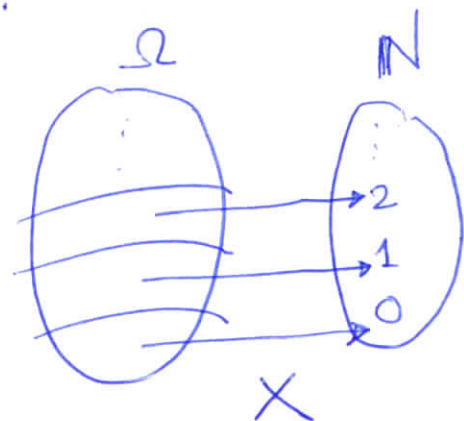


# Random Variables and Expectation.

Def A random variable on space  $(\Omega, \Pr)$  is a map  $X: \Omega \rightarrow \mathbb{R}$ .

Typically  $X: \Omega \rightarrow \{0, 1, 2, \dots\}$ .

Def  $\Pr[X=j] = \Pr[X^{-1}(j)]$ .



Def Expectation of a r.v.,

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

$$= \sum_{j=0,1,2,\dots} j \cdot \Pr[X=j]$$

Note If  $(\Omega, \Pr)$  is uniform, then  $\mathbb{E}[X]$

$$= \frac{1}{|\Omega|} \sum_{\omega \in \Omega} X(\omega) \text{ is just the average,}$$

## Examples

①  $\Omega = \text{Students.}$

$X = \text{Height.}$

$\mathbb{E}[X] = \text{Average height of student in class.}$

②  $\Omega = \{H, T\}^n.$

$X(\omega) = \# \text{ heads in the seq. of tosses } \omega.$

$$\mathbb{E}[X] = \sum_{j=0}^n j \cdot \Pr[X=j]$$

$$= \sum_{j=0}^n j \cdot \frac{\binom{n}{j}}{2^n}$$

$$= \frac{n}{2} \quad !$$

## Linearity of Expectation

Fact If  $X, Y$  are r.v.'s on  $(\Omega, \Pr)$   
then so is  $X+Y.$

$$\mathbb{E}[X+Y] = \mathbb{E}[X] + \mathbb{E}[Y].$$

Proof 
$$\begin{aligned} \mathbb{E}[X+Y] &= \sum_{\omega \in \Omega} (X+Y)(\omega) \\ &= \sum_{\omega \in \Omega} X(\omega) + Y(\omega) \\ &= \sum_{\omega \in \Omega} X(\omega) + \sum_{\omega \in \Omega} Y(\omega) \\ &= \mathbb{E}[X] + \mathbb{E}[Y]. \end{aligned}$$



Example  $\Omega = \{H, T\}^n$        $X = \# \text{heads.}$

Let  $X_i$  be a random variable ( $1 \leq i \leq n$ , "indicator")  
 st. 
$$X_i(\omega) = \begin{cases} 1 & \text{if } i^{\text{th}} \text{ toss is Heads} \\ 0 & \text{otherwise.} \end{cases}$$

$\therefore X = X_1 + X_2 + \dots + X_n$

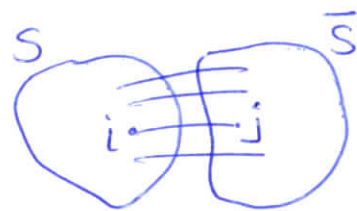
$\therefore \mathbb{E}[X] = \sum_{i=1}^n \mathbb{E}[X_i] = n \cdot \frac{1}{2}.$

$\mathbb{E}[X_i] = 1 \cdot P_X[X_i=1] + 0 \cdot P_0[X_i=0] = \frac{1}{2}.$



# Finding Cuts in Graphs (MAX-CUT)

Def In an undirected graph  $G(V, E)$ ,  $V = \{1, \dots, n\}$ , a cut  $(S, \bar{S})$  is a partition of  $V$  into disjoint sets  $S$  and  $\bar{S}$ .



Def "Edges cut".

$$e(S, \bar{S}) = \{ (i, j) \in E \mid i \in S, j \in \bar{S} \}$$

Problem Given  $G(V, E)$ , find a cut so as to maximize  $|e(S, \bar{S})|$ . NP-complete.

Theorem In a graph  $G(V, E)$ ,  $|E| = m$ , there always exists a cut  $(S, \bar{S})$  s.t.

$$|e(S, \bar{S})| \geq \frac{m}{2}.$$

Such a cut can be found in polynomial time by "a randomized algorithm."

Idea ④ We show that <sup>for</sup> a "random cut", expected #edges cut is  $\frac{m}{2}$ .

"Random cut" = Each vertex placed on  $S$   
or  $\bar{S}$  side w.p.  $\frac{1}{2}$ , independently

- ② Then we show that the expectation is achieved with reasonable prob. ( $\geq \frac{1}{n^2}$ ).
- ③ With (say)  $n^3$  repeated trials, one can get such a cut (i.e. of size  $\geq \frac{m}{2}$ ) with probability  $1 - 2^{-n}$ .

### Algorithm / Analysis

Construct a cut  $(S, \bar{S})$  as follows.

