $n + 1$ .

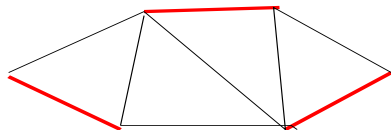
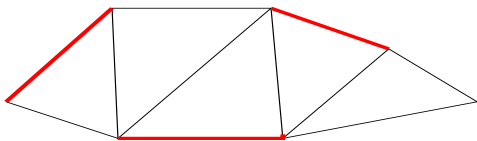
Let  $A = \{(i_t, a_{i_t}) : 1 \leq t \leq n+1\}$  where  $i_1 < i_2 \leq \dots < i_{n+1}$ .

Observe that  $a_{i_t} > a_{i_{t+1}}$  for  $1 \leq t \leq n$ , for otherwise  $(i_t, a_{i_t}) \preceq (i_{t+1}, a_{i_{t+1}})$  and  $A$  is not an anti-chain.

Thus  $A$  defines a monotone decreasing sequence of length  $n+1$ . □

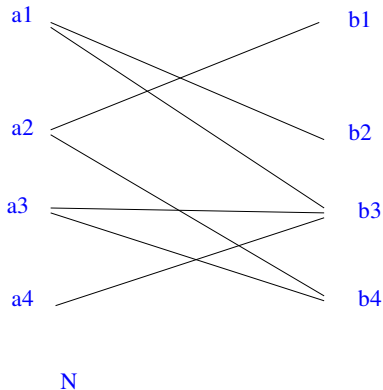
## Matchings in bipartite graphs

Re-call that a matching is a set of vertex disjoint edges.



P

Let  $G = (A \cup B, E)$  be a bipartite graph with bipartition  $A, B$ .  
For  $S \subseteq A$  let  $N(S) = \{b \in B : \exists a \in S, (a, b) \in E\}$ .

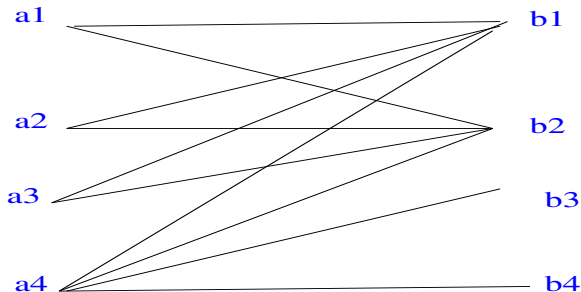


Clearly,  $|M| \leq |A|, |B|$  for any matching  $M$  of  $G$ .

## Theorem

(Hall)  $G$  contains a matching of size  $|A|$  iff

$$|N(S)| \geq |S| \quad \forall S \subseteq A.$$



$N(\{a_1, a_2, a_3\}) = \{b_1, b_2\}$  and so at most 2 of  $a_1, a_2, a_3$  can be saturated by a matching.

If  $G$  contains a matching  $M$  of size  $|A|$  then  
 $M = \{(a, f(a)) : a \in A\}$ , where  $f : A \rightarrow B$  is a 1-1 function.

But then,

$$|N(S)| \geq |f(S)| = S$$

for all  $S \subseteq A$ .

Let  $G = (A \cup B, E)$  be a bipartite graph which satisfies Hall's condition. Define a poset  $P = A \cup B$  and define  $<$  by  $a < b$  only if  $a \in A, b \in B$  and  $(a, b) \in E$ .

Suppose that the largest anti-chain in  $P$  is  $A = \{a_1, a_2, \dots, a_h, b_1, b_2, \dots, b_k\}$  and let  $s = h + k$ .

Now

$$N(\{a_1, a_2, \dots, a_h\}) \subseteq B \setminus \{b_1, b_2, \dots, b_k\}$$

for otherwise  $A$  will not be an anti-chain.

From Hall's condition we see that

$$|B| - k \geq h \text{ or equivalently } |B| \geq s.$$

Now by Dilworth's theorem,  $P$  is the union of  $s$  chains:

A matching  $M$  of size  $m$ ,  $|A| - m$  members of  $A$  and  $|B| - m$  members of  $B$ .

But then

$$m + (|A| - m) + (|B| - m) = s \leq |B|$$

and so  $m \geq |A|$ .





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A *network* consists of a **loopless** digraph  $D = (V, A)$  plus a function  $c : A \rightarrow \mathbf{R}_+$ . Here  $c(x, y)$  for  $(x, y) \in A$  is the *capacity* of the edge  $(x, y)$ .

We use the following notation: if  $\phi : A \rightarrow \mathbf{R}$  and  $S, T$  are (not necessarily disjoint) subsets of  $V$  then

$$\phi(S, T) = \sum_{\substack{x \in S \\ y \in T}} \phi(x, y).$$

Let  $s, t$  be distinct vertices. An  $s - t$  flow is a function  $f : A \rightarrow \mathbf{R}$  such that

$$f(v, V \setminus \{v\}) = f(V \setminus \{v\}, v) \quad \text{for all } v \neq s, t.$$

In words: flow into  $v$  equals flow out of  $v$ .

An  $s - t$  flow is *feasible* if

$$0 \leq f(x, y) \leq c(x, y) \quad \text{for all } (x, y) \in A.$$

An  $s - t$  *cut* is a partition of  $V$  into two sets  $S, \bar{S}$  such that  $s \in S$  and  $t \in \bar{S}$ .

The *value*  $v_f$  of the flow  $f$  is given by

$$v_f = f(s, V \setminus \{s\}) - f(V \setminus \{s\}, s).$$

Thus  $v_f$  is the net flow leaving  $s$ .

The *capacity* of the cut  $S : \bar{S}$  is equal to  $c(S, \bar{S})$ .

# Max-Flow Min-Cut Theorem

## Theorem

$$\max v_f = \min c(S, \bar{S})$$

where the maximum is over feasible  $s - t$  flows and the minimum is over  $s - t$  cuts.

**Proof** We observe first that

$$\begin{aligned} f(S, \bar{S}) - f(\bar{S}, S) &= (f(S, V) - f(S, S)) - (f(V, S) - f(S, S)) \\ &= f(S, V) - f(V, S) \\ &= v_f + \sum_{v \in S \setminus \{s\}} (f(v, V) - f(V, v)) \\ &= v_f. \end{aligned}$$

So,

$$v_f \leq f(S, \bar{S}) \leq c(S, \bar{S}).$$

This implies that

$$\max v_f \leq \min c(S, \bar{S}). \quad (11)$$

Given a flow  $f$  we define a *flow augmenting path*  $P$  to be a sequence of distinct vertices  $x_0 = s, x_1, x_2, \dots, x_k = t$  such that for all  $i$ , either

- Ⓕ1  $(x_i, x_{i+1}) \in A$  and  $f(x_i, x_{i+1}) < c(x_i, x_{i+1})$ , or
- Ⓕ2  $(x_{i+1}, x_i) \in A$  and  $f(x_{i+1}, x_i) > 0$ .

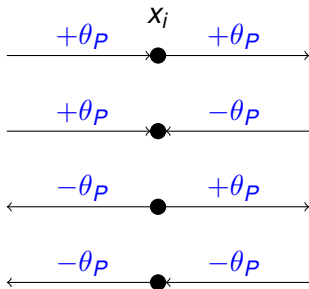
If  $P$  is such a sequence, then we define  $\theta_P > 0$  to be the minimum over  $i$  of  $c(x_i, x_{i+1}) - f(x_i, x_{i+1})$  (Case (F1)) and  $f(x_{i+1}, x_i)$  (Case (F2)).

**Claim 1:**  $f$  is a maximum value flow, iff there are no flow augmenting paths.

**Proof** If  $P$  is flow augmenting then define a new flow  $f'$  as follows:

- 1  $f'(x_i, x_{i+1}) = f(x_i, x_{i+1}) + \theta_P$  or
- 2  $f'(x_{i+1}, x_i) = f(x_{i+1}, x_i) - \theta_P$
- 3 For all other edges,  $(x, y)$ , we have  $f'(x, y) = f(x, y)$ .

We can see  
that the flow  
stays balanced at  $x_i$ .



We can see then that if there is a flow augmenting path then the new flow satisfies

$$V_{f'} = V_f + \theta_P > V_f.$$

Let  $S_f$  denote the set of vertices  $v$  for which there is a sequence  $x_0 = s, x_1, x_2, \dots, x_k = v$  which satisfies F1, F2 of the definition of flow augmenting paths.

If  $t \in S_f$  then the associated sequence defines a flow augmenting path. So, assume that  $t \notin S_f$ . Then we have,

- 1  $s \in S_f$ .
- 2 If  $x \in S_f, y \in \bar{S}_f, (x, y) \in A$  then  $f(x, y) = c(x, y)$ , else we would have  $y \in S_f$ .
- 3 If  $x \in S_f, y \in \bar{S}_f, (y, x) \in A$  then  $f(y, x) = 0$ , else we would have  $y \in S_f$ .

We therefore have

$$\begin{aligned}v_f &= f(S_f, \bar{S}_f) - f(\bar{S}_f, S) \\ &= c(S, \bar{S}_f).\end{aligned}$$

We see from this and (11) that  $f$  is a flow of maximum value and that the cut  $S_f : \bar{S}_f$  is of minimum capacity.

This finishes the proof of Claim 1 and the Max-Flow Min-Cut theorem.

Note also that we can construct  $S_f$  by beginning with  $S_f = \{s\}$  and then repeatedly adding any vertex  $y \notin S_f$  for which there is  $x \in S_f$  such that F1 or F2 holds. (A simple inductive argument based on sequence length shows that all of  $S_f$  is constructed in this way.)



Note also that we can construct  $S_f$  by beginning with  $S_f = \{s\}$  and then repeatedly adding any vertex  $y \notin S_f$  for which there is  $x \in S_f$  such that F1 or F2 holds.

This defines an algorithm for finding a maximum flow. The construction either finishes with  $t \in S_f$  and we can augment the flow.

Or, we find that  $t \notin S_f$  and we have a maximum flow.

Note, that if all the capacities  $c(x, y)$  are integers and we start with the all zero flow then we find that  $\theta_f$  is always a positive integer (formally one can use induction to verify this).

It follows that in this case, there is always a maximum flow that only takes integer values on the edges.

# Hall's Theorem.

Let  $G = (A, B, E)$  be a bipartite graph with  $A = \{a_1, \dots, a_n\}$  and  $B = \{b_1, \dots, b_n\}$ . A matching  $M$  is a set of edges that meets each vertex at most once. A matching is perfect if it meets each vertex.

Hall's theorem:

## Theorem

$G$  contains a perfect matching iff  $|N(S)| \geq |S|$  for all  $S \subseteq A$ .

