

I<sub>b</sub>) Probability Law assigns a non-negative value  <sup>$P(A) \geq 0$  called probability of  $A$</sup>  to an event  <sup>$(\omega) A$</sup>  that encodes our knowledge or belief about <sup>the collective</sup> likelihood of the <sup>elements</sup> (outcomes) of the event  $A$ .

$A \subset \Omega$  ( $A \in 2^\Omega$ )

Def Probability Law

$(E, \Omega, P)$   $\Omega$  is the sample space i.e.

<sup>the set of</sup> all the possible outcomes of experiment  $E$

Then ~~the~~ Probability Law ~~assigns~~ :

- specifies the likelihood of any outcome or/and of any set of possible outcomes (i.e. of an event)
- or alternatively, assigns to any event  $A$  a non-negative number  $P(A)$ , called the probability of  $A$ , satisfying the following axioms.

1. Nonnegativity  $P(A) \geq 0, \forall A \subset \Omega$

2. Additivity :  $A, B$  disjoint events (i.e.  $A \cap B = \emptyset$ )

Then  $P(A \cup B) = P(A) + P(B)$

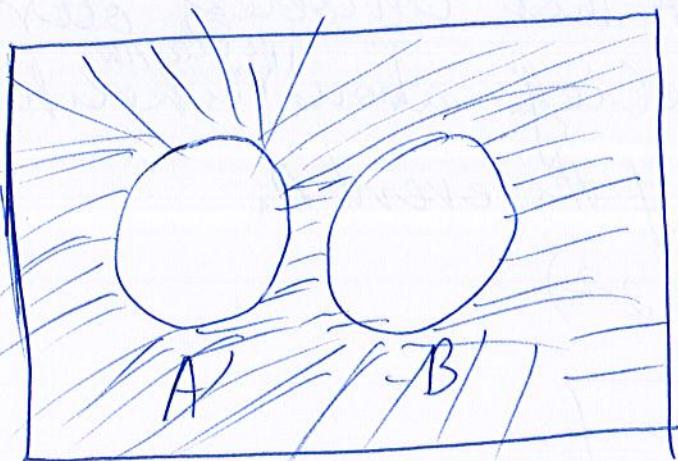
That is true for an infinite sequence of <sup>disjoint</sup> events  $A_1, A_2, A_3, \dots$

$P(A_1 \cup A_2 \cup A_3 \cup \dots) = P(A_1) + P(A_2) + \dots$

3. Normalization  $P(\Omega) = 1$



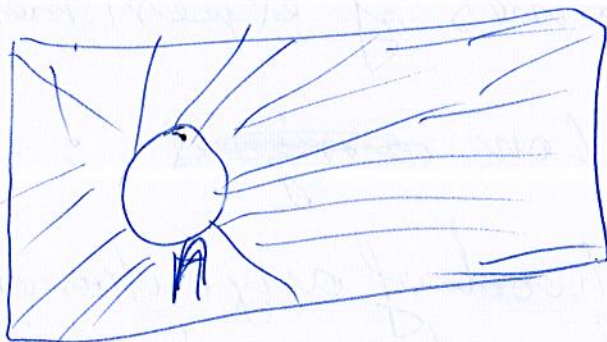
②



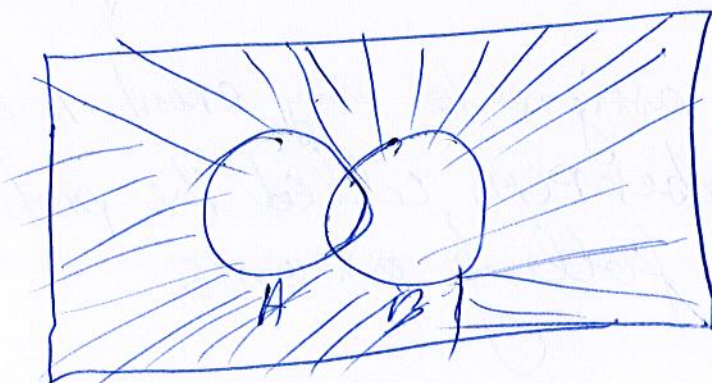
$$A^c \cap B^c$$

$$\left\{ \begin{array}{l} A \not\subset A^c \Rightarrow A \not\subset A^c \cap B^c \\ B \not\subset B^c \Rightarrow B \not\subset A^c \cap B^c \end{array} \right.$$

