

# Chap 1 Power of randomization

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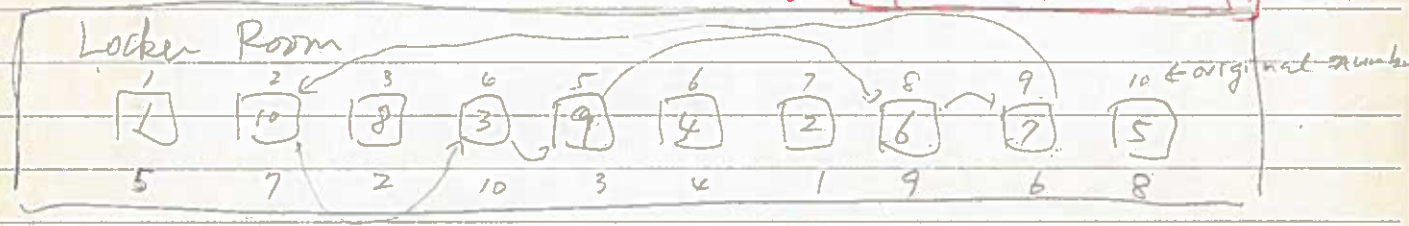
## Chap 6 1.1 The Locker Game

Power of Randomization  
only when used appropriately

P  
P ⇒ V  
A hiding info.

### Examp 1 The Locker Game (Puzzle)

2 → 8 → 5 → 1  
3: 9 → 6 → 7 → 10 → 3 4 → 4



name 1 ~ 10

Team Game A team of 10 people + Reference

- Each locker has a name of a player
- Each player enters the room one by one
- He/she can open at most 5 (1/2 of the whole boxes) lockers
- If he/she finds her own name, she wins
- All (10) players win! ... successful ⇒ the team wins

using some alg.

② The team can discuss its strategy before the game, but once it started, no communication is allowed

notes, computers, ...

otherwise, player 1 opens the first 5 lockers, ...

our evaluation is for the worst case.

Strategy 1 All players open the first 5 lockers

Strategy 2 Even (odd) number players open even (odd) number lockers. worst-case

Not deterministic strategies work.

Strategy 3 Each player opens 5 lockers at random

$$\left(\frac{1}{2}\right)^{10} = 1/4024 > 0$$

randomization works

but # of lockers increases, then ...

# Strategy 4. "Shared randomness"

A random permutation, say 5, 7, 2, 10, 3, 4, 1, 9, 6, 8, is shared by all the players. Then --

## Analysis of the winning (losing) probability

Shared randomness  $\Rightarrow$



Names inside  $\equiv$  a random permutation of 1-10.

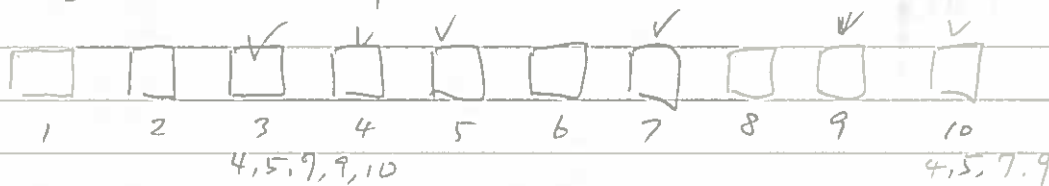
The players lose if  $\exists$  a cycle with length 6 or more

Let's calculate its probability or

# of permutations having such a cycle  
 $L(6), L(7), \dots, L(10)$

Case 1: length 6

To fix one such permutation



(1) Fix 6 positions  $\binom{10}{6}$

(2) Fix the order of the cycle  $5 \times 4 \times \dots \times 1$

(3) The other 4 numbers  $4!$

$$L(6) = \binom{10}{6} \cdot 5! \cdot 4! = \frac{10! \cdot 5! \cdot 4!}{6! \cdot 4!} = 10! / 6$$

