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Variable & Correlation.

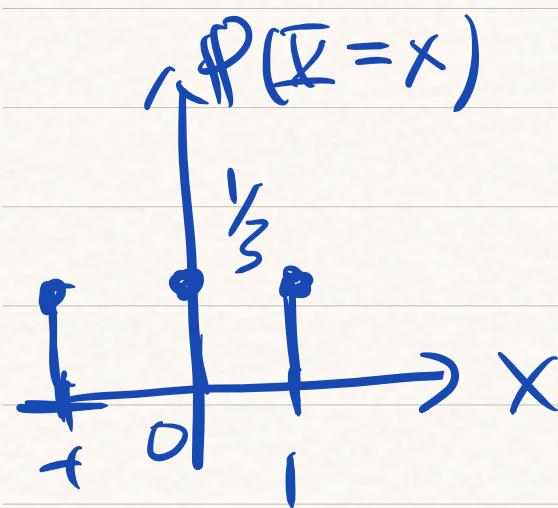
Function of Random Variables

Ex ① $Y = cX$ $c \in \mathbb{R}$ ($c \neq 0$). find $E(Y)$

$$E(f(X)) = \sum_x f(x) P_X(X=x)$$

$$E(Y) = \sum_x cx P_X(X=x) = c E(X)$$

Ex ②. $\Sigma \sim U[-1, 1]$, $Y = \Sigma^2$ find P_Y & $E[Y]$



$y = x^2$	-1	0	1
x	-1	0	1
$P(\Sigma=x)$	$\frac{1}{3}$	$\frac{1}{3}$	$\frac{1}{3}$

$$P(Y=0) = \frac{1}{3} \quad P(Y=1) = \frac{2}{3}$$

$$E(Y) = 0 \cdot \frac{1}{3} + 1 \cdot \frac{2}{3} = \frac{2}{3}$$

Ex(3) $\Sigma \sim U[-N, N]$, $Y = \Sigma^2$ find P_Y and $E(Y)$

$$\left[\sum_{k=1}^n k^2 = \frac{n(n+1)(2n+1)}{6} \right]$$

$$P_\Sigma(\Sigma=x) = \begin{cases} \frac{1}{2N+1} & x \in [-N, N] \\ 0 & \text{o.w.} \end{cases}$$

$$P_Y(Y=N^2+N) = 0$$

$$P_Y(Y=y=x^2) = \begin{cases} \frac{2}{2N+1} & y \in [1, N^2] \\ \frac{1}{2N+1} & y = 0 \\ 0 & \text{o.w.} \end{cases} \leftarrow$$

$$E(Y) = 0 \cdot \frac{1}{2N+1} + \sum_{k=1}^N k^2 \frac{2}{2N+1} + 0 = \frac{2}{2N+1} \sum_{k=1}^N k^2 = \frac{N(N+1)}{3}$$

Variance of Random Variable (r.v) $\Sigma : \Omega \rightarrow \mathbb{R}$

Def. a. r.v. Σ w/ $E(\Sigma) = \mu$. the variance of Σ is

$$\text{Var}(\Sigma) = E((\Sigma - \mu)^2)$$

$\sigma(\Sigma) := \sqrt{\text{Var}(\Sigma)}$ is standard deviation of Σ

$$\begin{aligned}\text{Var}(\Sigma) &= E((\Sigma - \mu)^2) = E(\Sigma^2 + \mu^2 - 2\mu\Sigma) = E(\Sigma^2) + \mu^2 - 2\mu E(\Sigma) \\ &= E(\Sigma^2) - \mu^2 = E(\Sigma^2) - [E(\Sigma)]^2 \quad (\text{long - short})\end{aligned}$$

Ex 1 . $Y = c\Sigma$: $E(Y) = cE(\Sigma)$ if Σ : var (Σ). what is var (Y)?

$$\text{var}(Y) = E(Y^2) - [E(Y)]^2 = c^2 E(\Sigma^2) - c^2 [E(\Sigma)]^2 = c^2 \text{Var}(\Sigma)$$

1. 6-faced dice Σ , find $\text{Var}(\Sigma)$

$$\mathbb{E}(\Sigma) = \frac{1}{6} (1+2+\dots+6) = \frac{(1+6) \cdot 6}{6 \cdot 2} = \frac{7}{2}$$

$$\mathbb{E}(\Sigma^2) = \frac{1}{6} \cdot 1^2 + \frac{1}{6} \cdot 2^2 + \dots + \frac{1}{6} \cdot 6^2 = \frac{91}{6}$$

$$\text{Var}(\Sigma) = \mathbb{E}(\Sigma^2) - \mathbb{E}(\Sigma)^2 = \frac{91}{6} - \frac{49}{4} = \frac{35}{12}$$