↪



Add to the diagram  
a B set



$$\left\{ \begin{array}{l} A \not\subset A^c \Rightarrow A \not\subset A^c \cap B^c \\ B \not\subset B^c \Rightarrow B \not\subset A^c \cap B^c \end{array} \right.$$

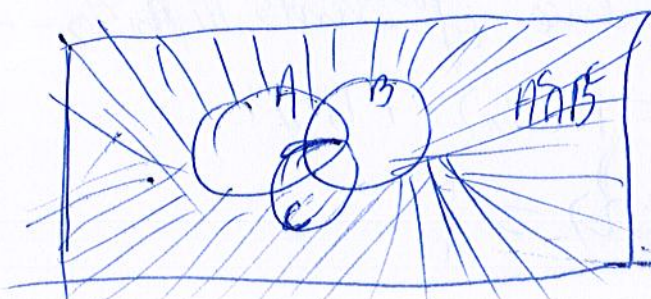
⇓

$$A \cap B \not\subset A^c \cap B^c$$

$$A \cup B \not\subset A^c \cap B^c$$

$$(A \cup B)^c = A^c \cap B^c$$

$$A^c \cap B^c = (A \cup B)^c$$



$$\rightarrow (A^c \cap B^c) \cap C$$

or

$$(A \cup B)^c \cap C$$



③  $P(\Omega) = 1$   $P(\emptyset) = 0$

Theorem:

a)  $A \subset B \Rightarrow P(A) \leq P(B)$

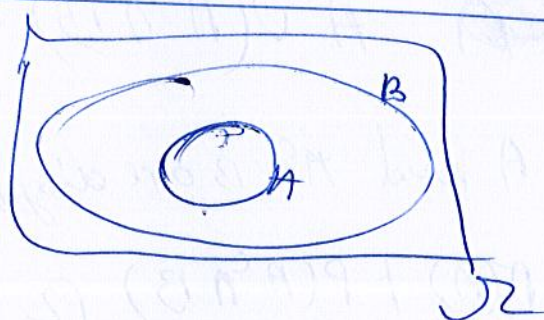
b)  $P(A \cup B) = P(A) + P(B) - P(A \cap B)$

c)  $P(A \cup B) \leq P(A) + P(B)$

d)  $P(A \cup B \cup C) = P(A) + P(A^c \cap B) + P(A^c \cap B^c \cap C)$

Proof:

a)



$A \subset B \subset \Omega$

(1)  $P(B) = P(A) + P(A^c \cap B)$  because  $A, A^c \cap B$  are disjoint

(2)  $P(B) - P(A) = P(A^c \cap B) \geq 0 \Rightarrow$

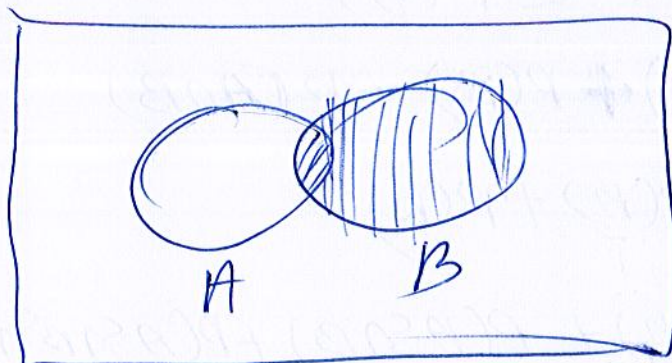
(3)  $P(B) \geq P(A) \Leftrightarrow P(A) \leq P(B)$

or from (1)  $P(B) = P(A) + P(A^c \cap B) \geq P(A)$



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b)  $P(A \cup B) \stackrel{?}{=} P(A) + P(B) - P(A \cap B)$



