

# Review of Prob. Theory

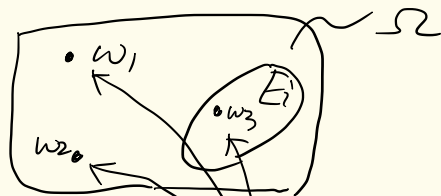
## Axioms of Prob.:

1.  $P(\Omega) = 1$

2.  $P(E_i) \geq 0$

3.  $P(E_1 \cup E_2 \cup \dots)$   
 $= \sum_{i=1}^{\infty} P(E_i),$

$E_i$ 's are mutually exclusive,  
i.e.,  $E_i \cap E_j = \emptyset, \forall i \neq j$

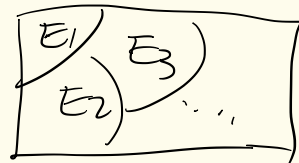


$\Omega$ : "sample space"  
(collection of all possible "outcomes")

$E_i$ : "event" (a subset of  $\Omega$ )

$$E_i \subset \Omega$$

$w$ : "outcome" .  $w \in \Omega$



Propositions:  $A' = \Omega \setminus A$

1.  $P(A) + P(A') = 1$



2.  $P(A) \leq 1$  , 3.  $P(A \cap B) + P(A \cup B) = P(A) + P(B)$

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

Ex: Consider games ("random experiments")

{ finite outcomes: (1), (2)

{ countably infinite outcomes

(1) Flip a coin:  $\Omega = \{H, T\}$



$$P(\{H\}) = \frac{1}{2}$$

$\uparrow \frac{1}{2}$

(2) Roll a die:  $\Omega = \{ "1", "2", "3", \dots, "6" \}$

$$P(\{ "6" \}) = \frac{1}{6}$$

• • •

(3) Roll a die until "6" appears.

$\Omega = \{ \omega_1, \leftarrow 6 \text{ on } 1^{\text{st}} \text{ roll}$

$\omega_2, \leftarrow 6 \text{ on } 2^{\text{nd}} \text{ roll}$

$\vdots$

$\omega_n, \leftarrow 6 \text{ on } n^{\text{th}} \text{ roll}$

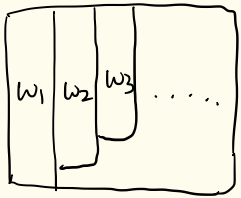
$\vdots$

}

$$P(\{\omega_1\}) = \frac{1}{6}$$

$$P(\{\omega_2\}) = \frac{5}{6} \cdot \frac{1}{6} = \frac{5}{6^2}$$

$$P(\{\omega_n\}) = \left(\frac{5}{6}\right)^{n-1} \cdot \frac{1}{6}$$



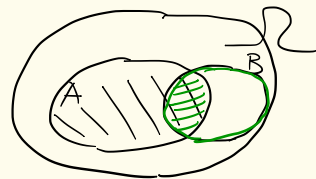
$$\lim_{n \rightarrow \infty} \sum_{i=1}^n P(\{\omega_i\}) = \lim_{n \rightarrow \infty} \sum_{i=1}^n \left(\frac{5}{6}\right)^{i-1} \cdot \frac{1}{6} = \frac{1}{1 - \frac{5}{6}} \cdot \frac{1}{6} = 1$$

$$\hookrightarrow = P(\underbrace{\{\omega_1\} \cup \{\omega_2\} \cup \dots}_{\Omega}) = 1$$

Conditional Prob:

$$P(A|B) \triangleq \frac{P(A \cap B)}{P(B)} = \frac{P(A, B)}{P(B)}, \quad P(B) \neq 0$$

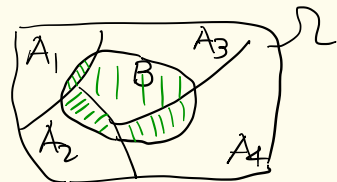
$$P(A) = P(A|\Omega)$$



Laws of total prob:

$$P(B) = \sum_{i=1}^n P(B, A_i), \quad \bigcup_{i=1}^n A_i = \Omega, \quad A_i \cap A_j = \emptyset, \quad \forall i \neq j$$

$$= \sum_{i=1}^n P(B|A_i) \cdot P(A_i)$$



Bayes' Theorem:

$$P(A_{\tilde{j}}|B) = \frac{P(A_{\tilde{j}}, B)}{P(B)} = \frac{P(B|A_{\tilde{j}}) \cdot P(A_{\tilde{j}})}{\sum_{i=1}^n P(B|A_i) P(A_i)}, \quad \tilde{j} = 1, \dots, n$$

