

30/11 LECTURE 18: MORE ON RANDOM VARIABLES.

Eg 18.1: (Coupon Collector problem) There are n ^{distinct} coupons. Each day a coupon is selected indep. & UAR. All coupons are available for selection every day. (Sampling with replacement).

T = # of days needed to get all n ^{distinct} coupons. (r.v.)

Fix $m \geq 1$. $P(T > m)$ = Prob. that in m days, I haven't got

$P(T = m)$ = Prob. that I collect the n ^{distinct} coupons. m^{th} distinct coupon on

$$\{T = m\} = \{T > m-1, T \leq m^{\text{th}} \text{ day}\} = \{T > m-1\} \cap \{T > m\}^c$$

$$P(T = m) = P(T > m-1) - P(T > m)$$

$T > m$ - depends on the coupons received on the first m' days.

A_j = j^{th} coupon isn't selected at all on the $- m'$ days.

$\{T > m\}$ = atleast one coupon isn't selected at all $- m'$ days

$$= \bigcup_{j=1}^n A_j^c$$

$$P(T > m) = P\left(\bigcup_{j=1}^n A_j^c\right) = S_1 - S_2 + S_3 - \dots + (-1)^{n+1} S_n$$

$$S_k = \sum_{i_1, \dots, i_k}^{\neq} P(A_{i_1}^c \cap \dots \cap A_{i_k}^c)$$

$$P(A_i^c) = P(i^{\text{th}} \text{ coupon isn't selected on day } 1, \dots, m)$$

$$\text{(indep. of selections)} = \prod_{l=1}^m P(i^{\text{th}} \text{ on day } l)$$

$$= \prod_{l=1}^m \left(\frac{n-1}{n}\right) = \left(\frac{n-1}{n}\right)^m$$

$$P(A_i^c \cap A_j^c) = P(i^{\text{th}} \& j^{\text{th}} \text{ coupon aren't } \dots \text{ on days } 1, \dots, m)$$

$$= \left(\frac{n-2}{n}\right)^m$$

$$P(A_{i_1} \cap \dots \cap A_{i_k}) = \left(\frac{n-k}{n}\right)^m$$

$$\Rightarrow S_k = \binom{n}{k} \left(\frac{n-k}{n}\right)^m$$

$$P(T > m) = \sum_{k=1}^n (-1)^{k-1} \binom{n}{k} \left(\frac{n-k}{n}\right)^m.$$



Job of prob - Compute prob. $\rightarrow$ Compute $E[f(x)]$.

Propn
18.2

(MARKOV'S INEQ.) Let X be a r.v. & $X \geq 0$ (i.e., $P_X(x) = 0 \forall x < 0$).

then $P(X \geq t) \leq \frac{EX}{t} \quad \forall t > 0$.

[Check if $X(\omega) \geq 0 \forall \omega \in \Omega$ then $P_X(x) = 0 \forall x < 0$].

Proof:

T.S.T.

$$t P(X \geq t) \leq E[X]$$

$$t E[\mathbb{1}_{[X \geq t]}] \leq E[X] \quad \square$$

linearity

$$\xRightarrow{\uparrow} E[t \mathbb{1}_{[X \geq t]}] \leq E[X]$$

Observe that $t \mathbb{1}[X \geq t] \leq X \quad \forall t > 0$.

$$\left[\begin{array}{l} X(\omega) \geq t \Rightarrow \text{LHS} = t \quad ; \quad \text{RHS} = X(\omega) \text{ \& LHS} \leq \text{RHS} . \\ X(\omega) < t \Rightarrow \text{LHS} = 0 \quad ; \quad \text{RHS} = X(\omega) \geq 0 \text{ \& LHS} \leq \text{RHS} \end{array} \right]$$

So by monotonicity of $\mathbb{E}$, we have that

$$\mathbb{E}[t \mathbb{1}[X \geq t]] \leq \mathbb{E}[X]$$

$$\& \text{ so } P(X \geq t) \leq \frac{\mathbb{E}[X]}{t} \quad \forall t > 0. \quad \square$$

Cor 18.3 Let X be a r.v. & let $f: \mathbb{R} \rightarrow \mathbb{R}_+ = (0, \infty)$ & f is non-decreasing.
Then $\forall t \in \mathbb{R}$

$$P(X \geq t) \leq \frac{\mathbb{E} f(X)}{f(t)} \quad \forall t \in \mathbb{R}.$$

Proof: If $X \geq t$ then $f(X) \geq f(t)$ as f is non-decreasing.

$$\text{So } \{X \geq t\} = \{\omega: X(\omega) \geq t\} \subseteq \{f(X) \geq f(t)\}$$

By monotonicity, $P(X \geq t) \leq P(f(X) \geq f(t))$
 (M. Ineq. as $f(X) \geq 0$) $\leq \frac{E[f(X)]}{f(t)}$ $\square$

Ex 18.4 let $f(x) = e^{sx}$, $s > 0$, $x \in \mathbb{R}$, $f(x) > 0$ & non-decreasing.

By above cor $P(X \geq t) \leq e^{-st} E[e^{sx}] \quad \forall t \in \mathbb{R}.$
 (Chernoff's bound).

