

Observe $Y \rightsquigarrow X$?
 (X, Y)



$$\text{Joint cdf } F_{X,Y}(x,y) = P(X \leq x, Y \leq y)$$

$$\text{- discrete : pmf } p_{X,Y}(x,y) = P(X=x, Y=y) = P((X,Y) \in A)$$

$$F_{X,Y}(x,y) = \sum_{\substack{x' \leq x \\ y' \leq y}} p_{X,Y}(x',y') = \sum_{(x',y') \in A} p_{X,Y}(x',y')$$

$$\text{marginal pmf } p_X(x) = \sum_y p_{X,Y}(x,y)$$

$$p_{X|Y}(x|y) = \frac{p_{Y|X}(y|x) p_X(x)}{p_Y(y)}$$

\downarrow
posterior

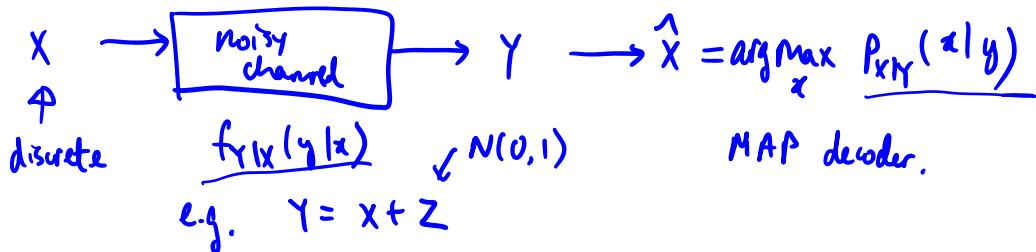
likelihood model
Bayes' rule
prior

$$\text{- Continuous joint pdf } f_{X,Y}(x,y) = f_X(x) f_{Y|X}(y|x)$$

$$\text{marginal pdf } f_X(x) = \int_{-\infty}^{\infty} f_{X,Y}(x,y) dy$$

$$f_{X|Y}(x|y) = \frac{f_{Y|X}(y|x) f_X(x)}{f_Y(y)}$$

- MAP decoder "a posterior prob."



Expectation $\int_X: \Omega \rightarrow \mathbb{R}$

Assume pdf/pmt "mean" $E[X] = \int_{-\infty}^{\infty} x f_X(x) dx$ constant

discrete $\sum_x x p_X(x)$ constant

" \int " = " \bar{z} "

E - expectation operator: $X \mapsto X'$

"mean"
 "average"
 constant
 $\xrightarrow{\text{constant}}$
 operator: $X \mapsto X'$
 r.v. r.v.
 $g(x) \rightarrow \text{r.v.}$
 $\mathbb{E}[g(x)] = \int g(x) f_X(x) dx$
 $x=c \text{ w.p. } 1 \quad P_X(x) = \begin{cases} 1 & \text{if } x=c \\ 0 & \text{o.w.} \end{cases}$

$$X = c \rightarrow \text{constant.} \quad \mathbb{E}[c] = \sum_x x P_X(x) = c$$

$$\rightarrow \mathbb{E}[ax + by] = a \mathbb{E}[x] + b \mathbb{E}[y]$$

Linearity of \int or \sum .

$$Y = g(x) \sim R_Y(y)$$

$$\begin{aligned} \mathbb{E}[Y] &= \sum_y y P_Y(y) = \sum_y y \left[\sum_{x: g(x)=y} P_X(x) \right] \\ &= \left[\sum_y \sum_{x: g(x)=y} y P_X(x) \right] = \left[\sum_x y P_X(x) \right] \end{aligned}$$

$$\begin{aligned} x &\rightarrow g(x) = y_1 & \sum_{x: g(x)=y_1} &= y_1 \\ &\rightarrow g(x) = y_2 & \sum_{g(x)=y_2} &= y_2 \end{aligned}$$

$$= \int_{-\infty}^{\infty} g(x) f_X(x) dx \quad \sum_y \sum_{x: g(x)=y} = \sum_x$$