Here  $N(S) = \{b \in B : \exists a \in A \text{ s.t. } \{a, b\} \in E\}$ .

Define a digraph  $\Gamma$  by adding vertices  $s, t \notin A \cup B$ . Then add edges  $(s, a_i)$  and  $(b_i, t)$  of capacity 1 for  $i = 1, 2, \dots, n$ . Orient the edges  $E$  for  $A$  to  $B$  and give them capacity  $\infty$ .

$G$  has a matching of size  $m$  iff there is an  $s - t$  flow of value  $m$ .  
An  $s - t$  cut  $X : \bar{X}$  has capacity

$$|A \setminus X| + |B \cap X| + |\{a \in X \cap A, b \in B \setminus X : \{a, b\} \in E\}| \times \infty.$$

It follows that to find a minimum cut, we need only consider  $X$  such that

$$\{a \in X \cap A, b \in B \setminus X : \{a, b\} \in E\} = \emptyset. \quad (12)$$

For such a set, we let  $S = A \cap X$  and  $T = X \cap B$ . Condition (12) means that  $T \supseteq N(S)$ . The capacity of  $X : \bar{X}$  is now  $(n - |S|) + |T|$  and for a fixed  $S$  this is minimised for  $T = N(S)$ .

Thus, by the Max-Flow Min-Cut theorem

$$\max\{|M|\} = \min_X \{c(X : \bar{X})\} = \min_S \{n - |S| + |N(S)|\}.$$

This implies Hall's theorem.

10/15/2021

# Graph orientation problem

Let  $G = (V, E)$  be a graph. When is it possible to orient the edges of  $G$  to create a digraph  $\Gamma = (V, A)$  so that every vertex has out-degree at least  $d$ . We say that  $G$  is  $d$ -orientable.

## Theorem

$G$  is  $d$ -orientable iff

$$|\{e \in E : e \cap S \neq \emptyset\}| \geq d|S| \text{ for all } S \subseteq V. \quad (13)$$

**Proof** If  $G$  is  $d$ -orientable then

$$|\{e \in E : e \cap S \neq \emptyset\}| \geq |\{(x, y) \in A : x \in S\}| \geq d|S|.$$

Suppose now that (13) holds. Define a network  $D$  as follows; the vertices are  $s, t, V, E$  – yes,  $D$  has a vertex for each edge of  $G$ .

There is an edge of capacity  $d$  from  $s$  to each  $v \in V$  and an edge of capacity one from each  $e \in E$  to  $t$ . There is an edge of infinite capacity from  $v \in V$  to each edge  $e$  that contains  $v$ .

Consider an integer flow  $f$ . Suppose that  $e = \{v, w\} \in E$  and  $f(e, t) = 1$ . Then either  $f(v, e) = 1$  or  $f(w, e) = 1$ . In the former we interpret this as orienting the edge  $e$  from  $v$  to  $w$  and in the latter from  $w$  to  $v$ .

Under this interpretation,  $G$  is  $d$ -orientable iff  $D$  has a flow of value  $d|V|$ .

Let  $X : \bar{X}$  be an  $s - t$  cut in  $N$ . Let  $S = X \cap V$  and  $T = X \cap E$ .

To have a finite capacity, there must be no  $x \in S$  and  $e \in E \setminus T$  such that  $x \in e$ .

So, the capacity of a finite capacity cut is at least

$$d(|V| - |S|) + |\{e \in E : e \cap S \neq \emptyset\}|$$

And this is at least  $d|V|$  if (13) holds.

**Example 1** A disc lies in a plane. Its centre is fixed but it is free to rotate. It has been divided into  $n$  sectors of angle  $2\pi/n$ . Each sector is to be colored Red or Blue. How many different colorings are there?

One could argue for  $2^n$ .

On the other hand, what if we only distinguish colorings which cannot be obtained from one another by a rotation. For example if  $n = 4$  and the sectors are numbered 0,1,2,3 in clockwise order around the disc, then there are only 6 ways of coloring the disc – 4R, 4B, 3R1B, 1R3B, RRBB and RBRB.



## Example 2

Now consider an  $n \times n$  “chessboard” where  $n \geq 2$ . Here we color the squares Red and Blue and two colorings are different only if one cannot be obtained from another by a rotation or a reflection. For  $n = 2$  there are 6 colorings.

The general scenario that we consider is as follows: We have a set  $X$  which will stand for the set of colorings when transformations are not allowed. (In example 1,  $|X| = 2^n$  and in example 2,  $|X| = 2^{n^2}$ ).

In addition there is a set  $G$  of permutations of  $X$ . This set will have a **group structure**:

Given two members  $g_1, g_2 \in G$  we can define their composition  $g_1 \circ g_2$  by  $g_1 \circ g_2(x) = g_1(g_2(x))$  for  $x \in X$ . We require that  $G$  is *closed* under composition i.e.  $g_1 \circ g_2 \in G$  if  $g_1, g_2 \in G$ .

We also have the following:

A1 The *identity* permutation  $1_X \in G$ .

A2  $(g_1 \circ g_2) \circ g_3 = g_1 \circ (g_2 \circ g_3)$  (Composition is associative).

A3 The inverse permutation  $g^{-1} \in G$  for every  $g \in G$ .

(A set  $G$  with a binary relation  $\circ$  which satisfies **A1,A2,A3** is called a **Group**).

In example 1  $D = \{0, 1, 2, \dots, n-1\}$ ,  $X = 2^D$  and the group is  $G_1 = \{e_0, e_1, \dots, e_{n-1}\}$  where  $e_j * x = x + j \pmod n$  stands for rotation by  $2j\pi/n$ .

In example 2,  $X = 2^{[n]^2}$ . We number the squares 1,2,3,4 in clockwise order starting at the upper left and represent  $X$  as a sequence from  $\{r, b\}^4$  where for example rrbrr means color 1,2,4 Red and 3 Blue.  $G_2 = \{e, a, b, c, p, q, r, s\}$  is in a sense independent of  $n$ .  $e, a, b, c$  represent a rotation through 0, 90, 180, 270 degrees respectively.  $p, q$  represent reflections in the vertical and horizontal and  $r, s$  represent reflections in the diagonals 1,3 and 2,4 respectively.

	e	a	b	c	p	q	r	s
rrrr	rrrr	rrrr	rrrr	rrrr	rrrr	rrrr	rrrr	rrrr
brrr	brrr	rbrr	rrbr	rrrb	rbrr	rrrb	brrr	rrbr
rbrr	rbrr	rrbr	rrrb	brrr	brrr	rrbr	rrrb	rbrr
rrbr	rrbr	rrrb	brrr	rbrr	rrrb	rbrr	rrbr	brrr
rrrb	rrrb	brrr	rbrr	rrbr	rrbr	brrr	rbrr	rrrb
bbrr	bbrr	rbbr	rrbb	brrb	bbrr	rrbb	brrb	rbbr
rbbr	rbbr	rrbb	brrb	bbrr	brrb	rbbr	rrbb	bbrr
rrbb	rrbb	brrb	bbrr	rbbr	rrbb	bbrr	rbbr	brrb
brrb	brrb	bbrr	rbbr	rrbb	rbbr	brrb	bbrr	rrbb
rbrb	rbrb	brbr	rbrb	brbr	brbr	brbr	rbrb	rbrb
brbr	brbr	rbrb	brbr	rbrb	rbrb	rbrb	brbr	brbr
bbbr	bbbr	rbbb	brbb	bbrb	bbrb	rbbb	brbb	bbbr
bbrb	bbrb	bbbr	rbbb	brbb	bbbr	brbb	bbrb	rbbb
brbb	brbb	bbrb	bbbr	rbbb	brbb	bbrb	bbbr	brbb
rbbb	rbbb	brbb	bbrb	bbbr	brbb	bbbr	rbbb	bbrb
bbbb	bbbb	bbbb	bbbb	bbbb	bbbb	bbbb	bbbb	bbbb

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# Theorem 2

(Frobenius, Burnside)

$$\nu_{X,G} = \frac{1}{|G|} \sum_{g \in G} |\text{Fix}(g)|.$$

**Proof** Let  $A(x, g) = 1_{g \cdot x = x}$ . Then

$$\begin{aligned} \nu_{X,G} &= \frac{1}{|G|} \sum_{x \in X} |S_x| \\ &= \frac{1}{|G|} \sum_{x \in X} \sum_{g \in G} A(x, g) \\ &= \frac{1}{|G|} \sum_{g \in G} \sum_{x \in X} A(x, g) \\ &= \frac{1}{|G|} \sum_{g \in G} |\text{Fix}(g)|. \end{aligned}$$

Let us consider example 1 with  $n = 6$ . We compute

$g$	$e_0$	$e_1$	$e_2$	$e_3$	$e_4$	$e_5$
$ Fix(g) $	64	2	4	8	4	2

Applying Theorem 2 we obtain

$$\nu_{X,G} = \frac{1}{6}(64 + 2 + 4 + 8 + 4 + 2) = 14.$$



# Cycles of a permutation

Let  $\pi : D \rightarrow D$  be a permutation of the finite set  $D$ . Consider the digraph  $\Gamma_\pi = (D, A)$  where  $A = \{(i, \pi(i)) : i \in D\}$ .  $\Gamma_\pi$  is a collection of vertex disjoint cycles. Each  $x \in D$  being on a unique cycle. Here a cycle can consist of a loop i.e. when  $\pi(x) = x$ .

Example:  $D = [10]$ .

$i$	1	2	3	4	5	6	7	8	9	10
$\pi(i)$	6	2	7	10	3	8	9	1	5	4

The cycles are  $(1, 6, 8)$ ,  $(2)$ ,  $(3, 7, 9, 5)$ ,  $(4, 10)$ .

In general consider the sequence  $i, \pi(i), \pi^2(i), \dots$ .

Since  $D$  is finite, there exists a first pair  $k < \ell$  such that  $\pi^k(i) = \pi^\ell(i)$ . Now we must have  $k = 0$ , since otherwise putting  $x = \pi^{k-1}(i) \neq y = \pi^{\ell-1}(i)$  we see that  $\pi(x) = \pi(y)$ , contradicting the fact that  $\pi$  is a permutation.

So  $i$  lies on the cycle  $C = (i, \pi(i), \pi^2(i), \dots, \pi^{k-1}(i), i)$ .

If  $j$  is not a vertex of  $C$  then  $\pi(j)$  is not on  $C$  and so we can repeat the argument to show that the rest of  $D$  is partitioned into cycles.