Case 2 Length 7  $L(7) = 10! / 7 \dots$

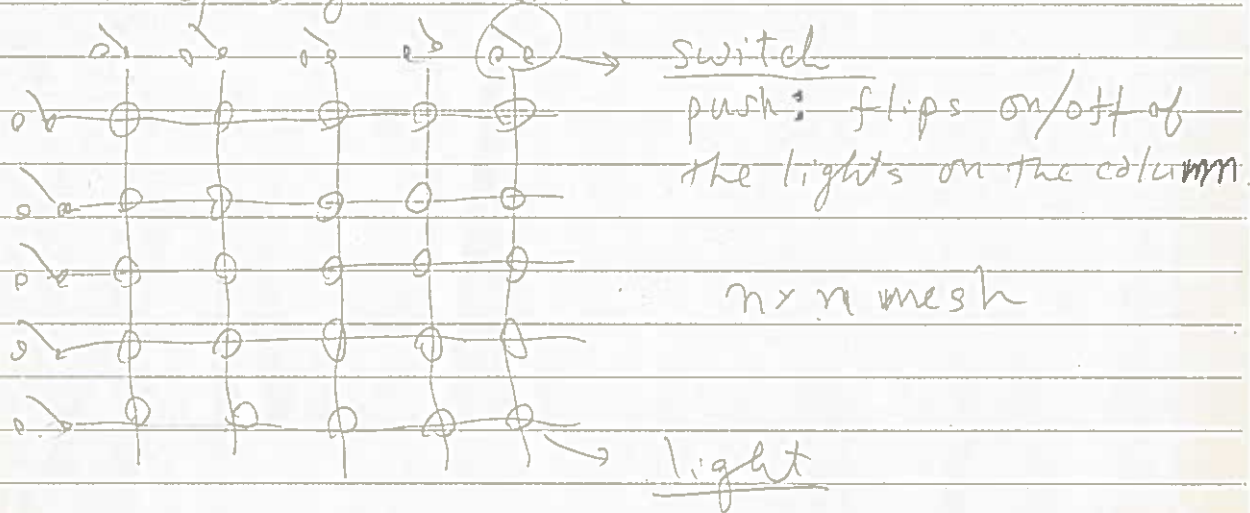
The losing prob =  $\frac{1}{6} + \frac{1}{7} + \dots + \frac{1}{10} \approx 0.645$

The winning prob.  $\geq 0.35$  (random choice  $1/1024$  for a simple)

For a general  $2n$  lockers  $\frac{1}{n+1} + \dots + \frac{1}{2n} \approx \int_n^{2n} \frac{1}{x} dx = \ln 2 < 0.7$

ⓐ proper use of randomness

## 1.2 Unbalancing Lights Game



Game? 100x100 mesh  
prize money (# of on lights - # of off lights)  
Game charge: 1,000 CZK  $\times 10$  CZK  
you can manipulate the switches for 10 minutes

ⓐ Initial settings are important.

Thm 1.1 For any  $a_{ij} = \pm 1$  ( $1 \leq i, j \leq n$ ), we can select the values of  $x_i, y_j$  ( $= \pm 1$ ) such that

$$\sum_{i,j} a_{ij} x_i y_j \geq \left( \sqrt{\frac{1}{2\pi}} n^{\frac{3}{2}} \right) \quad (\approx 4,000 \text{ if } n=100)$$

$n \approx 2.4$

$a_{ij}$ : initial on/off of the lights  $+1$ : on  $-1$ : off

$x_i, y_j$ :  $-1$ : push the switch

$+1$ : does not

(proof) consider the following randomized alg.

Select  $y_1, \dots, y_m$  ( $\pm 1$ ) at random  $x_i$   $\begin{matrix} y_1, y_2, \dots, y_m \\ \boxed{0 \ 0 \ \dots \ 0} \\ \pm 1 \ \ " \ \ " \end{matrix}$

$$\text{Let } R_i = \sum_{j=1}^m a_{ij} y_j \quad R = \sum_{i=1}^n |R_i|$$

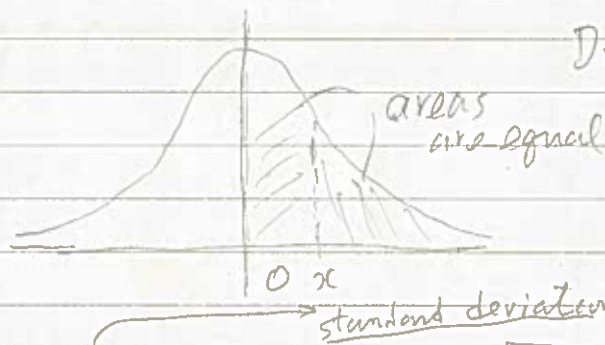
$\hookrightarrow$  sum of the  $\underbrace{\quad}_m$  values ( $\pm 1$  at random) in row  $i$