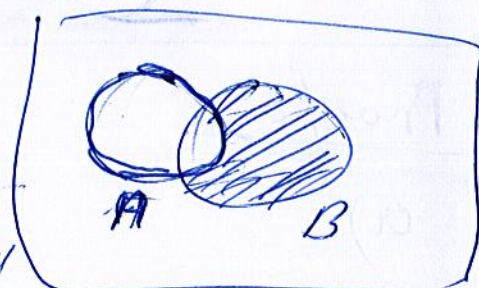
$$B = (A \cap B) \cup (A^c \cap B)$$

axiom 2  $\downarrow$  disjoint

$$P(B) = P(A \cap B) + P(A^c \cap B)$$

$$A \cup B = \cancel{A \cup B} A \cup (A^c \cap B)$$

axiom 2  $\downarrow$   $A$  and  $A^c \cap B$  are disjoint



$$P(A \cup B) = P(A) + P(A^c \cap B) \quad (2)$$

$$P(A \cup B) = P(A) + P(B) - P(A \cap B) \quad (3)$$

(1)  $P(B) = P(A \cup B) + P(A^c \cap B) - P(A \cap B)$

$$P(A) = P(A \cap B) + P(A \cup B)$$

$$P(B) = P(A \cup B) + P(A^c \cap B) - P(A \cap B)$$

c) prove that  $P(A \cup B) \leq P(A) + P(B)$

(1)  $\Rightarrow P(A \cup B) = P(B) - P(A^c \cap B)$

(2)  $\Rightarrow P(A \cup B) = P(A) + P(A^c \cap B)$

$$2 P(A \cup B) = P(B) + P(A)$$

$$P(A \cup B) = \frac{P(B) + P(A)}{2} \leq P(B) + P(A)$$

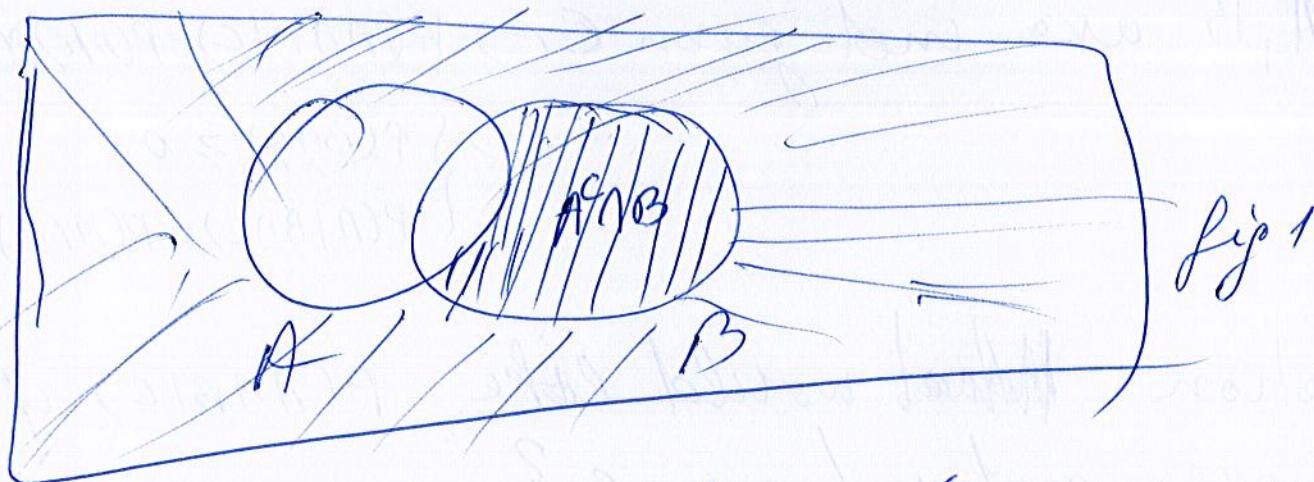
one  $\downarrow$   
as  $P(A \cap B) \geq 0$   
 $\Rightarrow$   
 $P(A \cup B) \leq P(A) + P(B)$



3.