Place each  $i \in V$  into  $S$  or  $\bar{S}$  w.p.  $\frac{1}{2}$  each independently.

Let  $X_{(i,j)}$  be indicator random var.

$$X_{(i,j)} = \begin{cases} 1 & \text{if edge } (i,j) \text{ is cut} \\ 0 & \text{otherwise} \end{cases}$$

$$\mathbb{E}[X_{(i,j)}] = \Pr[X_{(i,j)} = 1] = \frac{1}{2}.$$

Now we look at the r.v.  $|E(S, \bar{S})|$ .

$$|e(s, \bar{s})| = \sum_{(i,j) \in E} X_{(i,j)}$$

$$\therefore \mathbb{E}[|e(s, \bar{s})|] = \sum_{(i,j) \in E} \mathbb{E}[X_{(i,j)}]$$

$$= \frac{m}{2}$$

$$|E| = m$$

Claim With probability at least  $\geq \frac{1}{n^2}$ ,  
the cut  $(s, \bar{s})$  output has  $|e(s, \bar{s})| \geq \frac{m}{2}$ .

Proof let  $p = \text{prob that } |e(s, \bar{s})| \geq \frac{m}{2}$ .

$\therefore$  With prob  $1-p$ ,  $|e(s, \bar{s})| \leq \frac{m}{2} - \frac{1}{2}$ .

with prob  $p$ ,  $|e(s, \bar{s})| \leq m$

$$\therefore \frac{m}{2} = \mathbb{E}[|e(s, \bar{s})|]$$

$$\leq (1-p) \left( \frac{m}{2} - \frac{1}{2} \right) + p \cdot m$$

$$\therefore \frac{m}{2} \leq \frac{m}{2} - \frac{1}{2} p m - \frac{1}{2} + \frac{p}{2} + p m$$

$$\therefore \frac{1}{2} - \frac{p}{2} \leq \frac{1}{2} p m$$



$$\therefore \frac{1}{1+m} \leq p \quad m \leq \binom{n}{2} = \frac{n(n-1)}{2}$$

$$\therefore \frac{1}{n^2} \leq p$$

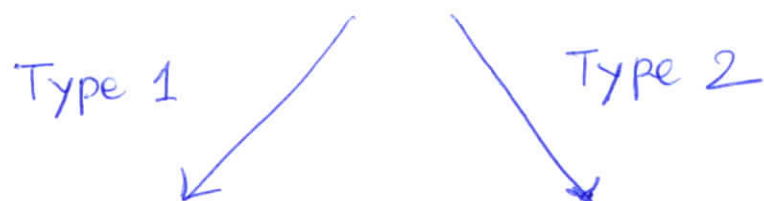


### Repeated trials

- Repeat the algorithm  $n^3$  times.
- "Prob of success" in each trial is  $\geq \frac{1}{n^2}$ .
- $\therefore$  Prob that it fails in all trials is at most  $(1 - \frac{1}{n^2})^{n^3} \leq 2^{-n}$ .



### Randomized Algorithms



- Outputs correct answer w.p.  $\geq 99\%$   
May err.
- Runs in polytime.

- Outputs correct answer w.p.  $\geq 99\%$  or does not answer.
- Runs in expected polytime.

## Randomized Quicksort (RQ)

- Given  $(a_1, a_2, \dots, a_n)$
- Pick "pivot"  $a_i$  at random,  $1 \leq i \leq n$ .
- $S_1 = \{a_j \mid a_j < a_i\}$   
 $S_2 = \{a_j \mid a_j > a_i\}$ .
- Output  $RQ(S_1) \cdot a_i \cdot RQ(S_2)$ .

Theorem Let  $T(n)$  be expected running time. Then  $T(n)$  is  $O(n \log n)$ .

Proof We show, by induction, that  $T(n) \leq D \cdot n \log n$  for some constant  $D$ .

Note.

$$\begin{aligned} T(n) &= Cn + \frac{1}{n} \sum_{i=1}^n T(i) + T(n-i) \\ &= Cn + \frac{2}{n} \sum_{i=1}^{n-1} T(i). \quad T(0) = 0. \end{aligned}$$



$$\therefore T(n) \leq cn + \frac{2}{n} \left( \sum_{i=1}^{\frac{n}{2}} T(i) + \sum_{i=\frac{n}{2}+1}^{n-1} T(i) \right)$$

$$\leq cn + \frac{2}{n} \cdot D \cdot \left( \sum_{i=1}^{\frac{n}{2}} i (\log n - 1) + \sum_{i=\frac{n}{2}+1}^{n-1} i \log n \right)$$

$$= cn + \frac{2D}{n} \left( \sum_{i=1}^{n-1} i \right) \log n - \frac{2D}{n} \left( \sum_{i=1}^{\frac{n}{2}} i \right)$$

$$= cn + \frac{2D}{n} \cdot \frac{n(n-1)}{2} \cdot \log n - \frac{2D}{n} \cdot \frac{n}{2} \left( \frac{n}{2} + 1 \right)$$

$$\leq cn + Dn \log n - \frac{Dn}{4}$$

$$\leq Dn \log n \quad \left( \text{provided } \frac{D}{4} \geq c \right).$$