$= m$  values of fair coin

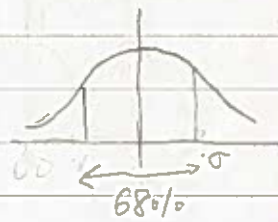
(head = +1, tail = -1)



It follows the binomial distribution



Distribution of  $|R_i|$



mean deviation

$$E[|R_i|] \doteq \sqrt{\frac{2}{\pi}} \sigma = \sqrt{\frac{2}{\pi}} \sqrt{m p q} = \sqrt{\frac{1}{2\pi}} \sqrt{m}$$

$$E[R] = \sum_i E[|R_i|] = \sqrt{\frac{1}{2\pi}} \sqrt{m} \cdot M$$

$\Rightarrow \exists$  selection of  $y_1, \dots, y_m$  that achieves

$$|R_i| \geq \sqrt{\frac{1}{2\pi}} \sqrt{m}$$

Now set  $x_i = -1$  iff  $R_i < 0$ .

Note that we can use this "algorithm" whose success probability is very high

$\Rightarrow$  play it!

### 1.3 Probabilistic Methods

Note that Th. 1.1. has nothing to do with probability

The probability that an event  $E$  happens  $> 0$   
 $\Rightarrow E$  can actually happen.

The " " " " does not  $< 1.0$

$\Rightarrow E$  can actually happen

Usually we design a probabilistic algorithm  $A$   
that makes  $E$  happen (if lucky)

If the success prob of  $A > 0$ , ---

If the failure " "  $A < 1.0$ , ---

Namely, we have more than 0.35 as the winning prob.

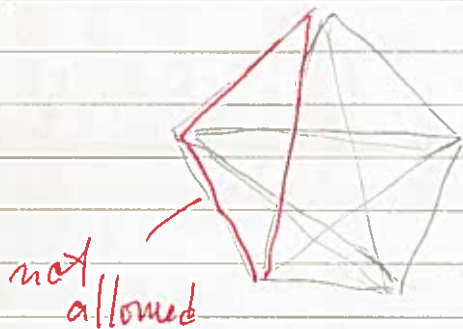
If we have more lockers,  $\frac{1}{n} + \frac{1}{n+1} + \dots + \frac{1}{2n} \approx \int_n^{2n} \frac{1}{x} dx = \ln 2 \approx 0.7$

We still have more than 0.3.

Example 2

① Ramsey Number

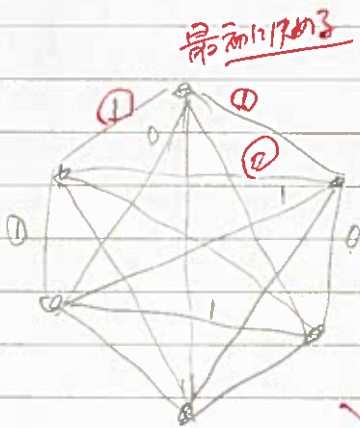
② Edge Coloring



$K_5$ : a complete graph with 5 vertices

Color each edge <sup>in</sup> with "blue" or "red", so that no triangle of the same color appears.

Possible? Yes!



$K_6$ : Do the same thing.

Possible? NO!

Prove this

↳ Then obviously NO for  $K_7, \dots$

$R(k, k)$ : The minimum  $n$  such that for any coloring of  $K_n$ , there must exist a single-colored  $K_k$ .   
 ↳ no successful coloring exists

$R(3, 3) = 6$

Triangle =  $K_3$

$n=5 \rightarrow$  successful coloring  
 $n=6 \neq$  " "

↳ 6 people  $\exists$  3 people ~~mutually~~ all know each other or mutually unknown no one knows each other

$$R(6,6) > 8$$

$K_6$  can be colored without monochromatic  $K_6$

No.

nothing to do with probability, but ...  
"Another power of randomization"

Thm 1.2  $R(k,k) > 2^{k/2}$  for all  $k \geq 3$

We prove:  $n = 2^{k/2} \Rightarrow \exists$  successful coloring (Goal: avoid a single-colored  $K_k$ )

Proof

Consider the following simple algorithm: and its analysis

• Just color each edge <sup>m</sup> with red or blue uniformly at random of  $K_m$

$\Rightarrow$  Compute the probability  $P$  that this algorithm fails, i.e., that a single-colored  $K_k$  appears

$\Rightarrow$  It turns out that  $P < 1.0$  if  $n = 2^{k/2}$  if  $n = 2^{k/2}$

$\Rightarrow E =$  "A single-colored  $K_k$  appears by random coloring" with probability  $P$

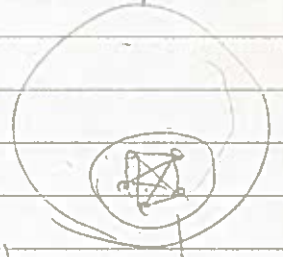
$\Rightarrow$  The probability that  $E$  happens is  $< 1.0$  coloring /  $\exists$  coloring  $E$  does not happen

① Computing probability  $P$ .