$$y = \begin{cases} 1 & \sum_{x: g(x)=1} \\ 2 & \vdots \\ 3 & \end{cases} + \sum_{x: g(x)=2} + \sum_{x: g(x)=3}$$

$$\begin{aligned} x & \\ X_1, X_2, \dots, X_n & \\ \frac{1}{n} \sum_{i=1}^n X_i & \xrightarrow[n \rightarrow \infty]{} \mathbb{E}[X] \quad \text{SLLN} \end{aligned}$$

$$\begin{aligned} & \uparrow \\ & 1, 3, 5 \\ & 2, 4, 10 \\ & 6, 7, 8, 9 \\ & \uparrow \\ & g(1) = g(3) = g(5) \\ & g(2) = g(4) = g(10) \\ & = 2 \end{aligned}$$

$$\mathbb{E}[X^2] = \int x^2 f(x) dx$$

$$\begin{aligned} 0 \leq \text{Var } X &= \mathbb{E}[(X - \mathbb{E}[X])^2] = \mathbb{E}[X^2 - 2X\mathbb{E}[X] + (\mathbb{E}[X])^2] \\ &= \mathbb{E}[X^2] - 2\mathbb{E}[X] \cdot \mathbb{E}[X] + (\mathbb{E}[X])^2 \end{aligned}$$

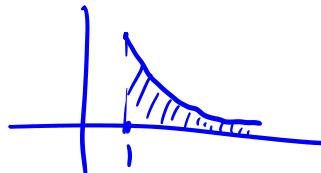
$$I - r.v.$$

$$\mathbb{E}[X^2] \geq (\mathbb{E}[X])^2$$

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

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

$$s.d. = \sqrt{\text{var}(X)}$$



$$f_X(x) = \begin{cases} \frac{1}{\pi} & 1 \leq x < \infty \\ 0 & \text{o.w.} \end{cases}$$

$$\mathbb{E}X = \underbrace{\int_{-\infty}^{\infty} xf_X(x) dx}_{\parallel} = \int_1^{\infty} \frac{x}{x^2} dx = \int_1^{\infty} \frac{1}{x} dx$$

$$= \log x \Big|_1^{\infty}$$

$$= \int_{-\infty}^0 xf_X(x) dx + \int_0^{\infty} xf_X(x) dx = \infty$$

$$= \underbrace{\int_{-\infty}^0 |x| f_X(x) dx}_{\geq 0} + \underbrace{\int_0^{\infty} |x| f_X(x) dx}_{\geq 0} = \overbrace{a - b}^{+\infty - +\infty}$$

$$f(x) = \frac{1}{\pi(1+x^2)} \quad \mathbb{E}X = \int_{-\infty}^{\infty} \frac{x}{\pi(1+x^2)} dx = \frac{1}{2\pi} \int_{-\infty}^{\infty} \frac{d(1+x^2)}{1+x^2}$$

$$- \int_{-\infty}^0 = \frac{1}{2\pi} \log(1+x^2) \Big|_{-\infty}^0 = \infty$$

$$\int_0^{\infty} = \frac{1}{2\pi} \log(1+x^2) \Big|_0^{\infty} = \infty$$

$\mathbb{E}[|X|] < \infty$ "integrable" = expectation $\mathbb{E}[X]$ exists.

$$X \quad g(x) \quad \boxed{\int \mathbf{1}_{\{X \in A\}}(\omega)} = \begin{cases} 1 & \text{if } X(\omega) \in A \\ 0 & \text{o.w.} \end{cases}$$

$$I_{\{X \geq a\}} = \begin{cases} 1 & \text{if } X \geq a \\ 0 & \text{o.w.} \end{cases}$$

indicator function

r.v. $I_{\{X \geq a\}} = \begin{cases} 1 & \text{if } X \geq a \\ 0 & \text{if } X < a \end{cases}$

$$\begin{aligned} \mathbb{E}[I_{\{X \geq a\}}] &= \int_{-\infty}^{\infty} I_{\{X \geq a\}} f_X(x) dx \\ &= \int_a^{\infty} f_X(x) dx \\ \mathbb{E}[I_{\{X \in A\}}] &= \mathbb{P}(X \in A) = \mathbb{P}(X \geq a) \end{aligned}$$

