

# UGA L3 MIAASH: Econometrics 1

## Review of Probability

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## References

- Wooldridge (2013) appendix B
- Stock and Watson (2009) chapter 2
- Abbring (2001) sections 2.1-2.3
- Diez, Barr, and Cetinkaya-Rundel (2012) chapter 2
- Grinstead and Snell (2003) chapters 1-7
  
- These lectures are mostly based on Wooldridge, so that will generally be the best reference to read.
- The others are some freely available alternatives that cover similar material.
- Abbring (2001) is most similar to Wooldridge in terms of length and its focus on econometrics.
- Diez, Barr, and Cetinkaya-Rundel (2012) and Grinstead and Snell (2003) also cover similar material, but explain things in more depth.

# Probability

- Purpose: system for quantifying chance and making predictions about future events
- Interpretations:
  - Relative frequency of an event in many repeated trials
  - Subjective assessment of the likelihood of an event

## Basic definitions

- The properties of probability are fairly intuitive, and you can work with probabilities without worrying too much about how probability is formally defined.
- However, if we want to be mathematically rigorous, we should carefully define what we mean by probability.
- This slide formally defines probability. Do not worry if it seems very abstract.

## Basic definitions

- **Random experiment**: procedure that has well-defined set of outcomes and “could” be infinitely repeated
- **Sample space**: set of possible outcomes of an experiment,  
 $S = \{a_1, a_2, \dots, a_J\}$
- **Event**: any subset of sample space,  $A \subseteq S$
- **Probability**: function from all subsets of the sample space,  $S$ , to  $[0, 1]$  such that
  1.  $\mathbb{P}(S) = 1$
  2.  $1 \geq \mathbb{P}(A) \geq 0$  for all  $A \subseteq S$
  3. If  $A_1, A_2, \dots$  are disjoint events then  $\mathbb{P}(A_1 \cup A_2 \cup \dots) = \mathbb{P}(A_1) + \mathbb{P}(A_2) + \dots$
- **Random variable**: function from  $S$  to a number

## Basic definitions

- An example of a random experiment is rolling a dice. In this case the sample space is  $S = \{1, 2, 3, 4, 5, 6\}$ .
- Some events are: get a 1 =  $\{1\}$ , get more than 4,  $\{5, 6\}$ , etc.
- A random variable is a numeric representation of a random experiment.
- For example, our experiment could be flipping a coin. Then  $S = \{\text{heads}, \text{tails}\}$ .
- A function assigns a number to each element of  $S$ , so one example of a random variable is  $X = \begin{cases} 1 & \text{if heads} \\ 0 & \text{if tails} \end{cases}$ .
- Another is  $Y = \begin{cases} 50 & \text{if heads} \\ -257 & \text{if tails} \end{cases}$ .
- A third example is  $Z = \begin{cases} 1 & \text{if heads} \\ 1 & \text{if tails} \end{cases}$ .

# Properties of probability

1.  $\mathbb{P}(\emptyset) = 0$
2.  $\mathbb{P}(A^c) = 1 - \mathbb{P}(A)$
3.  $A \subseteq B$  implies  $\mathbb{P}(A) \leq \mathbb{P}(B)$
4.  $\mathbb{P}(A \cup B) = \mathbb{P}(A) + \mathbb{P}(B) - \mathbb{P}(A \cap B)$



## Conditional probability

- **Conditional probability:**  $\mathbb{P}(A|B) = \mathbb{P}(A \cap B)/\mathbb{P}(B)$ 
  - Satisfies axioms of unconditional probability (and so also has properties on previous slide)
- **Bayes' Rule**  $\mathbb{P}(A|B) = \frac{\mathbb{P}(B|A)\mathbb{P}(A)}{\mathbb{P}(B)}$
- $A$  is **independent** of  $B$  iff  $\mathbb{P}(A \cap B) = \mathbb{P}(A)\mathbb{P}(B)$ , denote as  $A \perp\!\!\!\perp B$ 
  - $A \perp\!\!\!\perp B$  implies  $\mathbb{P}(A|B) = \mathbb{P}(A)$  and  $\mathbb{P}(B|A) = \mathbb{P}(B)$ .