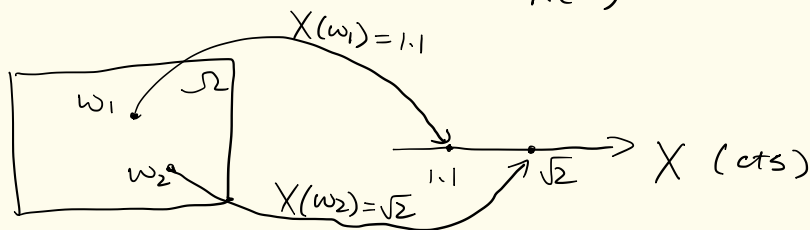
Independence :

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

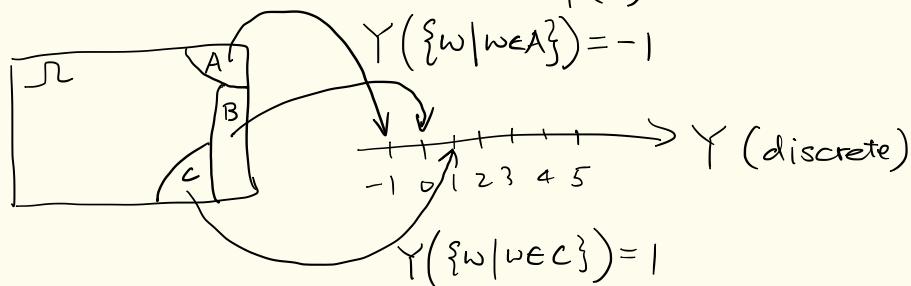
"condition - drop" test

Random Variables (r.v.s.): A map from sample space to  $\mathbb{R}$

Ex: ① continuous r.v.  $X: \Omega \rightarrow \mathbb{R}$   
 $\omega \mapsto X(\omega)$



② discrete r.v.  $Y: \Omega \rightarrow \mathbb{Z}$   
 $\omega \mapsto Y(\omega)$



Ex: Let  $X$  be the parity of roll of a die

$$X = \begin{cases} 0, & \text{even outcomes,} \\ 1, & \text{odd} \end{cases}$$

"1"	"2"	"3"
"4"	"5"	"6"

$$\text{Event } \{X=0\} \triangleq \{\omega \mid X(\omega)=0\} = \{2, 4, 6\}$$

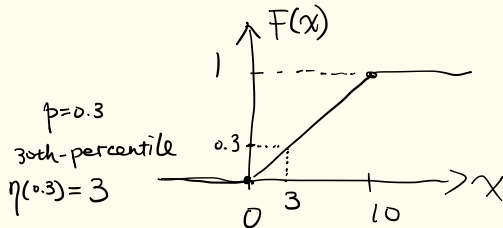
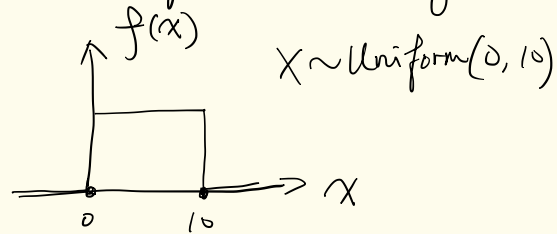
$$\begin{aligned} \mathbb{P}[X=0] &= \mathbb{P}[\{\omega \mid X(\omega)=0\}] = \mathbb{P}[\{2, 4, 6\}] \\ &= \mathbb{P}[\{2\}] + \mathbb{P}[\{4\}] + \mathbb{P}[\{6\}] = \frac{1}{6} \times 3 = \frac{1}{2} \end{aligned}$$

# Cumulative Distribution/Density Function (CDF)

$$F(c) \triangleq \mathbb{P}[X \leq c] = \mathbb{P}[\{\omega \in \Omega \mid X(\omega) \leq c\}]$$