2: Uniform dist. $\Sigma \sim U\{1, 2, \dots, n\}$. $P(\Sigma=x) = \frac{1}{n} \quad x \in \{1, n\}$

$$\mathbb{E}(\Sigma) = \frac{n+1}{2} = \frac{1}{n} (1+2+\dots+n) = \frac{1}{n} \cdot \frac{n(1+n)}{2}$$

$$\begin{aligned} \text{Var}(\Sigma) &= \underline{\mathbb{E}(\Sigma^2)} - \underline{\mathbb{E}(\Sigma)^2} = \underline{\frac{1}{n}(1^2+2^2+\dots+n^2)} - \underline{\frac{(n+1)^2}{4}} \\ &= \frac{n^2-1}{12} \end{aligned}$$

$$\mathbb{E}(X^2) = \frac{1}{n} (1^2 + 2^2 + \dots + n^2)$$

$$= \frac{1}{n} \frac{n(n+1)(2n+1)}{6} = \frac{(n+1)(2n+1)}{6}$$

$$\sum_{n=1}^N n^2 = \frac{N(N+1)(2N+1)}{6}$$

$$\mathbb{E}(X)^2 = \frac{(n+1)^2}{4}$$

$$\text{Var}(X) = \mathbb{E}(X^2) - \mathbb{E}(X)^2 = \frac{2(n+1)(2n+1) - 3(n+1)^2}{12}$$

$$= \frac{4n^2 + 6n + 2 - 3n^2 - 3 - 6n}{12}$$

$$\approx \frac{n^2 - 1}{12}$$

two r.v. \bar{X}, Y , $E(X)$, $E(Y)$, $\text{Var}(X)$, $\text{Var}(Y)$

new r.v. $Z = \bar{X} + Y$. find $\text{Var}(Z)$.

$$\begin{aligned}\text{Var}(Z) &= \text{Var}(\bar{X} + Y) = E((\bar{X} + Y)^2) - (E(\bar{X} + Y))^2 \\ &= E(\bar{X}^2 + Y^2 + 2\bar{X}Y) - (E(\bar{X}) + E(Y))^2 \\ &= \underbrace{E(\bar{X}^2)}_{\text{orange}} + \underbrace{E(Y^2)}_{\text{green}} + 2\underbrace{E(\bar{X}Y)}_{\text{yellow}} - \underbrace{E(\bar{X})^2}_{\text{orange}} - \underbrace{E(Y)^2}_{\text{green}} \rightarrow E(\bar{X})E(Y) \\ &= \text{Var}(\bar{X}) + \text{Var}(Y) + 2(E(\bar{X}Y) - E(\bar{X})E(Y))\end{aligned}$$

if $\bar{X} \perp Y$ then $E(\bar{X}Y) - E(\bar{X})E(Y) = 0$

meaning \bar{X} independent of Y

(the other way is not true!)

Covariance

$$\begin{aligned} \text{Def.: } \text{cov}(X, Y) &= E((X - E(X))(Y - E(Y))) \\ &= E(XY) - E(Y)E(X) - E(X)E(Y) + E(X)E(Y) \\ &= E(XY) - E(X)E(Y) \end{aligned}$$

$E(Y|E(X)) = E(X)E(Y)$

facts: 1) $\text{var}(X+Y) = \text{var}(X) + \text{var}(Y) + 2 \text{cov}(X, Y)$

2) $\text{cov}(X, X) = \text{var}(X)$

3) $\text{cov}(aX, bY) = ab \text{cov}(X, Y)$

4) $X \perp Y \Rightarrow \text{cov}(X, Y) = 0$

Correlation (Coefficient)

Def. r.v. X, Y . $\sigma(X) > 0$. $\sigma(Y) > 0$

$$\text{corr}(X, Y) = \frac{\text{cov}(X, Y)}{\sigma(X)\sigma(Y)} \in [-1, 1]$$

→ normalized cov.

Ex. show $\text{corr}(\cdot, \cdot) \in [-1, 1]$

proof: let $\tilde{X} = \frac{(X - \mathbb{E}(X))}{\sigma(X)}$

$$\begin{aligned}\mathbb{E}(\tilde{X}) &= 0 \\ \text{var}(\tilde{X}) &= 1\end{aligned}$$

$$\begin{aligned}\mathbb{E}(\tilde{X}^2) &= \mathbb{E}\left(\frac{X^2 + \mathbb{E}^2(X) - 2X\mathbb{E}(X)}{\sigma^2(X)}\right) \\ &= \frac{\mathbb{E}(X^2) - \mathbb{E}(X)^2}{\text{var}(X)} = 1\end{aligned}$$

$$\mathbb{E}(\tilde{X}^2) = 1$$

$$\tilde{\Sigma} = (\Sigma - \mathbb{E}(\Sigma)) / \sigma(\Sigma)$$

$$\mathbb{E}(\tilde{\Sigma}) = 0, \text{Var}(\tilde{\Sigma}) = 1$$

$$\tilde{Y} = (Y - \mathbb{E}(Y)) / \sigma(Y)$$

$$\mathbb{E}(\tilde{Y}) = 0, \text{Var}(\tilde{Y}) = 1$$

for $\tilde{\Sigma} + \tilde{Y}$ $\tilde{\Sigma} - \tilde{Y}$

$$0 \leq \mathbb{E}((\tilde{\Sigma} + \tilde{Y})^2) = \mathbb{E}(\tilde{\Sigma}^2) + \mathbb{E}(\tilde{Y}^2) + 2\mathbb{E}(\tilde{\Sigma}\tilde{Y}) \leq 2 + 2\mathbb{E}(\tilde{\Sigma}\tilde{Y})$$

