

# STAT 516

Using conditioning to evaluate mean and variance

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- ▶ The following approach often helps to find an expectation with respect to a complicated joint pmf:
  1. Condition  $X$  on the value  $y$  of a suitable random variable  $Y$
  2. Compute the conditional expectation  $E(X|y)$
  3. Average the resulting conditional expectation over  $y$
- ▶ The choice of  $Y$  is crucial

# The law of iterated expectations

- ▶ Let  $X$  and  $Y$  be random variables on the same probability space  $\Omega$ .
- ▶ Suppose  $E(X)$  and  $E(X|Y = y)$  exist for each  $y$ .
- ▶ Then,

$$E(X) = E_Y[E(X|Y = y)]$$

- ▶ In the discrete case, this amounts to

$$E(X) = \sum_y \mu_X(y) p_Y(y)$$

- ▶ The discrete case is very simple: by definition,

$$\mu_X(y) = \frac{\sum_x xp(x, y)}{p_Y(y)}$$

- ▶ Thus,

$$\begin{aligned}\sum_y \mu_X(y)p_Y(y) &= \sum_y \sum_x xp(x, y) = \sum_x \sum_y xp(x, y) \\ &= \sum_x x \sum_y p(x, y) = \sum_x xp_X(x) = E(X)\end{aligned}$$

# Iterated Variance formula

- ▶ Let  $X$  and  $Y$  be random variables defined on the same probability space  $\Omega$
- ▶ Suppose  $\text{Var}(X)$  and  $\text{Var}(X|Y = y)$  exist for each  $y$ .
- ▶ Then,

$$\text{Var}(X) = E_Y[\text{Var}(X|Y = y)] + \text{Var}_Y[E(X|Y = y)]$$

- ▶ This formula is valid for all types of variables (not just continuous ones)

# Some simple implications of the iterated variance formula



$$\text{Var}(g(X)|X = x) = 0$$



$$\text{Var}(g(X)h(Y)|Y = y) = h^2(y)\text{Var}(g(X)|Y = y)$$

## Example I: a two-stage experiment

- ▶ Suppose  $n$  fair dice are rolled. Those that show a six are rolled again
- ▶ What are the mean and the variance of the number of sixes obtained in the second round of this experiment?
- ▶ Suppose  $Y$  is the number of dice in the first round that show a six
- ▶ Let  $X$  be the number of dice in the second round that show a six

## Example I: a two-stage experiment

- ▶ Given  $Y = y$ ,  $X \sim \text{Bin}(y, \frac{1}{6})$
- ▶ Also,  $Y \sim \text{Bin}(n, \frac{1}{6})$
- ▶ Therefore,

$$E(X) = E[E(X|Y = y)] = E_Y \left[ \frac{y}{6} \right] = \frac{n}{36}$$



## Example I: a two-stage experiment

- ▶ Moreover,

$$\begin{aligned} \text{Var}(X) &= E_Y[\text{Var}(X|Y = y)] + \text{Var}_Y[E(X|Y = y)] \\ &= E_Y \left[ y \frac{1}{6} \frac{5}{6} \right] + \text{Var}_Y \left[ \frac{y}{6} \right] \\ &= \frac{5}{36} \frac{n}{6} + \frac{1}{36} n \frac{1}{6} \frac{5}{6} \\ &= \frac{35n}{1296} \end{aligned}$$