

COMPSCI 514: Algorithms for Data Science

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Lecture 13 (Midterm Review)

Summary

Last Class:

- Introduced the idea of low-distortion embeddings and the JL Lemma. $\|x-y\|_2$ $\|y\|$
 - Reduction of JL Lemma to the Distributional JL Lemma via union bound.
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- We will finish the proof of the JL Lemma after the midterm.
Ignore any practice questions on this topic.

This Class:

- Midterm review.

Midterm Format

Rough Outline: (subject to small changes)

✓ 1 pt for answer *1 pt for explanation*

- Question 1: 4-5 always, sometimes, nevers or true/false.
- Question 2: 3-4 short answers, sort of like quiz questions.
- Question 3-4: Multipart questions, similar to core competency problems.
- Question 5: Extra credit question. Similar to a harder core competency problem.

25
—
30

6

31
—
30

Questions

Content, Format, or Logistics Questions?

- Bernoulli
- Überhoff
- binom. Aft. FR

$$\binom{m}{2} \approx \frac{m^2}{2}$$

mid-term concepts, pdf

$$\text{Var}(X) = \mathbb{E}X^2 - (\mathbb{E}X)^2$$

$$\sum_{i=0}^1 i^2 \binom{m}{i} p^i (1-p)^{m-i}$$

$$(1+x) < e^x$$

$$\frac{\exp(-k/n)}{k} \leq \frac{1}{k}$$

Questions

$\sum_{i=1}^n \mathbb{E} S_i^2$
 X_1
 \vdots
 X_n

$S_1^2 = 0$
 $S_2^2 = 0$
 $S_3^2 = 0$
 \vdots
 $S_n^2 = 0$

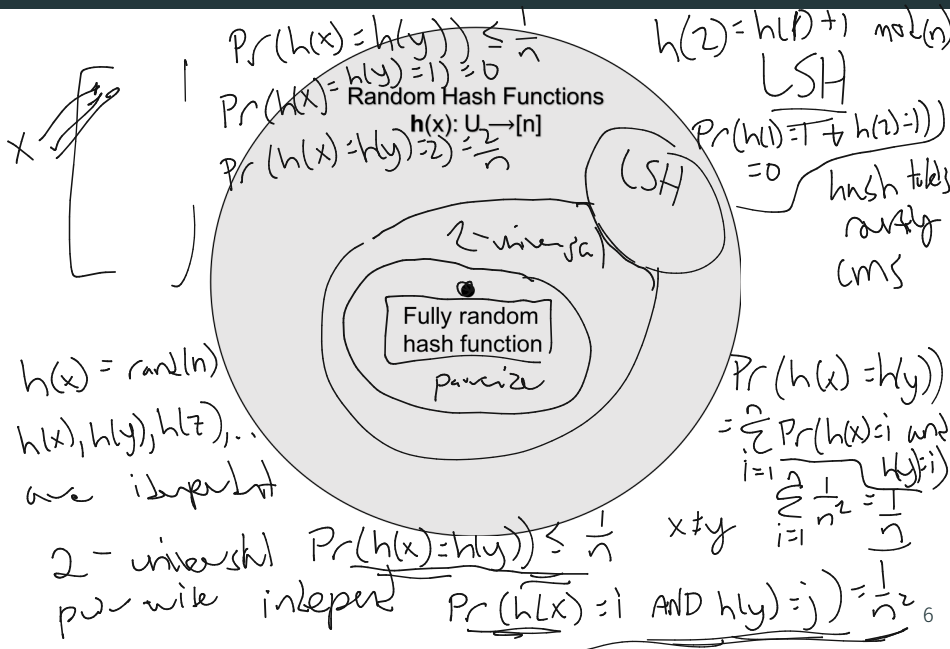
$S_{1,2,4}$
 S_2
 S_n^2

no collision
 $S_i^2 = (\sum X_{ij})^2$
 $\mathbb{E} \sum_k \sum_j X_{ij} \cdot X_{ik}$
 $\sum_k \sum_j \mathbb{E} X_{ij} X_{ik}$
 (pairwise ind)

$\sum_{i=1}^n \mathbb{E} S_i^2$
 collisions
 $\frac{1}{n^2} = \Pr(h(x_j) = i \text{ AND } h(x_k) = i)$

$\sum_k \mathbb{E} X_{ik}$
 $\sum_{i,k} \mathbb{E} X_{ik}$
 $\Pr(h(x_j) = h(x_k))$
 $\frac{1}{n}$