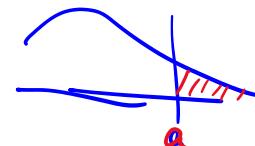
$$\begin{aligned} X > 0, a > 0 & \quad a I_{\{X \geq a\}} \leq I_{\{X \geq a\}} X \leq X \\ a \mathbb{P}(X \geq a) &\leq \mathbb{E}[X] \end{aligned}$$

$$X \leq Y \\ \mathbb{E}X \leq \mathbb{E}Y$$

$$\mathbb{E}[a I_{\{X \geq a\}}] \leq \mathbb{E}[X]$$

$$a \mathbb{P}(X \geq a) \leq \mathbb{E}[X]$$

$$\begin{aligned} X > 0 & \quad \mathbb{P}(X \geq a) \leq \frac{1}{a} \mathbb{E}[X] \\ X > 0, \varepsilon > 0 & \end{aligned}$$



Markov Inequality

$$\text{let } a = \varepsilon \mathbb{E}[X] > 0$$

$$\mathbb{P}(X \geq \varepsilon \mathbb{E}[X]) \leq \frac{1}{\varepsilon \mathbb{E}[X]} \mathbb{E}[X] = \frac{1}{\varepsilon}$$

$$X = \begin{cases} \frac{\varepsilon \mathbb{E}[X]}{\varepsilon} & \text{w.p. } \frac{1}{\varepsilon} \\ 0 & \text{w.p. } 1 - \frac{1}{\varepsilon} \end{cases}$$

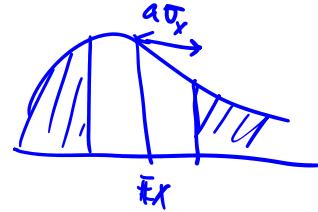
$$\mathbb{P}(X \geq \varepsilon \mathbb{E}[X]) = \frac{1}{\varepsilon}$$

Chebyshov Inequality $\sigma^2 = \text{Var } X$



Chebyshov Inequality $\sigma_x^2 = \text{Var } X$

$$\mathbb{P}(|X - \mathbb{E}X| \geq a\sigma_x) \leq \frac{1}{a^2}$$

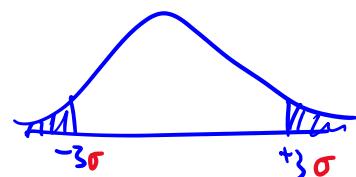


$$Y = |X - \mathbb{E}X|^2 \geq 0 \quad \mathbb{E}Y = \text{Var } X = \sigma_x^2$$

$$\mathbb{P}(Y \geq a^2 \sigma_x^2) \stackrel{\text{II}}{\leq} \frac{1}{a^2 \sigma_x^2} \mathbb{E}[Y] = \frac{1}{a^2}$$

$$\mathbb{P}(|X - \mathbb{E}X|^2 \geq a^2 \sigma_x^2)$$

$$|X - \mathbb{E}X| \geq a\sigma_x$$

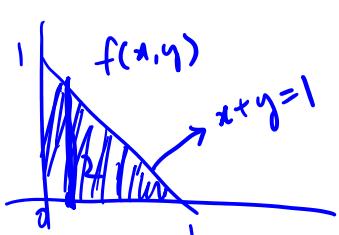


$$\mathbb{E}[g(x, y)] = \iint g(x, y) f_{x,y}(x, y) dx dy$$

Correlation $\mathbb{E}[XY]$

$$\begin{aligned} \text{Cov}(X, Y) &= \mathbb{E}[(X - \mathbb{E}X)(Y - \mathbb{E}Y)] \\ &= \mathbb{E}[XY - X\mathbb{E}Y - Y\mathbb{E}X + \mathbb{E}X \cdot \mathbb{E}Y] \\ &= \mathbb{E}[XY] - \mathbb{E}[X]\mathbb{E}[Y] - \mathbb{E}[Y]\mathbb{E}[X] + \mathbb{E}[X]\mathbb{E}[Y] \\ &= \mathbb{E}[XY] - \mathbb{E}[X]\mathbb{E}[Y] \end{aligned}$$

