

Lecture 3 - Objective and Outcomes

Moments provide summary information about distributions. Appropriate conditioning allows simple solutions to apparently intractable problem.

- moments and moment generating functions
- cumulant generating functions
- functions of random variables

After reviewing the notes you should:

- be able to work of mgf and cgf for a variety of different distributions
- understand a simple method for generating draws from random variables
- be able to get marginal distributions from joint distributions

Properties of moments

1. $m_1 = E(X) = \mu$ (mean); $\mu_1 = 0$.
2. $\mu_2 = E(X^2) - E(X)^2 = \text{var}(X) = \sigma^2$ (variance).
3. Coefficient of skewness:
 $\gamma_1 = E[(X - \mu)^3]/\sigma^3 = \mu_3/\mu_2^{\frac{3}{2}}$.
4. Coefficient of kurtosis:
 $\gamma_2 = (E[(X - \mu)^4]/\sigma^4) - 3 = (\mu_4/\mu_2^2) - 3$.

Bivariate distribution function

1.

$$F(-\infty, y) = \lim_{x \rightarrow -\infty} F(x, y) = 0,$$

$$F(x, -\infty) = \lim_{y \rightarrow -\infty} F(x, y) = 0,$$

$$F(\infty, \infty) = \lim_{x \rightarrow \infty, y \rightarrow \infty} F(x, y) = 1.$$

2. $P(x_1 < X \leq x_2, y_1 < Y \leq y_2)$
 $= F(x_2, y_2) - F(x_1, y_2) - [F(x_2, y_1) - F(x_1, y_1)].$

3. Right continuous in x :

$$\lim_{h \downarrow 0} F(x + h, y) = F(x, y),$$

Right continuous in y :

$$\lim_{h \downarrow 0} F(x, y + h) = F(x, y).$$

Marginal distribution functions

$$F_X(x) = \lim_{y \rightarrow \infty} F_{X,Y}(x, y) = F_{X,Y}(x, \infty),$$

$$F_Y(y) = \lim_{x \rightarrow \infty} F_{X,Y}(x, y) = F_{X,Y}(\infty, y).$$

Discrete bivariate distributions

1. Joint mass function:

$$f(x, y) = f_{X,Y}(x, y) = P(X = x, Y = y).$$

2. Probabilities:

$$\begin{aligned} &P(x_1 < X \leq x_2, y_1 < Y \leq y_2) \\ &= \sum_{x_1 < x \leq x_2} \sum_{y_1 < y \leq y_2} f(x, y). \end{aligned}$$

3. Marginal mass functions:

$$\begin{aligned} f_X(x) &= \sum_y f_{X,Y}(x, y), \\ f_Y(y) &= \sum_x f_{X,Y}(x, y). \end{aligned}$$

Continuous bivariate distributions

1. Joint density function: is a positive real-valued function f such that

$$F(x, y) = \int_{-\infty}^y \int_{-\infty}^x f(u, v) du dv \quad \text{for each } x, y \in \mathbb{R}.$$

2. Probabilities:

$$P(x_1 < X \leq x_2, y_1 < Y \leq y_2) = \int_{y_1}^{y_2} \int_{x_1}^{x_2} f(x, y) dx dy.$$

3. Marginal density functions:

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

$$f_Y(y) = \int_{-\infty}^{\infty} f_{X,Y}(x,y)dx,$$

Expectation of a function of two variables

$$E[g(X, Y)] = \begin{cases} \sum_y \sum_x g(x, y) f_{X,Y}(x, y), & \text{discrete case,} \\ \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} g(x, y) f_{X,Y}(x, y) dx dy, & \text{continuous case.} \end{cases}$$

Discrete conditional distributions

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1. Conditional distribution of Y given $X = x$:

$$F_{Y|X}(y|x) = P(Y \leq y | X = x).$$

2. Conditional mass of Y given $X = x$:

$$f_{Y|X}(y|x) = P(Y = y | X = x) = f_{X,Y}(x, y) / f_X(x).$$

3. Relationship between distribution and mass:

$$F_{Y|X}(y|x) = \sum_{y_i \leq y} f_{Y|X}(y_i|x).$$

Continuous conditional distributions

1. Conditional distribution of Y given

$X = x$:

$$F_{Y|X}(y|x) = \int_{-\infty}^y (f_{X,Y}(x, v) / f_X(x)) dv.$$

2. Conditional density of Y given $X = x$:

$$f_{Y|X}(y|x) = f_{X,Y}(x, y) / f_X(x).$$

Conditional, joint and marginal densities

1. Conditional mass/density

$$f_{Y|X}(y|x) = \frac{f_{X,Y}(x,y)}{f_X(x)}$$
$$= \begin{cases} \frac{f_{X,Y}(x,y)}{\sum_y f_{X,Y}(x,y)}, & \text{discrete case,} \\ \frac{f_{X,Y}(x,y)}{\int_{-\infty}^{\infty} f_{X,Y}(x,y)dy}, & \text{continuous case.} \end{cases}$$

2. Joint mass/density:

$$f_{X,Y}(x,y) = f_{Y|X}(y|x)f_X(x).$$

3. Marginal mass/density

$$f_Y(y) = \begin{cases} \sum_x f_{Y|X}(y|x)f_X(x), & \text{discrete case,} \\ \int_{-\infty}^{\infty} f_{Y|X}(y|x)f_X(x)dx, & \text{continuous case.} \end{cases}$$