

Today

- Is Randomness needed
- Method of conditional expectations
- Limited Independence

CSS.413.1

Pseudorandomness

Lecture 03 (2021-08-31)

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Is randomness needed?

Recall: MAXCUT

Input: A graph $G = (V, E)$.

Output: Cut $(S, V \setminus S)$ that cuts as many edges as possible.

Algorithm:

Output a random $S \subseteq V$.

(Formally, for each $i \in [n]$, ($N = n$)

toss R_i . If $R_i = 1$, add v_i to S
 else don't .)

$$\begin{aligned} \mathbb{E}[\text{cut}(S)] &= \mathbb{E}\left[\sum_{e \in E} \mathbb{1}[S \text{ cuts } e]\right] \\ &= \sum_{e \in E} \frac{1}{2} = |E|/2. \end{aligned}$$

Qn: How can one eliminate/reduce randomness in the above alg?

Technique 1: Enumeration.

- Enumerate over all possible choices of the random coins

😊 - Eliminates Randomness.

☹ - Expensive (#choices = 2^m
 $m = \# \text{random coins}$)
fixed

Technique: Method of Conditional Expectations

$R_1 \dots R_n$ - random coins.

$x_1 \dots x_n$ - particular choice for
s.t. $x_i \in \{R_i, \bar{R}_i\}$

Qn: Fixed first i coins to $x_1 \dots x_i$, what is the best choice for the $(i+1)$ st coin?

Cond. Exp

$$e(x_1 \dots x_i) = E[|cut(B)| \mid R_1=x_1, R_2=x_2, \dots, R_i=x_i]$$

Note: $e(\cdot) = E[|cut(B)|] = E|Z|$

$$\begin{aligned}
 e(x_1 \dots x_n) &= \mathbb{E}_{R_{i+1} \dots R_n} [e(x_1 \dots x_i, R_{i+1} \dots R_n)] \\
 &= \mathbb{E}[e(x_1 \dots x_n, R_{i+1})] \\
 &= \frac{1}{2} e(x_1 \dots x_i, 0) + \frac{1}{2} e(x_1 \dots x_i, 1).
 \end{aligned}$$

$$e(i) = \frac{e(0) + e(1)}{2}$$

- $e(i) = |E|/2$.

- For $i=0$ to $n-1$

- $\left\{ \begin{array}{l} \text{Compute } e(x_1 \dots x_i, 0) \\ \phantom{\text{Compute }} e(x_1 \dots x_i, 1) \\ \text{If } e(x_1 \dots x_i, 0) \geq e(x_1 \dots x_i, 1) \\ \phantom{\text{If }} \text{set } x_{i+1} \leftarrow 0 \\ \phantom{\text{If }} \text{else } x_{i+1} \leftarrow 1 \end{array} \right.$

- Output S corresponding to $x_1 \dots x_n$.

$$|E|/2 \leq e(i) \leq e(x_1) \leq e(x_1, x_2) \leq \dots \leq e(x_1 \dots x_n)$$

Qn: Are the conditional exp $e(x_1 \dots x_i)$ easily computable?

$x_1 \dots x_i$



$$\begin{aligned}
 V &= S_i \cup T_i \cup U_i \\
 S_i \cup T_i &= \{1, \dots, e\}
 \end{aligned}$$

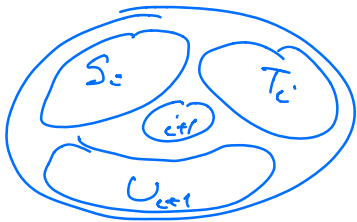
$$S_i = \{j \leq i \mid x_j = 1\} \quad U_i = [n] \setminus (S_i \cup T_i)$$

$$T_i = \{j \leq i \mid x_j = 0\}$$

$$|e(x_1 \dots x_i)| = |\text{cut}(S_i, T_i)| + \frac{1}{2} |\text{cut}(U_i, [n])|$$

