

CS 473 Algorithms: Lecture 12 (2022-03-01)

logistics: - pers on F17

last lecture: - flow - reductions to max flow  
- circulations w/ demands  
- \_\_\_\_\_, w/ lower bounds  
- bipartite matching w/ forced matches

today: randomized algo

Q: what is the "most realistic" model of computers?

- can run on actual computers
- captures behavior of actual computers

↳ rand()

↳ randomized algo on  $[a, b]$  rand(k)

-  $O(1)$  time

- returns  $a \leq i < b$  w/p  $\frac{1}{k}$

Q: who will win the next US presidential election?

A: ask everyone

~ 328 million ppl

~ 235 million eligible voters

~ 138 million actual voters

fact: a poll of 738

uniformly random voters will

- estimate actual vote up to

error  $\leq 5\%$

- with probability  $\geq 99\%$

Q: why rand algo?

A: - simpler also, flow analysis can

- faster also, flow nonzero false probability

sometimes only efficient also than is randomized



Q: how to model randomized algo?

models = deterministic: deterministic algo  $f$   
worst case input  $x$

$$x \mapsto f(x)$$

$$\text{complexity } T(n) = \max_{|x|=n} T(x)$$

def:  $\Omega$  finite / countably-infinite set.

$P_r: \Omega \rightarrow [0,1]$ . then  $(\Omega, P_r)$  is

a discrete probability space if  $\sum_{\omega \in \Omega} P_r[\omega] = 1$

For an event  $E \subseteq \Omega$   $P_r[E] = \sum_{\omega \in E} P_r[\omega]$

A random variable is a function  $X: \Omega \rightarrow \mathbb{R}$ ,  $n \in \mathbb{N}$

expectation 
$$E[X] = \sum_{\omega \in \Omega} X(\omega) \cdot P_r[\omega]$$

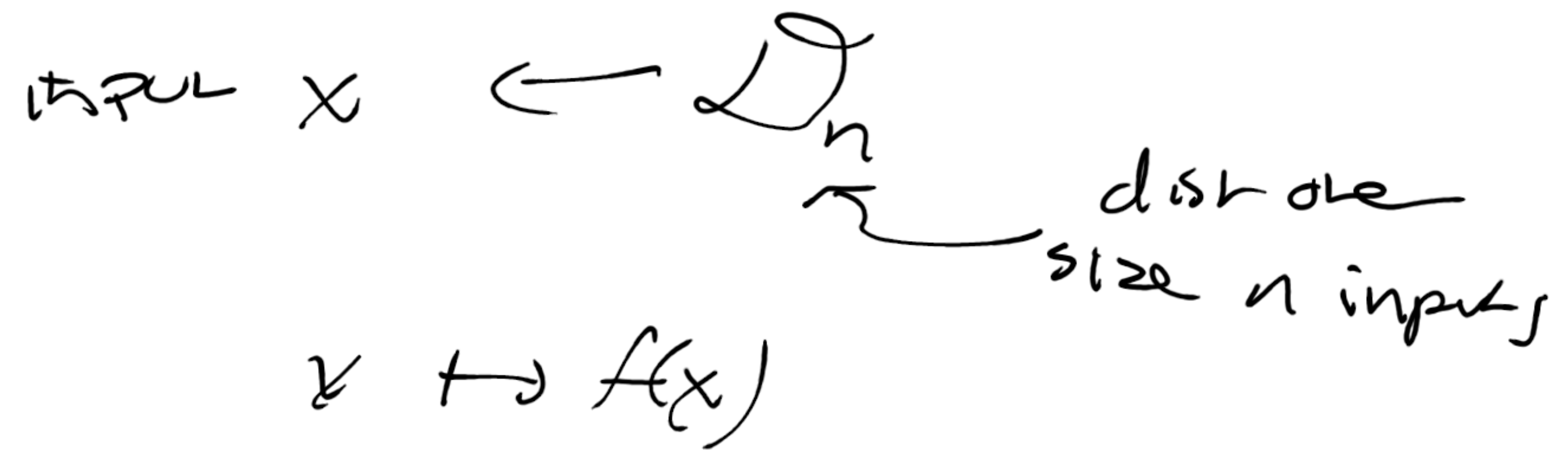
how to analyze rand algo:

