### Bush 631-607: Quantitative Methods

Lecture 9 (10.26.2021): Probability vol. I

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Fall 2021

# What is today's plan?

- Calculating uncertainty: probability
- What is probability? why should we learn it?
- ▶ Probability theory (some equations. . . )
- How we use probability in the real world?
- R work: prop.table(), addmargins(),

### Learning from data

#### Our 8-week quest:

- How to estimate causal effects.
- Understand measurement challenges.
- Assess correlations.
- Generate prediction about unknown quantities.

The question now?

How do we know our estimates are 'real' or just due to random chance?

# We have findings!!!

- ▶ Data patters are systematic? Or noise?
- Our estimates → real relationship or random?

#### Solutions:

- Select (at random) a different treatment/control group.
- Select (at random) a different sample.
- Method to quantify the degree of statistical uncertainty of empirical findings.

# **Probability**



### Intro to probabilities

#### PROBABILITY:

- Set of tools to measure uncertainty in world (and our data).
- Method to formalize uncertainty or chance variation.
- Define odds for all (defined) possible outcomes.

#### What's the chance?

#### January 28, 1986: Challenger shuttle







#### Probabilites translated

#### Challenger accident - chance of failure?

- ► Experts: 100-1.
- ▶ NASA management: 100,000-1.
- ▶ What is 100,000 in 1?
- Repeated testing and odds of event (failure).
- ▶ Enough events? we can calculate probabilities. . .

## Probability explained

- ▶ Probability → measure randomness.
- ▶ Random ≠ complete unpredictability:
  - Short-term: unpredictable (very hard to calculate).
  - ► Long-term: predictable (multiple repetitions).

# Probability explained

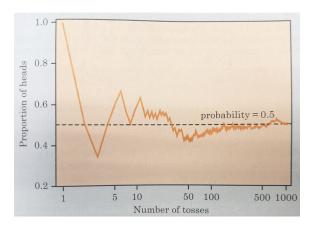


- ► Odds for heads? and tails?
- ▶ Overall: 0.5 probability H/T.

### Coin toss chances

▶ 5 flips: HHHHT

► How 0.5 exactly?



The secret?

#### Repetition - multiple iterations

- Estimate probability.
- ▶ Why only estimate? "toss again..."
- Mathematical probability ideal in infinite series of trails.
- Explain long-term regularity of random event (behavior).

Why?







Law of Averages: When?

#### Rare event and our behavior

TABLE 1.1 How Dangerous Is Terrorism?

Cause of Death	Times more likely to kill an American compared to a terrorist attack	
Heart disease	35,079	
Cancer	33,842	
Alcohol-related death	4,706	
Car accident	1,048	
Risky sexual behavior	452	
Fall	353	
Starvation	187	
Drowning	87	
Railway accident	13	
Accidental suffocation in bed	12	
Lethal force by a law enforcement officer	8	
Accidental electrocution	8	
Hot weather	6	

	% Critical threat	% Important but not critical threat
International terrorism	79	18
Development of nuclear weapons by Iran	75	18

Solve this:

Linda is 31 years old, single, outspoken, and very bright. She majored in philosophy. As a student, she was deeply concerned with issues of discrimination and social justice, and also participated in anti-nuclear demonstrations.

## Schools of thought

#### Frequentist

- ▶ The *limit* of relative frequency.
- Ratio of number of events occur and total number of trails.
- Challenge: same conditions??

#### Bayesian

- Measure of subjective belief about an event occurring.
- Challenge: how to conduct science?

### Probability theory

#### Concepts, axions and definitions

- ▶ Sample space  $(\Omega)$ : set of all possible outcomes.
- Event: any subset of outcomes in sample space.
- ► Card deck: 52 cards (13 rank) x (4 suits)
- Trial: pick a card at-random

## Probability

Calculate probability of event:

$$P(A) = \frac{Elements(A)}{Elements(\Omega)}$$

Example:  $coin toss \times 3$ 

Sample space  $(\Omega)$ : {HHH,HHT,HTH,HTT,THH.THT,TTH,TTT}.

Get an least two heads?

Event A: {HHH,HHT,HTH,THH}.