## Discrete random variables

- Recall: a **random variable** is a function from  $S$  to a number
- A random variable is **discrete** if it can only take countably many different values
- E.g.  $X \in \{x_1, \dots, x_m\}$
- **Probability mass function (PMF)** of  $X$ ,  $p_i = \mathbb{P}(X = x_i)$  gives the probability that  $X$  equals each of its possible values
- **Cumulative distribution function (CDF)**:  
$$F(x) = \mathbb{P}(X \leq x) = \sum_{i=1}^m p_i 1\{x_i \leq x\}$$

## Continuous random variables

- A random variable is **continuous** if it takes on any single real value with zero probability
- Set of possible values is uncountably infinite (e.g. real line or line segment)
- CDF is continuous and differentiable
- **Probability density function (PDF)**: the derivative of the CDF

$$f(x) = \frac{dF}{dx}(x) \quad \text{and} \quad F(x) = \int_{-\infty}^x f(t)dt$$

## Bivariate distributions

### ■ Discrete:

- Joint PMF  $f_{X,Y}(x,y) = \mathbb{P}(X = x, Y = y)$
- Marginal PMF  $f_X(x) = \mathbb{P}(X = x) = \sum_{y \in \text{all values of } Y} \mathbb{P}(X = x, Y = y)$
- Conditional PMF  $f_{X|Y}(x|y) = \frac{f_{X,Y}(x,y)}{f_Y(y)}$

### ■ Continuous:

- joint CDF  $F_{X,Y}(x,y) = \mathbb{P}(X \leq x, Y \leq y) = \int_{-\infty}^x \int_{-\infty}^y f_{X,Y}(u,t) du dt$
- joint PDF  $f_{X,Y}(x,y) = \frac{\partial^2 F_{X,Y}}{\partial x \partial y}(x,y)$
- Marginal CDF for  $X$ ,  $F_X(x) = \mathbb{P}(X \leq x) = F_{X,Y}(x, \infty)$
- Marginal PDF for  $X$ ,  $f_X(x) = \frac{dF_X}{dx}(x) = \int_{-\infty}^{\infty} f_{X,Y}(x,y) dy$
- Conditional PDF  $f_{X|Y}(x|y) = \frac{f_{X,Y}(x,y)}{f_Y(y)}$

# Independence

- Random variables  $X, Y$  are independent if  $f_{X,Y}(x, y) = f_X(x)f_Y(y)$  for all  $x, y$
- $X \perp\!\!\!\perp Y$  implies  $f_{X|Y}(x|y) = f_X(x)$

# Expectation

- The **expectation** of  $X$  is

$$\mathbb{E}[X] = \begin{cases} \sum_{i=1}^m x_i f_X(x_i) & \text{if } X \text{ discrete} \\ \int_{-\infty}^{\infty} x f_X(x) dx & \text{if } X \text{ continuous} \end{cases}$$

- Is a constant
- a.k.a. expected value, population average, population mean, first moment
- Expectation is linear:

$$\mathbb{E}[aX + bY] = a\mathbb{E}[X] + b\mathbb{E}[Y]$$

- Expectation of function  $g$ :

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

## Variance and other moments

- **$k$ th moment** of  $X$ :  $\mathbb{E}[X^k]$
- **$k$ th central moment** of  $X$ :  $\mathbb{E}[(X - \mathbb{E}[X])^k]$
- **Variance** is the 2nd central moment

$$\begin{aligned}\mathbb{V}(X) &= \mathbb{E}[(X - \mathbb{E}[X])^2] \\ &= \mathbb{E}[X^2] - \mathbb{E}[X]^2\end{aligned}$$

- $\mathbb{V}(a + bX) = b^2\mathbb{V}(X)$
- **Standard deviation** is  $\sqrt{\mathbb{V}(X)}$

