Lecture 11 Probability

Ryan $McWay^{\dagger}$

 $^{\dagger}Applied\ Economics,$ University of Minnesota

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LAST LECTURE REVIEW

- ► (Inequality) Constrained Optimization:
 - ► Kuhn Tucker Conditions
 - ► Corner Solutions
- ► Comparative Statics & Envelope Theorem:
 - ► The Multiplier
 - ► Comparative Statics
 - ► Envelope Theorem

REVIEW ASSIGNMENT

- 1. Problem Set 10 solutions are available on Github.
- 2. Any issues or problems **You** would like to discuss?

DAILY ICEBREAKER

- ► Attendance via prompt:
 - ► Name
 - ▶ Daily Icebreaker: What crime would you commit if you would either never be caught for it or had complete immunity from conviction?



Probability

MOTIVATION

- ► General background
 - ▶ How we evaluate the likelihood of events occurring.
- ▶ Why do economists' care?
 - ▶ This provides a foundation for 'empirical' economics to work with real-world data.
- ► Application in this career
 - ▶ In working with estimators and throughout econometrics.

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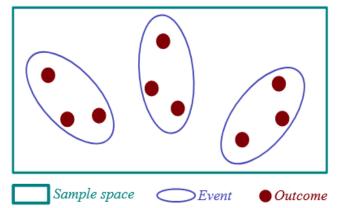
OVERVIEW

- 1. Outcomes & Events
- 2. Probability
- 3. Probability Limits
- 4. Independence
- 5. Law of Total Probability
- 6. Conditional Probability
- 7. Cumulative Distribution Function
- 8. Probability Distribution Function
- 9. Conditional Probability
 Distribution Function

- 10. Joint & Marginal Distributions
- 11. Gaussian (Normal)
 Distribution
- 12. Other Distributions
- 13. Bayes Rules
- 14. Moments of a Distribution
- 15. Variance & Standard Deviation
- 16. Covariance
- 17. Correlation

1. OUTCOMES AND EVENTS

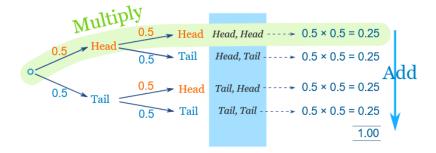
- ▶ Outcomes: All possible values that may be realized given the domain.
- ► Sample space: The set of all possible outcomes.
- ► Event: A subset of the outcomes in the sample space.



2. Probability

- ► Experiment: A procedure that could be infinitely repeated with a well-defined set of outcomes.
- ▶ Random Trail: One run of the experiment.
- ▶ Relative Frequency: Fraction of random trails in which an event occurs.
- ▶ Probability Pr(A): The relative frequency approached in the limit as the experiment is repeated infinitely.
 - ► How likely an outcome will occur in any given random trail.
- ▶ Probability Tree: A diagram of potential outcomes determining the probability of occurrence.
- ightharpoonup Complement A^c : All other events except those that occur in event A.
 - $ightharpoonup Pr(A^c) = 1 Pr(A)$

PROBABILITY TREE



2. Probability

- ▶ Probabilities are ranged 0-1, with the sum of all possible events equaling 1.
 - ightharpoonup Pr(A) < 1
- ▶ Non-existent Event: $Pr(\emptyset) = 0$
- ▶ Monotone Probability Inequality: If $A \subset B \implies Pr(A) \leq Pr(B)$
- ► Inclusion-Exclusion Principle: $Pr(A \cup B) = Pr(A) + Pr(B) Pr(A \cap B)$
- ▶ Boole's Inequality: $Pr(A \cup B) \le Pr(A) + Pr(B)$
- ▶ Bonferroni's Inequality: $Pr(A \cap B) \ge Pr(A) + Pr(B) 1$

3. PROBABILITY LIMITS

- ▶ Probability limits are used to test large sample properties of a distribution or estimator.
- ▶ This relies on asymptotics of the distribution as $n \to \infty$.
- ightharpoonup Probability limit approaching c:

$$\lim_{n \to \infty} Pr([x_n - c] > \varepsilon) = 0$$

$$x_n = c$$

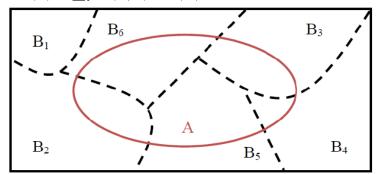
$$x_n \xrightarrow{p} c$$

4. Independence

- ▶ Independence: Events where the probability of an event is unrelated to the outcome of another event.
- ightharpoonup Pr(A|B) = Pr(A)
- $ightharpoonup Pr(A \wedge B) = Pr(A|B) * Pr(B) = Pr(A) * Pr(B)$
- ▶ Mutually Exclusive: Each outcome have nothing in common (do not share the sample space).
- $ightharpoonup E_i \cap E_j = \emptyset$