## Example 2

It is straightforward to check that when  $n$  is even, we have

$g$	e	a	b	c	p	q	r	s
$ Fix(g) $	$2^{n^2}$	$2^{n^2/4}$	$2^{n^2/2}$	$2^{n^2/4}$	$2^{n^2/2}$	$2^{n^2/2}$	$2^{n(n+1)/2}$	$2^{n(n+1)/2}$

For example, if we divide the chessboard into 4  $n/2 \times n/2$  sub-squares, numbered 1,2,3,4 then a coloring is in  $Fix(a)$  iff each of these 4 sub-squares have colorings which are rotations of the coloring in square 1.

## Polya's Theorem

We now extend the above analysis to answer questions like:  
How many *distinct* ways are there to color an  $8 \times 8$  chessboard with 32 white squares and 32 black squares?

The scenario now consists of a set  $D$  (*Domain*), a set  $C$  (colors) and  $X = \{x : D \rightarrow C\}$  is the set of colorings of  $D$  with the color set  $C$ .  $G$  is now a group of permutations of  $D$ .

We see first how to extend each permutation of  $D$  to a permutation of  $X$ . Suppose that  $x \in X$  and  $g \in G$  then we define  $g * x$  by

$$g * x(d) = x(g^{-1}(d)) \quad \text{for all } d \in D.$$

**Explanation:** The color of  $d$  is the color of the element  $g^{-1}(d)$  which is mapped to it by  $g$ .

Consider Example 1 with  $n = 4$ . Suppose that  $g = e_1$  i.e. rotate clockwise by  $\pi/2$  and  $x(1) = b, x(2) = b, x(3) = r, x(4) = r$ .

Then for example

$$g * x(1) = x(g^{-1}(1)) = x(4) = r, \text{ as before.}$$

Now associate a **weight**  $w_c$  with each  $c \in C$ .

If  $x \in X$  then

$$W(x) = \prod_{d \in D} w_{x(d)}.$$

Thus, if in Example 1 we let  $w(r) = R$  and  $w(b) = B$  and take  $x(1) = b, x(2) = b, x(3) = r, x(4) = r$  then we will write  $W(x) = B^2 R^2$ .

For  $S \subseteq X$  we define the **inventory** of  $S$  to be

$$W(S) = \sum_{x \in S} W(x).$$

The problem we discuss now is to compute the **pattern inventory**  $PI = W(S^*)$  where  $S^*$  contains one member of each orbit of  $X$  under  $G$ .

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For example, in the case of Example 2, with  $n = 2$ , we get

$$PI = R^4 + R^3B + 2R^2B^2 + RB^3 + B^4.$$

To see that the definition of  $PI$  makes sense we need to prove  
**Lemma 3** If  $x, y$  are in the same orbit of  $X$  then  $W(x) = W(y)$ .

**Proof** Suppose that  $g * x = y$ . Then

$$\begin{aligned} W(y) &= \prod_{d \in D} w_y(d) \\ &= \prod_{d \in D} w_{g*x}(d) \\ &= \prod_{d \in D} w_x(g^{-1}(d)) \end{aligned} \tag{14}$$

$$\begin{aligned} &= \prod_{d \in D} w_x(d) \\ &= W(x) \end{aligned} \tag{15}$$

Note, that we can go from (14) to (15) because as  $d$  runs over  $D$ ,  $g^{-1}(d)$  also runs over  $d$ .



Let  $\Delta = |D|$ . If  $g \in G$  has  $k_i$  cycles of length  $i$  then we define

$$ct(g) = x_1^{k_1} x_2^{k_2} \cdots x_{\Delta}^{k_{\Delta}}.$$

The **Cycle Index Polynomial** of  $G$ ,  $C_G$  is then defined to be

$$C_G(x_1, x_2, \dots, x_{\Delta}) = \frac{1}{|G|} \sum_{g \in G} ct(g).$$

In Example 2 with  $n = 2$  we have

$g$	e	a	b	c	p	q	r	s
$ct(g)$	$x_1^4$	$x_4$	$x_2^2$	$x_4$	$x_2^2$	$x_2^2$	$x_1^2 x_2$	$x_1^2 x_2$

and so

$$C_G(x_1, x_2, x_3, x_4) = \frac{1}{8}(x_1^4 + 3x_2^2 + 2x_1^2 x_2 + 2x_4).$$

In Example 2 with  $n = 3$  we have

$g$	e	a	b	c	p	q	r	s
$ct(g)$	$x_1^9$	$x_1 x_4^2$	$x_1 x_2^4$	$x_1 x_4^2$	$x_1^3 x_2^3$	$x_1^3 x_2^3$	$x_1^3 x_2^3$	$x_1^3 x_2^3$

and so

$$C_G(x_1, x_2, x_3, x_4) = \frac{1}{8}(x_1^9 + x_1 x_4^2 + 4x_1^3 x_2^3 + 2x_1 x_4^2).$$

## Theorem (Polya)

$$PI = C_G \left( \sum_{c \in C} w_c, \sum_{c \in C} w_c^2, \dots, \sum_{c \in C} w_c^\Delta \right).$$

**Proof** In Example 2, we replace  $x_1$  by  $R + B$ ,  $x_2$  by  $R^2 + B^2$  and so on. When  $n = 2$  this gives

$$\begin{aligned} PI &= \frac{1}{8}((R + B)^4 + 3(R^2 + B^2)^2 + \\ &\quad 2(R + B)^2(R^2 + B^2) + 2(R^4 + B^4)) \\ &= R^4 + R^3B + 2R^2B^2 + RB^3 + B^4. \end{aligned}$$

Putting  $R = B = 1$  gives the number of distinct colorings. Note also the formula for  $PI$  tells us that there are 2 distinct colorings using 2 reds and 2 Blues.

## Proof of Polya's Theorem

Let  $X = X_1 \cup X_2 \cup \dots \cup X_m$  be the equivalence classes of  $X$  under the relation

$$x \sim y \text{ iff } W(x) = W(y).$$

By Lemma 2,  $g * x \sim x$  for all  $x \in X, g \in G$  and so we can think of  $G$  acting on each  $X_i$  individually i.e. we use the fact that  $x \in X_i$  implies  $g * x \in X_i$  for all  $i \in [m], g \in G$ . We use the notation  $g^{(i)} \in G^{(i)}$  when we restrict attention to  $X_i$ .

Let  $m_i$  denote the number of orbits  $\nu_{X_i, G^{(i)}}$  and  $W_i$  denote the common PI of  $G^{(i)}$  acting on  $X_i$ . Then

$$\begin{aligned} PI &= \sum_{i=1}^m m_i W_i \\ &= \sum_{i=1}^m W_i \left( \frac{1}{|G|} \sum_{g \in G} |Fix(g^{(i)})| \right) && \text{by Theorem 2} \\ &= \frac{1}{|G|} \sum_{g \in G} \sum_{i=1}^m |Fix(g^{(i)})| W_i \\ &= \frac{1}{|G|} \sum_{g \in G} W(Fix(g)) \end{aligned} \tag{16}$$

Note that (16) follows from  $Fix(g) = \bigcup_{i=1}^m Fix(g^{(i)})$  since  $x \in Fix(g^{(i)})$  iff  $x \in X_i$  and  $g * x = x$ .

Suppose now that  $ct(g) = x_1^{k_1} x_2^{k_2} \cdots x_{\Delta}^{k_{\Delta}}$  as above. Then we claim that

$$W(\text{Fix}(g)) = \left( \sum_{c \in C} w_c \right)^{k_1} \left( \sum_{c \in C} w_c^2 \right)^{k_2} \cdots \left( \sum_{c \in C} w_c^{\Delta} \right)^{k_{\Delta}}. \quad (17)$$

Substituting (17) into (16) yields the theorem.

To verify (17) we use the fact that if  $x \in \text{Fix}(g)$ , then the elements of a cycle of  $g$  must be given the same color. A cycle of length  $i$  will then contribute a factor  $\sum_{c \in C} w_c^i$  where the term  $w_c^i$  comes from the choice of color  $c$  for every element of the cycle. □

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# Combinatorial Games

**Game 1** Start with  $n$  chips. Players A,B alternately take 1,2,3 or 4 chips until there are none left. The winner is the person who takes the last chip:

Example

	A	B	A	B	A	
$n = 10$	3	2	4	1		B wins
$n = 11$	1	2	3	4	1	A wins

What is the optimal strategy for playing this game?



# Combinatorial Games

**Game 2** Chip placed at point  $(m, n)$ . Players can move chip to  $(m', n)$  or  $(m, n')$  where  $0 \leq m' < m$  and  $0 \leq n' < n$ . The player who makes the last move and puts the chip onto  $(0, 0)$  wins.

What is the optimal strategy for this game?

**Game 2a** Chip placed at point  $(m, n)$ . Players can move chip to  $(m', n)$  or  $(m, n')$  or to  $(m - a, n - a)$  where  $0 \leq m' < m$  and  $0 \leq n' < n$  and  $0 \leq a \leq \min\{m, n\}$ . The player who makes the last move and puts the chip onto  $(0, 0)$  wins.

What is the optimal strategy for this game?

# Combinatorial Games

**Game 3**  $W$  is a set of words. A and B alternately remove words  $w_1, w_2, \dots$ , from  $W$ . The rule is that the first letter of  $w_{i+1}$  must be the same as the last letter of  $w_i$ . The player who makes the last legal move wins.

## Example

$W = \{ \textit{England, France, Germany, Russia, Bulgaria, \dots} \}$

What is the optimal strategy for this game?

# Abstraction

Represent each position of the game by a vertex of a digraph  $D = (X, A)$ .

$(x, y)$  is an arc of  $D$  iff one can move from position  $x$  to position  $y$ .

We assume that the digraph is finite and that it is **acyclic** i.e. there are no directed cycles.

The game starts with a token on vertex  $x_0$  say, and players alternately move the token to  $x_1, x_2, \dots$ , where  $x_{i+1} \in N^+(x_i)$ , the set of out-neighbours of  $x_i$ . The game ends when the token is on a **sink** i.e. a vertex of out-degree zero. The last player to move is the winner.

# Abstraction

Example 1:  $V(D) = \{0, 1, \dots, n\}$  and  $(x, y) \in A$  iff  $x - y \in \{1, 2, 3, 4\}$ .

Example 2:  $V(D) = \{0, 1, \dots, m\} \times \{0, 1, \dots, n\}$  and  $(x, y) \in N^+((x', y'))$  iff  $x = x'$  and  $y > y'$  or  $x > x'$  and  $y = y'$ .

Example 2a:  $V(D) = \{0, 1, \dots, m\} \times \{0, 1, \dots, n\}$  and  $(x, y) \in N^+((x', y'))$  iff  $x = x'$  and  $y > y'$  or  $x > x'$  and  $y = y'$  or  $x - x' = y - y' > 0$ .

