

UGA M1: Econometrics 1 **Probability**

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- 1. Definitions
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References

- ? appendix B
- ? chapter 2
- ? sections 2.1-2.3
- ? chapter 2
- ? 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.
- ? is most similar to Wooldridge in terms of length and its focus on econometrics
- ? and ? 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, $S = \{a_1, a_2, ..., a_J\}$
- **Event**: any subset of sample space, $A \subseteq S$
- ullet Probability: function from all subsets of the sample space, S, to [0,1] such that
 - 1. $\mathbb{P}(S) = 1$
 - 2. $1 > \mathbb{P}(A) > 0$ for all $A \subseteq S$
 - 3. If $A_1, A_2, ...$ are disjoint events then $\mathbb{P}(A_1 \cup A_2 \cup ...) = \mathbb{P}(A_1) + \mathbb{P}(A_2) + ...$
- 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 = {heads, 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. $X \in \{x_1, ..., x_m\}$
- 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):

$$F(x) = \mathbb{P}(X \le x) = \sum_{i=1}^{m} p_i \mathbb{1}\{x_i \le x\}$$

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

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

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:

- \blacksquare joint CDF $F_{X,Y}(x,y) = \mathbb{P}(X \le x, Y \le y) = \int_{-\infty}^{x} \int_{-\infty}^{y} f_{X,Y}(u,t) du dt$
- joint PDF $f_{X,Y}(x,y) = \frac{\partial^2 F_{X,Y}}{\partial x \partial y}(x,y)$
- Marginal CDF for X, $F_X(x) = \mathbb{P}(X \le x) = F_{X,Y}(x,\infty)$
- 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
- $X \perp \!\!\!\perp Y \text{ implies } f_{X|Y}(x|y) = f_X(x)$

Expectation

 \blacksquare 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

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

$$Var(X) = \mathbb{E}\left[(X - \mathbb{E}[X])^2\right]$$
$$= \mathbb{E}[X^2] - \mathbb{E}[X]^2$$

- $Var(a+bX) = b^2 Var(X)$
- Standard deviation is $\sqrt{\operatorname{Var}(X)}$

Covariance

Covariance of X and Y:

$$Cov(X, Y) = \mathbb{E} [(X - \mathbb{E}[X]) (Y - \mathbb{E}[Y])]$$

$$= \mathbb{E}[XY] - \mathbb{E}[X]\mathbb{E}[Y]$$

$$= \mathbb{E}[(X - \mathbb{E}[X]) Y] = \mathbb{E}[X (Y - \mathbb{E}[Y])]$$

- $ightharpoonup \operatorname{Cov}(X,X) = \operatorname{Var}(X)$
- $Var(a+bX+cY) = b^2 Var(X) + c^2 Var(Y) + 2bc Cov(X,Y)$

Correlation

- Correlation of X and Y is $Corr(X, Y) = \frac{Cov(X, Y)}{\sqrt{Var(X)Var(Y)}}$
- Cauchy-Schwartz inequality: $|Cov(X, Y)| \le \sqrt{Var(X)Var(Y)}$
 - Implies $-1 \le \operatorname{Corr}(X, Y) \le 1$

Conditional expectation

• Conditional expectation of Y given X = x:

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

- $\blacksquare \mathbb{E}[Y|X]$ is a function of X
- Has all properties of unconditional expectation
- Properties:

 - $$\begin{split} & \quad \mathbb{E}[g_1(X) + g_2(X)Y|X] = g_1(X) + g_2(X)\mathbb{E}[Y|X] \\ & \quad \text{Law of iterated expectations: } \mathbb{E}_X\left[\mathbb{E}_{Y|X}[Y|X]\right] = \mathbb{E}[Y] \end{split}$$
 - $\mathbb{E}[X(Y \mathbb{E}[Y|X])] = 0$

Conditional variance

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

$$\operatorname{Var}(Y) = \mathbb{E}_X \left[\operatorname{Var}(Y|X) \right] + \operatorname{Var} \left(\mathbb{E}[Y|X] \right)$$

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

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