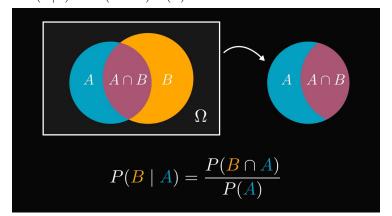
5. LAW OF TOTAL PROBABILITY

- Suppose the sample space is partitioned into n disjoint events: B_1, \ldots, B_n .
- $ightharpoonup Pr(A) = \sum_{i}^{n} Pr(A|B_i) * Pr(B_i)$



6. CONDITIONAL PROBABILITY

- ightharpoonup Pr(A|B): The probability of A occurring given B occurs.
- $ightharpoonup Pr(A|B) = Pr(A \wedge B)Pr(B)$



DEMONSTRATION: CONDITIONAL PROBABILITY

Question:

The probability a product breaks down is $P(T \ge t) = e^{-t/5} \forall t \ge 0$. I have the product for two years and didn't breakdown. What is the probability it breaks down in year 3?

Answer

$$P(B) = P(T \ge 2) = e^{-2/5}$$

$$P(A) = P(A \cap B) = P(2 \le T \le 3) = e^{-2/5} - e^{-3/5}$$

$$P(A|B) = \frac{P(A \cap B)}{P(B)}$$

$$= \frac{e^{-2/5} - e^{-3/5}}{e^{-2/5}} \approx 0.1813$$

DEMONSTRATION: CONDITIONAL PROBABILITY

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7. CUMULATIVE DISTRIBUTION FUNCTION

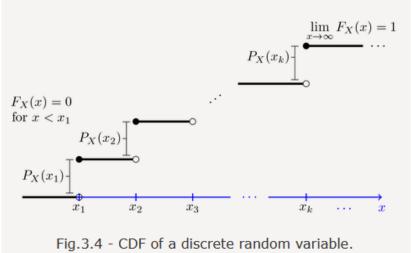
- ightharpoonup F(x): Describes the probability that a random variable x is less than or equal to that particular realization.
- ▶ $0 \le F(X) \le 1$
- ▶ Discrete:

$$F(x_j) = Pr(X \le x_j)$$

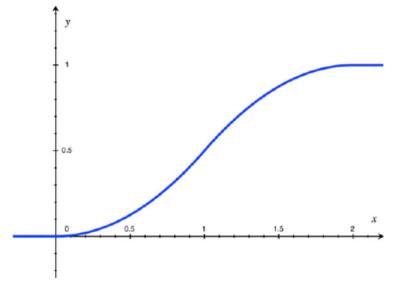
► Continuous:

$$F(x_j) = \int_{-\infty}^{x_j} f(x) dx$$

7. CUMULATIVE DISTRIBUTION FUNCTION



7. CUMULATIVE DISTRIBUTION FUNCTION



8. Probability Distribution Function

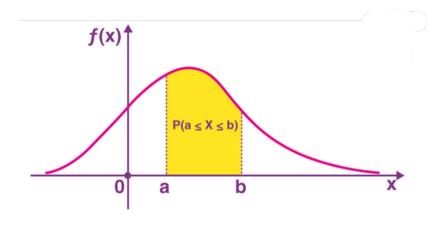
- ▶ f(x): The frequency of a random variable outcomes across the sample space.
- $ightharpoonup 0 \le f(x) \le 1$
- ▶ Discrete:

$$f(x_j) = \begin{cases} p_j & \text{where } j = 1, \dots, k \\ 0 & \text{otherwise} \end{cases}$$

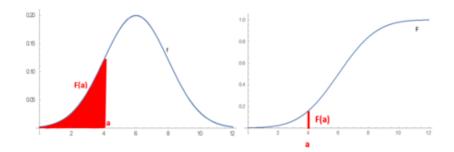
► Continuous:

$$Pr(a \le X \le b) = \int_a^b f(x)dx$$

8. Probability Distribution Function



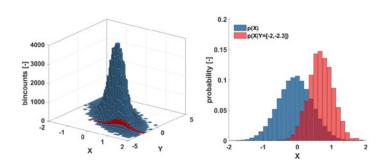
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9. CONDITIONAL PROBABILITY DISTRIBUTION FUNCTION