Example 3:  $V(D) = \{(W', w) : W' \subseteq W \setminus \{w\}\}$ .  $w$  is the last word used and  $W'$  is the remaining set of unused words.  $(X', w') \in N^+((X, w))$  iff  $w' \in X$  and  $w'$  begins with the last letter of  $w$ . Also, there is an arc from  $(W, \cdot)$  to  $(W \setminus \{w\}, w)$  for all  $w$ , corresponding to the games start.

# Abstraction

We will first argue that such a game must eventually end.

A **topological numbering** of digraph  $D = (X, A)$  is a map  $f : X \rightarrow [n]$ ,  $n = |X|$  which satisfies  $(x, y) \in A$  implies  $f(x) < f(y)$ .

## Theorem

*A finite digraph  $D = (X, A)$  is acyclic iff it admits at least one topological numbering.*

**Proof**     Suppose first that  $D$  has a topological numbering. We show that it is acyclic.

Suppose that  $C = (x_1, x_2, \dots, x_k, x_1)$  is a directed cycle. Then  $f(x_1) < f(x_2) < \dots < f(x_k) < f(x_1)$ , contradiction.

# Abstraction

Suppose now that  $D$  is acyclic. We first argue that  $D$  has at least one sink.

Thus let  $P = (x_1, x_2, \dots, x_k)$  be a longest simple path in  $D$ . We claim that  $x_k$  is a sink.

If  $D$  contains an arc  $(x_k, y)$  then either  $y = x_i, 1 \leq i \leq k - 1$  and this means that  $D$  contains the cycle  $(x_i, x_{i+1}, \dots, x_k, x_i)$ , contradiction or  $y \notin \{x_1, x_2, \dots, x_k\}$  and then  $(P, y)$  is a longer simple path than  $P$ , contradiction.

# Abstraction

We can now prove by induction on  $n$  that there is at least one topological numbering.

If  $n = 1$  and  $X = \{x\}$  then  $f(x) = 1$  defines a topological numbering.

Now assume that  $n > 1$ . Let  $z$  be a sink of  $D$  and define  $f(z) = n$ . The digraph  $D' = D - z$  is acyclic and by the induction hypothesis it admits a topological numbering,  $f : X \setminus \{z\} \rightarrow [n - 1]$ .

The function we have defined on  $X$  is a topological numbering. If  $(x, y) \in A$  then either  $x, y \neq z$  and then  $f(x) < f(y)$  by our assumption on  $f$ , or  $y = z$  and then  $f(x) < n = f(z)$  ( $x \neq z$  because  $z$  is a sink).



# Abstraction

The fact that  $D$  has a topological numbering implies that the game must end. Each move increases the  $f$  value of the current position by at least one and so after at most  $n$  moves a sink must be reached.

The positions of a game are partitioned into 2 sets:

- $P$ -positions: The next player cannot win. The **previous** player can win regardless of the current player's strategy.
- $N$ -positions: The **next** player has a strategy for winning the game.

Thus an  $N$ -position is a **winning** position for the next player and a  $P$ -position is a **losing** position for the next player.

The main problem is to determine  $N$  and  $P$  and what the strategy is for winning from an  $N$ -position.



Let the vertices of  $D$  be  $x_1, x_2, \dots, x_n$ , in topological order.

## Labelling procedure

- 1  $i \leftarrow n$ , Label  $x_n$  with  $P$ .  $N \leftarrow \emptyset$ ,  $P \leftarrow \emptyset$ .
- 2  $i \leftarrow i - 1$ . If  $i = 0$  STOP.
- 3 Label  $x_i$  with  $N$ , if  $N^+(x_i) \cap P \neq \emptyset$ .
- 4 Label  $x_i$  with  $P$ , if  $N^+(x_i) \subseteq N$ .
- 5 goto 2.

The partition  $N, P$  satisfies

$$x \in N \text{ iff } N^+(x) \cap P \neq \emptyset$$

To play from  $x \in N$ , move to  $y \in N^+(x) \cap P$ .

# Abstraction

In Game 1,  $P = \{5k : k \geq 0\}$ .

In Game 2,  $P = \{(x, x) : x \geq 0\}$ .

## Lemma

*The partition into  $N, P$  satisfying  $x \in N$  iff  $N^+(x) \cap P \neq \emptyset$  is unique.*

**Proof** If there were two partitions  $N_i, P_i, i = 1, 2$ , let  $x_i$  be the vertex of highest topological number which is not in  $(N_1 \cap N_2) \cup (P_1 \cap P_2)$ . Suppose that  $x_i \in N_1 \setminus N_2$ .

But then  $x_i \in N_1$  implies  $N^+(x_i) \cap P_1 \cap \{x_{i+1}, \dots, x_n\} \neq \emptyset$  and  $x_i \in P_2$  implies  $N^+(x_i) \cap P_2 \cap \{x_{i+1}, \dots, x_n\} = \emptyset$ .

But  $P_1 \cap \{x_{i+1}, \dots, x_n\} = P_2 \cap \{x_{i+1}, \dots, x_n\}$ .



10/29/2021

# Sums of games

Suppose that we have  $p$  games  $G_1, G_2, \dots, G_p$  with digraphs  $D_i = (X_i, A_i)$ ,  $i = 1, 2, \dots, p$ .

The sum  $G_1 \oplus G_2 \oplus \dots \oplus G_p$  of these games is played as follows. A position is a vector

$(x_1, x_2, \dots, x_p) \in X = X_1 \times X_2 \times \dots \times X_p$ . To make a move, a player chooses  $i$  such that  $x_i$  is not a sink of  $D_i$  and then replaces  $x_i$  by  $y \in N_i^+(x_i)$ . The game ends when each  $x_i$  is a sink of  $D_i$  for  $i = 1, 2, \dots, n$ .

Knowing the partitions  $N_i, P_i$  for game  $i = 1, 2, \dots, p$  does not seem to be enough to determine how to play the sum of the games.

We need more information. This will be provided by the **Sprague-Grundy Numbering**

# Sums of games

## Example

**Nim** In a one pile game, we start with  $a \geq 0$  chips and while there is a positive number  $x$  of chips, a move consists of deleting  $y \leq x$  chips. In this game the  $N$ -positions are the positive integers and the unique  $P$ -position is 0.

In general, Nim consists of the sum of  $n$  single pile games starting with  $a_1, a_2, \dots, a_n > 0$ . A move consists of deleting some chips from a non-empty pile.

Example 2 is Nim with 2 piles.

## Sprague-Grundy (*SG*) Numbering

For  $S \subseteq \{0, 1, 2, \dots\}$  let

$$\text{mex}(S) = \min\{x \geq 0 : x \notin S\}.$$

Now given an acyclic digraph  $D = X, A$  with topological ordering  $x_1, x_2, \dots, x_n$  define  $g$  iteratively by

- 1  $i \leftarrow n, g(x_n) = 0.$
- 2  $i \leftarrow i - 1.$  If  $i = 0$  STOP.
- 3  $g(x_i) = \text{mex}(\{g(x) : x \in N^+(x_i)\}).$
- 4 goto 2.

## Lemma

$$x \in P \leftrightarrow g(x) = 0.$$

**Proof** Because

$$x \in N \text{ iff } N^+(x) \cap P \neq \emptyset$$

all we have to show is that

$$g(x) > 0 \text{ iff } \exists y \in N^+(x) \text{ such that } g(y) = 0.$$

But this is immediate from  $g(x) = \text{mex}(\{g(y) : y \in N^+(x)\})$   $\square$

# Sums of games

Another one pile subtraction game.

- A player can remove any even number of chips, but not the whole pile.
- A player can remove the whole pile if it is odd.

The terminal positions are 0 or 2.

## Lemma

$g(0) = 0$ ,  $g(2k) = k - 1$  and  $g(2k - 1) = k$  for  $k \geq 1$ .



# Sums of games

**Proof** 0,2 are terminal positions and so  $g(0) = g(2) = 0$ .  
 $g(1) = 1$  because the only position one can move to from 1 is 0. We prove the remainder by induction on  $k$ .

Assume that  $k > 1$ .

$$\begin{aligned}g(2k) &= \text{mex}\{g(2k-2), g(2k-4), \dots, g(2)\} \\ &= \text{mex}\{k-2, k-3, \dots, 0\} \\ &= k-1.\end{aligned}$$

$$\begin{aligned}g(2k-1) &= \text{mex}\{g(2k-3), g(2k-5), \dots, g(1), g(0)\} \\ &= \text{mex}\{k-1, k-2, \dots, 0\} \\ &= k.\end{aligned}$$



# Sums of games

We now show how to compute the **SG** numbering for a sum of games.

For binary integers  $a = a_m a_{m-1} \cdots a_1 a_0$  and  $b = b_m b_{m-1} \cdots b_1 b_0$  we define  $a \oplus b = c_m c_{m-1} \cdots c_1 c_0$  by

$$c_i = \begin{cases} 1 & \text{if } a_i \neq b_i \\ 0 & \text{if } a_i = b_i \end{cases}$$

for  $i = 1, 2, \dots, m$ .

So  $11 \oplus 5 = 14$ .

# Sums of games

## Theorem

If  $g_i$  is the SG function for game  $G_i$ ,  $i = 1, 2, \dots, p$  then the SG function  $g$  for the sum of the games  $G = G_1 \oplus G_2 \oplus \dots \oplus G_p$  is defined by

$$g(x) = g_1(x_1) \oplus g_2(x_2) \oplus \dots \oplus g_p(x_p)$$

where  $x = (x_1, x_2, \dots, x_p)$ .

For example if in a game of Nim, the pile sizes are  $x_1, x_2, \dots, x_p$  then the SG value of the position is

$$x_1 \oplus x_2 \oplus \dots \oplus x_p$$

# Sums of games

**Proof** It is enough to show this for  $p = 2$  and then use induction on  $p$ .

Write  $G = H \oplus G_p$  where  $H = G_1 \oplus G_2 \oplus \cdots \oplus G_{p-1}$ . Let  $h$  be the SG numbering for  $H$ . Then, if  $y = (x_1, x_2, \dots, x_{p-1})$ ,

$$\begin{aligned}g(x) &= h(y) \oplus g_p(x_p) \quad \text{assuming theorem for } p = 2 \\ &= g_1(x_1) \oplus g_2(x_2) \oplus \cdots \oplus g_{p-1}(x_{p-1}) \oplus g_p(x_p)\end{aligned}$$

by induction.