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



$$\begin{aligned} \mathbb{E}[X] &= \int x f_X(x) dx \\ &= \int x \int f(x,y) dy dx \\ &= \int_0^1 \int_0^{1-x} x \cdot 2 dy dx \end{aligned}$$

$$\text{Cov}(X, Y) = \frac{1}{2} \cdot \frac{1}{2} = \frac{1}{4}$$

$$\text{Cov}(X, Y) = \frac{1}{12} - \frac{1}{9} = \frac{1}{3}$$

$$\iint xy f_{X,Y}(x,y) dy dx = \int_0^1 \int_0^{1-x} 2xy dy dx = \frac{1}{12}$$

$\text{Cov}(X, Y) = 0$ X, Y are uncorrelated

$$X \perp\!\!\!\perp Y \Rightarrow \text{Cov}(X, Y) = 0 \quad \text{have: } \mathbb{E}[XY] = \mathbb{E}[X]\mathbb{E}[Y]$$

$$\mathbb{E}[XY] = \iint xy f_{X,Y}(x,y) dx dy$$

$$= \iint \underline{x} \underline{y} f_{X,Y}(x,y) dx dy$$

$$= \int x f_X(x) dx \cdot \int y f_Y(y) dy$$

$$= \mathbb{E}[X] \cdot \mathbb{E}[Y]$$

$\text{Cov}(X, Y) = 0 \Rightarrow X \perp\!\!\!\perp Y ? \text{ FALSE.}$

$$p(x,y) = \begin{cases} \frac{2}{5} & (x,y) = (+1, +1), (-1, -1) \\ \frac{1}{10} & (x,y) = (+2, -2), (-2, +2) \\ 0 & \text{o.w.} \end{cases} \quad |X| = |Y| \quad p(x,y) \neq p(x)p(y)$$

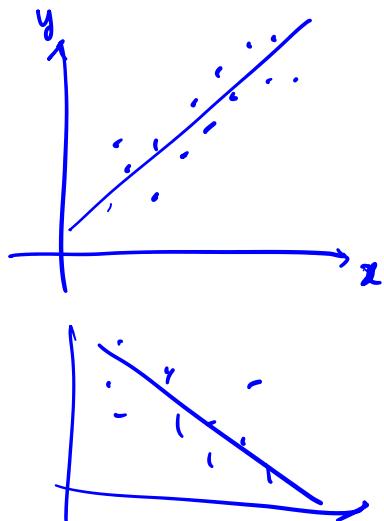
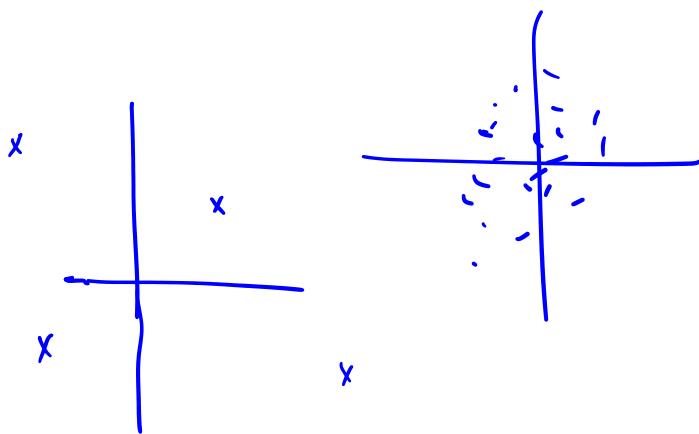
$$\begin{aligned} \mathbb{E}[X] &= \sum_x x p(x) = \sum_x x \sum_y p(x,y) \\ &= 1 \cdot \frac{2}{5} - 1 \cdot \frac{2}{5} + 2 \cdot \frac{1}{10} - 2 \cdot \frac{1}{10} \\ &= 0 \end{aligned}$$

