

Randomness in Computing



CS
537

LECTURE 20

Last time

- Probabilistic method
 - The Counting Argument
 - The Expectation Argument
 - Derandomization using conditional expectations

Today

- Probabilistic method
 - Sample and Modify
 - The Second Moment Method

To prove that an object of required value exists:

1. Define a **distribution** on objects.
2. **Sample** an object from the distribution.
 - Compute the **expected** value of the sampled object.
3. Conclude that there exists an object with value equal to **at least** (**at most**) the expectation.

Example: Large sum-free subset

- Given a set A of positive integers, a **sum-free** subset $S \subseteq A$ contains no three elements $i, j, k \in S$ satisfying $i + j = k$.
- **Goal:** find as large sum-free subset S as possible.
- **Examples:** $A = \{2, 3, 4, 5, 6, 8, 10\}$
 $A = \{1, 2, 3, 4, 5, 6, 8, 9, 10, 18\}$

Theorem

Every set A of n positive integers contains a sum-free subset of size greater than $n/3$.

Finding a large sum-free subset

Algorithm

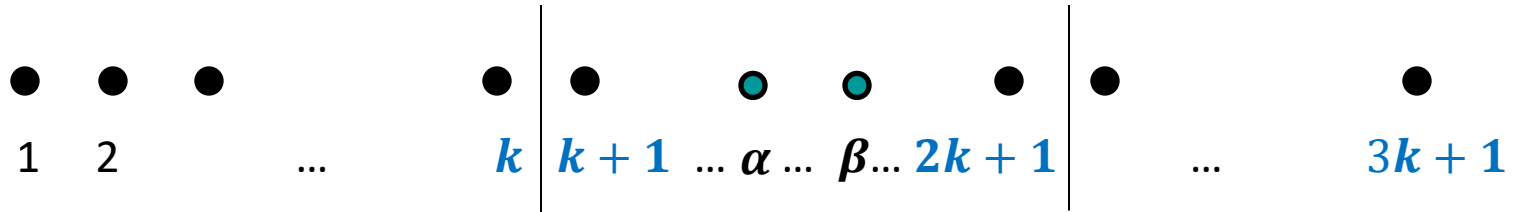
1. Let $p > \max$ element of A be a prime, where $p = 3k + 2$.
//The other choice, $3k + 1$, would also work.
2. Select a number q uniformly at random from $[p - 1]$.
3. Map each element $t \in A$ to $tq \bmod p$.
4. $S \leftarrow$ all elements of A that got mapped to $\{k + 1, \dots, 2k + 1\}$.
5. Return S .

Need to prove:

- S is sum-free
- The expected number of elements from A that are mapped to $\{k + 1, \dots, 2k + 1\}$ is $> n/3$.

Showing that S is sum-free

- Let i and j be any two elements in S .
- Say i is mapped to α ; j is mapped to β ; $\alpha, \beta \in [k + 1, 2k + 1]$



- Then $\alpha = iq \bmod p$ and $\beta = jq \bmod p$
- We need to show that $i + j$, if present in A , is not mapped to $[k + 1, 2k + 1]$.
- $i + j$ is mapped to $(\alpha + \beta) \bmod p$

Argue that

- $(\alpha + \beta)$ must be greater than $2k + 1$.
- If $(\alpha + \beta) > p$, then $(\alpha + \beta) \bmod p$ is at most k .

The expected size of S

1. Let $p > \max$ element of A be a prime, where $p = 3k + 2$.
2. Select a number q uniformly at random from $[p - 1]$.
3. Map each element $t \in A$ to $tq \bmod p$.
4. $S \leftarrow$ all elements of A that got mapped to $\{k + 1, \dots, 2k + 1\}$.