$$0 \leq \mathbb{E}((\tilde{\Sigma} - \tilde{Y})^2) = \mathbb{E}(\tilde{\Sigma}^2) + \mathbb{E}(\tilde{Y}^2) - 2\mathbb{E}(\tilde{\Sigma}\tilde{Y}) \leq 2 - 2\mathbb{E}(\tilde{\Sigma}\tilde{Y})$$

$$\mathbb{E}(\tilde{\Sigma}\tilde{Y}) \in [-1, 1] \quad (\text{see next page})$$

$$-1 \leq \text{corr}(\Sigma, Y) = \frac{\text{cov}(\tilde{\Sigma}, \tilde{Y})}{\sqrt{\text{var}(\tilde{\Sigma})\text{var}(\tilde{Y})}} = \frac{\mathbb{E}(\tilde{\Sigma}\tilde{Y})}{\sqrt{2}} \leq 1$$

* next page

$$0 \leq 2 + 2\mathbb{E}(\tilde{X}\tilde{Y}) \Rightarrow \mathbb{E}(\tilde{X}\tilde{Y}) \geq -1$$

$$0 \leq 2 - 2\mathbb{E}(\tilde{X}\tilde{Y}) \Rightarrow \mathbb{E}(\tilde{X}\tilde{Y}) \leq 1 \quad \left. \begin{array}{l} \\ \end{array} \right\} \Rightarrow -1 \leq \mathbb{E}(\tilde{X}\tilde{Y}) \leq 1$$

when $\Sigma = Y$ $\text{corr}(\Sigma, Y) = 1$

when $\Sigma = -Y$, $\text{corr}(\Sigma, Y) = -1$

* $\text{corr}(\Sigma, Y) = \frac{\text{cov}(\Sigma, Y)}{\sigma(\Sigma)\sigma(Y)} = \text{cov}\left(\frac{\Sigma}{\sigma(\Sigma)}, \frac{Y}{\sigma(Y)}\right)$

$$= \mathbb{E}\left(\left(\frac{\Sigma}{\sigma(\Sigma)} - \mathbb{E}\left(\frac{\Sigma}{\sigma(\Sigma)}\right)\right)\left(\frac{Y}{\sigma(Y)} - \mathbb{E}\left(\frac{Y}{\sigma(Y)}\right)\right)\right)$$

$$= \mathbb{E}\left(\left(\frac{\Sigma - \mathbb{E}(\Sigma)}{\sigma(\Sigma)}\right)\left(\frac{Y - \mathbb{E}(Y)}{\sigma(Y)}\right)\right) = \mathbb{E}(\tilde{\Sigma}\tilde{Y}) = \mathbb{E}(\tilde{\Sigma}\tilde{Y}) - \mathbb{E}(\tilde{\Sigma})\mathbb{E}(\tilde{Y})$$

$$= \text{cov}(\tilde{\Sigma}\tilde{Y})$$

$$\uparrow \tilde{\Sigma} = \frac{\Sigma - \mathbb{E}(\Sigma)}{\sigma(\Sigma)}, \tilde{Y} = \frac{Y - \mathbb{E}(Y)}{\sigma(Y)}$$

if $\Sigma = Y$ then $\text{corr}(\Sigma, Y) = 1$

↑ there's a glitch here
I can't remove it :(

Independence \Rightarrow uncorrelated
 \Leftarrow

indep. $X \perp Y \Leftrightarrow P_{XY}(x, y) = P_X(x) P_Y(y)$

uncorr. $\text{cov}(X, Y) = 0 \Leftrightarrow E(XY) = E(X)E(Y)$

two r.v. can be uncorrelated but dependent,
but if they are independent, they must be uncorrelated.

Ex ④. $\Sigma \sim U[-1, 1]$, $Y = \Sigma^2$

uncorrelated \nRightarrow independent

\hookrightarrow they are dependent

$$E(\Sigma) = 0, E(Y) = \frac{2}{3} \quad \underline{E(\Sigma Y) = 0} \quad \underline{\text{cov}(\Sigma, Y) = 0}$$

Σ	-1	0	1
Y	1	0	1
ΣY	-1	0	1
P	$\frac{1}{3}$	$\frac{1}{3}$	$\frac{1}{3}$

$$\begin{aligned} E(\Sigma Y) - E(\Sigma)E(Y) \\ = 0 - 0 \cdot \frac{2}{3} \\ = 0 \end{aligned}$$

$$0 = E(\Sigma Y) - E(\Sigma)E(Y)$$

\downarrow def of uncorrelated

$$E(\Sigma Y) = -1 \cdot \frac{1}{3} + 0 \cdot \frac{1}{3} + 1 \cdot \frac{1}{3} = 0$$