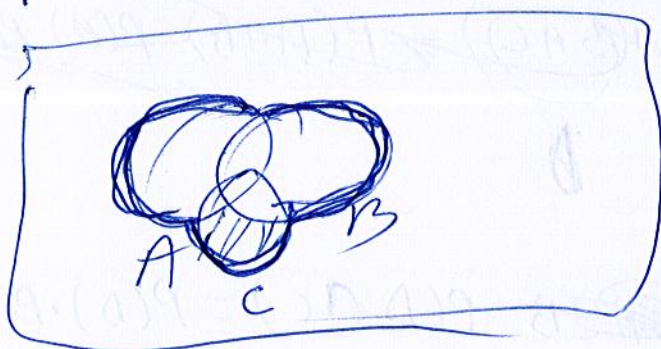
Prove that

$$P(A \cup B \cup C) = P(A) + P(A^c \cap B) + P(A^c \cap B^c \cap C)$$



~~$P(A \cup B)$~~

$$(A \cup B)^c = A^c \cap B^c$$



$$A \cup B \cup C = A \cup \underbrace{A^c \cap B}_{\text{fig 1}} \cup (A^c \cap B^c \cap C)$$

$$A \cup B = A \cup (A^c \cap B)$$

$$(A \cup B) \cup C = \underbrace{A \cup (A^c \cap B)}_{\text{fig 1}} \cup (A^c \cap B^c \cap C)$$

$A, A^c \cap B, A^c \cap B^c \cap C$  are disjoint, Hence

~~$P(A \cup B \cup C) = P(A) + P(B) + P(C)$~~

$$P(A \cup B \cup C) = P(A) + P(A^c \cap B) + P(A^c \cap B^c \cap C)$$

∴



$$(6) P(A \cap B) = P(A|B) \cdot P(B)$$

Again on conditional independency

Def

$A, B$  are ind given  $C \Leftrightarrow P(A \cap B | C) = P(A | C) P(B | C)$

$$\Leftrightarrow \begin{cases} P(B | C) > 0 \\ P(A | B \cap C) = P(A | C) \end{cases}$$

Question What would like  $P(A \cap B | C)$  if  $A, B$  are not ind given  $C$ ?

$$P(A \cap B \cap C) = \cancel{P(A \cap B | C)} \cdot P(C) \neq \cancel{P(A \cap B)} \cdot \cancel{P(C)}$$

$$\neq \cancel{P(A)} \cdot \cancel{P(A | B | C)} \quad (1)$$

$$= P((A \cap B) \cap C) = \cancel{P(A \cap B)} P(C | D) = P(D) \cdot P(D | C) \\ = P(C | D) \cdot P(D) \quad (2)$$

$$(1) \neq \cancel{P(A \cap B)} \cdot \cancel{P(A | B | C)} \Rightarrow P(A \cap B | C) = \frac{P(A \cap B \cap C)}{P(C)}$$

$$(2) \neq \cancel{P(C | A \cap B)} \cdot P(A \cap B) = \frac{P(C | A \cap B) \cdot P(A \cap B)}{P(A \cap B)} = \frac{P(C \cap A \cap B)}{P(A \cap B)} = \frac{P(C \cap A | B) \cdot P(B)}{P(A | B) \cdot P(B)} \\ = \frac{P(C \cap A | B)}{P(A | B)} \\ P(A \cap B \cap C) = P(A \cap B | C) \cdot P(C)$$



$$P(A \cap B | C) = \frac{P(A \cap B \cap C)}{P(C)} = \frac{P(A \cap B | C) \cdot P(C)}{P(C)}$$

$$P(A \cap B | C) = \frac{P(A \cap B \cap C)}{P(C)} = \frac{P(A \cap B \cap C)}{P(C)}$$

$$P(A \cap B \cap C) = \frac{P(A \cap B | C) \cdot P(C)}{P(C)}$$

$$P(A \cap B \cap C) = P(A \cap (B \cap C)) = P(A | B \cap C) \cdot P(B \cap C) \\ = P(A | B \cap C) \cdot P(B | C) \cdot P(C)$$

$$P(A \cap B | C) = \frac{P(A | B \cap C) \cdot P(B | C) \cdot P(C)}{P(C)} \quad (1)$$

$$P(B \cap A | C) = \frac{P(B \cap A \cap C)}{P(C)} = \frac{P(B | A \cap C) \cdot P(A | C) \cdot P(C)}{P(C)}$$