Ex 18.5 let X be a r.v. $P(|X| \geq t) \leq \frac{E[X^2]}{t^2} \quad \forall t > 0.$

Ex 18.6: $\Omega = S_n$, $P(\sigma) = \frac{1}{n!}$ $\sigma \in S_n$.

Fix $k \in \{2, \dots, n\}$

$P(\exists \text{ a increasing subseq. of length } k) = ?$

Increasing subseq. of length k is $i_1 < i_2 < \dots < i_k$
 $\exists \sigma(i_1) < \sigma(i_2) < \dots < \sigma(i_k)$

Ex. $n=3, k=2$. $\sigma = (\sigma(1), \sigma(2), \sigma(3)) = (2, 3, 1)$.

only one incr. subseq. of length 2, $\sigma(1) < \sigma(2)$.

$\sigma = (1, 3, 2)$, 2 incr. subseq. of length 2 $\sigma(1) < \sigma(2)$

$\sigma = (1, 2, 3)$, 3 - - - of 2 $\sigma(1) < \sigma(3)$.

$X_k(\sigma) = \#$ of increasing subsequences of length k . (r.v.)

$P(X_k \geq m) = ?$ i.e., Prob. $\exists$ m increasing subsequence of length k .

$L(\sigma) =$ Length of longest increasing subsequence. (r.v.)
 $= \max \{ k : \exists \text{ an incr. subsequence of length } k \}$

$$L: S_n \rightarrow [n] \quad ; \quad X_k: S_n \rightarrow \{0\} \cup \left[\binom{n}{k} \right]$$

(check $X_k(\sigma) \leq \binom{n}{k} \forall \sigma \in S_n$)

$$L(\sigma) = \max \{k : X_k(\sigma) \geq 1\}$$

$$\{L(\sigma) \geq k\} = \{X_k(\sigma) \geq 1\}.$$

$$\{L \geq k\} := \{\sigma : L(\sigma) \geq k\} \quad \{X_k \geq 1\} = \{\sigma : X_k(\sigma) \geq 1\}$$

$$P(L \geq k) = P(X_k \geq 1)$$

$$\text{(Markov's inequality)} \rightarrow \leq E[X_k].$$

$$X_k(\sigma) = \# \{ (i_1, \dots, i_k) : i_1 < \dots < i_k \text{ \& } \sigma(i_1) < \dots < \sigma(i_k) \}$$

$$= \sum_{i_1 < \dots < i_k} \mathbb{1}[\sigma(i_1) < \dots < \sigma(i_k)]$$

$$\begin{aligned}
 X_{i_1, \dots, i_k}(\sigma) &= \mathbb{1}[\sigma(i_1) < \dots < \sigma(i_k)] \\
 \text{Use linearity} \quad \mathbb{E}[X_k] &= \sum_{i_1 < \dots < i_k} \mathbb{E}[X_{i_1, \dots, i_k}] \\
 (\text{check}) \quad &= \sum_{i_1 < \dots < i_k} \mathbb{P}(\{\sigma : \sigma(i_1) < \dots < \sigma(i_k)\}) \\
 &= \sum_{i_1 < \dots < i_k} \frac{|\{\sigma : \sigma(i_1) < \dots < \sigma(i_k)\}|}{n!} \\
 (\text{check}) \quad &= \sum_{i_1 < \dots < i_k} \frac{1}{k!} = \binom{n}{k} \frac{1}{k!}
 \end{aligned}$$

$\mathbb{P}(L \geq k) \leq \binom{n}{k} \frac{1}{k!}$ Find seq. $k_n \Rightarrow \text{RHS} \rightarrow 0$ as $n \rightarrow \infty$. 

Dan Romik — "The beautiful mathematics of LIS".

Exo 1.50 union bound on combato. $\mathbb{P}(\forall i, v_i > 1)$.

Defn
18.7

Define variance of X ($\text{VAR}[X]$) by

$$\text{VAR}[X] := \mathbb{E}[(X - \mathbb{E}[X])^2]$$

(STD. DEVN.) $\text{SD}[X] := \sqrt{\text{VAR}[X]}$

Obs:

$$\text{VAR}[X] = \mathbb{E}[X^2] - (\mathbb{E}[X])^2$$

$$\text{VAR}[X] = \mathbb{E}[(X - \mathbb{E}[X])^2] = \mathbb{E}[X^2 - 2X\mathbb{E}[X] + (\mathbb{E}[X])^2]$$

$$\stackrel{\text{(lin. of exp.)}}{=} \mathbb{E}[X^2] - 2\mathbb{E}[X\mathbb{E}[X]] + \mathbb{E}[(\mathbb{E}[X])^2]$$

$$\stackrel{\downarrow \text{linearity}}{=} \mathbb{E}[X^2] - 2\mathbb{E}[X]\mathbb{E}[X] + (\mathbb{E}[X])^2$$

$$= \mathbb{E}[X^2] - (\mathbb{E}[X])^2. \quad \square$$

Exo If X is const r.v. (i.e. $P_X(a) = 1$ for some $a \in \mathbb{R}$) then $\mathbb{E}[X] = a$.
