

EE660

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Discussion #1

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Submission of HW:

use the D2L

only accept PDF

2 pdf (1 pdf for programming code
1 pdf HW)

Review on Probability:

event A , prob. of event A

$$0 \leq P(A) \leq 1$$

Discrete RV:

prob. of event $X=x$ denoted as

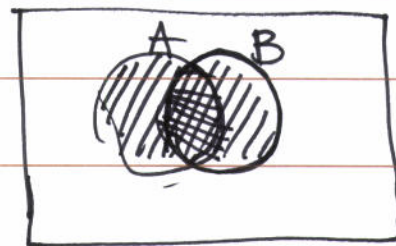
$$P(X=x)$$

Prob. Mass Func. PMF

Fundamental Rules:

① prob. of union:

A and B



$$P(A \text{ or } B) = P(A) + P(B) - P(A, B)$$

$$= P(A) + P(B)$$

A and B are mutually exclusive

$$P(AB) = 0 \quad X$$

②

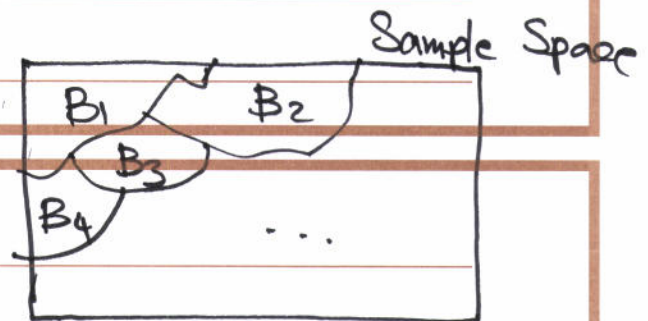
② prob. of joint

$$P(A, B) = P(A)P(B|A) = P(B)P(A|B)$$

extend $\left(\begin{aligned} P(X_1, X_2, \dots, X_n) &= P(X_1)P(X_2|X_1)P(X_3|X_1, X_2) \\ &\dots P(X_n|X_1, \dots, X_{n-1}) \end{aligned} \right.$

Suppose ^{we} are given the joint prob. $P(A, B)$
how to find marginal prob. ?

$P(A)$



$$P(A) = \sum_i P(A|B_i)P(B_i)$$

(Total Prob. Law)

③ Conditional Prob.

$$P(A|B) = \frac{P(AB)}{P(B)} \quad P(B) \neq 0$$

④ Baye's Rule :

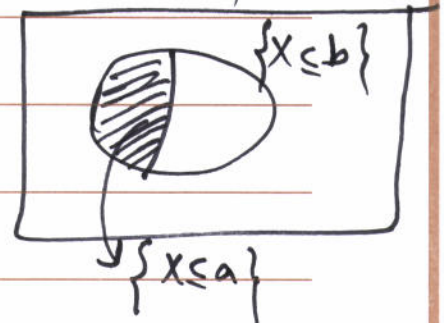
$$\begin{aligned} P(X=x | Y=y) &= \frac{P(Y=y | X=x) P(X=x)}{P(Y=y)} \\ &= \frac{P(X=x, Y=y)}{P(Y=y)} = \frac{P(Y=y | X=x) P(X=x)}{P(Y=y)} \\ &\rightarrow = \frac{P(Y=y | X=x) P(X=x)}{\sum_x P(Y=y | X=x) P(X=x)} \end{aligned}$$

Continuous RV :

X can take any values

$$P(X=a) = 0$$

$$P(\underbrace{a < X \leq b}_{\{X \leq b\} - \{X \leq a\}}) = P(X \leq b) - P(X \leq a)$$



$$P(X \leq x) \triangleq F_x(x)$$

cumulative dist.
function (CDF)

$$f_X(x) \triangleq \frac{d}{dx} F_X(x) \quad , \quad \text{prob. density func. (PDF)}$$

$$P(a < X \leq b) = \int_a^b f_X(x) dx$$

Note: $f_X(x) \geq 0 \quad \forall x$

$f(x)$ is NOT nece. ≤ 1

$$\int_{-\infty}^{+\infty} f_X(x) dx = 1$$

(disc. $\sum_x P(X=x) = 1$)

Properties of distributions:

1. Average: $E(X) = \sum x P_X(x)$
 $= \int x f_X(x) dx$

2. Variance: $\text{Var}(X) = E((X - E(X))^2)$
 $= E(X^2 - 2XE(X) + E^2(X))$
 $= E(X^2) - E^2(X)$

3. Covariance between X and Y

$$\begin{aligned}\text{Cov}(X, Y) &\triangleq E((X - E(X))(Y - E(Y))) \\ &= E(XY) - E(X)E(Y)\end{aligned}$$