Main idea: Every element $t \in A$ gets mapped to $tq \bmod p$, which is a uniformly random element of $\{1, \dots, 3k + 1\}$.

$$\Pr[t \text{ is selected to be in } S] = \frac{|\{k + 1, \dots, 2k + 1\}|}{|\{1, \dots, 3k + 1\}|} > 1/3$$

Example: Large sum-free subset

- Given a set A of positive integers, a **sum-free** subset $S \subseteq A$ contains no three elements $i, j, k \in S$ satisfying $i + j = k$.
- **Goal:** find as large as S as possible.
- **Examples:** $A = \{2, 3, 4, 5, 6, 8, 10\}$
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To prove that an object of required value exists:

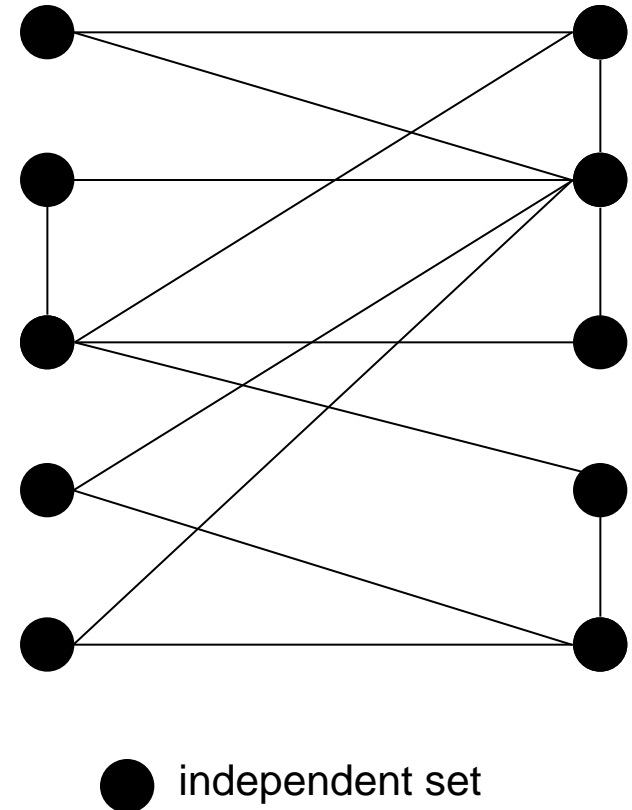
1. Define a **distribution** on objects.
2. **Sample** an object from the distribution.
3. *Modify the sampled object.*
 - Compute the *expected* value of the modified object.
4. Conclude that there exists an object with value equal to **at least** (**at most**) the expectation.

Example: Finding an independent set

An **independent set** in an undirected graph G is a set of nodes that includes at most one endpoint of every edge.

- What is the size of the largest independent set in this graph?

Finding a largest independent set in a given graph is NP-hard.



Example: a large independent set

Theorem

Let G be a connected graph with n nodes and m edges.

Then G has an independent set of size $\geq \frac{n^2}{4m}$.

Proof: Let $d = \frac{2m}{n}$ be the average degree in G .

Since G is connected, $d \geq 1$.

Algorithm

1. Delete each node in G (together with adjacent edges) independently w.p. $1 - 1/d$.
2. For each remaining edge: remove it and one (arbitrary) adjacent node.
3. Output remaining nodes.

Analysis: Algorithm returns an independent set.

Claim

The expected size of the returned set is $\geq \frac{n^2}{4m}$.

Example: a large independent set

Claim

The expected size of the returned set is $\geq \frac{n^2}{4m}$.

1. Delete each node w.p. $1 - 1/d$.
2. Remove each edge with one adjacent node.

Proof: Recall: $d = \frac{2m}{n}$ is the average degree in G .

- Let X = the number of nodes that remain after Step 1.

$$\mathbb{E}[X] = n \cdot \frac{1}{d}$$

- Let Y = the number of edges that remain after Step 1.