It is enough now to show, for  $p = 2$ , that

- A1 If  $x \in X$  and  $g(x) = b > a$  then there exists  $x' \in N^+(x)$  such that  $g(x') = a$ .
- A2 If  $x \in X$  and  $g(x) = b$  and  $x' \in N^+(x)$  then  $g(x') \neq g(x)$ .
- A3 If  $x \in X$  and  $g(x) = 0$  and  $x' \in N^+(x)$  then  $g(x') \neq 0$

# Sums of games

A1. Write  $d = a \oplus b$ . Then

$$a = d \oplus b = d \oplus g_1(x_1) \oplus g_2(x_2). \quad (18)$$

Now suppose that we can show that either

$$(i) d \oplus g_1(x_1) < g_1(x_1) \text{ or } (ii) d \oplus g_2(x_2) < g_2(x_2) \text{ or both.} \quad (19)$$

Assume that (i) holds.

Then since  $g_1(x_1) = \text{mex}(N_1^+(x_1))$  there must exist  $x'_1 \in N_1^+(x_1)$  such that  $g_1(x'_1) = d \oplus g_1(x_1)$ .

Then from (18) we have

$$a = g_1(x'_1) \oplus g_2(x_2) = g(x'_1, x_2).$$

Furthermore,  $(x'_1, x_2) \in N^+(x)$  and so we will have verified A1.

# Sums of games

Let us verify (19).

Suppose that  $2^{k-1} \leq d < 2^k$ .

Then  $d$  has a 1 in position  $k$  and no higher.

Since  $d_k = a_k \oplus b_k$  and  $a < b$  we must have  $a_k = 0$  and  $b_k = 1$ .

So either (i)  $g_1(x_1)$  has a 1 in position  $k$  or (ii)  $g_2(x_2)$  has a 1 in position  $k$ . Assume (i).

But then  $d \oplus g_1(x_1) < g_1(x_1)$  since  $d$  “destroys” the  $k$ th bit of  $g_1(x_1)$  and does not change any higher bit.

# Sums of games

A2. Suppose without loss of generality that  $g(x'_1, x_2) = g(x_1, x_2)$  where  $x'_1 \in N^+(x_1)$ .

Then  $g_1(x'_1) \oplus g_2(x_2) = g_1(x_1) \oplus g_2(x_2)$  implies that  $g_1(x'_1) = g_1(x_1)$ , contradiction. □

A3. Suppose that  $g_1(x_1) \oplus g_2(x_2) = 0$  and  $g_1(x'_1) \oplus g_2(x_2) = 0$  where  $x'_1 \in N^+(x_1)$ .

Then  $g_1(x_1) = g_1(x'_1)$ , contradicting  $g_1(x_1) = \text{mex}\{g_1(x) : x \in N^+(x_1)\}$ .

# Sums of games

If we apply this theorem to the game of Nim then if the position  $x$  consists of piles of  $x_i$  chips for  $i = 1, 2, \dots, p$  then

$$g(x) = x_1 \oplus x_2 \oplus \dots \oplus x_p.$$

In our first example,  $g(x) = x \bmod 5$  and so for the sum of  $p$  such games we have

$$g(x_1, x_2, \dots, x_p) = (x_1 \bmod 5) \oplus (x_2 \bmod 5) \oplus \dots \oplus (x_p \bmod 5).$$



11/1/2021

# Geography

Start with a chip sitting on a vertex  $v$  of a graph or digraph  $G$ . A move consists of moving the chip to a neighbouring vertex.

In edge geography, moving the chip from  $x$  to  $y$  deletes the edge  $(x, y)$ . In vertex geography, moving the chip from  $x$  to  $y$  deletes the vertex  $x$ .

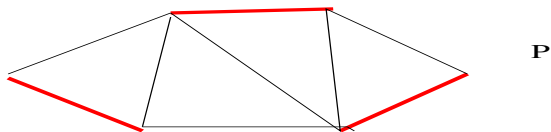
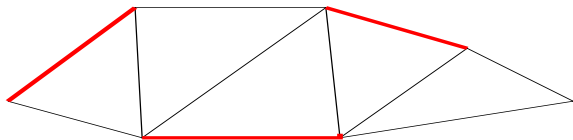
The problem is given a position  $(G, v)$ , to determine whether this is a  $P$  or  $N$  position.

**Complexity** Both edge and vertex geography are Pspace-hard on digraphs. Edge geography is Pspace-hard on an undirected graph. Only vertex geography on a graph is polynomial time solvable.

# Undirected Vertex Geography

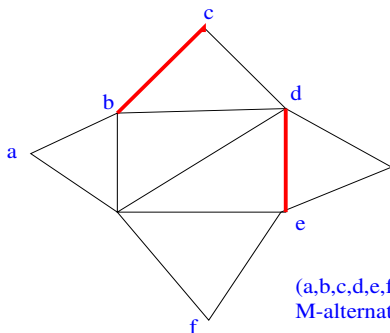
We need some simple results from the theory of matchings on graphs.

A *matching*  $M$  of a graph  $G = (V, E)$  is a set of edges, no two of which are incident to a common vertex.



# Undirected Vertex Geography

$M$ -alternating path



$(a,b,c,d,e,f)$  is an  
 $M$ -alternating path

An  $M$ -alternating path joining 2  $M$ -unsaturated vertices is called an  $M$ -augmenting path.

# Undirected Vertex Geography

$M$  is a *maximum* matching of  $G$  if no matching  $M'$  has more edges.

## Theorem

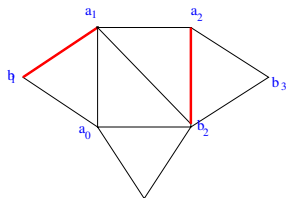
$M$  is a maximum matching iff  $M$  admits no  $M$ -augmenting paths.

**Proof** Suppose  $M$  has an augmenting path

$P = (a_0, b_1, a_1, \dots, a_k, b_{k+1})$  where

$e_i = (a_{i-1}, b_i) \notin M, 1 \leq i \leq k+1$  and

$f_i = (b_i, a_i) \in M, 1 \leq i \leq k.$



Let  $M' = M - \{f_1, f_2, \dots, f_k\} + \{e_1, e_2, \dots, e_{k+1}\}.$

# Undirected Vertex Geography

- $|M'| = |M| + 1$ .
- $M'$  is a matching

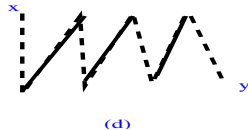
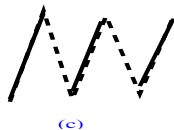
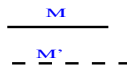
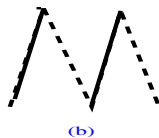
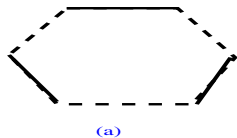
For  $x \in V$  let  $d_M(x)$  denote the degree of  $x$  in matching  $M$ , So  $d_M(x)$  is 0 or 1.

$$d_{M'}(x) = \begin{cases} d_M(x) & x \notin \{a_0, b_1, \dots, b_{k+1}\} \\ d_M(x) & x \in \{b_1, \dots, a_k\} \\ d_M(x) + 1 & x \in \{a_0, b_{k+1}\} \end{cases}$$

So if  $M$  has an augmenting path it is not maximum.

# Undirected Vertex Geography

Suppose  $M$  is not a maximum matching and  $|M'| > |M|$ .  
Consider  $H = G[M \nabla M']$  where  $M \nabla M' = (M \setminus M') \cup (M' \setminus M)$  is the set of edges in *exactly* one of  $M, M'$ .  
Maximum degree of  $H$  is 2 –  $\leq 1$  edge from  $M$  or  $M'$ . So  $H$  is a collection of vertex disjoint alternating paths and cycles.



$x, y$   $M$ -unsaturated

$|M'| > |M|$  implies that there is at least one path of type (d).  
Such a path is  $M$ -augmenting

# Undirected Vertex Geography

## Theorem

$(G, v)$  is an  $N$ -position in UVG iff every maximum matching of  $G$  covers  $v$ .

**Proof** (i) Suppose that  $M$  is a maximum matching of  $G$  which covers  $v$ . Player 1's strategy is now: Move along the  $M$ -edge that contains the current vertex.

If Player 1 were to lose, then there would exist a sequence of edges  $e_1, f_1, \dots, e_k, f_k$  such that  $v \in e_1$ ,  $e_1, e_2, \dots, e_k \in M$ ,  $f_1, f_2, \dots, f_k \notin M$  and  $f_k = (x, y)$  where  $y$  is the current vertex for Player 1 and  $y$  is not covered by  $M$ .

But then if  $A = \{e_1, e_2, \dots, e_k\}$  and  $B = \{f_1, f_2, \dots, f_k\}$  then  $(M \setminus A) \cup B$  is a maximum matching (same size as  $M$ ) which does not cover  $v$ , contradiction.



# Undirected Vertex Geography

(ii) Suppose now that there is some maximum matching  $M$  which does not cover  $v$ . If  $(v, w)$  is Player 1's move, then  $w$

must be covered by  $M$ , else  $M$  is not a maximum matching.

Player 2's strategy is now: Move along the  $M$ -edge that contains the current vertex. If Player 2 were to lose then there exists  $e_1 = (v, w), f_1, \dots, e_k, f_k, e_{k+1} = (x, y)$  where  $y$  is the current vertex for Player 2 and  $y$  is not covered by  $M$ .

But then we have defined an augmenting path from  $v$  to  $y$  and so  $M$  is not a maximum matching, contradiction.  $\square$

# Undirected Vertex Geography

Note that we can determine whether or not  $v$  is covered by all maximum matchings as follows: Find the size  $\sigma$  of the maximum matching  $G$ .

This can be done in  $O(n^3)$  time on an  $n$ -vertex graph. Find the size  $\sigma'$  of a maximum matching in  $G - v$ . Then  $v$  is covered by all maximum matchings of  $G$  iff  $\sigma \neq \sigma'$ .

# Tic Tac Toe

We consider the following multi-dimensional version of Tic Tac Toe (Noughts and Crosses to the English).

The *board* consists of  $[n]^d$ . A point on the board is therefore a vector  $(x_1, x_2, \dots, x_d)$  where  $1 \leq x_i \leq n$  for  $1 \leq i \leq d$ .

A *line* is a set points  $(x_j^{(1)}, x_j^{(2)}, \dots, x_j^{(d)})$ ,  $j = 1, 2, \dots, n$  where each sequence  $x^{(i)}$  is either (i) of the form  $k, k, \dots, k$  for some  $k \in [n]$  or is (ii)  $1, 2, \dots, n$  or is (iii)  $n, n-1, \dots, 1$ . Finally, we cannot have Case (i) for all  $i$ .

Thus in the (familiar)  $3 \times 3$  case, the top row is defined by  $x^{(1)} = 1, 1, 1$  and  $x^{(2)} = 1, 2, 3$  and the diagonal from the bottom left to the top right is defined by  $x^{(1)} = 3, 2, 1$  and  $x^{(2)} = 1, 2, 3$

## Lemma

The number of winning lines in the  $(n, d)$  game is  $\frac{(n+2)^d - n^d}{2}$ .