- ▶ Note that $Pr(A|B) = \frac{Pr(A \land B)}{Pr(B)}$.
- ▶ Implies $Pr(A \land B) = Pr(A|B) \times Pr(B)$.
- ▶ Using this logic: $f_{Y|X}(y|x) = f(y|x) = Pr(Y = y|X = x)$.
- ► Implies $f(y|x) = \frac{f(x,y)}{f(x)}$
- ▶ If X and Y are independent, then $f(y|x) = f(y) \forall x, y$.
- ► Implies $f_{X,Y}(x,y) = f_X(x) \times f_Y(y) \forall x, y$

9. CONDITIONAL PROBABILITY DISTRIBUTION FUNCTION



10. JOINT AND MARGINAL DISTRIBUTIONS

- ▶ Joint Distribution $Pr(A \land B)$: The joint probability density functions.
- ► Discrete:

$$f_{X,Y}(x,y) = Pr(X = x \land Y = y)$$

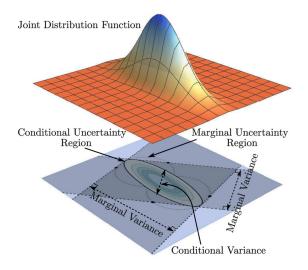
► Continuous:

$$Pr(a \le x \le b \land c \le y \le d) = \int_{c}^{d} \int_{a}^{b} f(x, y) dx dy$$

ightharpoonup Marginal Distribution f(x): The joint distribution for x conditioning on all values of y.

$$f(x) = \sum_{y} f(x, y)$$

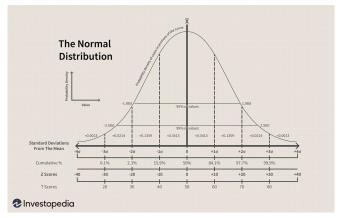
10. JOINT AND MARGINAL DISTRIBUTIONS



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11. Gaussian (Normal) Distribution

$$f(x|\mu, \sigma^2) = \frac{1}{\sigma\sqrt{2\pi}}e^{-\frac{1}{2}(\frac{(x-\mu)^2}{\sigma^2})}$$
$$X \sim N(\mu, \sigma^2)$$



12. OTHER DISTRIBUTIONS

► Chi-Squared Distribution

$$X = \sum_{i=1}^{n} Z_i^2 \sim \mathcal{X}_n^2$$

► T Distribution

$$T = \frac{Z}{\sqrt{\frac{x}{n}}} \sim t_n$$

► F Distribution

$$F = \frac{X_1/n_1}{X_2/n_2} \sim F_{n_1,n_2}$$

► Logistical Distribution

$$f(x) = \frac{1}{1 + e^{-x}} (1 - \frac{1}{1 + x^{-x}})$$

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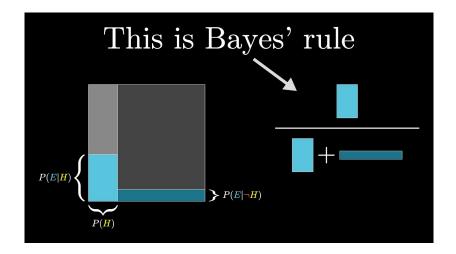
13. BAYES RULE

- ▶ As compared to a frequentest approach (what we have been discussing), Bayesian probabilities condition the frequency of an event occurring on the prior information known about the other events occurring.
- \blacktriangleright For two events A and B:

$$Pr(B|A) = \frac{Pr(B) \times Pr(A|B)}{Pr(A)}$$

$$Pr(B|A) = \frac{Pr(B) \times Pr(A|B)}{Pr(A|B) \times Pr(B) + Pr(A|B^c) \times Pr(B^c)}$$

13. BAYES RULE



DEMONSTRATION: BAYES RULE

Ouestion:

A patient's probability of having a disease is P(A) = 0.1. The patient's probability of having an underlying condition is P(B) = 0.05. Being informed, you know P(B|A) = 0.07 is the probability the patient has the condition given the disease. What is the risk a patient has the disease given they have the condition P(A|B)?

Answer

$$P(A|B) = \frac{P(B|A)P(A)}{P(B)} = \frac{0.07 \times 0.1}{0.05} = 0.14$$

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Answer:

$$P(A|B) = \frac{P(B|A)P(A)}{P(B)} = \frac{0.07 \times 0.1}{0.05} = 0.14$$

PRACTICE: BAYES RULE

1. 1% of women over 50 have breast cancer. 90% of women with breast cancer had a positive test (e.g., mammogram). 8% of all women get a false positive test (i.e., positive when there is **no** cancer). What is the probability of a woman having cancer if she has a positive mammogram test?