$$\underline{\mathbb{E}[X]\mathbb{E}[Y]} = 0$$

$$\underline{\mathbb{E}[XY]} = (+1)(+1)\frac{2}{5} + (-1)(-1)\frac{2}{5} + (+2)(-2)\frac{1}{10} + (-2)(+2)\frac{1}{10}$$

$$= 0$$

$$\Rightarrow \text{Cov}(X, Y) = 0$$



$$\rho_{X,Y} = \frac{\text{Cov}(X,Y)}{\sqrt{\text{Var}(X)\text{Var}(Y)}} \quad -1 \leq \rho_{X,Y} \leq +1$$

$$P_{X|Y}(x|y) = \frac{P_{X,Y}(x,y)}{P_Y(y)} \quad \checkmark \quad \{x=x\} \cap \{x \in A\}$$

discrete

$$P_{X|A}(x) = \frac{P(x=x | x \in A)}{P(x \in A)} = \frac{P(x=x \cap x \in A)}{P(x \in A)}$$
$$= \begin{cases} \frac{P(x=x)}{P(x \in A)} & \text{if } x \in A \\ 0 & \text{if } x \notin A \end{cases}$$

$$f_{X|A}(x) = \begin{cases} \frac{f_X(x)}{P(x \in A)} & \text{if } x \in A \\ 0 & \text{o.w.} \end{cases}$$

$$A = \{x > a\} \quad P(x \in A) = P(x > a) = e^{-\lambda a}$$

$$f_{X|A}(x) = \lambda e^{-\lambda(x-a)}, \quad x > a \quad f_X(x) = \begin{cases} \lambda e^{-\lambda x} & \text{if } x > 0 \\ 0 & \text{o.w.} \end{cases}$$

$$\underline{f_{X|A}(x)} = \begin{cases} \lambda e^{-\lambda(x-a)}, & x > a \\ 0 & \text{o.w.} \end{cases} \quad f_X(x) = \begin{cases} \lambda e^{-\lambda x} & \text{if } x > 0 \\ 0 & \text{o.w.} \end{cases}$$

$$\underline{\mathbb{E}[g(x) | A]} = \int_{-\infty}^{\infty} g(x) f_{X|A}(x) dx$$



$$\Omega = \bigcup_{i=1}^n A_i \quad A_i \cap A_j = \emptyset, \quad i \neq j$$

$$\underline{\mathbb{E}[g(x)]} = \sum_{i=1}^n P(x \in A_i) \underline{\mathbb{E}[g(x) | A_i]}$$

$$\sum_{i=1}^n P(x \in A_i) \int g(x) f_{X|A_i}(x) dx$$

$$\int g(x) f_{X|A_i}(x) dx$$

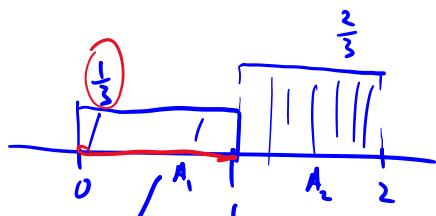
$$= \int g(x) \sum_{i=1}^n \underbrace{P(x \in A_i)}_{f_X(x)} f_{X|A_i}(x) dx$$

$$f_{X|A_i}(x) = \frac{f_X(x)}{P(x \in A_i)} \mathbf{1}_{\{x \in A_i\}}$$

$$= \int g(x) \sum_{i=1}^n f_X(x) \underbrace{\mathbf{1}_{\{x \in A_i\}}}_{\sum_{i=1}^n \mathbf{1}_{\{x \in A_i\}} = 1} dx$$

