

Session 11

conditional expectation

discrete R.V

$$E(X|Y=y) = \sum_x x P_{X|Y}(x|y)$$

$$E(g(X)|Y=y) = \sum_x g(x) P_{X|Y}(x|y)$$

continuous R.V

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

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

and also we have: $\text{Var}(X|Y) = E(X^2|Y) - (E(X|Y))^2$

Q1. The joint PDF of X and Y is given by $f(x, y) = (2m + 6)x^m y$; $0 < y < x < 1$
Find $\text{Var}(Y^2|X = x)$.

Moment Generating Function (MGF)

$$M_X(t) = E(e^{tx}) = \begin{cases} \sum_x e^{tx} \cdot p(x) & ; \text{ if } x \text{ is discrete} \\ \int_{-\infty}^{+\infty} e^{tx} f_X(x) dx & ; \text{ if } x \text{ is continuous.} \end{cases}$$

Note 1. $E(X^n) = \frac{d^n M_X(t)}{dt^n} \Big|_{t=0}$, for example, $\begin{cases} E(X) = M'_X(0) \\ E(X^2) = M''_X(0) \end{cases}$

Note 2. MGF of the sum of indep. r.v's equals to the product of individual MGFs, i.e., if $X_1, X_2, \dots, X_n$ are indep, then:

$$M_{X_1+X_2+\dots+X_n}(t) = M_{X_1}(t) \cdot M_{X_2}(t) \cdot \dots \cdot M_{X_n}(t)$$

Q2. If Z is a standard normal r.v. and if Y is defined by $Y = a + bZ + cZ^2$, then find $\rho(x, y)$.

Q3. Let $M_X(t) = e^{2e^t - 2}$ and $M_Y(t) = \left(\frac{3}{4}e^t + \frac{1}{4}\right)^{10}$. If X and Y are independent, find $E(XY)$.

Limit Theorems

* The central Limit theorem (CLT)

Let $X_1, X_2, \dots, X_n$ be a sequence of independent and identically distributed R.V.'s, each having mean " μ " and variance σ^2 .

$$\text{Then, } Y = \frac{\sum_{i=1}^n X_i - n\mu}{\sigma\sqrt{n}} = \frac{\bar{X} - \mu}{\sigma/\sqrt{n}} \overset{n \geq 30}{\sim} N(0,1)$$

In other words, as n gets larger the r.v Y , which is defined based on X_i 's, would have standard normal distribution.

Q4. Let $X_1, X_2, \dots, X_{20}$ be independent Poisson R.V.s with mean 1 and variance 1. Use CLT theorem to approximate $P\{\sum_{i=1}^{20} X_i > 15\}$