$$= (P(B | A \cap C) \cdot P(A | C)) \quad (2)$$

$$(1) \quad P(A \cap B | C) = P(A | B \cap C) \cdot P(B | C) \quad \text{if ind} \\ = P(A | C) \cdot P(B | C)$$

$$(2) \quad P(A \cap B | C) = P(B | A \cap C) \cdot P(A | C) = P(B | C) \cdot P(A | C)$$

$$\text{Recall } P(A \cap B) = P(A | B) \cdot P(B) = P(A) \cdot P(B) \\ P(A \cap B) = P(B | A) \cdot P(A) = P(B) \cdot P(A)$$



(8)

$X \rightarrow$  Geometric R.V. ~~the~~ representing the number of attempts until he first succeeds ~~head~~ come up.  $\downarrow$   
pass the test

PMF:  $P_X(X) = (1-p)^{x-1} \cdot p$

Let  $A =$  the event ~~that~~ student passes the test within  $n$  attempts  $A = \{X \leq n\}$

$$P(A) = \sum_{k=1}^n (1-p)^{k-1} p$$

$$P_{X|A}(X) = \frac{P(\{X=x\} \cap A)}{P(A)} = \frac{P(X=x, A)}{P(A)} = \frac{P_X(x)}{P(A)}$$

$$P_X(x) = (1-p)^{x-1} p$$

$$P_X(k) = (1-p)^{k-1} p$$

$$P_X(k) = (1-p)^{k-1} p \quad \text{where } 1 \leq k \leq n$$

$$P_{X|A}(k) = \frac{(1-p)^{k-1} p}{\sum_{m=1}^n (1-p)^{m-1} p} \quad \text{for } k = 1, 2, 3, \dots, n$$

$P(A)$

0

$$P_{X|A}(x) = \frac{P(\{X=x\} \cap A)}{P(A)}$$

$$P_{X|A}(k) = \frac{P(\{X=k\} \cap A)}{P(A)}$$

$$P_{X|A}(k) = (1-p)^{k-1} p \quad \underbrace{1 \leq k \leq n}_A$$



(9) A → event that student succeeds within n attempts

$$\text{Val}(X) = \{1, 2, 3, \dots\}$$

$$P_X(k) = (1-p)^{k-1} p$$

~~$$P_X(k) = (1-p)^{k-1} p, 1 \leq k \leq n$$~~

$$P(\{X=k\} \cap A) = (1-p)^{k-1} p \text{ and } 1 \leq k \leq n$$

$$P_{X|A}(k) = \begin{cases} \frac{(1-p)^{k-1} p}{\sum_{m=1}^n (1-p)^{m-1} p} & \text{for } 1 \leq k \leq n \\ 0, & \text{otherwise} \end{cases}$$

$$P(A) = \sum_{m=1}^n (1-p)^{m-1} p$$

•  $A_1$  event that  $\{X=1\}$

•  $A_2$  event that  $\{X>1\}$

$$E[X] = ?$$

$$E[X] = \sum_{i=1}^2 E[X|A_i] \cdot P(A_i)$$

$$E[X] = E[X|A_1]P(A_1) + E[X|A_2]P(A_2)$$

$$P(A_1) = \sum_{m=1}^n (1-p)^{m-1} p = p$$

$$P(A_2) = P(\{X>1\}) = 1 - P(\{X \leq 1\}) = 1 - P(\{X=1\}) = 1 - p$$

$$E[X|A_1] = \sum x \cdot P_{X|A_1}(x)$$



(10)

$$E[X|A_1] = \sum_{x=1}^{\infty} x \cdot P_{X|A_1}(x)$$

$$E[X|A_2] = \sum_{x=1}^{\infty} x \cdot P_{X|A_2}(x)$$

$$P_{X|A_1}(x) = \frac{(1-p)^{x-1} \cdot p}{P(A_1)} \quad \text{also} = \frac{P(\{X=x\} \cap A_1)}{P(A_1)} \quad A_1 = \{X=1\}$$

$$= \frac{(1-p)^{1-1} \cdot p}{P(A_1)} = \frac{p}{p} = 1 = \begin{cases} 1, & x=1 \\ 0, & \text{for the rest of } \mathbb{N} \end{cases}$$

