

# MATH 451/551

## Chapter 6. Joint Distribution

### 6.3 Expected Values

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# Example 1



## Example 1

For the random variables  $X$  and  $Y$  with joint probability density function  $f(x, y) = 2$ ,  $x > 0$ ,  $y > 0$ ,  $x + y < 1$ , find  $V(E(Y|X))$ .

# Theorem



## Theorem 6.11

If  $X$  and  $Y$  are random variables, then

$$E(X) = E\{E(X|Y)\}$$

when the expectations exist.

# Theorem



## Theorem 6.12

If  $X$  and  $Y$  are random variables, then

$$V(X) = E\{V(X|Y)\} + V\{E(X|Y)\}$$

when the expectations exist.

## Example 2



### Example 2

The binomial distribution was introduced as the number of successes in  $n$  independent Bernoulli trials, each with a probability of success  $p$ . The population mean and variance of  $X \sim \text{Bin}(n, p)$  are

$$E(X) = np \quad \text{and} \quad V(X) = np(1 - p).$$

But what if an application arises where  $n$  is fixed but  $p$  is unknown? Such an application might arise in an area of quality control known as **acceptance sampling**, where a fixed number of items are sampled from a large lot. To be more specific, assume that  $p$  is no longer a fixed probability, but is itself a random variable. Furthermore, assume that you have so little information about the probability of success on each Bernoulli trial that you can only state that the probability of success  $P$  is a random variable that is equally likely to fall between 0 and 1. That is to say,  $P \sim U(0, 1)$ . The problem here is to find  $E(X)$  and  $V(X)$  when  $X \sim \text{Bin}(n, P)$ , where  $P \sim U(0, 1)$ .

# Thank You



# THANK YOU!