correctness: output always correct

complexity: expected runtime as function of input size

model: probabilistic

det algo  $f$



complexity:  $T(n) = \mathbb{E}_{x \leftarrow \mathcal{D}_n} T(x)$

rank:  $\mathcal{D}$  is often unknown in real life

$\Rightarrow$  theory with

randomized algo: randomized algo  
worst case input  $x \xrightarrow{\text{random in algo}}$   
 $x \mapsto f(x)$

complexity  $T(n) = \max_x \mathbb{E}[T(x)]$   
 $|x|=n$

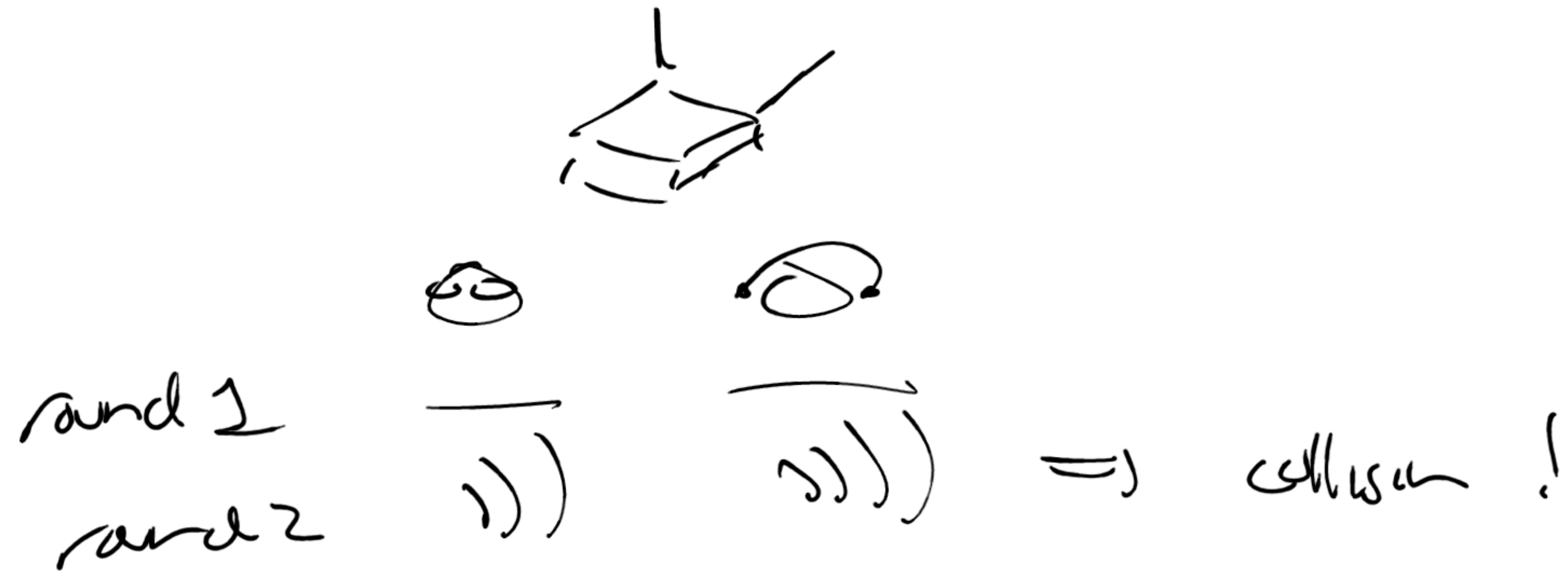
rank: "right" notion

Q: how does multi user work?



Q: how to avoid collision?

idea = take turns



idea = assign turns

↳ requires communication to assign!