0, otherwise

konkret value (we could write  $P_{X|A}(k)$  (so a konkret value  $k$  from  $\{2, 3, \dots\}$ )

$$P_{X|A_2}(x) = \frac{P(\{X=x\} \cap A_2)}{P(A_2)} = \frac{P(\{X=x\} \cap \{X>1\})}{P(A_2)}$$

$$= \frac{(1-p)^{x-1} \cdot p}{P(A_2)} = \frac{(1-p)^{x-1} \cdot p}{1-p} = (1-p)^{x-2} \cdot p, \text{ for } x>1$$

0, otherwise  
i.e. for  $x=1$

$$= (1-p) \cdot p + (1-p)^2 \cdot p + (1-p)^3 \cdot p + \dots$$

$$= p + (1-p)p + (1-p)^2 p + (1-p)^3 p + \dots + (1-p)^4 p + \dots$$

$$= p + q \cdot p + q^2 p + q^3 p + q^4 p + \dots = p(1 + q + q^2 + q^3 + \dots) =$$

$$= p \cdot \frac{1}{1-q} = \frac{p}{1-q} = \frac{p}{1-(1-p)} = \frac{p}{p} = 1$$



(41)

$$E[X] = E[X|A_1] \cdot P(A_1) + E[X|A_2] \cdot P(A_2)$$

$$E[X|A_1] = \sum_{x=1}^{\infty} x \cdot P_{X|A_1}(x) = 1 \cdot P_{X|A_1}(1) + \sum_{x=2}^{\infty} x \cdot 0 = 1 \cdot 1 + 0 = 1$$

$$E[X|A_2] = \sum_{x=1}^{\infty} x \cdot P_{X|A_2}(x) = 1 \cdot 0 + \sum_{x=2}^{\infty} x \cdot P_{X|A_2}(x) =$$

$$= \underbrace{1 \cdot 0}_{x=1} + \sum_{x=2}^{\infty} x \cdot (1-p)^{x-2} \cdot p \quad \begin{matrix} x'=x-1 \\ x'=1 \end{matrix} = \sum_{x'=1}^{\infty} (x'+1) \cdot (1-p)^{x'-1} \cdot p =$$

$$= \sum_{x'=1}^{\infty} x' \cdot (1-p)^{x'-1} \cdot p + \sum_{x'=1}^{\infty} (1-p)^{x'-1} \cdot p \quad \begin{matrix} x=x'+1 \\ x'=1 \end{matrix} = 1 =$$

$$\sum_x P_X(x) = 1$$

$$= \cancel{E[X]} \sum_{x=1}^{\infty} x (1-p)^x p + 1 = \underline{E[X] + 1}$$

Hence  $E[X] = 1 \cdot P(A_1) + (E[X] + 1) \cdot P(A_2)$

$$= 1 \cdot p + (E[X] + 1) \cdot (1-p) =$$

$$= p \cancel{+ p E[X]} - p + E[X] + 1 =$$

$$E[X] = 1 + (1-p) \cdot E[X] \Rightarrow E[X](1+p-1) = 1 \Rightarrow$$

$$\therefore \boxed{E[X] = \frac{1}{p}}$$



(12)

$$\sum_{x=2}^{\infty} f(x) = f(\overset{2}{\underset{\downarrow}{2}}) + f(\overset{3}{\underset{\downarrow}{3}}) + f(\overset{4}{\underset{\downarrow}{4}}) + f(\overset{5}{\underset{\downarrow}{5}}) + \dots$$

$$\sum_{x'=1}^{\infty} f(\overset{x'=1}{\cancel{x}}) = \underset{\substack{\uparrow \\ \text{we want}}}{\sqrt{(2)^2}} + f(\overset{2}{\underset{\downarrow}{3}}) + f(\overset{3}{\underset{\downarrow}{4}}) + f(\overset{4}{\underset{\downarrow}{5}}) + \dots$$

