## Lecture 10

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## 1 Expectation

For a random variable X with distribution f(x),

$$E[X] = \int_{-\infty}^{\infty} x f(x) dx$$

or

$$E[X] = \sum_{x} x f(x)$$

where we denote

$$\mu = E[X]$$

## 2 Variance

Let X be a random variable with f(x). Then

$$var(X) = E[(X - \mu)^2]$$

which is

$$\int_{-\infty}^{\infty} (x - \mu)^2 f(x) dx$$

or

$$\sum_{x} (x - \mu)^2 f(x)$$

We write

$$\sigma^2 = \text{var}(x)$$

and define

$$\sigma = \sqrt{\sigma^2}$$

to be the standard deviation (which is non-negative).

Example 2.1. Consider the uniform distribution

$$f(x) = \begin{cases} \frac{1}{\alpha} & 0 \le x \le \alpha \\ 0 & \text{else} \end{cases}$$

Then the mean is  $\frac{\alpha}{2}$ . The variance is then

$$\int_{-\infty}^{\infty} \left( x - \frac{\alpha}{2} \right)^2 f(x) dx = \int_{0}^{\alpha} (x^2 - \alpha x + \alpha^2) \frac{1}{\alpha} dx$$
$$= \frac{\alpha^2}{3} - \frac{\alpha^2}{2} + \frac{\alpha^2}{4}$$
$$= \frac{\alpha^2}{12}$$

For  $\alpha \to 0$ , we get the Dirac Delta function (not a function).

A useful formula is

$$\sigma^{2} = \int_{-\infty}^{\infty} (x - \mu)^{2} f(x) dx$$

$$= \int_{-\infty}^{\infty} (x^{2} - 2\mu x + \mu^{2}) f(x) dx$$

$$= E[X^{2}] - 2\mu E[X] + \mu^{2}$$

$$= E[X^{2}] - 2\mu^{2} + \mu^{2}$$

$$= E[X^{2}] - \mu^{2}$$

## 3 Covariance

Consider the random variables X and Y, distribution f(x, y) with means  $\mu_x$  and  $\mu_y$ .

**Definition 3.1.** The covariance of X and Y is

$$\sigma_{X,Y} = \operatorname{cov}(X,Y) = E[(X - \mu_x)(Y - \mu_y)]$$

Observe

$$var(X) = \sigma_{xx}$$

If X is above its mean when Y is below its mean, we get a negative covariance. Similarly, if X and Y are both above/below their means, we get a positive covariance. We can consider covariance as a correlation.