Percentile:  $p \in [0, 1]$ .  $(100 \cdot p)$ -th-percentile of the dist of r.v.  $X$  denoted by  $\eta(p)$  is

$$\eta(p) = F^{-1}(p)$$



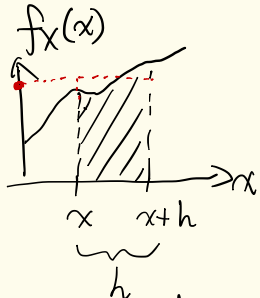


Prob density function (pdf) | Prob mass func (pmf)

$$f_X(x) \stackrel{\Delta}{=} \lim_{h \rightarrow 0} \frac{P[X \leq X \leq x+h]}{h}$$

$$= \lim_{h \rightarrow 0} \frac{P[\{\omega \in \Omega \mid X(\omega) \in [x, x+h]\}]}{h}$$

$$\geq 0$$



Trapezoidal rule

Expectation value (mean)

$$E[X] = \int_{-\infty}^{\infty} x \underbrace{f_X(x)}_{\text{pdf}} dx$$

$$E[h(X)] = \int_{-\infty}^{\infty} h(x) \underbrace{f_X(x)}_{\text{pdf}} dx$$

$$E[aX+b] = a \cdot E[X] + b$$

$$P_Y(y) \stackrel{\Delta}{=} P[Y=y]$$

$$= P[\{\omega \in \Omega \mid Y(\omega)=y\}]$$

$$\in [0, 1]$$

$$E[X] = \mu_X = \sum_x x \underbrace{P_X(x)}_{\text{pmf}}$$

$$E[X] = \sum_x h(x) P_X(x)$$

Variance  
or  $\text{Var}(X)$

$$V(X) = \int_{-\infty}^{\infty} (x - \mu_x)^2 f_X(x) dx$$

$$V(X) = \sum_x (x - \mu_x)^2 P_X(x)$$

$$= \underbrace{E[(X - \mu_x)^2]}_{\sigma^2} = E[X^2] - \underbrace{(EX)^2}_{\mu^2} \Rightarrow E[X^2] = \mu^2 + \sigma^2$$

$$V(aX + b) = a^2 V(X)$$

Moments:  $\mu_k = E[X^k]$  :  $k^{\text{th}}$  moment

$$= \int_{-\infty}^{\infty} x^k f_X(x) dx$$

$$= \sum_x x^k P_X(x)$$

Joint distribution:  $\iint f_{XY}(x,y) dx dy = 1$  |  $\sum_y \sum_x P_{XY}(x,y) = 1$

$$f_X(x) = \int_{y \in \mathbb{R}} f_{XY}(x,y) dy$$

$$P_X(x) = \sum_y P_{XY}(x,y)$$

$P_X(x)$

$Y \backslash X$	1	...	N
1	0.1		0.02
2	0.2		0.01

Sum along the row/column to marginalize

(Joint) Independent pdf/cdf/pmf factors under  $\mathbb{P}_2$ , e.g.,  
f, F, P

$$f(x_1, \dots, x_n) = f(x_1) \cdot \dots \cdot f(x_n) = \prod_{i=1}^n f(x_i) \quad \left( \prod_i X_i \right)$$

independence

Conditional dist:  $f_{Y|X}(y|x) = \frac{f(x,y)}{f(x)}$  |  $\underbrace{P_{Y|X}(y|x)}_{P[Y=y|X=x]} = \frac{P(x,y)}{P(x)}$

Covariance  $C_{xy} / \text{Cov}(X, Y) \triangleq E[(X - E[X])(Y - E[Y])]$   
 $= E[XY] - (E[X])(E[Y])$

Def: Uncorrelatedness:  $\text{Cov}(X, Y) = 0$  or  $E[XY] = E[X]E[Y]$

①  $X \perp\!\!\!\perp Y \Rightarrow \text{Cov}(X, Y) = 0$

②  $\text{Cov}(X, Y) = 0$   
 $X, Y$  are joint Gaussian/normal  $\} \Rightarrow X \perp\!\!\!\perp Y$

Proof for " $X \perp\!\!\!\perp Y \Rightarrow \text{Cov}(X, Y) = 0$ ":

$$\text{Cov}(X, Y) = \underline{E[XY]} - E[X]E[Y]$$

$$= \iint xy f(x, y) dx dy - \left( \int x f(x) dx \right) \cdot \left( \int y f(y) dy \right)$$

$$\stackrel{X \perp\!\!\!\perp Y}{=} \iint xy f(x) f(y) dx dy - \left( \int x f(x) dx \right) \cdot \left( \int y f(y) dy \right) = 0 \quad \square$$

Correlation:  $r_{xy} \triangleq E[XY]$

Def: Orthogonality (in prob/stat):  $E[XY] = 0$