After the random coloring, the probability that a fixed  $K_k$  is single-colored is

$$\left(\frac{1}{2}\right)^{\binom{k}{2}} \cdot 2 = 2^{1 - \binom{k}{2}}$$

(red or blue)



The # of ways of selecting  $K_k$  is  $\binom{n}{k}$ , arbitrary  $K_k$

so the prob. that at least one single-colored  $K_k$  appears is at most  $\binom{n}{k} 2^{1 - \binom{k}{2}}$

$$\Rightarrow P \leq \binom{n}{k} 2^{1 - \binom{k}{2}}$$

② This value  $< 1$  if  $n = 2^{k/2}$

Union bound event  $A_1, \dots, A_m$

$$P(A_1 \cup \dots \cup A_m) \leq P(A_1) + \dots + P(A_m)$$

$$1 - \binom{k}{2} = 1 - \frac{k(k-1)}{2} = -\frac{k^2}{2} + \frac{k}{2} + 1$$

$$\begin{aligned} \binom{n}{k} 2^{1 - \binom{k}{2}} &= \frac{n!}{k!(n-k)!} \cdot 2^{-\frac{k^2}{2} + \frac{1}{2}(k+2)} \\ &\leq \frac{n^k}{k!} \cdot 2^{-\frac{k^2}{2} + \frac{1}{2}(k+2)} \\ &= \frac{n^k}{k!} \cdot \frac{2^{\frac{k}{2}}}{2^{\frac{k^2}{2}}} \end{aligned}$$

 $\uparrow$ 
 $\uparrow$ 
if  $n = 2^{\frac{k}{2}}$ 

1

1

 $\ll 1.0$ 

//

In fact  $P \ll 1.0$  if  $k$  is large, namely the algorithm succeeds with high probability, i.e., it is a good algorithm.



Define new random variables  $X_i(\sigma)$

$$X_i(\sigma) = \begin{cases} 1 & \text{if } \sigma(i) = i \\ 0 & \text{otherwise} \end{cases}$$

~~For~~  $(2, 1, 3, 4, 5)$   $X_1(\sigma) = X_2(\sigma) = 0, X_3(\sigma) = X_4(\sigma) = X_5(\sigma) = 1$

Obviously  $X(\sigma) = X_1(\sigma) + X_2(\sigma) + \dots + X_n(\sigma)$ .

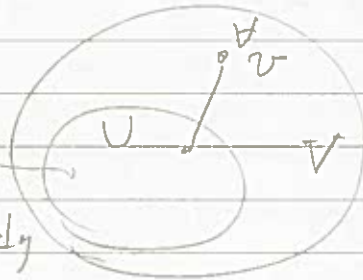
Also,  $X_i(\sigma) = 1$  with prob  $1/n$ .  $\overbrace{1 \dots n}^{d_i}$

$$\begin{aligned} \text{So, } E[X(\sigma)] &= E[X_1(\sigma)] + E[X_2(\sigma)] + \dots + E[X_n(\sigma)] \\ &= 1/n + \dots + 1/n = 1. \end{aligned}$$

① Dominating set of a graph  $G = (V, E)$

$U \subseteq V$  is a dominating set iff for any  $v \in V - U$ , there is a vertex  $u \in U$  that is adjacent to  $v$ .

Set of representatives  
s.t. everybody in  $V$   
is known by somebody  
in  $U$ .

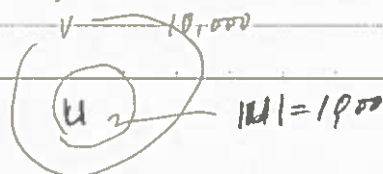


Thm 1.3 Suppose that  $G = (V, E)$  has a minimum degree of  $\delta$  ( $\geq 1$ ). Then there exists a dominating set whose size is at most  $n[1 + \ln(\delta + 1)] / (\delta + 1)$

nothing to do with probability

$n = 10,000, \delta = 100 \Rightarrow \text{this value} \approx 455$

Proof Trick: not evaluating the value directly.



