

Random Variables

A **random variable** X is a map $X : \Omega \rightarrow \mathbb{R}$. For $A \subset \mathbb{R}$ we write

$$\mathbb{P}(X \in A) = \mathbb{P}(\{w \in \Omega : X(w) \in A\}) \quad (1)$$

The **cdf** F_X of X is

$$F_X(x) = \mathbb{P}(X \leq x) \quad (2)$$

For continuous X , the **pdf** f_X is a function satisfying

$$\int_A f_X(x) dx = \mathbb{P}(X \in A) \quad (3)$$

Note that $f_X = F'_X$.

Transformations

Let $Y = g(X)$, $\mathcal{X} = \{x : f_X(x) > 0\}$, and $\mathcal{Y} = \{y : y = g(x), x \in \mathcal{X}\}$ (\mathcal{X} and \mathcal{Y} called the *support* of X and Y). Then $\forall A \subset \mathcal{Y}$

$$\mathbb{P}(Y \in A) = \mathbb{P}(X \in \{x : g(x) \in A\}) \quad (4)$$

For the cdf F_Y

$$F_Y(y) = \mathbb{P}(X \in \{x : g(x) \leq y\}) = \int_{\{x: g(x) \leq y\}} f_X(x) dx \quad (5)$$

For g monotonic

$$F_Y(y) = \begin{cases} F_X(g^{-1}(y)) & \text{if } g \text{ increasing} \\ 1 - F_X(g^{-1}(y)) & \text{if } g \text{ decreasing} \end{cases} \quad (6)$$

Additionally, for g monotonic, if $g^{-1}(y)$ has a continuous derivative on \mathcal{Y}

$$f_Y(y) = f_X(g^{-1}(y)) \left| \frac{dg^{-1}(y)}{dy} \right| \quad \text{for } y \in \mathcal{Y} \quad (7)$$

Expected Values

The **mean** or **expected value** of $g(X)$ is

$$\mathbb{E}(g(X)) = \int g(x) dF(x) = \int g(x) dP(x) \quad (8)$$

Related properties and definitions:

$$(a) \quad \mu = \mathbb{E}(X) \quad (9)$$

$$(b) \quad \mathbb{E}(\sum_i c_i g_i(X_i)) = \sum_i c_i \mathbb{E}(g_i(X_i)) \quad (10)$$

$$(c) \quad \mathbb{E}\left(\prod_i X_i\right) = \prod_i \mathbb{E}(X_i), \quad X_1, \dots, X_n \text{ indep't} \quad (11)$$

$$(d) \quad Var(X) = \sigma^2 = \mathbb{E}((X - \mu)^2) \quad \text{is the } \mathbf{variance} \text{ of } X \quad (12)$$

$$(e) \quad Var(X) = \mathbb{E}(X^2) - \mu^2 \quad (13)$$

$$(f) \quad Var\left(\sum_i a_i X_i\right) = \sum_i a_i^2 Var(X_i), \quad X_1, \dots, X_n \text{ indep't} \quad (14)$$

$$(g) \quad Cov(X, Y) = \mathbb{E}((X - \mu_X)(Y - \mu_Y)) \quad \text{is the } \mathbf{covariance} \quad (15)$$

$$(h) \quad Cov(X, Y) = \mathbb{E}(XY) - \mu_X \mu_Y \quad (16)$$

$$(i) \quad \rho(X, Y) = Cov(X, Y) / \sigma_X \sigma_Y, \quad -1 \leq \rho(X, Y) \leq 1 \quad (17)$$

The **conditional expectation** of Y given X is the random variable $g(X) = \mathbb{E}(Y|X)$, where

$$\mathbb{E}(Y|X = x) = \int y f(y|x) dy \quad (18)$$

$$\text{and } f(y|x) = f_{X,Y}(x, y) / f_X(x) \quad (19)$$

The *Law of Total/Iterated Expectation* is

$$\mathbb{E}(Y) = \mathbb{E}[\mathbb{E}(Y|X)] \quad (20)$$

The *Law of Total Variance* is

$$Var(Y) = Var[\mathbb{E}(Y|X)] + \mathbb{E}[Var(Y|X)] \quad (21)$$

The *Law of Total Covariance* is

$$Cov(X, Y) = \mathbb{E}(Cov(X, Y|Z)) + Cov(\mathbb{E}(X|Z), \mathbb{E}(Y|Z)) \quad (22)$$

Moment Generating Function

The **mgf** of X is

$$M_X(t) = \mathbb{E}(e^{tX}) \quad (23)$$

Properties:

$$(a) \quad M_X^{(n)}(t)|_{t=0} = \mathbb{E}(X^n) \quad \text{is the } \mathbf{n^{th} \text{ moment}} \text{ of } X \quad (24)$$

$$(b) \quad M_x(t) = M_Y(t) \quad \forall t \text{ around } 0 \implies X \stackrel{d}{=} Y \quad (25)$$

$$(c) \quad M_{aX+b}(t) = e^{bt} M_X(at) \quad (26)$$

$$(d) \quad M_{\sum_i X_i}(t) = \prod_i M_{X_i}, \quad X_1, \dots, X_n \text{ indep't} \quad (27)$$

Independence

Random variables X and Y are **independent**, written $X \perp\!\!\!\perp Y$, iff

$$\mathbb{P}(X \in A, Y \in B) = \mathbb{P}(X \in A) \mathbb{P}(Y \in B) \quad (28)$$

If (X, Y) is a random vector with pdf $f_{X,Y}$, then

$$X \perp\!\!\!\perp Y \iff f_{X,Y}(x, y) = f_X(x) f_Y(y) \quad (29)$$

Distributions

Discrete distributions:

$$(a) \quad \text{Bernoulli} \quad f(x|p) = p^x (1-p)^{1-x}, \quad x \in \{0, 1\} \quad (30)$$

$$(b) \quad \text{Binomial} \quad f(x|n, p) = \binom{n}{x} p^x (1-p)^{n-x}, \quad x \in \{0, 1, \dots, n\} \quad (31)$$

$$(c) \quad \text{Poisson} \quad f(x|\lambda) = \frac{e^{-\lambda} \lambda^x}{x!}, \quad x \in \{0, 1, 2, \dots\} \quad (32)$$

Continuous distributions:

$$(a) \quad \text{Uniform} \quad f(x|a, b) = \frac{1}{b-a}, \quad x \in [a, b] \quad (33)$$

$$(b) \quad \text{Normal} \quad f(x|\mu, \sigma^2) = \frac{1}{\sigma\sqrt{2\pi}} e^{-(x-\mu)^2/(2\sigma^2)}, \quad x \in \mathbb{R} \quad (34)$$

$$(c) \quad \text{Gamma} \quad f(x|\alpha, \beta) = \frac{1}{\Gamma(\alpha)\beta^\alpha} x^{\alpha-1} e^{-x/\beta}, \quad x \in \mathbb{R}_+, \alpha, \beta > 0 \quad (35)$$

Miscellaneous

To include: MGFs and/or CDFs for above distributions. More distributions. Misc. useful things: L'Hôpital's rule, Taylor approximation, definitions of e , Leibnitz's Rule.