**Proof** In the definition of a line there are  $n$  choices for  $k$  in (i) and then (ii), (iii) make it up to  $n + 2$ . There are  $d$  independent choices for each  $i$  making  $(n + 2)^d$ .

Now delete  $n^d$  choices where only Case (i) is used. Then divide by 2 because replacing (ii) by (iii) and vice-versa whenever Case (i) does not hold produces the same set of points (traversing the line in the other direction). □

# Tic Tac Toe

The game is played by 2 players. The Red player (X player) goes first and colours a point red. Then the Blue player (O player) colours a different point blue and so on.

A player wins if there is a line, all of whose points are that player's colour. If neither player wins then the game is a draw. The second player does not have a winning strategy:

## Lemma

*Player 1 can always get at least a draw.*

**Proof** We prove this by considering *strategy stealing*.

Suppose that Player 2 did have a winning strategy. Then Player 1 can make an arbitrary first move  $x_1$ . Player 2 will then move with  $y_1$ . Player 1 will now win playing the winning strategy for Player 2 against a first move of  $y_1$ .

This can be carried out until the strategy calls for move  $x_1$  (if at all). But then Player 1 can make an arbitrary move and continue, since  $x_1$  has already been made. □

The Hales-Jewett Theorem of Ramsey Theory implies that there is a winner in the  $(n, d)$  game, when  $n$  is large enough with respect to  $d$ . The winner is of course Player 1.

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# Tic Tac Toe

$$\begin{bmatrix} 11 & 1 & 8 & 1 & 12 \\ 6 & 2 & 2 & 9 & 10 \\ 3 & 7 & * & 9 & 3 \\ 6 & 7 & 4 & 4 & 10 \\ 12 & 5 & 8 & 5 & 11 \end{bmatrix}$$

The above array gives a strategy for Player 2 in the  $5 \times 5$  game ( $d = 2, n = 5$ ).

For each of the 12 lines there is an associated pair of positions. If Player 1 chooses a position with a number  $i$ , then Player 2 responds by choosing the other cell with the number  $i$ .

This ensures that Player 1 cannot take line  $i$ . If Player 1 chooses the \* then Player 2 can choose any cell with an unused number.



# Tic Tac Toe

So, later in the game if Player 1 chooses a cell with  $j$  and Player 2 already has the other  $j$ , then Player 2 can choose an arbitrary cell.

Player 2's strategy is to ensure that after all cells have been chosen, he/she will have chosen one of the numbered cells associated with each line. This prevents Player 1 from taking a whole line. This is called a *pairing* strategy.

# Tic Tac Toe

We now generalise the game to the following: We have a family  $\mathcal{F} = A_1, A_2, \dots, A_N \subseteq A$ . A move consists of one player, taking an uncoloured member of  $A$  and giving it his colour.

A player wins if one of the sets  $A_i$  is completely coloured with his colour.

A pairing strategy is a collection of distinct elements  $X = \{x_1, x_2, \dots, x_{2N-1}, x_{2N}\}$  such that  $x_{2i-1}, x_{2i} \in A_i$  for  $i \geq 1$ .

This is called a *draw forcing pairing*. Player 2 responds to Player 1's choice of  $x_{2i+\delta}$ ,  $\delta = 0, 1$  by choosing  $x_{2i+3-\delta}$ . If Player 1 does not choose from  $X$ , then Player 2 can choose any uncoloured element of  $X$ .

# Tic Tac Toe

In this way, Player 2 avoids defeat, because at the end of the game Player 2 will have coloured at least one of each of the pairs  $x_{2i-1}, x_{2i}$  and so Player 1 cannot have completely coloured  $A_i$  for  $i = 1, 2, \dots, N$ .

## Theorem

If

$$\left| \bigcup_{A \in \mathcal{G}} A \right| \geq 2|\mathcal{G}| \quad \forall \mathcal{G} \subseteq \mathcal{F} \quad (20)$$

then there is a draw forcing pairing.

**Proof** We define a bipartite graph  $\Gamma$ .  $A$  will be one side of the bipartition and  $B = \{b_1, b_2, \dots, b_{2N}\}$ . Here  $b_{2i-1}$  and  $b_{2i}$  both represent  $A_i$  in the sense that if  $a \in A_i$  then there is an edge  $(a, b_{2i-1})$  and an edge  $(a, b_{2i})$ .

A draw forcing pairing corresponds to a complete matching of  $B$  into  $A$  and the condition (20) implies that Hall's condition is satisfied. □

## Corollary

*If  $|A_i| \geq n$  for  $i = 1, 2, \dots, n$  and every  $x \in A$  is contained in at most  $n/2$  sets of  $\mathcal{F}$  then there is a draw forcing pairing.*

**Proof** The degree of  $a \in A$  is at most  $2(n/2)$  in  $\Gamma$  and the degree of each  $b \in B$  is at least  $n$ . This implies (via Hall's condition) that there is a complete matching of  $B$  into  $A$ .  $\square$

# Tic Tac Toe

Consider Tic tac Toe when  $d = 2$ . If  $n$  is even then every array element is in at most 3 lines (one row, one column and at most one diagonal) and if  $n$  is odd then every array element is in at most 4 lines (one row, one column and at most two diagonals).

Thus there is a draw forcing pairing if  $n \geq 6$ ,  $n$  even and if  $n \geq 9$ ,  $n$  odd. (The cases  $n = 4, 7$  have been settled as draws.  $n = 7$  required the use of a computer to examine all possible strategies.)

# Tic Tac Toe

In general we have

## Lemma

*If  $n \geq 3^d - 1$  and  $n$  is odd or if  $n \geq 2^d - 1$  and  $n$  is even, then there is a draw forcing pairing of  $(n, d)$  Tic tac Toe.*

**Proof** We only have to estimate the number of lines through a fixed point  $\mathbf{c} = (c_1, c_2, \dots, c_d)$ .

If  $n$  is odd then to choose a line  $L$  through  $\mathbf{c}$  we specify, for each index  $i$  whether  $L$  is (i) constant on  $i$ , (ii) increasing on  $i$  or (iii) decreasing on  $i$ .

This gives  $3^d$  choices. Subtract 1 to avoid the all constant case and divide by 2 because each line gets counted twice this way.

# Tic Tac Toe

When  $n$  is even, we observe that once we have chosen in which positions  $L$  is constant,  $L$  is determined.

Suppose  $c_1 = x$  and 1 is not a fixed position. Then every other non-fixed position is  $x$  or  $n - x + 1$ . Assuming w.l.o.g. that  $x \leq n/2$  we see that  $x < n - x + 1$  and the positions with  $x$  increase together at the same time as the positions with  $n - x + 1$  decrease together.

Thus the number of lines through  $\mathbf{c}$  in this case is bounded by  $\sum_{i=0}^{d-1} \binom{d}{i} = 2^d - 1$ . □



# Quasi-probabilistic method

We now prove a theorem of Erdős and Selfridge.

## Theorem

If  $|A_i| \geq n$  for  $i \in [N]$  and  $N < 2^{n-1}$ , then Player 2 can get a draw in the game defined by  $\mathcal{F}$ .

**Proof** At any point in the game, let  $C_j$  denote the set of elements in  $A$  which have been coloured with Player  $j$ 's colour,  $j = 1, 2$  and  $U = A \setminus C_1 \cup C_2$ . Let

$$\Phi = \sum_{i: A_i \cap C_2 = \emptyset} 2^{-|A_i \cap U|}.$$

Suppose that the players choices are  $x_1, y_1, x_2, y_2, \dots$ . Then we observe that immediately after Player 1's first move,  $\Phi < N2^{-(n-1)} < 1$ .

# Quasi-probabilistic method

We will show that Player 2 can keep  $\Phi < 1$  through out. Then at the end, when  $U = \emptyset$ ,  $\Phi = \sum_{i:A_i \cap C_2 = \emptyset} 1 < 1$  implies that  $A_i \cap C_2 \neq \emptyset$  for all  $i \in [N]$ .

So, now let  $\Phi_j$  be the value of  $\Phi$  after the choice of  $x_1, y_1, \dots, x_j$ . then if  $U, C_1, C_2$  are defined at precisely this time,

$$\begin{aligned}\Phi_{j+1} - \Phi_j &= - \sum_{\substack{i:A_i \cap C_2 = \emptyset \\ y_j \in A_i}} 2^{-|A_i \cap U|} + \sum_{\substack{i:A_i \cap C_2 = \emptyset \\ y_j \notin A_i, x_{j+1} \in A_i}} 2^{-|A_i \cap U|} \\ &\leq - \sum_{\substack{i:A_i \cap C_2 = \emptyset \\ y_j \in A_i}} 2^{-|A_i \cap U|} + \sum_{\substack{i:A_i \cap C_2 = \emptyset \\ x_{j+1} \in A_i}} 2^{-|A_i \cap U|}\end{aligned}$$

# Quasi-probabilistic method

We deduce that  $\Phi_{j+1} - \Phi_j \leq 0$  if Player 2 chooses  $y_j$  to maximise  $\sum_{\substack{i: A_i \cap C_2 = \emptyset \\ y \in A_i}} 2^{-|A_i \cap U|}$  over  $y$ .

In this way, Player 2 keeps  $\Phi < 1$  and obtains a draw. □

In the case of  $(n, d)$  Tic Tac Toe, we see that Player 2 can force a draw if

$$\frac{(n+2)^d - n^d}{2} < 2^{n-1}$$

which is implied, for  $n$  large, by

$$n \geq (1 + \epsilon)d \log_2 d$$

where  $\epsilon > 0$  is a small positive constant.

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# Hereditary Families

Given a **Ground Set**  $E$ , a **Hereditary Family**  $\mathcal{A}$  on  $E$  is collection of subsets  $\mathcal{I} = \{I_1, I_2, \dots, I_m\}$  (the **independent sets**) such that

$$I \in \mathcal{I} \text{ and } J \subseteq I \text{ implies that } J \in \mathcal{I}.$$

- 1 The set  $\mathcal{M}$  of matchings of a graph  $G = (V, E)$ .
- 2 The set of (edge-sets of) forests of a graph  $G = (V, E)$ .
- 3 The set of **stable** sets of a graph  $G = (V, E)$ . We say that  $S$  is stable if it contains no edges.
- 4 If  $G = (A, B, E)$  is a bipartite graph and  $\mathcal{I} = \{S \subseteq B : \exists \text{ a matching } M \text{ that covers } S\}$ .
- 5 Let  $\mathbf{c}_1, \mathbf{c}_2, \dots, \mathbf{c}_n$  be the columns of an  $m \times n$  matrix  $\bar{A}$ . Then  $E = [n]$  and  $\mathcal{I} = \{S \subseteq [n] : \{\mathbf{c}_i, i \in S\} \text{ are linearly independent}\}$ .