An edge remains iff both of its endpoints remain, i.e. w.p. $1/d^2$.

$$\mathbb{E}[Y] = m \cdot \frac{1}{d^2} = \frac{nd}{2} \cdot \frac{1}{d^2} = \frac{n}{2d}$$

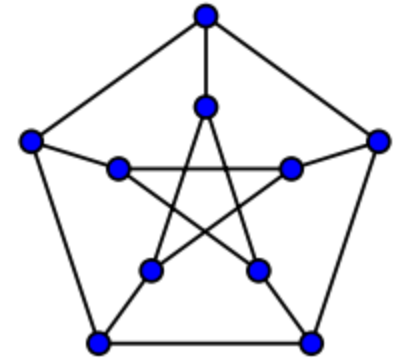
- Step 2 removes at most Y nodes.
- Let Z = the number of nodes in the output: $Z \geq X - Y$

$$\mathbb{E}[Z] \geq \mathbb{E}[X] - \mathbb{E}[Y] = \frac{n}{d} - \frac{n}{2d} = \frac{n}{2d} = \frac{n}{2} \cdot \frac{n}{2m} = \frac{n^2}{4m}$$

Example 2: Graphs with large girth

The **girth** of an undirected graph G is the length of the shortest cycle contained in G .

- What is the girth of this graph?



Ex. 2: Dense graphs with large girth

Theorem

\forall integer $k \geq 3$, for sufficiently large n , there is a graph with n nodes, at least $\frac{n^{1+1/k}}{4}$ edges and girth at least k .

Proof:

Algorithm

1. Sample a graph $G \sim G_{n,p}$ with $p = n^{1/k-1}$.
2. Delete an (arbitrary) edge in G from each cycle of length $\leq k - 1$.
3. Return G .

Analysis: G has n nodes and girth at least k .

- Let X = number of edges in the graph sampled in Step 1.

1. Sample a graph $G \sim G_{n,p}$ with $p = n^{1/k-1}$.
2. Delete an edge from each cycle of length $\leq k - 1$.

Let X = number of edges in the graph sampled in Step 1.

What is the expectation of X ?

- A. np
- B. $\binom{n}{2} p$
- C. $n^2 p(1 - p)$
- D. None of the above.

Ex. 2: Dense graphs with large girth

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3. Return G .

Analysis: G has n nodes and girth at least k .

- Let X = number of edges in the graph sampled in Step 1.

$$\begin{aligned}\mathbb{E}[X] &= p \cdot \binom{n}{2} = n^{1/k-1} \cdot \frac{n(n-1)}{2} = \frac{1}{2} n^{1+1/k} \left(1 - \frac{1}{n}\right) \\ &\geq \frac{1}{3} n^{1+1/k} \quad \text{for } n \geq 3\end{aligned}$$

Ex. 2: Dense graphs with large girth

Claim

G has at least $\frac{n^{1+1/k}}{4}$ edges

1. Sample a graph $G \sim G_{n,p}$ with $p = n^{1/k-1}$.
2. Delete an edge from each cycle of length $\leq k-1$.

Proof: Recall: $\mathbb{E}[X] \geq \frac{1}{3}n^{1+1/k}$ for sufficiently large n .

- Let Y = the number of cycles of length $\leq k-1$ in the sampled graph.
- For $i \in [3, k-1]$, there are $\binom{n}{i} \cdot \frac{(i-1)!}{2}$ possible cycles of length i , each occurring w.p. p^i

$$\begin{aligned} \mathbb{E}[Y] &= \sum_{i=3}^{k-1} \binom{n}{i} \cdot \frac{(i-1)!}{2} \cdot p^i \leq \sum_{i=3}^{k-1} (np)^i = \sum_{i=3}^{k-1} (n^{1/k})^i < k \cdot n^{\frac{k-1}{k}} \\ &\leq \frac{1}{12} n^{1+1/k} \text{ for sufficiently large } n \end{aligned}$$

- Let Z = the number of edges remaining in G : $Z \geq X - Y$

$$\mathbb{E}[Z] \geq \mathbb{E}[X] - \mathbb{E}[Y] > \frac{1}{3}n^{1+1/k} - \frac{1}{12}n^{1+1/k} = \frac{n^{1+1/k}}{4}$$

