

Midterm

Jing Rao (e0522068)

P1 (a). True. Use Markov's Inequality.

$$\Pr(X \geq 5) \leq \frac{E(X)}{5} = 1$$

(b) True. Because $E(X) = 1$ and X must be non-negative integers. And $\text{Var}(X) = 2$, so we must have some values > 1 and some < 1 . Thus, $\Pr(X > 1) > 0$, which means $\Pr(X \geq 2) > 0$.

(c) False. If ~~sample space~~ ~~one~~ random process are $\Pr(X=0) = \frac{2}{3}$ $\Pr(X=3) = \frac{1}{3}$

(d) False. Because X can only pick 0, 1, 2, 3, ..., which means to keep $E(X) = 1$, there must be ~~more~~ 0's ~~than~~ with bigger probability.

Midterm

Bao Jinge (20522065)

P2. Let A denotes not lose and X denotes the score.

$$\Pr(X=1|A) = \frac{\Pr(\{X=1\} \cap A)}{\Pr(A)} = \frac{\overbrace{\Pr(A=1)}^{\Pr(X=1)}}{\Pr(A)} = \frac{\frac{1}{6}}{\Pr(A)}$$

Because $\Pr(A) =$ ~~$\frac{1}{6} + \frac{1}{6} + \frac{1}{6} + \dots + \frac{1}{6} + \dots$~~

$$= \sum_{k=1}^{\infty} \left(\frac{1}{6}\right)^k = 1$$

so $\Pr(X=1|A) = \frac{\frac{1}{6}}{\Pr(A)} = \frac{1}{6}$

P4. If there are one edge between two vertices in I , we ~~say~~ let two vertices v_i and v_j and $i < j$. Because v_i is inserted first than v_j , so v_j is v_i 's neighbor. However v_i has been inserted ~~yet~~ before v_j will be inserted. So v_j ~~is~~ all neighbors can't be all after v_j in order. ^{because v_i} So False. In other way, I is an independent set

In other way, what vertex inserted into S is a ~~center~~ vertex in a ~~graph~~ ^{graph} associated with all other vertices in ~~graph~~ (like a center)

For each ~~node~~ vertex i , ~~if~~ let I_i ~~denote~~ ^{indicate} vertex i is in I

$$|I| = \sum_{i \in V} I_i \quad E(|I|) = \sum_{i \in V} E(I_i)$$

$$\Pr(I_i = 1) = E(I_i) = \frac{1}{n!}$$

$$E(I_i) = \Pr(I_i = 1) = \frac{(n-1)!}{(n-1)! \deg(v_i)!} = \frac{1}{\deg(v_i) + 1}$$

$$E(I_i) = \Pr(I_i = 1) = \frac{(n-1)!}{(n-1)! \deg(v_i)!} = \frac{1}{\deg(v_i) + 1}$$

use ~~Exp~~ Linearity of expectation

$$E(|I|) = \sum_{i \in V} \frac{1}{\deg(v_i) + 1}$$

correct $(n-1)!$ permutations except vertex i

To illustrate, ~~for each~~ there are $n!$ permutations, to let $I_i = 1$, we need make all his neighbors, $\deg(v_i)$ neighbors to fall his back in order there are $\binom{n-1}{\deg(v_i)}$ choices. \neq

Midterm

Jing Bao (e0522065)

PS Here we use Yao's minimax principle. If I design a deterministic algorithm to compress any binary string of 10^6 . Then the randomized algorithm will get best accuracy than this deterministic algorithm.

Here we choose binary string in uniformly.

Suppose we design algorithm ~~compress~~ keep only first 90% bits of string, then when recovering.

When means 0.9×10^6 will be kept and 0.1×10^6 bits on the tail will ~~be~~ removed.

~~There~~ There are 10^5 bits we don't know when recovering. Thus we can recover right is ~~is~~ ~~is~~ w.p. $(\frac{1}{2})^{10^5}$, but we can't recover from this compression is $1 - (\frac{1}{2})^{10^5}$

Use Yao's Principle, we know that randomized algorithm can't get better accuracy than deterministic, so ~~this~~ her claims cannot be true.

Midterm

Jing Bao (20522065)

P6 (a) suppose for each ball i X_{ij} indicate color is j

$$X_j = \sum_i X_{ij} \quad E(X_j) = E\left(\sum_i X_{ij}\right) = \sum_i E(X_{ij}) = \sum_i p_i$$

Use Chernoff Bound

$$\Pr(|X_j - mp_j|) = \Pr(m|q_i^{(m)} - p_i|)$$

$$\Pr(m|q_i^{(m)} - p_i| \geq m\epsilon) = \Pr(|X_j - mp_j| \geq m\epsilon) \leq 2e^{-\frac{m p_i (\frac{\epsilon}{p_i})^2}{3}} = 2e^{-\frac{m \epsilon^2}{3 p_i}}$$

$$\text{let } m = \frac{3 p_i}{\epsilon^2} \ln \frac{2}{\delta}, \text{ because } m = O(\epsilon^{-2} \log \delta^{-1})$$

$$\text{we get } \Pr(|X_j - mp_j| \geq m\epsilon) \leq 2e^{-\frac{m \epsilon^2}{3 p_i}} = \delta$$

so proof

$$(b) \text{ Because } \max_{S \subseteq [n]} \left| \sum_{i \in S} q_i^{(m)} - \sum_{i \in S} p_i \right| \leq \epsilon$$

which means for every $S \subseteq [n]$ (a) is true

$$\text{so } \Pr\left(\left| \sum_{i \in S} q_i^{(m)} - \sum_{i \in S} p_i \right| \geq \epsilon\right) \leq \frac{\delta}{2^n}$$

which δ for each element in $[n]$, S can pick or not pick, so 2^n choices.

$$\text{let } m = \frac{3 p_i}{\epsilon^2} (n + \log \frac{2}{\delta}) \text{ we get this answer.}$$

$$\text{we get } \Pr\left(\left| \sum_{i \in S} q_i^{(m)} - \sum_{i \in S} p_i \right| \geq \epsilon\right) \leq \frac{\delta}{2^n}$$

use union bound ~~max~~ ~~Pr~~ ~~for~~ ~~e~~

$$\Pr(\forall S \subseteq [n] \left| \sum_{i \in S} q_i^{(m)} - \sum_{i \in S} p_i \right| \geq \epsilon) \leq \delta$$

so we get the answer.

(c) from (a) we get

$$\Pr(|X_i - \mu p_i| \geq m\epsilon) \leq 2e^{-\frac{m\epsilon^2}{3p_i}} \leq \frac{2e^{-\frac{m\epsilon^2}{3}}}{3}$$

Here we let $m = \frac{3p_i}{\epsilon^2} (n\delta^{-1})$

$$\text{we get } \Pr(|X_i - \mu p_i| \geq m\epsilon) \leq \frac{\delta}{n}$$

use union bound

$$\Pr(\max_{i \in [n]} |q_i^{(m)} - p_i| \geq \epsilon) \leq \Pr(\exists i \in [n] |m|q_i^{(m)} - p_i| \geq m\epsilon) \leq \frac{\delta}{n} \cdot n$$

we get proof