$$\sum_{i=1}^n \mathbf{1}_{\{x \in A_i\}} = 1$$

$$= \int g(x) f_X(x) dx = \mathbb{E}[g(X)]$$



$$\mathbb{E}[X | A_1] = \frac{1}{2}$$

$$\mathbb{E}[X^2 | A_1] =$$

$$\mathbb{E}[X | A_2] = \frac{1+2}{2} = \frac{3}{2}$$

$$\mathbb{E}[X^2 | A_2] =$$

$$\mathbb{E}[X^2 | A_1] \quad \mathbb{E}[X^2 | A_2] = \frac{7}{3}$$

$$= \int x^2 f_{X|A_1}(x) dx \quad (\text{Ex})$$

$$= \int_0^1 x^2 \left[\frac{1}{3} \right] dx = \frac{1}{3}$$

$$f_{X|A_1}(x) = \frac{f_X(x)}{P(x \in A_1)} \quad x \in A_1$$

$$\begin{aligned} \mathbb{E}[X] &= P(x \in A_1) \mathbb{E}[X | A_1] + P(x \in A_2) \mathbb{E}[X | A_2] \\ &= \frac{1}{3} \cdot \frac{1}{2} + \frac{2}{3} \cdot \frac{3}{2} = \frac{7}{6} \end{aligned}$$

$$\mathbb{E}[X^2] = P(x \in A_1) \mathbb{E}[X^2 | A_1] + P(x \in A_2) \mathbb{E}[X^2 | A_2]$$

$$\begin{aligned} s & \leftarrow s \leftarrow s \\ E[X^2] &= P(X \in A_1) E[X^2 | A_1] + P(X \in A_2) E[X^2 | A_2] \\ &= \frac{1}{3} \cdot \frac{1}{3} + \frac{2}{3} \cdot \frac{7}{3} = \frac{15}{9} \end{aligned}$$

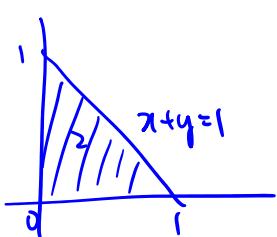
$$\underbrace{f_{X|Y}(x|y)}_{\substack{\text{const.} \\ \downarrow}} = \frac{f_{X,Y}(x,y)}{f_Y(y)} \quad \underbrace{\mathbb{E}[g(x) | Y=y]}_{\substack{\text{const.} \\ \downarrow}} = \int g(x) f_{X|Y}(x|y) dx$$

$$\begin{aligned} \mathbb{E}[g(x, Y) | Y=y] &= \int g(x, y) f_{X|Y}(x|y) dx \\ \mathbb{E}[x | Y=y] &= \int x \downarrow y f_{X|Y}(x|y) dx \\ &= \underset{\text{nm}}{y} \int x f_{X|Y}(x|y) dx \\ &= y \mathbb{E}[x | Y=y] \end{aligned}$$

$$\mathbb{E}[x | Y=y] = h(y) \quad Y \text{ r.v.}$$

$$\underbrace{P(Y=y)}_{\substack{\text{r.v.}}} \rightarrow \mathbb{E}[x | Y=y]$$

$$X \rightarrow \mathbb{E}[x | Y] \quad \begin{matrix} \uparrow \\ \text{r.v.} \end{matrix} \quad \begin{matrix} h(Y) = \mathbb{E}[x | Y] \\ \rightarrow \text{r.v.} \end{matrix} \quad \begin{matrix} \uparrow \\ \text{expectation operator} \end{matrix}$$



$$\begin{aligned} \mathbb{E}[x | Y=y] &= \int x f_{X|Y}(x|y) dx \\ &= \int_0^{1-y} \frac{x}{1-y} dx \quad 1-y \end{aligned}$$

$$\begin{aligned}
 & \int_0^{1-y} x \, dx \\
 &= \frac{1}{2} \int_0^{1-y} x^2 \, dx \\
 &= \frac{1}{2} \frac{(1-y)^2}{2} \\
 &= \frac{1-y}{2} \quad 0 \leq y < 1
 \end{aligned}$$

$$\mathbb{E}[x|Y=y] = y \mathbb{E}[x|Y=y] = \frac{y(1-y)}{2}$$

$\mathbb{E}[x|Y]$ r.v.
pdf?