# Matroids

An independence system is a **matroid** if whenever  $I, J \in \mathcal{I}$  with  $|J| = |I| + 1$  there exists  $e \in J \setminus I$  such that  $I \cup \{e\} \in \mathcal{I}$ . We call this the **Independent Augmentation Axiom – IAA**.

Matroid independence is a generalisation of linear independence in vector spaces. Only Examples 2,4 and 5 above are matroids.

To check Example 5, let  $\bar{A}_I$  be the  $m \times |I|$  sub-matrix of  $\bar{A}$  consisting of the columns in  $I$ . If there is no  $e \in J \setminus I$  such that  $I \cup \{e\} \in \mathcal{I}$  then  $\bar{A}_J = \bar{A}_I \mathbf{M}$  for some  $|I| \times |J|$  matrix  $\mathbf{M}$ .

Matrix  $\mathbf{M}$  has more columns than rows and so there exists  $\mathbf{x} \neq \mathbf{0}$  such that  $\mathbf{M}\mathbf{x} = \mathbf{0}$ . But then  $\bar{A}_J \mathbf{x} = \mathbf{0}$ , implying that the columns of  $\bar{A}_J$  are linearly dependent. Contradiction.

These are called **Representable Matroids**.

# Cycle Matroids/Graphic Matroids

To check Example 2 we define the vertex-edge incidence matrix  $\bar{A}_G$  of graph  $G = (V, E)$  over  $GF_2$ .

$\bar{A}_G$  has a row for each vertex  $v \in V$  and a column for each edge  $e \in E$ . There is a 1 in row  $v$ , column  $e$  iff  $v \in e$ .

We verify that a set of columns  $\mathbf{c}_i, i \in I$  are linearly dependent iff the corresponding edges contain a cycle.

If the edges contain a cycle  $(v_1, v_2, \dots, v_k, v_1)$  then the sum of the columns corresponding to the vertices of the cycle is  $\mathbf{0}$ .

To show that a forest  $F$  defines a linearly independent set of columns  $I_F$ , we use induction on the number of edges in the forest. This is trivial if  $|E(F)| = 1$ .

# Cycle Matroids/Graphic Matroids

Let  $\bar{A}_F$  denote the submatrix of  $\bar{A}$  made up of the columns corresponding  $F$ .

Now a forest  $F$  must contain a vertex  $v$  of degree one. This means that the row corresponding to  $v$  in  $\bar{A}_F$  has a single one, in column  $e$  say.

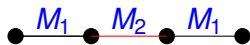
Consider the forest  $F' = F \setminus \{e\}$ . Its corresponding columns  $I_{F'}$  are linearly independent, by induction. Adding back  $e$  adds a row with a single one and preserves independence. Let  $\mathbf{B}$  denote  $\bar{A}_{F'}$  minus row  $e$ .

$$\bar{A}_F = \begin{bmatrix} 1 & \mathbf{0} \\ & \mathbf{B} \end{bmatrix}.$$



# Transversal Matroids

We now check Example 4. These are called **Transversal Matroids**. If  $M_1, M_2$  are two matchings in a graph  $G$  then  $M_1 \oplus M_2 = (M_1 \setminus M_2) \cup (M_2 \setminus M_1)$  consists of **alternating paths and cycles**.



Suppose now that we have two matchings  $M_1, M_2$  in bipartite graph  $G = (A, B, E)$ . Let  $I_j, j = 1, 2$  be the vertices in  $B$  covered by  $M_j$ . Suppose that  $|I_1| > |I_2|$ .

Then  $M_1 \oplus M_2$  must contain an alternating path  $P$  with end points  $b \in I_1 \setminus I_2, a \in A$ . Let  $E_1$  be the  $M_1$  edges in  $P$  and let  $E_2$  be the  $M_2$  edges of  $P$ . Then  $(M_1 \cup E_1) \setminus E_2$  is a matching that covers  $I_1 \cup \{b\}$ .

# Representable Matroids

A matroid is **binary** if it is representable by a matrix over  $GF_2$ .

So a graphic matroid is binary.

A matroid is **regular** if it can be represented by a matrix of elements in  $\{0, \pm 1\}$  for which every square sub-matrix has determinant  $0, \pm 1$ . These are called **totally unimodular matrices**

A matrix with 2 non-zeros in each column, one equal to +1 and the other equal to -1 is totally unimodular. This implies that graphic matroids are regular. (Take the vertex-edge incidence matrix and replace one of the ones in each column by a -1.)

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## Theorem

A collection  $\mathcal{B} = \{B_1, B_2, \dots, B_m\}$  of subsets of  $E$  form the bases of a matroid on  $E$  iff for all  $i, j$  and  $e \in B_i \setminus B_j$  there exists  $f \in B_j \setminus B_i$  such that  $(B_i \cup \{f\}) \setminus \{e\} \in \mathcal{B}$ .

**Proof:** Suppose first that  $\mathcal{B}$  are the bases of a matroid with independent sets  $\mathcal{I}$  and that  $e \in B_i$  and  $e \notin B_j$ . Then  $B'_i = B_i \setminus \{e\} \in \mathcal{I}$  and  $|B'_i| < |B_j|$ . So there exists  $f \in B_j \setminus B'_i$  such that  $B''_i = B'_i \cup \{f\} \in \mathcal{I}$ . Now  $f \neq e$  since  $e \notin B_j$  and  $|B''_i| = |B_i|$ . So  $B''_i$  must be a basis.

# Rank

If  $S \subseteq E$  then its **rank**

$$r(S) = \max |\{I \in \mathcal{I} : I \subseteq S\}|.$$

So  $S \in \mathcal{I}$  iff  $r(S) = |S|$ . We show next that  $r$  is **submodular**.

## Theorem

If  $S, T \subseteq E$  then  $r(S \cup T) + r(S \cap T) \leq r(S) + r(T)$ .

**Proof:** Let  $I_1$  be a maximal independent subset of  $S \cap T$  and let  $I_2$  be a maximal independent subset of  $S \cup T$  that contains  $I_1$ . (Such a set exists because of the IAA.)

But then

$$r(S \cap T) + r(S \cup T) = |I_1| + |I_2| = |I_2 \cap S| + |I_2 \cap T| \leq r(S) + r(T).$$

# Rank

For representable matroids this corresponds to the usual definition of rank.

For the cycle matroid of graph  $G = (V, E)$ , if  $S \subseteq E$  is a set of edges and  $G_S$  is the graph  $(V, S)$  then  $r(S) = |V| - \kappa(G_S)$ , where  $\kappa(G_S)$  is the number of components of  $G_S$ .

This clearly true for connected graphs and so if  $C_1, C_2, \dots, C_s$  are the components of  $G_S$  then  $r(S) = \sum_{i=1}^s |C_i| - 1 = |V| - s$ .

For a partition matroid as defined above,

$$r(S) = \sum_{i=1}^m \min\{k_i, |S \cap E_i|\}.$$

# Circuits

A **circuit** of a matroid  $\mathcal{M}$  is a minimal dependent set. If a set  $S \subseteq E, S \notin \mathcal{I}$  then  $S$  contains a circuit.

So the circuits of the cycle matroid of a graph  $G$  are the cycles.

## Theorem

*If  $C_1, C_2$  are circuits of  $\mathcal{M}$  and  $e \in C_1 \cap C_2$  then there is a circuit  $C \subseteq (C_1 \cup C_2) \setminus \{e\}$ .*

**Proof:** We have  $r(C_i) = |C_i| - 1, i = 1, 2$ . Also,  $r(C_1 \cap C_2) = |C_1 \cap C_2|$  since  $C_1 \cap C_2$  is a proper subgraph of  $C_1$ .

If  $C' = (C_1 \cup C_2) \setminus \{e\}$  contains no circuit then  $r(C_1 \cup C_2) \geq r(C') = |C_1 \cup C_2| - 1$ . But then

$$\begin{aligned} |C_1 \cup C_2| - 1 &\leq r(C_1 \cup C_2) \leq r(C_1) + r(C_2) - r(C_1 \cap C_2) \\ &= (|C_1| - 1) + (|C_2| - 1) - |C_1 \cap C_2|. \end{aligned}$$

Contradiction.

## Theorem

If  $B$  is a basis of  $\mathcal{M}$  and  $e \in E \setminus B$  then  $B' = B \cup \{e\}$  contains a unique circuit  $C(e, B)$ . Furthermore, if  $f \in C(e, B)$  then  $(B \cup \{e\}) \setminus \{f\}$  is also a basis of  $\mathcal{M}$ .

**Proof:**  $B' \notin \mathcal{I}$  because  $B$  is maximal. So  $B'$  must contain at least one circuit.

Suppose it contains distinct circuits  $C_1, C_2$ . Then  $e \in C_1 \cap C_2$  and so  $B'$  contains a circuit  $C_3 \subseteq (C_1 \cup C_2) \setminus \{e\}$ .

But then  $C_3 \subseteq B$ , contradiction. □



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# Dual Matroid

## Theorem

If  $\mathcal{B}$  denotes the set of bases of a matroid  $\mathcal{M}$  on ground set  $E$  then  $\mathcal{B}^* = \{E \setminus B : B \in \mathcal{B}\}$  is the set of bases of a matroid  $\mathcal{M}^*$ , the dual matroid.

**Proof:** Suppose that  $B_1^*, B_2^* \in \mathcal{B}^*$  and  $e \in B_1^* \setminus B_2^*$ .

Let  $B_i = E \setminus B_i^*, i = 1, 2$ . Then  $e \in B_2 \setminus B_1$ .

So there exists  $f \in B_1 \setminus B_2$  such that  $(B_2 \cup \{e\}) \setminus \{f\} \in \mathcal{B}$ .

This implies that  $(B_2^* \cup \{f\}) \setminus \{e\} \in \mathcal{B}^*$ . □

# Greedy Algorithm

Suppose that each  $e \in E$  is given a weight  $w_e$  and that the weight  $w(I)$  of an independent set  $I$  is given by  $w(I) = \sum_{e \in I} c_e$ . The problem we discuss is

Maximize  $w(I)$  subject to  $I \in \mathcal{I}$ .