Random Hash Functions



Concentration Bounds

$\Pr(X > t) \leq \frac{\mathbb{E}X}{t}$ Markov's	Concentration Bound Requirements		$\text{Var}(X_i) = p - p^2$ $\leq \frac{1}{4}$
$\mathbb{E}[X]$ $X > 0$	Chebyshev's $\text{Var}(X)$ $\mathbb{E}(X)$	$m=1$ $m=2$ <u>Chernoff</u> $X = \sum X_i$ ind. $X_i \in \{0, 1\}$ $\mathbb{E}[X] = \mu$ $\Pr(X - \mathbb{E}X \geq \delta \mu)$	<u>Bernstein</u> $X = \sum X_i$ ind. $ X_i \leq M$ $\sigma^2 = \text{Var}(X)$ †

$$\Pr(|X - \mathbb{E}X| \geq t) \leq \frac{\text{Var}(X)}{t^2} \quad \frac{1}{10}$$

$$\exp\left(\frac{-1/2^2 \cdot \mu}{2 + 1/2}\right)$$

$$\delta = \frac{10}{\mu}$$

Example Problems

3. Consider an algorithm \mathcal{A} running in time $T(\mathcal{A})$, that with probability .6 outputs an estimate of the number of triangles in an input graph up to error ± 100 , and with probability .4 outputs some bad estimate with worse error. Describe an algorithm that outputs an estimate of the number of triangles in an input graph up to error ± 100 with probability $\geq .99$ and runs in time $O(T(\mathcal{A}))$.

The Chernoff bound states that for independent random variables X_1, \dots, X_n taking values in $\{0, 1\}$, letting $\mu = \mathbb{E} [\sum_{i=1}^n X_i]$, for any $\delta > 0$,

$$\Pr \left(\left| \sum_{i=1}^n X_i - \mu \right| > \delta \mu \right) \leq 2 \exp \left(-\frac{\delta^2 \mu}{2 + \delta} \right).$$

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Example Problems

2. Assume there are 1000 registered users on your site u_1, \dots, u_{1000} , and in a given day, each user visits the site with some probability p_i . The event that any user visits the site is independent of what the other users do. Assume that $\sum_{i=1}^{1000} p_i = 500$.
- (a) Let \mathbf{X} be the number of users that visit the site on the given day. What is $\mathbb{E}[\mathbf{X}]$.
 - (b) Apply a Chernoff bound to show that $\Pr[\mathbf{X} \geq 600] \leq .01$.
 - (c) Apply Markov's inequality and Chebyshev's inequality to bound the same probability. How do they compare?

The Chernoff bound states that for independent random variables X_1, \dots, X_n taking values in $\{0, 1\}$, letting $\mu = \mathbb{E}[\sum_{i=1}^n X_i]$, for any $\delta > 0$,

$$\Pr(|\sum_{i=1}^n X_i - \mu| > \delta\mu) \leq 2 \exp\left(-\frac{\delta^2 \mu}{2+\delta}\right).$$

Example Problems

2. Assume there are 1000 registered users on your site u_1, \dots, u_{1000} , and in a given day, each user visits the site with some probability p_i . The event that any user visits the site is independent of what the other users do. Assume that $\sum_{i=1}^{1000} p_i = 500$.

- (a) Let \mathbf{X} be the number of users that visit the site on the given day. What is $\mathbb{E}[\mathbf{X}]$.
- (b) Apply a Chernoff bound to show that $\Pr[\mathbf{X} \geq 600] \leq 1/5$.
- (c) Apply Markov's inequality and Chebyshev's inequality to bound the same probability. How do they compare?

$$a) X = \sum X_i \rightarrow \mathbb{E}X = \sum_{i=1}^{1000} \mathbb{E}X_i = \sum_{i=1}^{1000} p_i = 500$$

$$b) \Pr(\sum X_i - 500 \geq \delta \cdot 500) \leq 2 \exp\left(-\frac{\delta^2 \mu}{2\delta}\right) \quad \boxed{\delta = 1/5}$$

$$\Pr(\sum X_i - 500 \geq 100) \leq 2 \exp\left(-\frac{1/5^2 \cdot 500}{2 + 1/5}\right)$$

$$2 \exp\left(-\frac{20}{4}\right) \leq 2 \exp(-5) \leq \underline{\underline{.01}}$$

The Chernoff bound states that for independent random variables X_1, \dots, X_n taking values in $\{0, 1\}$, letting $\mu = \mathbb{E}[\sum_{i=1}^n X_i]$, for any $\delta > 0$,

$$\Pr(|\sum_{i=1}^n X_i - \mu| > \delta \mu) \leq 2 \exp\left(-\frac{\delta^2 \mu}{2 + \delta}\right).$$

Example Problems

ALWAYS, SOMETIMES, or NEVER:

$$(1 - \delta)^n = 1 - n\delta + \delta^2 \dots + \delta^n$$

2. $\Pr[\max(X_1, \dots, X_n) \geq t] \leq \sum_{i=1}^n \Pr[X_i \geq t]$ for any random variables X_1, \dots, X_n .

$$\Pr(X_1 \geq t \text{ OR } X_2 \geq t \text{ OR } \dots \text{ OR } X_n \geq t) \leq \sum \Pr(X_i \geq t)$$

union bound

Use which answers with prob $1 - \delta$
Use this to answer n questions

(c) $\Pr[\mathbf{X} = s \cap \mathbf{Y} = t] = \Pr[\mathbf{X} = s] \cdot \Pr[\mathbf{Y} = t]$.

independent

JL

$$\Pr(\text{all questions correct}) \geq 1 - n\delta$$

$$1 - \Pr(\text{fails at least one}) \leq n \cdot \delta = \sum_{i=1}^n \Pr(\text{fails question } i)$$

union bound