PRACTICE: BAYES RULE

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Answer: Show Work

$$Pr(A|B) \approx 8.65\%$$

PRACTICE: BAYES RULE

- 1. Female cancer rates.
- 2. 50% of email is spam. A spam filter can detect 99% of spam. The probability of a false positive (i.e., not spam) is 5%. What is the probability an email in your spam folder is not spam?

PRACTICE: BAYES RULE

- 1. Female cancer rates.
- 2. 50% of email is spam. A spam filter can detect 99% of spam. The probability of a false positive (i.e., not spam) is 5%. What is the probability an email in your spam folder is not spam?

Answer: Show Work

$$P(B^c|A) = \frac{5}{104}$$

▶ 1st Moment: Mean or average occurrence.

$$\bar{X} = \mu_X = \frac{1}{n} \sum_{i=1}^n X_i$$

▶ 2nd Moment: Variance or spread of distribution $Var(X) = \sigma_x = \mathbb{E}[(X - \mu_x)^2]$

▶ 3rd Moment: Skewness or lack of symmetry in the distribution

$$Skew(X) = \frac{\mu_x^3}{\sigma_x^3} = \mathbb{E}[(X - \mu_x)^3]$$

▶ 4th Moment: Kurtosis or relative weight (fatness) of the tails of the distribution

$$Kurt(X) = \frac{\mu_x^4}{\sigma_x^4} - 3 = \mathbb{E}[(X - \mu_x)^4]$$

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15. Variance & Standard Deviation

▶ Variance: The spread of the distribution. How typical is a value x given the mean \bar{X} .

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

► If the distribution is independently and identically distributed (I.I.D.)

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

► In a sample, variance is adjusted by sample size:

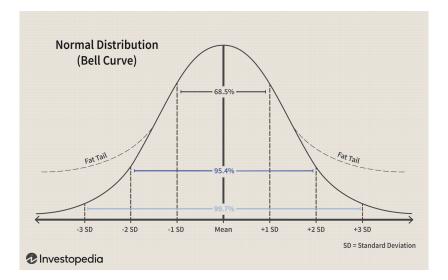
$$s^{2} = \frac{1}{n-1} \sum_{i=1}^{n} (X_{i} - \bar{X})^{2}$$

▶ Standard Deviation: The typical deviation from the mean.

$$sd(X) = \sigma_x = +\sqrt{Var(X)} \equiv +\sqrt{s^2}$$

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15. Variance & Standard Deviation



16. COVARIANCE

► Indicator for degree two variables move together (co-vary) as either variable moves about its own distribution.

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

17. CORRELATION

- ► Indicator for degree one variables move due to changes in another variable (e.g., Not their combined variance)
- ▶ -1 < Corr(X, Y) < 1
- ► Independence: Corr(X, Y) = 0

$$Corr(X, Y) = \frac{Cov(X, Y)}{sd(X) \times sd(Y)} = \frac{\sigma_{XY}}{\sigma_X \sigma_Y}$$





Review

REVIEW: PROBABILITY

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ASSIGNMENT

- ▶ Readings on Statistics before Lecture 12:
 - ► Hansen Metrics Ch. 6, 7, & 8
- ► Assignment:
 - ► Problem Set 11 (PS11)
 - ► Solution set will be available following end of Lecture 11
- ► Struggling?
 - 1. Read the 'Encouraged Reading'
 - 2. Review 'Supplementary material'
 - 3. Reach out directly

BAYES RULE QUESTION 1 ANSWER:

◆ QUESTION

$$Pr(B) = 0.01$$

$$Pr(B^{c}) = 0.99$$

$$Pr(A|B) = 0.9$$

$$Pr(A|B^{c}) = 0.096$$

$$Pr(B|A) = \frac{0.9 \cdot 0.01}{0.9 \cdot 0.01 + 0.096 \cdot 0.99}$$

BAYES RULE QUESTION 2 ANSWER:

◆ QUESTION

 $A = \text{spam detected. } B = \text{email is spam. } B^c = \text{email is not spam.}$

$$Pr(B) = P(B^{c}) = 0.5$$

$$Pr(A|B) = 0.99$$

$$Pr(A|B^{c}) = 0.05$$

$$Pr(B^{c}|A) = \frac{Pr(A|B^{c})Pr(B^{c})}{Pr(A|B)P(B) + Pr(A|B^{c})Pr(B^{c})}$$

$$= \frac{0.05 \cdot 0.5}{0.99 \cdot 0.5 + 0.05 \cdot 0.5}$$