$$= \sum_{x'=1}^{\infty} f(x'+1)$$

$$\begin{matrix} 1 & 2 & 3 & 4 & \dots \\ \downarrow & \downarrow & \downarrow & \downarrow & \\ 2 & 3 & 4 & 5 & \dots \end{matrix}$$

$$\text{Var}(Z) = E[(Z - E[Z])^2]$$

$\sigma_Z = \sqrt{\text{Var}(Z)}$

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

$$= E[(X+Y - E[X] - E[Y])^2] = E[(X - E[X] + Y - E[Y])^2]$$

$$\begin{matrix} \tilde{X} = X - E[X] \\ \tilde{Y} = Y - E[Y] \end{matrix} \quad E[(\tilde{X} + \tilde{Y})^2] = E[\tilde{X}^2 + 2\tilde{X}\tilde{Y} + \tilde{Y}^2] =$$

$$= E[\tilde{X}^2] + 2E[\tilde{X}\tilde{Y}] + E[\tilde{Y}^2] =$$

$$= E[(X - E[X])^2] + 2E[(X - E[X])(Y - E[Y])] + E[(Y - E[Y])^2] =$$

$$= \text{Var}(X) + 2E[X \cdot E[Y] - X \cdot E[Y] - E[X] \cdot Y + E[X]E[Y]]$$

$$= \text{Var}(X) + 2E[X]E[Y] - 2E[X]E[Y] - E[X]E[Y] + E[X]E[Y] + \text{Var}(Y)$$

$(E[E[Y]] = E[Y])$

$$\text{Var}(\bar{X} + \bar{Y}) = \text{Var}(\bar{X}) + \text{Var}(\bar{Y})$$



$$E[X] = \int_{-\infty}^{+\infty} x p_x(x) dx$$

$$E[X|Y=y] = \int_{-\infty}^{+\infty} x p_{x|y}(x|y) dx$$

$$E[g(X)|Y=y] = \int_{-\infty}^{+\infty} g(x) p_{x|y}(x|y) dx$$

$\nearrow$  small  $x$

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

$f(x)$  is pdf

$$E[X] = \sum_{i=1}^m E[X|A_i] P(A_i)$$

$$E[g(X)] = \sum_{i=1}^m E[g(X)|A_i] \cdot P(A_i)$$

joint      conditions

$$p_{x,y}(x,y) = p_{x|y}(x|y) \cdot f_y(y)$$

$$f_x(x) = \int_{-\infty}^{+\infty} p_{x,y}(x,y) dy = \int_{-\infty}^{+\infty} p_{x|y}(x|y) f_y(y) dy$$

Let  $g = f_{x|y}(x|y)$ , then

$$\int_{-\infty}^{+\infty} f_{x|y}(x|y) f_y(y) dy = E_{Y \sim P_Y}[f_{x|y}(x|y)] = E[f_{y|x}(y|x)]$$

$$f_x(x) = E_Y[f_{x|y}(x|y)]$$



$$p_x(x) = \int_y p(x,y) dy = \int_y \underbrace{p(x|y)}_{g(y)} \cdot p_y(y) dy = E_y [p_{x|y}(x|y)]$$

$$p_x(x) = E_y [p_{x|y}(x|y)]$$

$$p_x(x) = E_{y \sim p(y)} [p_{x|y}(x|y)]$$

$$y=0 \quad \downarrow$$

$$p_x(x) = E_{\theta \sim p(\theta)} [p_{x|\theta}(x|\theta)]$$

$$p(x|\theta)$$

$$p(\theta|x)$$

$$y=0$$

$$p_{\theta}(0) = E_{\theta} [p_{x|\theta}(x|\theta)]$$

$$p(\theta) = E_x [p_{\theta|x}(\theta|x)]$$

$$p(\theta) = E_{x \sim p(x)} [p_{\theta|x}(\theta|x)]$$

marginal of theta is expected value of x

$$\begin{cases} x=0 \\ y=x \end{cases}$$

$$p_{\theta}(0) = E_{y \sim p(y)} [p_{\theta|y}(0|y)]$$

if discrete

$$p_{\theta}(0) = \sum_y p_y(y) \cdot p_{\theta|y}(0|y)$$