The 2nd moment method

- Consider a R.V. X with $\mathbb{E}[X] > 0$.
- We want to give an upper bound on $\Pr[X = 0]$.
- By Chebyshev, for all $a > 0$,

$$\Pr[|X - \mathbb{E}[X]| \geq a] \leq \frac{\text{Var}[X]}{a^2}$$

$$\Pr[X = 0] \leq \Pr[|X - \mathbb{E}[X]| \geq \mathbb{E}[X]] \leq \frac{\text{Var}[X]}{(\mathbb{E}[X])^2}$$

Theorem

If X is a random variable with $\mathbb{E}[X] > 0$, then

$$\Pr[X = 0] \leq \frac{\text{Var}[X]}{(\mathbb{E}[X])^2}$$

Threshold behavior in random graphs

$$G \sim G(n, p)$$

For many properties \mathcal{P} , there exists function $f(n)$ s.t.

1. when $p \ll f(n)$, probability that G has $\mathcal{P} \rightarrow 0$ as $n \rightarrow \infty$
2. when $p \gg f(n)$, probability that G has $\mathcal{P} \rightarrow 1$ as $n \rightarrow \infty$

(It holds for all nontrivial monotone properties.)

Review question

What is the expected number of k -cliques in $G \sim G_{n,p}$?

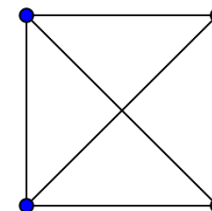
A. $\binom{n}{k} \cdot p^k$

B. $\binom{n}{k} \cdot p^{k(k-1)/2}$

C. $\binom{k}{2} \binom{n}{k} \cdot p^k$

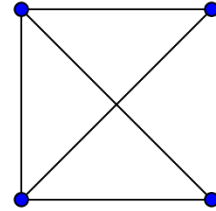
D. $n \cdot p^k (1 - p)$

E. None of the above



Review question

What is the expected number of copies of this graph in $G \sim G_{n,p}$?



- A. $\binom{n}{4} \cdot p^6$
- B. $4 \binom{n}{4} \cdot p^6$
- C. $4 \binom{n}{4} \cdot p^5 (1 - p)$
- D. $6 \binom{n}{4} \cdot p^5 (1 - p)$
- E. None of the above

Example: having a 4-clique

Theorem

Let $G \sim G_{n,p}$ and $p^* = \Pr[G \text{ has a } K_4]$.

1. If $p = o(n^{-2/3})$ then $p^* \rightarrow 0$ as $n \rightarrow \infty$
2. If $p = \omega(n^{-2/3})$ then $p^* \rightarrow 1$ as $n \rightarrow \infty$

Proof: Let $X =$ number of 4-cliques in G .

For every subset C of 4 nodes, let X_C be the indicator for C being a K_4 .

$$\mathbb{E}[X] = \sum_C \mathbb{E}[X_C] = \binom{n}{4} \cdot p^6$$

1. $p = o(n^{-2/3})$

Markov

$$p^* = \Pr[X \geq 1] \leq \frac{\mathbb{E}[X]}{1} = \mathbb{E}[X]$$

$$\leq \frac{n^4}{4!} \cdot p^6 = \frac{n^4}{4!} \cdot o(n^{-(2/3) \cdot 6}) = \frac{n^4}{4!} \cdot o(n^{-4}) = o(1)$$

Example: having a 4-clique

Theorem

Let $G \sim G_{n,p}$ and $p^* = \Pr[G \text{ has a } K_4]$.

