

# Two random variables

- ▶ An RV by itself is not that interesting
- ▶ In this course: **relationships** between two or more RVs
- ▶ Example: Do **college** graduates **earn** more than non-college graduates?
- ▶ Other examples: see intro slides

# Concepts

Tools for describing multiple random variables:

- ▶ **Joint, marginal** distribution
- ▶ **Conditional** distribution/expectation/variance
- ▶ **Independence**, covariance and correlation
- ▶ **Rules** for expectations of multiple RVs

## Example (SW, 2.2)

Commute	Rain ( $X = 0$ )	No rain ( $X = 1$ )	Total
Long ( $Y = 20$ )	0.15	0.07	[0.22]
Short ( $Y = 40$ )	0.15	0.63	[0.78]
Total	[0.30]	[0.70]	[1.00]

**Table:** 2.2 in SW, p. 26

# Joint, marginal, conditional

Two random variables,  $X$  and  $Y$

- ▶ Joint:

- ▶  $P(X = x, Y = y)$ ,
- ▶ probability that it rains and I have a long commute

- ▶ Marginal

- ▶  $P(X = x) = \sum_{i=1}^k P(X = x, Y = y_i)$
- ▶ probability that it rains

- ▶ Conditional

- ▶  $P(Y = y | X = x) = \frac{P(X=x, Y=y)}{P(X=x)}$
- ▶ probability of having a long commute **given** that it rains

# Conditional mean and variance

- ▶ Mean and variance for conditional distribution
- ▶ The **conditional expectation** of  $X$  given  $Y = y$  is the expectation of  $X$  if you already have the information that  $Y = y$
- ▶ Same for the **conditional variance**

## Conditional expectation: example

- ▶  $M$  is the number of times your computer crashes
- ▶  $A$  denotes computer is old ( $A = 0$ ) or new ( $A = 1$ )

	$M$				
	0	1	2	3	4
$A = 0$	0.35	0.065	0.05	0.025	0.01
$A = 1$	0.45	0.035	0.01	0.005	0.00

- ▶ Q: If I **know** the computer is old, how many crashes do I **expect**?

## Conditional expectation: formula

- ▶ The answer to the question is

$$E[M|A = 0] = \sum_{k=1}^K M_k P(M_k|A = 0)$$

- ▶ Compare this to the **unconditional** expectation

$$E[M] = \sum_{k=1}^K M_k P(M_k)$$

- ▶ Replaced **marginal** by **conditional** probability distribution

# Covariance, correlation

- ▶ Two more ways to measure the relationship between two r.v.'s are the **covariance** and **correlation**
- ▶ The **covariance** and **correlation** measure whether two variables are moving together, with respect to the mean
- ▶ Formulas: overhead



# Interpretation

- ▶ The sign of the **covariance** is easy to interpret
- ▶ The magnitude of the covariance is hard to interpret
- ▶ The sign and magnitude of the **correlation** are easy to interpret
- ▶ The correlation is always between -1 and 1
- ▶ The closer to 1 (-1), the more strongly positively (negatively) correlated the variables are

# Mean and variance, multiple RV

- ▶ Most important rules to remember from “The Mean and Variance of Sums of Random Variables”, pp. 32–35:
  - ▶  $E[X + Y] = E[X] + E[Y]$
  - ▶  $Var[X + Y] = Var[X] + Var[Y] + 2Cov(X, Y)$
- ▶ If the covariance between  $X$  and  $Y$  is zero ( $\sigma_{XY} = 0$ ),

$$Var[X + Y] = Var[X] + Var[Y]$$