Consider a random vertex set of  $n \frac{\ln(\delta+1)}{\delta+1}$

Let  $p = \ln(\delta+1)/(\delta+1)$ .

For each  $v \in V$ , put it into  $X$  with prob.  $p$ .

Let  $Y$  be a set of vertices  $v \in V - X$  such that  $v$  is not adjacent to any vertex  $u$  in  $X$ . (\*)

$$E[|X|] = np$$

The prob. that a vertex  $v \in V$  is in  $Y$   
 $= \Pr[\text{neither } v \text{ nor its neighbors comes into } X]$   
 $\leq (1-p)^{\delta+1}$

↓

$$E[|Y|] \leq n(1-p)^{\delta+1}$$

$$E[|X| + |Y|] \leq np + n(1-p)^{\delta+1}$$

$$= n \frac{\ln(\delta+1)}{\delta+1} + n \left(1 - \frac{\ln(\delta+1)}{\delta+1}\right)^{\delta+1}$$

$$\leq n \frac{1 + \ln(\delta+1)}{\delta+1}$$

$$\rightarrow \left(1 - \frac{\ln k}{k}\right)^k \leq \frac{1}{k}$$

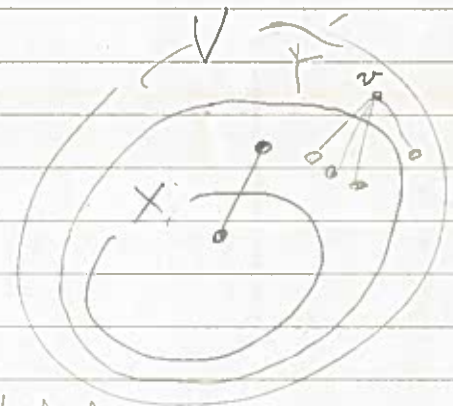
using  $\ln(1-x) \leq -x$

This means there exist

$X$  and  $Y$  such that  $|X| + |Y| \leq n \frac{1 + \ln(\delta+1)}{\delta+1}$  they satisfy (\*)

and it is easy to obtain such ones.

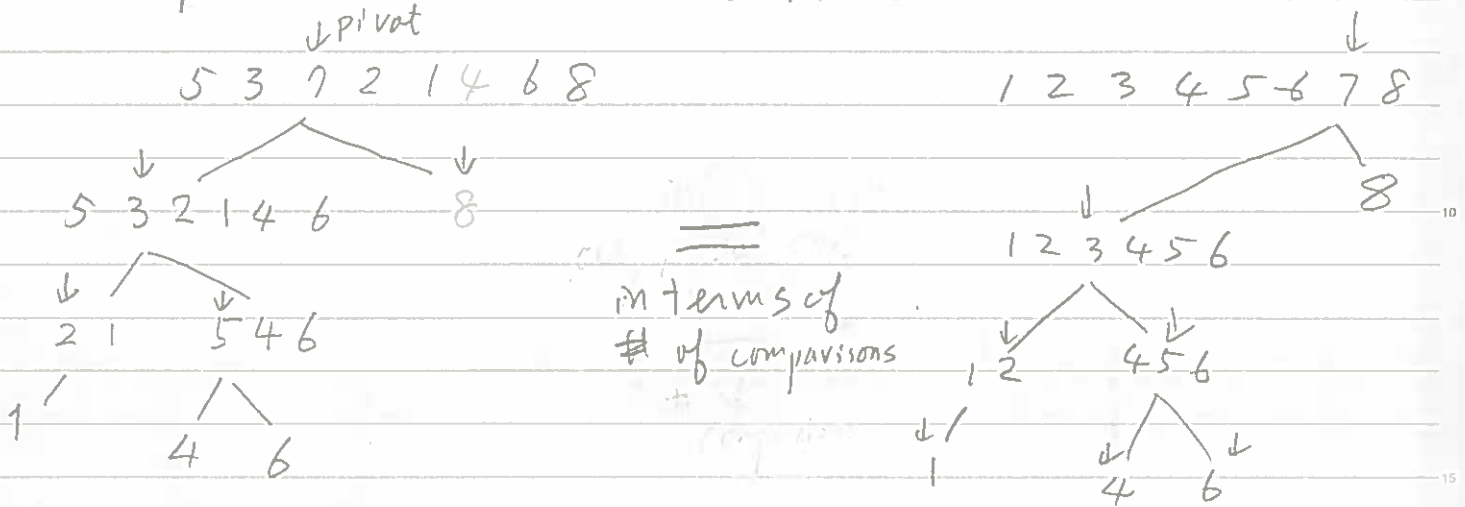
Note that  $X \cup Y$  is a dominating set. //