1. If $p = o(n^{-2/3})$ then $p^* \rightarrow 0$ as $n \rightarrow \infty$
2. If $p = \omega(n^{-2/3})$ then $p^* \rightarrow 1$ as $n \rightarrow \infty$

Proof: Expected number of 4-cliques: $\mathbb{E}[X] = \binom{n}{4} \cdot p^6$

2. $p = \omega(n^{-2/3})$

$$\mathbb{E}[X] \rightarrow \infty \text{ as } n \rightarrow \infty$$

Goal: Show $\text{Var}[X] \ll (\mathbb{E}[X])^2$

$$\begin{aligned} \text{Cov}(Y, Z) &= \mathbb{E}[(Y - \mu_Y) \cdot (Z - \mu_Z)] \\ &= \mathbb{E}[YZ] - \mu_Y \mu_Z \leq \mathbb{E}[YZ] \end{aligned}$$

$$\text{Var}[X] = \text{Var}\left[\sum_C X_C\right] = \sum_C \text{Var}[X_C] + \sum_{C \neq D} \text{Cov}[X_C, X_D]$$

$$\text{Var}[X_C] = \mathbb{E}[X_C^2] - (\mathbb{E}[X_C])^2 = \mathbb{E}[X_C] - (\mathbb{E}[X_C])^2 = p^6 - p^{12} \leq p^6$$

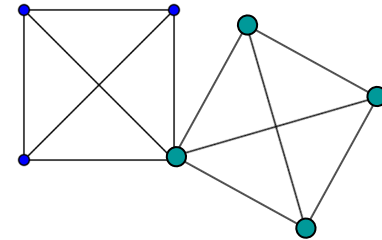
$$\sum_C \text{Var}[X_C] \leq \binom{n}{4} \cdot p^6 = O(n^4 p^6)$$

Bounding the covariance

$$\mathit{Cov}(X_C, X_D) \leq \mathbb{E}[X_C \cdot X_D]$$

Case 1: $|C \cap D|$ is 0 or 1

- Corresponding cliques do not share an edge.
- X_C and X_D are independent.
- $\mathit{Cov}(X_C, X_D) = 0$

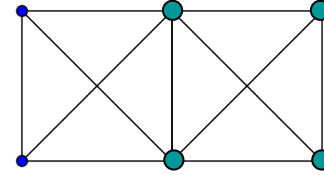


Bounding the covariance

$$\text{Cov}(X_C, X_D) \leq \mathbb{E}[X_C \cdot X_D]$$

Case 2: $|C \cap D| = 2$

$$\text{Cov}(X_C, X_D) \leq \mathbb{E}[X_C \cdot X_D]$$

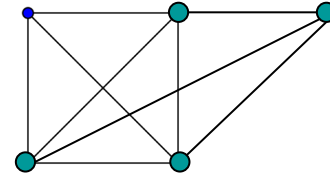


Bounding the covariance

$$\text{Cov}(X_C, X_D) \leq \mathbb{E}[X_C \cdot X_D]$$

Case 3: $|C \cap D| = 3$

$$\text{Cov}(X_C, X_D) \leq \mathbb{E}[X_C \cdot X_D]$$



Putting it all together

Theorem

Let $G \sim G_{n,p}$ and $p^* = \Pr[G \text{ has a } K_4]$.

2. If $p = \omega(n^{-2/3})$ then $p^* \rightarrow 1$ as $n \rightarrow \infty$

- $$\begin{aligned} \text{Var}[X] &\leq \text{Var}[\sum_C X_C] = \sum_C \text{Var}[X_C] + \sum_{C \neq D} \text{Cov}[X_C, X_D] \\ &= O(n^4 p^6 + n^6 p^{11} + n^5 p^9) \end{aligned}$$
- $$\begin{aligned} \Pr[X = 0] &\leq \frac{\text{Var}[X]}{(\mathbb{E}[X])^2} = O\left(\frac{n^4 p^6 + n^6 p^{11} + n^5 p^9}{n^8 p^{12}}\right) \\ &= O\left(\frac{1}{n^4 p^6} + \frac{1}{n^2 p^6} + \frac{1}{n^3 p^3}\right) \\ &= o(1) \text{ for } p = \omega(n^{-2/3}) \end{aligned}$$