Q: how to assign without centralized server?  
ie break symmetry

idea = use names

Mifan ~~has~~ → even length

⇒ talk in even order

Alice → odd length

⇒ talk in odd order

no collisions!

but needs more work

↳  $|bob| = 3 \Rightarrow$  collision

idea = use randomness!



def - the contention resolution process is, for

$n$  people  $P_1, \dots, P_n$

in round  $j$ : each person tries to communicate

w/p  $P$

(if  $P_i$  is lucky) person communicates, they succeed

else everyone fails

Q: how many rounds until everyone succeeds?

def:  $X_{ij} = \begin{cases} 1 & P_i \text{ attempts communication in round } j \\ 0 & \text{else} \end{cases}$

$\{X_{ij}\}_{ij}$  are independent

$$\hookrightarrow P_i [X_{ij} = a \wedge X_{ij} = b] = P_i [X_{ij} = a] \cdot P_i [X_{ij} = b]$$

$$\begin{aligned} \text{len} : \mathbb{E}[X_{ij}] &= 1 \cdot \underbrace{P_i[X_{ij}=1]}_P + 0 \cdot \underbrace{P_i[X_{ij}=0]}_{1-P} \\ &= P \cdot [X_{ij}=1] \\ &= P \end{aligned}$$

$$\begin{aligned} \text{len} : \mathbb{E}[\# \text{ people trying to communicate in round } j] \\ = \mathbb{E}\left[\sum_{i=1}^n X_{ij}\right] = \sum_{i=1}^n \mathbb{E}[X_{ij}] \end{aligned}$$

len [linearity of expectation] any var  $X, Y, \mathbb{E}[X+Y] = \mathbb{E}[X] + \mathbb{E}[Y]$

$$= n \cdot P \leftarrow \text{def: choose } P = \frac{1}{n}, \text{ so } = 1$$

lem =  $\{ P_i \text{ succeeds in round } j \}$   
 $= \{ X_{ij} = 1 \} \cap \{ X_{i'j} = 0 \}_{i' \neq i}$

co =  $P_r [ P_i \text{ succeeds round } j ]$   
 $= P_r [ X_{ij} = 1 \wedge \bigwedge_{i' \neq i} X_{i'j} = 0 ]$

independence  $\rightarrow$   $= \underbrace{P_r [ X_{ij} = 1 ]}_{= p = \frac{1}{n}} \cdot \prod_{i' \neq i} \underbrace{P_r [ X_{i'j} = 0 ]}_{= 1-p = 1 - \frac{1}{n}}$

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

idea: avoid exact expressions

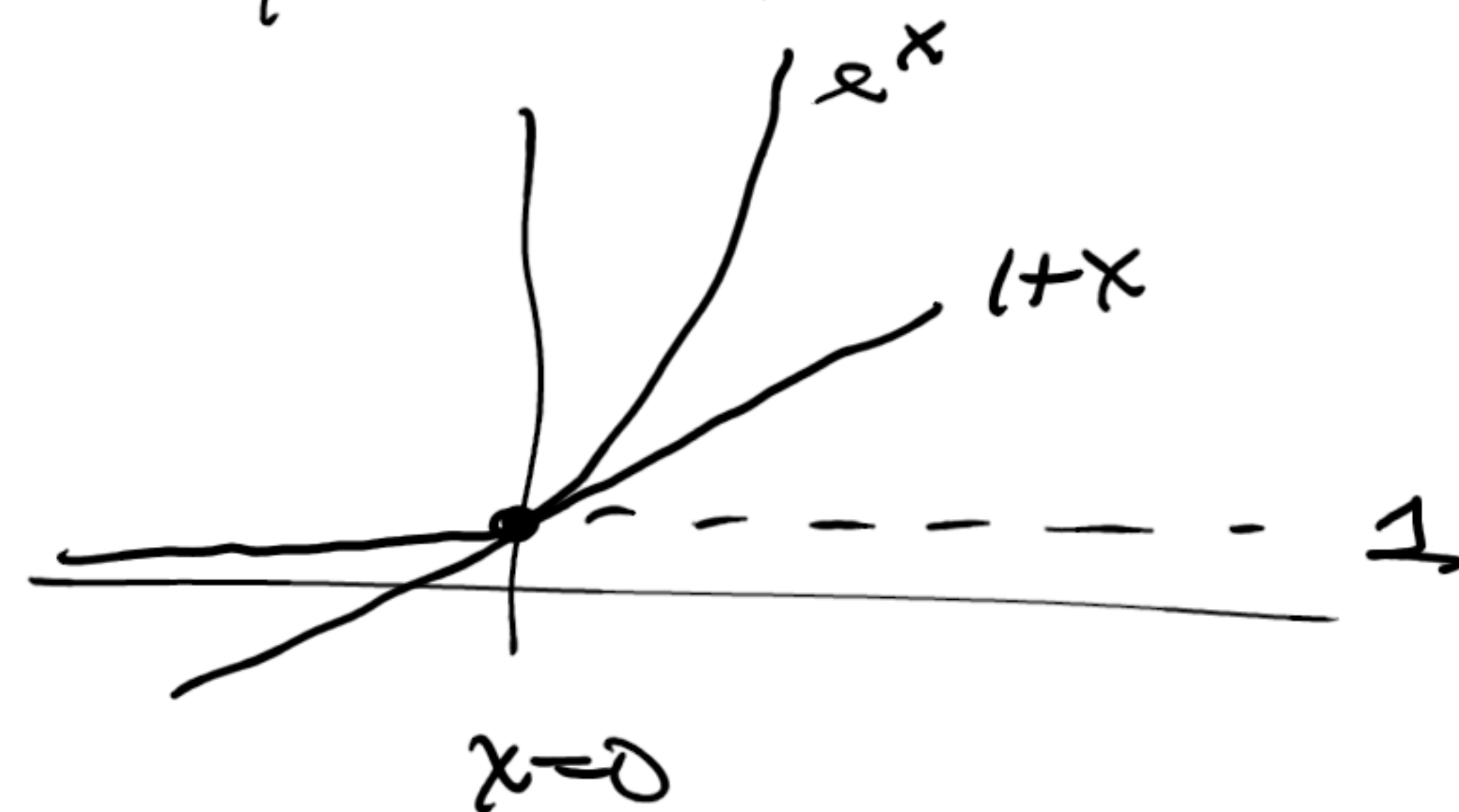
- exact expression are often difficult to obtain  
 - to utilize