1.4 How to prove the success prob. of randomized Algorithm

@ Randomized algs are usually VERY simple but not easy to prove their performances

Example 1 : Randomized Quick Sort



@ Complexity = # of comparisons

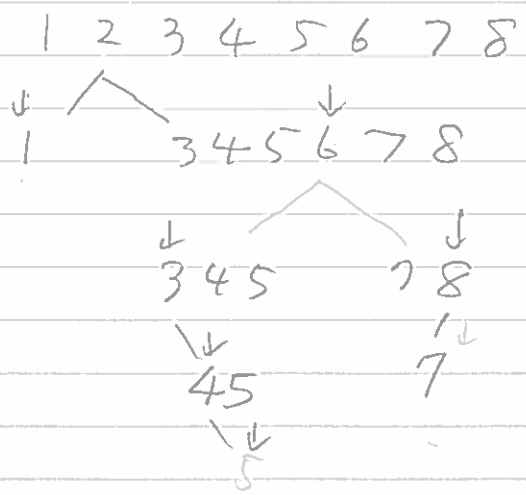
pivot sequence = 7 3 8 2 5 1 4 6

Th. 1.4 # of comparisons is  $O(n \log n)$  whp. next page

looks like a random permutation

Consequently:

A (random) permutation  $\sigma = 2 6 1 8 3 4 5 7$



determines the execution of QS.

determines # of comparisons  $7 + 5 + 3 + \dots$

Simple argument such as ---



if a random pivot falls in this interval,  
then the length of unsorted sequence becomes  
at most  $2/3$

The prob. that happens is  $1/3$



Do this  $\Theta(\log n)$  times  $\Rightarrow$  Above event happens  
at least  $2 \log n$  times w.h.p.

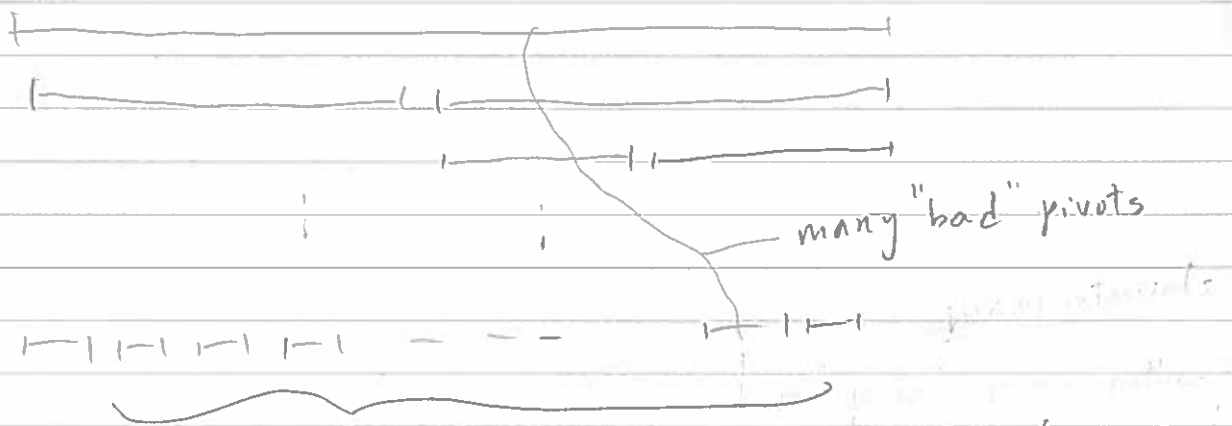


$$\left(\frac{2}{3}\right)^{2 \log n} \approx n^{-1.1}$$



The above length becomes 1 w.h.p.

However, there are many "intervals"



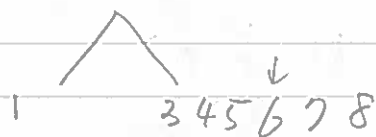
All of them must become single elements.