## Covariance

- **Covariance** of  $X$  and  $Y$ :

$$\begin{aligned}\mathbb{C}(X, Y) &= \mathbb{E}[(X - \mathbb{E}[X])(Y - \mathbb{E}[Y])] \\ &= \mathbb{E}[XY] - \mathbb{E}[X]\mathbb{E}[Y] \\ &= \mathbb{E}[(X - \mathbb{E}[X])Y] = \mathbb{E}[X(Y - \mathbb{E}[Y])]\end{aligned}$$

- $\mathbb{C}(X, X) = \mathbb{V}(X)$
- $\mathbb{C}(a_1 + b_1X + c_1Y, a_2 + b_2X + c_2Y) =$   
 $b_1b_2\mathbb{V}(X) + c_1c_2\mathbb{V}(Y) + (b_1c_2 + c_1b_2)\mathbb{C}(X, Y)$
- $\mathbb{V}(a + bX + cY) = b^2\mathbb{V}(X) + c^2\mathbb{V}(Y) + 2bc\mathbb{C}(X, Y)$
- $\mathbb{V}\left(\sum_{i=1}^N b_i X_i\right) = \sum_{i=1}^N \left(\sum_{j=1}^N b_i b_j \mathbb{C}(X_i, X_j)\right)$



# Correlation

- **Correlation** of  $X$  and  $Y$  is  $\text{Corr}(X, Y) = \frac{\mathbb{C}(X, Y)}{\sqrt{\mathbb{V}(X)\mathbb{V}(Y)}}$
- **Cauchy-Schwartz inequality**:  $|\mathbb{C}(X, Y)| \leq \sqrt{\mathbb{V}(X)\mathbb{V}(Y)}$ 
  - Implies  $-1 \leq \text{Corr}(X, Y) \leq 1$
- $\text{Corr}(X, Y) = \pm 1$  iff  $Y = a + bX$

## Conditional expectation

- **Conditional expectation** of  $Y$  given  $X = x$ :

$$\mathbb{E}[Y|X = x] = \begin{cases} \sum_{i=1}^m y_i f_{Y|X}(y_i|x) & \text{if discrete} \\ \int_{-\infty}^{\infty} y f_{Y|X}(y|x) dy & \text{if continuous} \end{cases}$$

- $\mathbb{E}[Y|X]$  is a function of  $X$
- Has all properties of unconditional expectation
- Properties:
  - $\mathbb{E}[g_1(X) + g_2(X)Y|X] = g_1(X) + g_2(X)\mathbb{E}[Y|X]$
  - **Law of iterated expectations:**  $\mathbb{E}_X [\mathbb{E}_{Y|X}[Y|X]] = \mathbb{E}[Y]$
  - $\mathbb{E}[X(Y - \mathbb{E}[Y|X])] = 0$

## Conditional variance

- **Conditional variance:**  $\mathbb{V}(Y|X) = \mathbb{E}[(Y - \mathbb{E}[Y|X])^2 | X]$
- Relation to variance:

$$\mathbb{V}(Y) = \mathbb{E}_X[\mathbb{V}(Y|X)] + \mathbb{V}(\mathbb{E}[Y|X])$$

- Analysis of variance (ANOVA)
- $\mathbb{E}_X[\mathbb{V}(Y|X)]$  is within- $X$  variance
- $\mathbb{V}(\mathbb{E}[Y|X])$  is between- $X$  variance

## References

- Abbring, Jaap. 2001. "An Introduction to Econometrics: Lecture notes." URL <http://jabbring.home.xs4all.nl/courses/b44old/lect210.pdf>.
- Diez, David M, Christopher D Barr, and Mine Cetinkaya-Rundel. 2012. *OpenIntro Statistics*. OpenIntro. URL <http://www.openintro.org/stat/textbook.php>.
- Grinstead, Charles M and J. Laurie Snell. 2003. *Introduction to Probability*. American Mathematical Society. URL [http://www.dartmouth.edu/~chance/teaching\\_aids/books\\_articles/probability\\_book/book.html](http://www.dartmouth.edu/~chance/teaching_aids/books_articles/probability_book/book.html).
- Stock, J.H. and M.W. Watson. 2009. *Introduction to Econometrics*, 2/E. Addison-Wesley.
- Wooldridge, J.M. 2013. *Introductory econometrics: A modern approach*. South-Western.