ex:  $\frac{1}{100}$  vs  $\frac{1}{\ln 100}$

lem [calculus]:

$\forall x \quad 1+x \leq e^x$

sketch: Taylor expansion



rank:  $1+x \geq e^x \quad \forall x \leq 0$

co:  $(1 - \frac{1}{n})^{n-1} \geq \frac{1}{e}$ ,  $n > 1$

pf:  $\frac{1}{(1 - \frac{1}{n})^{n-1}} = \left( \frac{n}{n-1} \right)^{n-1} = \left( \frac{n}{n-1} \right)^{n-1}$

$= \left( 1 + \frac{1}{n-1} \right)^{n-1} \leq \left( e^{\frac{1}{n-1}} \right)^{n-1}$

$= e$

co:  $P_r [ P_i \text{ succeeds in round } j ]$

$\geq \frac{1}{e \cdot n}$



$$\begin{aligned}
 \underline{Q} \rightarrow & \Pr [P_i \text{ fails in rounds } 1, \dots, t] \\
 &= \prod_{j=1}^t \underbrace{\Pr [P_i \text{ fails in round } j]}_{\leq (1 - \gamma_{en})} \\
 &\leq (1 - \gamma_{en})^t \\
 &\leq (e^{-\gamma_{en}})^t \\
 &= e^{-t \gamma_{en}}
 \end{aligned}$$

$$\underline{Q} = \Pr [P_i \text{ fails rounds } 1, \dots, c \cdot en] \leq \frac{1}{e^c}$$

$$\Rightarrow \text{in } \Theta(n) \text{ rounds } P_i \text{ succeeds w.p. } \geq 1 - \frac{1}{e^c}$$

$$= 1 - \Theta(1)$$

$$= \Theta(1)$$

$$\Rightarrow \text{in } \Theta(\ln n) \text{ rounds } \longrightarrow \geq 1 - \frac{1}{e^{\Theta(\ln n)}}$$