$T(\sigma)$ : # of comparisons determined by  $\sigma$

Random Variable  $X_{ij}(\sigma) = \begin{cases} 1 & ; i \text{ and } j \text{ compared under } \sigma \\ 0 & ; \text{ -- not --} \end{cases}$

$i \quad j$   
1 2 3 4 5 6 7 8

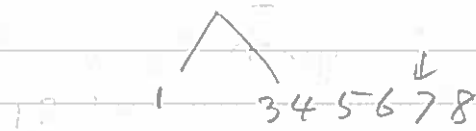
$\sigma = 2 6 1 8 3 4 7 5$



3 4 5 7 8

Not compared

$\sigma = 2 7 1 8 3 4 6 5$



3 4 5 6 7 8

Compared

$k$ : the first number in  $\sigma$  s.t.  $i \leq k \leq j$

$k = i$  or  $j \Rightarrow i$  and  $j$  are compared

$i < k < j \Rightarrow$  " NOT "

$$E(T(\sigma)) = E\left(\sum_{i < j} X_{ij}(\sigma)\right)$$

$$= \sum_{i < j} E(X_{ij}(\sigma)) = \sum_{i < j} \text{prob}(X_{ij} = 1)$$

$$= \sum_{i < j} 2/(j-i+1) \quad (k \text{ can be this many})$$

the first  $k$  is equal prob. for

$i, i+1, \dots, j$

$$= 2 \left( \frac{1}{2} + \frac{1}{3} + \dots + \frac{1}{n} \right) + 2 \left( \frac{1}{2} + \frac{1}{3} + \dots + \frac{1}{n-1} \right) + \dots$$

$i=1, j=2, 3, \dots$                        $i=2, j=3, 4, \dots$

$$\leq 2n \ln n = \mathcal{O}(n \log n) \quad 1 + \frac{1}{2} + \dots + \frac{1}{n} \approx \ln n$$

Q Markov Bound  $\rightarrow \mathcal{O}(n \log n)$  whp

Example 2: 3SAT

Input: 3CNF formula

$$f = (x_1 \vee \bar{x}_3 \vee x_4)(\bar{x}_2 \vee x_4 \vee x_5)(\dots)$$

Question: Is  $f$  satisfiable?

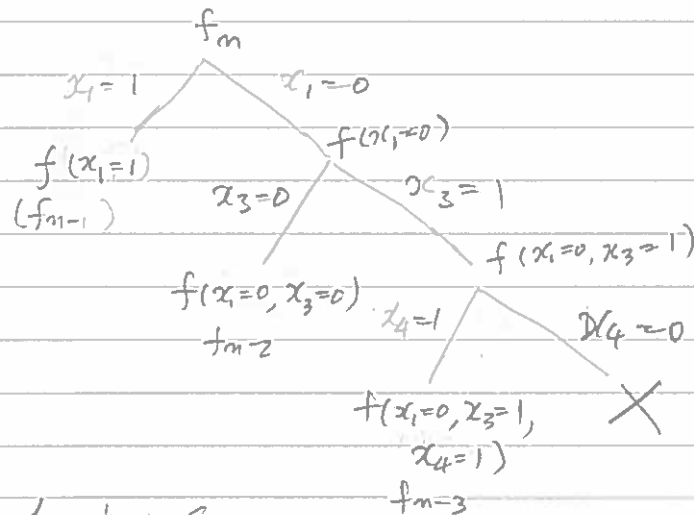
$\exists$  assignment making  $f$  true

Check  $(x_1, \dots, x_n)$  from  $(0, 0, \dots, 0)$

to  $(1, 1, \dots, 1)$

$\Rightarrow 2^n$  steps

But many "good" algorithms running faster



$T(n)$ : # of steps

$$\leftarrow T(n) = 2T(n-1)$$

$$T(n) = T(n-1) + T(n-2) + T(n-3)$$

$$\Rightarrow T(n) = 2^n$$

$$\Downarrow$$

$$T(n) = 1.84^n \ll 2^n$$

$$c^n = c^{n-1} + c^{n-2} + c^{n-3}$$

$$c^3 = c^2 + c + 1$$

1.5<sup>n</sup>  
No. ↓  
1.334<sup>n</sup>  
1.321<sup>n</sup>

# Local Search for 3SAT [Schöeningg 99]

Input  $f = (\alpha_1 \vee \alpha_2 \vee \alpha_3) (\ ) (\ ) \dots (\ )$

$x = (a_1, a_2, \dots, a_n) \leftarrow$  random assignment

Repeat  $3n$  times

If  $f(a_1, a_2, \dots, a_n) = 1$  Then output "yes"  
otherwise

select an arbitrary false clause.