Normalize $\text{Cov}(X, Y)$

$$\text{Corr}(X, Y) \triangleq \frac{\text{Cov}(X, Y)}{\sqrt{\text{Var}(X) \text{Var}(Y)}}$$

Some Discrete Distributions:

① Bernoulli (p)

$$X \in \{0, 1\}$$

$$P(X=1) = p$$

$$E(X) = p$$

$$\text{Var}(X) = p(1-p)$$

② Binomial (n, p)

$$X \in \{0, 1, 2, \dots, n\}$$

$$E(X) = np$$

$$\text{Var}(X) = np(1-p)$$

$$P(X=k) = \binom{n}{k} p^k (1-p)^{n-k}$$

$$X = \sum_{i=1}^n X_i$$

(6)

③ Poisson Distribution: (with parameter λ)

$$X \in \{0, 1, 2, \dots\}$$

$$P(X=x) = e^{-\lambda} \frac{\lambda^x}{x!}$$

④ Empirical Distribution:

$$\text{given } D = \{x_1, x_2, \dots, x_N\}$$

$$P_{\text{emp}}(A) \triangleq \frac{1}{N} \sum_{i=1}^N \delta_{x_i}(A)$$

$$\delta_x(A) \triangleq \begin{cases} 1 & \text{if } x \in A \\ 0 & \text{else} \end{cases}$$

Continuous Distribution:

① Gaussian (μ, σ^2)

$$f_X(x) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left\{-\frac{1}{2\sigma^2} (x-\mu)^2\right\}$$

$$E(X) = \mu$$

$$\text{Var}(X) = \sigma^2$$

② Exponential Distribution (λ)

$$f_X(x) = \begin{cases} \lambda e^{-\lambda x} & x \geq 0 \\ 0 & \text{else} \end{cases}$$

$$E(X) = \frac{1}{\lambda} \quad \text{Var}(X) = \frac{1}{\lambda^2}$$

Sofar we had only one RV, we can extend to multivariable scenario

$$P(x_1, x_2, \dots, x_D)$$

$$\begin{aligned} \text{Discrete } F_X(X_1=x_1, X_2=x_2, \dots, X_n=x_n) \\ = \text{Prob}(X_1 \leq x_1, \dots, X_n \leq x_n) \end{aligned}$$

$$\text{Cov}(\underline{X}) \triangleq E \left((\underline{X} - E(\underline{X})) (\underline{X} - E(\underline{X}))^T \right)$$

$$= \begin{pmatrix} \text{Var}(X_1) & \text{Cov}(X_1, X_2) & \dots \\ \text{Cov}(X_2, X_1) & \text{Var}(X_2) & \\ & & \ddots \end{pmatrix}$$

Correlation Matrix $R = \begin{pmatrix} \text{Corr}(X_1, X_1) & \text{Corr}(X_1, X_2) & \dots \\ \text{Corr}(X_2, X_1) & \text{Corr}(X_2, X_2) & \\ \vdots & & \ddots \end{pmatrix}$

$$\text{Corr}(X_1, X_2) = 1$$

\Rightarrow linear relationship between X_1 & X_2

$$X_2 = aX_1 + b$$

what if $X_1 \perp X_2 \Rightarrow \text{Corr}(X_1, X_2) = 0$

\rightarrow Proof: $\text{Corr}(X_1, X_2) \Rightarrow \text{Cov}(X_1, X_2) = \underbrace{E(X_1 X_2)}_{\text{?}} - \underbrace{E(X_1)}_{E(X_1)} \underbrace{E(X_2)}_{E(X_2)}$

$$E(f(x)g(y)) \stackrel{\Delta}{=} E(f(x))E(g(y)) \text{ indep.}$$

Transformation of RV:

if X is a multivariable RV & $\underline{y} = f(x)$

• what prop. does \underline{y} have?

$$\underline{Y} = A\underline{X} + b$$

$$E(\underline{Y}) = A E(X) + b$$

$$\begin{aligned} \text{Cov}(\underline{Y}) &= E((\underline{Y} - E(\underline{Y}))(\underline{Y} - E(\underline{Y}))^T) \\ &= E((A(X - E(X)) + b)(A(X - E(X)) + b)^T) \end{aligned}$$

$$= A \Sigma A^T$$

$$\downarrow$$
$$\text{Cov}(X)$$