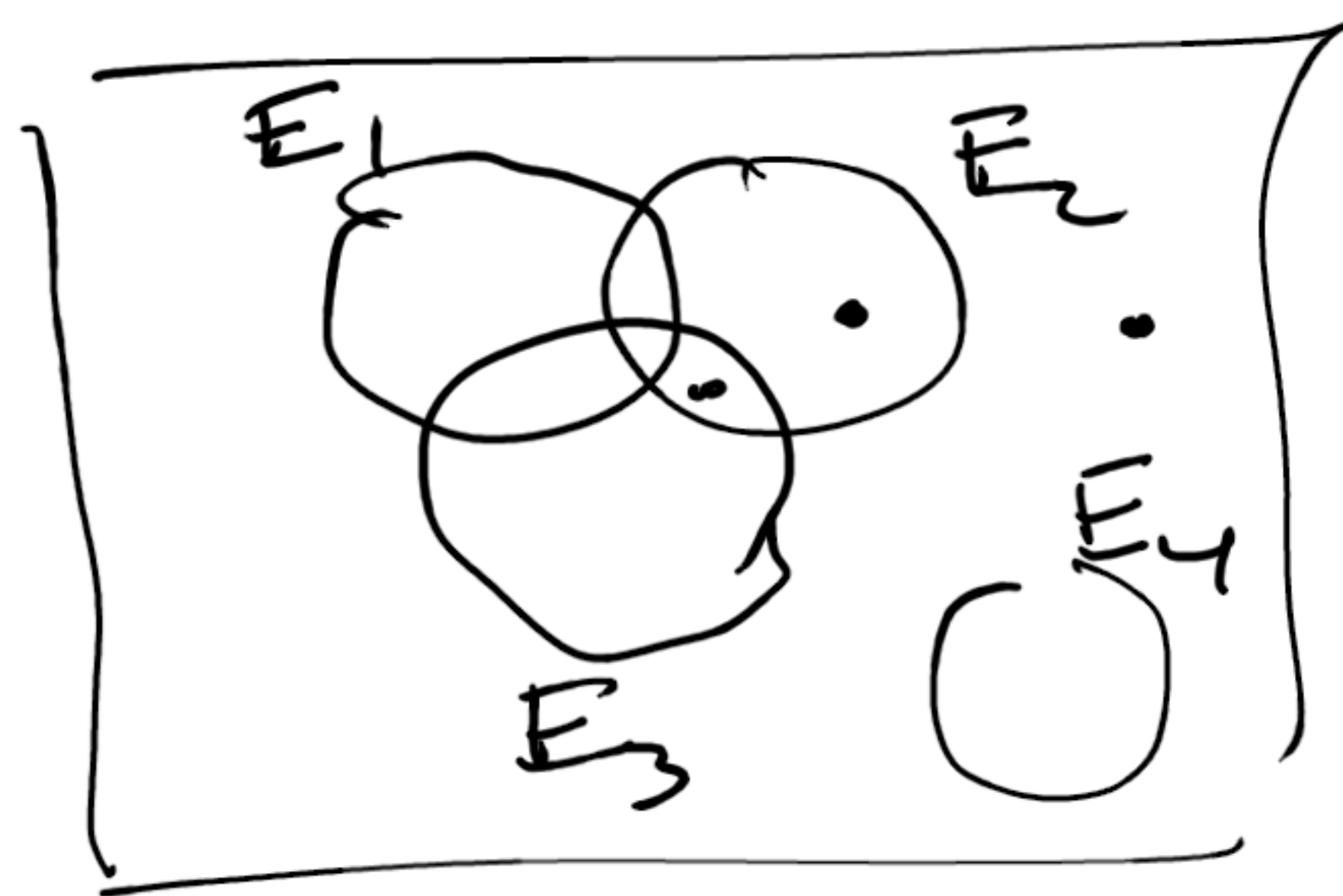
$$= 1 - \frac{1}{n^{\Theta(1)}}$$

Q: all players succeed?