$$\begin{aligned} e(x_1 \dots x_i) &= \mathbb{E} [|\text{cut}(S)| \mid \forall j \leq i, R_j = x_j] \\ &= \sum_{e \in E} \mathbb{E} [1_{[S \text{ cuts } e]} \mid \forall j \leq i, R_j = x_j] \\ &= |\text{cut}(S_i, T_i)| + \frac{1}{2} |\text{cut}(U_i, [n])| \end{aligned}$$

$$\begin{aligned} e(x_1 \dots x_i, 1) &= |\text{cut}(S_i, T_i)| + |\text{cut}(\{i+1\}, T_i)| \\ &\quad + \frac{1}{2} |\text{cut}(U_{i+1}, [n])| \end{aligned}$$



Similarly,

$$\begin{aligned} e(x_1 \dots x_i, 0) &= |\text{cut}(S_i, T_i)| \\ &\quad + |\text{cut}(\{i+1\}, S_i)| \\ &\quad + \frac{1}{2} |\text{cut}(U_{i+1}, [n])| \end{aligned}$$

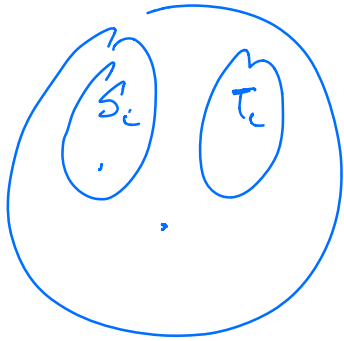
$$\begin{aligned} e(x_1 \dots x_i, 1) - e(x_1 \dots x_i, 0) &= |\text{cut}(\{i+1\}, T_i)| \\ &\quad - |\text{cut}(\{i+1\}, S_i)| \end{aligned}$$

$$S_0, T_0 \leftarrow \emptyset, U_0 \leftarrow [n]$$

Alg: For $i \leftarrow 0$ to $n-1$

$$\begin{cases} \text{compute } \text{cut}(\{i+1\}, T_i) \neq \text{cut}(\{i+1\}, S_i) \\ \text{if } \text{cut}(\{i+1\}, T_i) \geq \text{cut}(\{i+1\}, S_i) \end{cases}$$

Set $x_{u+1} \leftarrow 1$ else $x_{u+1} \leftarrow 0$



Greedy Algorithm

Technique 3: Limited Independence

$$\begin{aligned} \mathbb{E}[|\text{cut}(S)|] &= \sum_{e=(i,j) \in E} \Pr[S \text{ cuts } (i,j)] \\ &= \sum_{e=(i,j) \in E} \Pr[R_i \neq R_j] \\ &= \frac{|E|}{2} \end{aligned}$$

Obs: (1) For above argument, the following suffices.

$$\forall i \neq j \quad \Pr[R_i \neq R_j] = \frac{1}{2}.$$

(2) For the rand alg, it suffices if R_1, \dots, R_n - pairwise independent
(i.e., $\forall i \neq j$).

$$\forall b_i, b_j \in \{0,1\} \quad \Pr[R_i = b_i, R_j = b_j] = \frac{1}{4}$$

Qn: Can pairwise indep n -coins be

generated using fewer bits of randomness!

3 coins: $R_1, R_2, R_3 = R_1 \oplus R_2$
independently

N coins: $N = 2^k - 1$

$R_1 \dots R_k$ - independent random coins.

For each $\emptyset \neq S \subseteq [k]$

$$R_S = \bigoplus_{i \in S} R_i$$

Note: $R_{\{i\}} = R_i$

Claim: $\{R_S\}_{\emptyset \neq S \subseteq [k]}$ is a pairwise mod set of $(2^k - 1)$ coins.

