# EE 214 PROBABILITY AND RANDOM PROCESSES IN ELECTRICAL ENGINEERING

Lecture 1B: Expected Value and Variance

#### STATISTICS OF RANDOM VARIABLES

■ A random experiment is performed N times. Random variable X takes on values  $x_1, x_2, ..., x_m$ .  $x_1$  occurs  $N_1$  times,  $x_2$  occurs  $N_2$  times, etc.

$$\overline{x} = \frac{N_1 x_1 + N_2 x_2 + \dots + N_m x_m}{N_1 + N_2 + \dots + N_m}$$

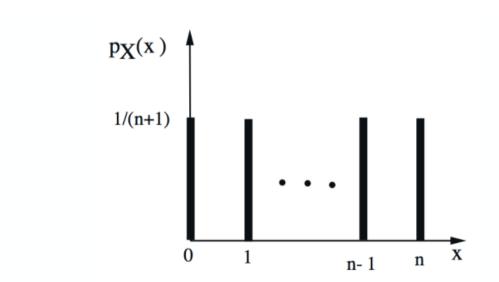
$$\overline{x} = \sum_{i} x_i \frac{N_i}{N}$$

$$\overline{x} = E[X] = \sum_{i} x_i p_X(x_i)$$

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# **EXPECTATION**

A discrete RV with uniform PMF



$$E[X] = 0 \times \frac{1}{n+1} + 1 \times \frac{1}{n+1} + \dots + n \times \frac{1}{n+1}$$

# **EXPECTATION**

In general, the expectation value of a random variable g(X), denoted by E[g(X)], is given by:

discrete

$$E[g(X)] = \sum_{i} g(x_i) p_X(x_i)$$

continuous

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

# PROPERTIES OF EXPECTATION

$$\blacksquare$$
  $E[c] = c$ 

 $\blacksquare E[cg(x)] = cE[g(x)]$ 

•  $E[g_1(x) + g_2(x)] = E[g_1(x)] + E[g_2(x)]$ 

# **DEFINITIONS**

Variance:

$$\sigma_X^2 = E[(X - \overline{X})^2] = \overline{(X - \overline{X})^2}$$

Standard Deviation

$$\sigma_X = \sqrt{E[(X - \overline{X})^2]}$$

Mean Square Value

$$E[X^2] = \overline{X^2}$$

## **EXAMPLE**

• Show that  $\sigma_X^2 = E[X^2] - (E[X])^2$ 

$$\sigma_X^2 = E[(X - \overline{X})^2] = E[X^2 - 2X\overline{X} + \overline{X}^2]$$

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

$$\sigma_X^2 = E[X^2] - 2\overline{X}(\overline{X}) + \overline{X}^2$$

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

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# **DEFINITIONS**

$$m_n = E[X^n] = \overline{X^n}$$

$$\mu_n = E[(X - \overline{X})^n] = (X - \overline{X})^n$$

#### BERNOULLI RANDOM VARIABLE

Probability mass function (PMF)

Outcome	X	$p_{X}(x)$
Failure	0	1 – p
Success	1	р

Statistics

$$E[X] = \sum x p_X(x) = (0)(1-p) + (1)(p) = p$$

$$E[X^2] = \sum x^2 p_X(x) = (0)^2 (1-p) + (1)^2 (p) = p$$

$$\sigma_X^2 = p - p^2$$

## BINOMIAL

Number of successes on n trials of a Bernoulli experiment

$$p_X(x) = \binom{n}{x} p^x (1-p)^{n-x}$$

Recall: Binomial theorem

$$\sum_{x=0}^{n} \binom{n}{x} p^x q^{n-x} = (p+q)^n$$

## **BINOMIAL**

mean

$$E[X] = np \sum_{y=0}^{n-1} {n-1 \choose y} p^{y} (1-p)^{n-1-y} = np$$

the mean square

$$E[X^2] = np(1-p) + n^2p^2$$

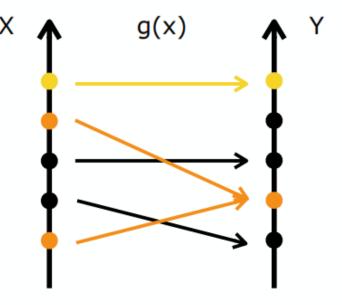
The variance

$$\sigma_X^2 = (np(1-p) + n^2p^2) - (np)^2$$
  
$$\sigma_X^2 = np(1-p)$$

# TRANSFORMATION OF RV

• Given 2 discrete RVs X and Y, and Y = g(X), how do we calculate  $p_Y(y)$ ?

$$p_Y(y) = \mathbf{P}(g(X) = y)$$
$$= \sum_{x:g(x)=y} p_X(x)$$



## **EXAMPLE**

• Consider a discrete random variable  $X = \{-1, 0, 1\}$  with probabilities  $p_{-1}$ ,  $p_0$  and  $p_1$  Determine the pmf of  $Y = X^2$ 

$$p_{X}(x) = \begin{cases} p_{-1} & x = -1 \\ p_{0} & x = 0 \\ p_{1} & x = 1 \end{cases}$$

Note: 
$$p_{-1} + p_0 + p_1 = 1$$

$$p_{Y}(y) = \begin{cases} p_{-1} & x = -1 \rightarrow y = 1 \\ p_{0} & x = 0 \rightarrow y = 0 \\ p_{1} & x = 1 \rightarrow y = 1 \end{cases}$$

$$p_{Y}(y) = \begin{cases} p_{0} & y = 0 \\ p_{-1} + p_{1} & y = 1 \end{cases}$$