Let [union bound]: events  $E_1, \dots, E_n$

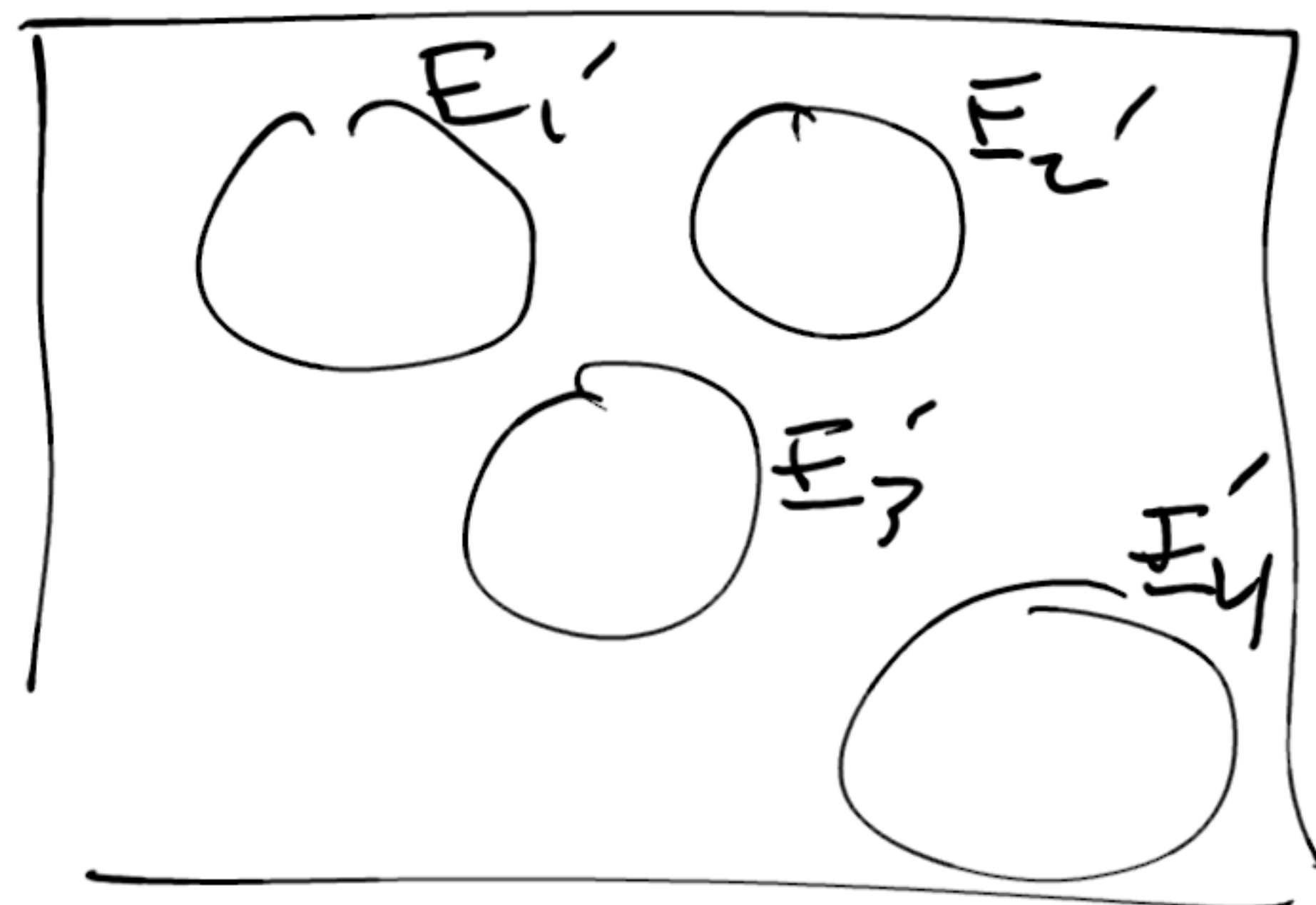
$$P\left[\bigcup_{i=1}^n E_i\right] \leq \sum_{i=1}^n P_i(E_i)$$

Sketch:



$\wedge$

vs



- note:
- $n$  rounds clearly required
  - can also analyze expected # rounds

con.  $P$  [any]  $P_i$  fails in rounds  $1, \dots, c \cdot \ln n$

$$= P\left[\bigcup_i P_i \text{ fails in } \uparrow\right]$$

$$\leq \sum_{i=1}^n \underbrace{P_i(P_i \text{ fails in } \uparrow)}_{\leq \frac{1}{e^c}}$$

$$\leq n \cdot \frac{1}{e^c}$$

con:  $P$  [any]  $P_i$  fails in rounds  $1, \dots, (c \ln n) \cdot \ln n \cdot \ln n$

$$\leq \frac{1}{n^c}$$

$\Rightarrow P$  [all]  $P_i$  communicate in first  $\underbrace{2 \ln n \ln n \text{ rounds}}_{\substack{\uparrow \\ c \ln n}} \geq 1 - \frac{1}{n}$

today: randomized algo

- mutation

- led

- crossover solution

next lesson: rand algo

logistics: per 5 over FIT