## Lecture 13

## We explore *expectation* in more detail.

**Proposition:** For a discrete  $\mathbf{rv} X$  with  $\mathbf{pmf} p(x)$ 

 $\mathcal{E}(g_1(x) + \dots + g_k(x)) = \mathcal{E}(g_1(x)) + \dots + \mathcal{E}(g_k(x))$ 

**Definition:** The *variance* of a discrete rv X with pmf p(x) is

$$\sigma^2 \equiv \mathcal{V}(X) \equiv E[(X - \mathcal{E}(X))^2]$$

- we call  $\sigma$  the standard deviation
- $\sigma$  and  $\sigma^2$  are measures of spread
- contrast sample quantities  $(\bar{x}, s)$  with popln quantities  $(\mu, \sigma)$

Example: Consider the experiment consisting of three flips of a coin. Let  $X \equiv$  the number of heads. Obtain V(X).

**Proposition:**  $V(X) = E(X^2) - (E(X))^2$ 

**Proposition:**  $V(aX + b) = a^2V(X)$ 

Example: Let X be the average january temperature in degrees Celsius where E(X) = 5C and  $V(X) = 3(C)^2$ . Find the expected value and the variance of Y where Y is the average january temperature in degrees Fahrenheit. **Problem:** Calculate  $\sigma$  and  $E(3X + 4X^2)$  corresponding to the rv X with pmf p(x) where

Example: In a game of chance, I bet x dollars. With probability p, I win y dollars. What should x be for this to be a fair game? Definition: A discrete rv X has a *Binomial* distribution denoted  $Bin(n, \theta)$  if it has pmf

$$p(x) = \begin{cases} \binom{n}{x} \theta^x (1-\theta)^{n-x} & x = 0, 1, \dots, n \\ 0 & \text{otherwise} \end{cases}$$

Motivation for the Binomial - the most important discrete distribution:

Consider performing an experiment n times where the probability of success in every trial is  $\theta$  and the n experiments are independent. We are interested in the probability of x successes.

The probability of getting x successes S and n-x failures F in the specific order

$$SS\cdots S$$
  $FF\cdots F$ 

is

$$\begin{array}{l} \theta\theta\cdots\theta \quad (1-\theta)(1-\theta)\cdots(1-\theta) \\ = \ \theta^x(1-\theta)^{n-x}. \end{array}$$

Therefore

$$P(x \text{ successes}) = {\binom{n}{x}} \theta^x (1-\theta)^{n-x}$$

for x = 0, 1, ..., n.

Key points for the Binomial distribution:

- 1. the n trials are independent
- 2. same probability of success  $\theta$  in each trial

Example: You roll a die 10 times and are interested in obtaining 5's or 6's. What is the probability that x rolls result in either 5's or 6's?

For 
$$X \sim \operatorname{Bin}(n, \theta)$$
  
•  $\operatorname{E}(X) = n\theta$   
•  $\operatorname{V}(X) = n\theta(1 - \theta)$ 

Proof (first result):

$$\begin{split} \mathbf{E}(X) &= \Sigma_{x=0}^{n} x \begin{pmatrix} n \\ x \end{pmatrix} \theta^{x} (1-\theta)^{n-x} \\ &= \Sigma_{x=1}^{n} x_{\overline{x!(n-x)!}}^{n!} \theta^{x} (1-\theta)^{n-x} \\ &= n\theta \ \Sigma_{x=1}^{n} \frac{(n-1)!}{(x-1)!(n-x)!} \ \theta^{x-1} (1-\theta)^{n-x} \\ &= n\theta \ \Sigma_{y=0}^{n-1} \frac{(n-1)!}{y!(n-1-y)!} \ \theta^{y} (1-\theta)^{n-1-y} \\ &= n\theta \ \Sigma_{y=0}^{n-1} \begin{pmatrix} n-1 \\ y \end{pmatrix} \theta^{y} (1-\theta)^{n-1-y} \\ &= n\theta \end{split}$$