$$\boxed{\mathbb{E}[x|Y]} + \boxed{\frac{1-y}{2} = z}$$

$$\begin{aligned}
 f_Y(y) &= \begin{cases} 2(1-y) & 0 \leq y < 1 \\ 0 & \text{else.} \end{cases} \\
 f_Z(z) &= \frac{f_Y(y)}{|g'(y)|} \\
 &= \frac{f_Y(1-z)}{\frac{1}{2}} \quad 0 < z < \frac{1}{2} \\
 &= 2 \cdot 2(1-1+2z) \\
 &= g_z
 \end{aligned}$$

$$\mathbb{E}[z] = \mathbb{E}[\mathbb{E}[x|Y]] = \mathbb{E}[x]$$

$\mathbb{E}[\mathbb{E}[g(x,Y) | Y]] = \mathbb{E}[g(x,Y)]$

$$= \int [\mathbb{E}[g(x,Y) | Y=y]] f_Y(y) dy$$

$$= \int \left[\int g(x,y) f_{X|Y}(x|y) dx \right] f_Y(y) dy$$

$$= \iint g(x,y) \underbrace{f_{X|Y}(x|y) f_Y(y)}_{\text{pdf}} dx dy$$

$$= \iint g(x,y) \underbrace{f_{X,Y}(x,y)}_{\text{joint pdf}} dx dy = \mathbb{E}[g(x,Y)]$$

$$P \sim f_p(p) = 2(1-p), 0 \leq p \leq 1$$

$$N = \# \text{ of heads} \quad X_i = \begin{cases} 1 & \text{if } i\text{th toss is head} \\ 0 & \text{o.w.} \end{cases}$$

$$X_1, X_2, \dots, X_n \quad N = \sum_{i=1}^n X_i$$

$$P(X_i=1 | P) = p \quad E[X_i | P] = p$$

$$\begin{aligned} E[N | P] &= E\left[\sum_{i=1}^n X_i | P\right] \\ &= \sum_{i=1}^n E[X_i | P] = nP \end{aligned}$$

$$E[N] = E[E[N | P]] = E[nP]$$

$$= n E[P]$$

$$= n \int_0^1 p^2(1-p) dp = \frac{n}{3}$$

$$E[X | Y] = Y^2 \quad Y \sim \text{Unif}[0, 1]$$

$$E[X] = E[E[X | Y]]$$

$$= E[Y^2] = \int_0^1 y^2 \cdot 1 dy = \frac{1}{3}$$

r.v.

$$\underbrace{\text{Var}(X | Y=y)}_{f_{X|Y}(x|y)} = E\left[\frac{(X - E[X | Y=y])^2}{f_{X|Y}(x|y)} | Y=y\right] = \frac{E[X^2 | Y=y] - (E[X | Y=y])^2}{f_{X|Y}(x|y)}$$

$$\rightarrow E[\text{Var}(X | Y)] = \underbrace{E[X^2]}_{\text{r.v.}} - \underbrace{E[(E[X | Y])^2]}_{\text{r.v.}}$$

$$\begin{aligned} \rightarrow \text{Var}(E[X | Y]) &= E[(E[X | Y] - E[X])^2] \\ &\stackrel{\uparrow \text{r.v.}}{=} E[\underbrace{E[X | Y]^2}_{\text{r.v.}} - 2 E[X] \underbrace{E[X | Y]}_{\text{r.v.}} + \underbrace{E[X]^2}_{\text{r.v.}}] \\ &= E[E[X | Y]^2] - \underbrace{E[E[X | Y]]^2}_{\text{r.v.}} \end{aligned}$$

$$\begin{aligned}
 &= \mathbb{E}[\mathbb{E}[X|Y]^2] - \underbrace{\mathbb{E}[\mathbb{E}[X|Y]] + \mathbb{E}[X]}_{\text{"bias"} \quad \text{"variance"} \downarrow} \\
 &\quad = \mathbb{E}[\mathbb{E}[X|Y]^2] - \mathbb{E}[X]^2 \\
 &\quad = \mathbb{E}[\text{var}(X|Y)] + \text{var}(\mathbb{E}[X|Y]) = \mathbb{E}[X^2] - \mathbb{E}[X]^2 = \text{var}(X)
 \end{aligned}$$

$X \rightarrow Y$ observed. $X|Y$?

$$\hat{x} = \mathbb{E}[X|Y]$$