In fact,

$$\forall S \neq T, \forall a, b \in \{0,1\} \quad \Pr[R_S = a, R_T = b] = \frac{1}{4}$$

Pf: $R_S = R_{S \cap T} \oplus R_{S \setminus T}$

$$R_T = R_{S \cap T} \oplus R_{T \setminus S}$$

$S \neq T \Rightarrow S, T$ non-empty, \Rightarrow at least 2

of S_{NT}, S_{IT}, T_{IS}
are non-empty

Also, $R_{SNT}, R_{SIT}, R_{TIS}$ - independent

Hence R_S, R_T are ind w/ uniform marginals \square

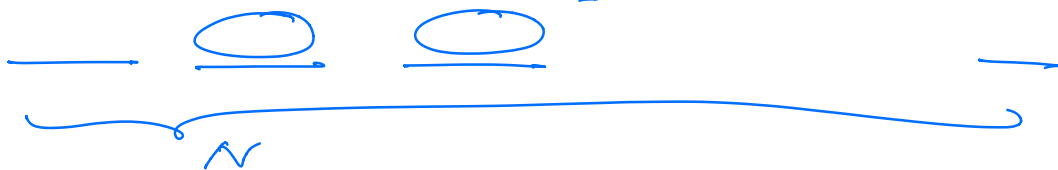
Concl: To obtain N pairwise independent
one only needs $\log(N+1)$
(fully) independent coins.

Pairwise Independent family of hash
functions

N random coins - pairwise independent

N random elements

(each in the range $[M]$)
 $\in [M]$



Qnr: Generate N random samples.

R_1, \dots, R_N s.t

$$(1) \forall i: \forall a \in [M] \Pr[R_i = a] = 1/M$$

(2) $\forall c \neq j$, $R_c \neq R_j$ are independent.

$$M = \{0, 1\}^m$$

Repeat the above (+) process
m times.

$$\begin{aligned} \# \text{ truly independent coins needed} \\ &= m \cdot (\log(N+1)) \\ &= (\log M) \cdot (\log(N+1)) \end{aligned}$$

Experiment:

\mathbb{F}_q - finite field of size q .

$$q \geq \max\{M, N\}.$$

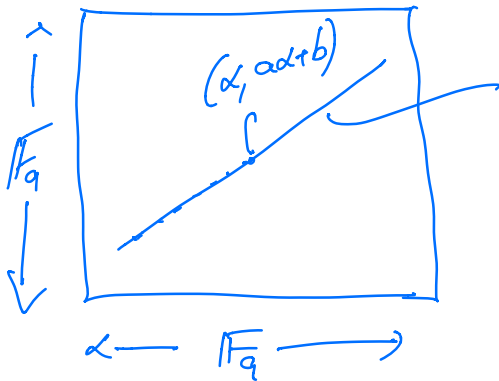
- $a, b \in \mathbb{F}_q$;
- $\forall \alpha \in \mathbb{F}_q$, $R_\alpha \leftarrow a\alpha + b$.

Claim: $\{R_\alpha\}_{\alpha \in \mathbb{F}_q}$ over random choice
of $(a, b) \in \mathbb{F}_q \times \mathbb{F}_q$
- pairwise independent.

In fact, $\forall (c, d) \in \mathbb{F}_q \times \mathbb{F}_q$
 $\exists \alpha \neq \beta \in \mathbb{F}_q$
/

$$P_{(a,b)} \left[P_{\alpha} = c \text{ and } P_{\beta} = d \right] = \frac{1}{|\mathbb{F}_q|^2}$$

$$\begin{aligned} \text{Pf: } P_{(a,b)} \left[\begin{array}{l} a\alpha + b = c \\ a\beta + b = d \end{array} \right] &= P_{(a,b)} \left[\begin{array}{l} a(\alpha - \beta) = c - d \\ a\alpha + \beta = c \end{array} \right] \\ &= P_a \left[a = \frac{c-d}{\alpha - \beta} \right] \cdot P_b \left[b = c - a\alpha \mid a = \frac{c-d}{\alpha - \beta} \right] \\ &= \frac{1}{|\mathbb{F}_q|} \cdot \frac{1}{|\mathbb{F}_q|} = \frac{1}{|\mathbb{F}_q|^2}. \end{aligned}$$



a, b
 $y = ax + b.$

$P_{\alpha} = a\alpha + b. \rightarrow (\alpha, a\alpha + b)$

□