$(\alpha_b \vee \bar{\alpha}_c \vee \alpha_d)$   
0 1 0  $\begin{cases} 1 \rightarrow 0 \\ 0 \rightarrow 1 \end{cases}$

Select one of the Three variables and flip it at random

## Analysis of the success prob.

$x = (1, 0, 0, 1, 1, 0, 1, 1)$

$x = (0, 0, 1, 0, 1, 0, 0, 0)$

satisfying assignment

Hamming distance  
k-1 k k+1

↑  
we are here

Assume there is only one sat. assignment

by a single flip:

← with at least prob 1/3  
→ with at most prob 2/3  
**why?**

$\alpha_b \vee \bar{\alpha}_c \vee \alpha_d$   
0 1 0 current  
1 1 0 sat  
1 0 0 or  
⋮

For example ( $k=5$ )



↪ good direction 7 times &  
bad " " twice

$\sum_{i=0}^n p_R(i) = e^{-n}$   
 $\Rightarrow$  repeating  $e^m$  times gives us a good success prob.



(we are successful)

This can happen with prob.  $\binom{9}{2} \left(\frac{1}{3}\right)^2 \left(\frac{2}{3}\right)^7$  approximately

In general: bad direction  $i$  times  
 good " "  $k+i$  "   
 We are also successful  
 good: 9 times  
 bad: 4 times

with prob.  $p_R(i) = \binom{k+2i}{i} \left(\frac{1}{3}\right)^{k+i} \left(\frac{2}{3}\right)^i$

Summing up from  $i=0$  to  $n$  ( $\rightarrow \binom{m}{n}$ )

$\sum_{i=0}^n p_R(i)$  Total success prob.

However, the value of  $p_R(i)$  <sup>is</sup> ~~changes~~ exponentially (due to the binomial coefficient), so it is <sup>small</sup>

$$\sum_{i=1}^n \binom{n}{i} \approx \binom{n}{n/2}$$

approximately OK to consider a maximum single term, i.e.,  $p_R(k)$  (ie, when  $i=k$ )

$$\sum_{i=0}^n p_R(i) \approx p_R(k) = \binom{3k}{k} \left(\frac{1}{3}\right)^{2k} \left(\frac{2}{3}\right)^k \approx \left(\frac{1}{2}\right)^k$$

(i) A useful formula

$$\binom{m}{dm} \approx \left(\frac{1}{d}\right)^{dm} \left(\frac{1}{1-d}\right)^{(1-d)m}$$

Recall that we use a random assignment at the beginning of training

At position (Hamming dist.)  $k$  with prob

correct  $(0, 0, 0, 0, 0)$   $\binom{m}{k} \left(\frac{1}{2}\right)^k \left(\frac{1}{2}\right)^{m-k} = \binom{m}{k} \left(\frac{1}{2}\right)^m$

init'd  $(0, x, x, 0, 0)$

From there we can get to the solution with prob.

$n=3k, d=1/3$

$$\binom{3k}{k} = 3^k \left(\frac{3}{2}\right)^{2k}$$

$$\left(\frac{1}{2}\right)^k$$



So the total success prob. =

$$\begin{aligned} & \sum_k \binom{n}{k} \left(\frac{1}{2}\right)^n \left(\frac{1}{2}\right)^k \\ & \approx \binom{n}{n/3} \left(\frac{1}{2}\right)^n \left(\frac{1}{2}\right)^{n/3} \quad (\text{max at } k = n/3) \\ & = \left(\frac{3}{4}\right)^n \end{aligned}$$

$$\boxed{1.334} \ll 2$$

Namely, if we repeat the procedure  $\left(\frac{4}{3}\right)^n$  times, we can get an answer with high prob.

$$\binom{n}{n/3} = 3^{n/3} \left(\frac{3}{2}\right)^{2n/3}$$

$$2^{0.416n}$$

$$\left(3^{1/3}\right)^n \left(\left(\frac{3}{2}\right)^{2/3}\right)^n \left(\frac{1}{2}\right)^n \left(\left(\frac{1}{2}\right)^{1/3}\right)^n$$

$$3^{\frac{1}{3} + \frac{2}{3}} \cdot \left(\frac{1}{2}\right)^{\frac{2}{3}} \cdot \left(\frac{1}{2}\right)^{\frac{4}{3}}$$

$$3^{\frac{1}{3}} \cdot \left(\frac{1}{2}\right)^2$$

$$\frac{3}{4}$$

Algorithm

Th 1.5

Schoening solves 3SAT in time  $O\left(\left(\frac{3}{4}\right)^n\right)$  whp.