**Greedy Algorithm:**

**begin**

Sort  $E = \{e_1, e_2, \dots, e_m\}$  so  $w(e_i) \geq w(e_{i+1})$  for  $1 \leq i < m$ ;

$S \leftarrow \emptyset$ ;

**for**  $i = 1, 2, \dots, m$ ;

**begin**

**if**  $S \cup \{e_i\} \in \mathcal{I}$  **then**;

**begin**;

$S \leftarrow S \cup \{e_i\}$ ;

**end**;

**end**;

**end**

# Greedy Algorithm

## Theorem

*The greedy algorithm finds a maximum weight independent set for all choices of  $w$  if and only if it is a matroid.*

Suppose first that the Greedy Algorithm always finds a maximum weight independent set. Suppose that  $\emptyset \neq I, J \in \mathcal{I}$  with  $|J| = |I| + 1$ . Define

$$w(e) = \begin{cases} 1 + \frac{1}{2|I|} & e \in I. \\ 1 & e \in J \setminus I. \\ 0 & e \notin I \cup J. \end{cases}$$

If there does not exist  $e \in J \setminus I$  such that  $I \cup \{e\} \in \mathcal{I}$  then the Greedy Algorithm will choose the elements of  $I$  and stop. But  $I$  does not have maximum weight. Its weight is  $|I| + 1/2 < |J|$ . So if Greedy succeeds, then the IAA holds.

# Greedy Algorithm

Conversely, suppose that our independence system is a matroid. We can assume that  $w(e) > 0$  for all  $e \in E$ . Otherwise we can restrict ourselves to the matroid defined by  $\mathcal{I}' = \{I \subseteq E^+\}$  where  $E^+ = \{e \in E : w(e) > 0\}$ .

Suppose now that Greedy chooses  $I_G = e_{i_1}, e_{i_2}, \dots, e_{i_k}$  where  $i_t < i_{t+1}$  for  $1 \leq t < k$ . Let  $I = e_{j_1}, e_{j_2}, \dots, e_{j_\ell}$  be any other independent set and assume that  $j_t < j_{t+1}$  for  $1 \leq t < \ell$ . We can assume that  $\ell \geq k$ , for otherwise we can add something from  $I_G$  to  $I$  to give it larger weight.

We show next that  $k = \ell$  and that  $i_t \leq j_t$  for  $1 \leq t \leq k$ . This implies that  $w(I_G) \geq w(I)$ .

# Greedy Algorithm

Suppose then that there exists  $t$  such that  $i_t > j_t$  and let  $t$  be as small as possible for this to be true.

Now consider  $I = \{e_{i_s} : s = 1, 2, \dots, t-1\}$  and  $J = \{e_{j_s} : s = 1, 2, \dots, t\}$ . Now there exists  $e_{j_s} \in J \setminus I$  such that  $I \cup \{e_{j_s}\} \in \mathcal{I}$ .

But  $j_s \leq j_t < i_t$  and Greedy should have chosen  $e_{j_s}$  before choosing  $e_{i_{t+1}}$ .

Also,  $i_k \leq j_k$  implies that  $k = \ell$ . Otherwise Greedy can find another element from  $I \setminus I_G$  to add.

# Minors

Given a graph  $G = (V, E)$  and an edge  $e$  we can get new graphs by **deleting**  $e$  or **contracting**  $e$ .

We describe a corresponding notion for matroids. Suppose that  $F \subseteq E$  then we define the matroid  $\mathcal{M}_{\setminus F}$  with independent sets  $\mathcal{I}_{\setminus F}$  obtained by **deleting**  $F$ :  $I \in \mathcal{I}_{\setminus F}$  if  $I \in \mathcal{I}$ ,  $I \cap F = \emptyset$ .

It is clear that the IAA holds for  $\mathcal{M}_{\setminus F}$  and so it is a matroid.

For **contraction** we will assume that  $F \in \mathcal{I}$ . Then contracting  $F$  defines  $\mathcal{M}_{/F}$  with independent sets  $\mathcal{I}_{/F} = \{I \in \mathcal{I} : I \cap F = \emptyset, I \cup F \in \mathcal{I}\}$ .

We argue next that  $\mathcal{M}_{/F}$  is also a matroid.

## Lemma

$$\mathcal{M}.F = (\mathcal{M}_{\setminus F}^*)^* \text{ and } \mathcal{M}_{\setminus F} = (\mathcal{M}^*.F)^* .$$

**Proof:**

$$\begin{aligned} I \in \mathcal{I}.F &\leftrightarrow \exists B \in \mathcal{B}_{\setminus F}, I \subseteq B \\ &\leftrightarrow \exists B^* \in \mathcal{B}_{\setminus F}^*, I \cap B^* = \emptyset \\ &\leftrightarrow I \in (\mathcal{I}_{\setminus F}^*)^* . \end{aligned}$$

For the second claim we use

$$\mathcal{M}^*.F = (\mathcal{M}_{\setminus F}^{**})^* = (\mathcal{M}_{\setminus F})^* .$$





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# Matroid Intersection

Suppose we are given two matroids  $\mathcal{M}_1, \mathcal{M}_2$  on the same ground set  $E$  with  $\mathcal{I}_1, \mathcal{I}_2$  and  $r_1, r_2$  etc. having their obvious meaning.

An **intersection** is a set  $I \in \mathcal{I}_1 \cap \mathcal{I}_2$ . We give a min-max relation for the size of the largest independent intersection. Let  $\mathcal{J}$  denote the set of intersections.

Theorem (Edmonds)

$$\max\{|J| : J \in \mathcal{J}\} = \min\{r_1(A) + r_2(E \setminus A) : A \subseteq E\}.$$

# Matroid Intersection

Before proving the theorem let us see a couple of applications:

Hall's Theorem: suppose we are given a bipartite graph  $G = (A, B, E)$ . Let  $\mathcal{M}_A, \mathcal{M}_B$  be the following two partition matroids.

For  $\mathcal{M}_A$  we define the partition  $E_a = \{e \in E : a \in e\}$ ,  $a \in A$ . We let  $k_a = 1$  for  $a \in A$ . We define  $\mathcal{M}_B$  similarly.

Intersections correspond to matchings and  $r_1(A)$  is the number of vertices in  $A$  that are incident with an edge of  $E$ . Similarly  $r_2(E \setminus A)$  is the number of vertices in  $B$  that are incident with an edge not in  $A$ .

# Matroid Intersection

For  $X \subseteq A$ , let

$$A_X = \{v \in A : v \in e \text{ for some } e \in X\}.$$

Define  $B_X$  similarly.

So

$$\max\{|M|\} = \min\{|A_X| + |B_{E \setminus X}| : X \subseteq E\}.$$

Now we can assume that if  $e \in E \setminus X$  then  $e \cap A_X = \emptyset$ , otherwise moving  $e$  to  $X$  does not increase the RHS of the above.

Let  $S = A \setminus A_X$ . Then  $|B_{E \setminus X}| = |N(A)|$  and so

$$\max\{|M|\} = \min\{|A| - |S| + |N(S)| : S \subseteq A\}.$$

# Matroid Intersection

**Rainbow Spanning Trees:** we are given a connected graph  $G = (V, E)$  where each edge  $e \in E$  is given a color  $c(e) \in [m]$  where  $m \geq n - 1$ . Let  $E_i = \{e : c(e) = i\}$  for  $i \in [m]$ .

A set of edges  $S$  is said to be **rainbow colored** if  $e, f \in S$  implies that  $c(e) \neq c(f)$ .

For a set  $A \subseteq E$ , we let

$$r_1(A) = c(A) = |\{i \in [m] : \exists e \in A \text{ s.t. } c(e) = i\}|$$
$$r_2(E \setminus A) = n - \kappa(G \setminus A).$$

So,  $G$  contains a rainbow spanning tree iff

$$c(A) + (n - \kappa(G \setminus A)) \geq n - 1 \text{ for all } A \subseteq E. \quad (21)$$

# Matroid Intersection

We simplify (21) to obtain

$$c(A) + 1 \geq \kappa(G \setminus A). \quad (22)$$

We can then further simplify (22) as follows: if we add to  $A$  all edges that use a color used by some edge of  $A$  then we do not change  $c(A)$  but we do not decrease  $\kappa(G \setminus A)$ .

Thus we can restrict our sets  $A$  to  $E_I = \bigcup_{i \in I} E_i$  for some  $I \subseteq [m]$ . Then (22) becomes

$$\kappa(E_{[m] \setminus I}) \leq |I| + 1 \text{ for all } I \subseteq [m]$$

or

$$\kappa(E_I) \leq m - |I| + 1 \text{ for all } I \subseteq [m]$$

If you think for a moment, you will see that this is obviously necessary.

# Matroid Intersection

Proof of the matroid intersection theorem.

For the upper bound consider  $J \in \mathcal{J}$  and  $A \subseteq E$ . Then

$$|J| = |J \cap A| + |J \setminus A| \leq r_1(A) + r_2(E \setminus A).$$

We assume that  $e \in \mathcal{J}$  for all  $e \in E$ . (Loops can be “ignored”.)

We proceed by induction on  $|E|$ . Let

$$k = \min\{r_1(A) + r_2(E \setminus A) : A \subseteq E\}.$$

Suppose that  $|J| < k$  for all  $J \in \mathcal{J}$ .

# Matroid Intersection

Then  $(\mathcal{M}_1) \setminus \{e\}$  and  $(\mathcal{M}_2) \setminus \{e\}$  have no common independent set of size  $k$ . This implies that if  $F = E \setminus \{e\}$  then

$$r_1(A) + r_2(F \setminus A) \leq k - 1 \text{ for some } A \subseteq F.$$

Similarly,  $\mathcal{M}_1 \cdot \{e\}$  and  $\mathcal{M}_2 \cdot \{e\}$  have no common independent set of size  $k - 1$ . This implies that

$$r_1(B) - 1 + r_2(E \setminus (B \setminus \{e\})) - 1 \leq k - 2 \text{ for some } e \in B \subseteq E.$$

This gives

$$r_1(A) + r_2(E \setminus (A \cup \{e\})) + r_1(B) + r_2(E \setminus (B \setminus \{e\})) \leq 2k - 1.$$



# Matroid Intersection

So, using submodularity and

$$(E \setminus (A \cup \{e\})) \cup (E \setminus (B \setminus \{e\})) = E \setminus (A \cap B)$$

and

$$(E \setminus (A \cup \{e\})) \cap (E \setminus (B \setminus \{e\})) = E \setminus (A \cup B).$$

We have used  $e \notin A$  and  $e \in B$  here. So,

$$r_1(A \cup B) + r_2(E \setminus (A \cup B)) + r_1(A \cap B) + r_2(E \setminus (A \cap B)) \leq 2k - 1.$$

But, by assumption,

$$r_1(A \cup B) + r_2(E \setminus (A \cup B)) \geq k, \quad r_1(A \cap B) + r_2(E \setminus (A \cap B)) \geq k,$$

contradiction.