Probability:  $P(A) = \frac{4}{8} = 0.5$ 

## Probability

- Define how likely/unlikely events are.
- Based on three axioms:
  - 1. Probability of any event A is nonnegative (P(A) >= 0).
  - 2. Normalization  $(P(\Omega) = 1)$ .
  - 3. Addition rule If events A and B are mutually exclusive then P(AorB) = P(A) + P(B)
- Axioms  $1\&2 \rightarrow 1 > P(event) > 0$

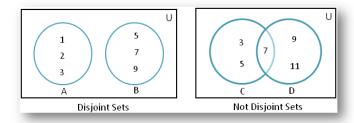
## Gambling 101

#### Probability of mutually exclusive events

- ▶ What is  $P(A) \rightarrow select Queen card at-random?$
- Any card selection: 1/52.
- ▶ Select queen event:  $\{Q\clubsuit, Q\diamondsuit, Q\heartsuit, Q\spadesuit\}$ .
- ▶ P (event) = union of mutually exclusive events  $\rightarrow$  addition rule
- $P(Q) = P(Q \clubsuit) + P(Q \diamondsuit) + P(Q \heartsuit) + P(Q \spadesuit) = \frac{4}{52} \approx 7.7\%$

# Events relationships

#### Mutually & not Mutally exclusive events

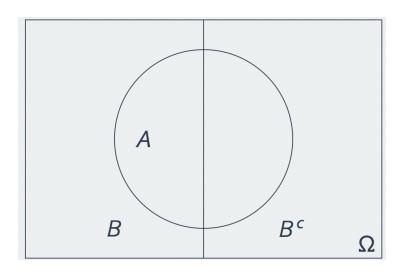


## Probability facts

- ▶ Probability of complement:  $P(A^C) = P(notA) = 1 P(A)$
- ▶ Probability of **not drawing** a Queen:  $1 \frac{4}{52} = \frac{48}{52}$
- Probability of events (not disjointed): the presidential race with 3rd candidate.
- ▶ General addition rule: P(AorB) = P(A) + P(B) P(A&B)
- ▶ Cards example: probability of Queen or ♣?
- ▶ Queen  $(\frac{4}{52}) + \clubsuit (\frac{13}{52}) Q \clubsuit (\frac{1}{52}) = \frac{16}{52}$

## Probability facts

▶ Law of total probability:  $P(A) = P(A\&B) + P(A\&B^C)$ 



## Calculating outcomes

- ▶ **Permutations**: enumerating all possible outcomes.
- Ordering three events (A/B/C): {ABC,ACB,BAC,BCA,CAB,CBA}.
- A short-cut??

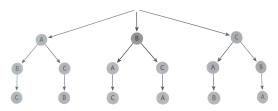


Figure 6.3. A Tree Diagram for Permutations. There are 6 ways to arrange 3 unique objects. Source: Adapted from example by Madit, http://texample.net.

## Calculating outcomes

General permutation formula:

$$_{n}P_{k} = n*(n-1)*...*(n-k+2)*(n-k+1) = \frac{n!}{(n-k)!}$$

How many ways to sit 6 students in our class?

```
# Use permutations formula
factorial(20)/factorial(14)
```

```
## [1] 27907200
```

#### Permutations

- ► The birthday problem:
  - ▶ What n so P(two people share birthday) > 0.5?
  - Easier route by looking at complement.
  - ▶ Find  $\rightarrow$  1 P(nobody has the same birthday).

```
bday <- function(k) {
  logdenom <- k * log(365) + lfactorial(365-k)
  lognumer <- lfactorial(365)
  pr <- 1 - exp(lognumer - logdenom)
  return(pr)
}
k <- 1:20
test_bday <- bday(k)
names(test_bday) <- k

test_bday[15:20]</pre>
```

```
## 15 16 17 18 19 20
## 0.2529013 0.2836040 0.3150077 0.3469114 0.3791185 0.4114384
```

### Sampling procedures

#### With replacement:

- Same unit can be 'selected' repeatedly.
- ► Replace card in stack after draw.
- Two people born on the same day.

#### Without replacement:

- Each unit can be sampled at most once.
- Card removed after draw.
- Procedure matters for probability calculations.
- ► Combinations: another counting method (ignore ordering).

### And...



### And...

#### Probabilities and the real-world





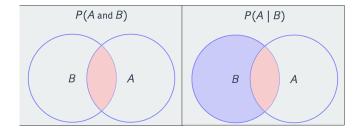
# Using probability in bars?

- ▶ Setting: 5 men, 5 women.
- Objective: get a dance.
- ▶ All go for blonde  $\rightarrow$  P(dance) =  $\frac{1}{4}$
- ▶ Each man  $\rightarrow$  non-blonde:  $P(dance) = \frac{1}{1}$
- ▶ Nash equilibrium: no incentive to deviate.
- Mutual cooperation: global trade, negotiations (prisoner's dilemma).

## Conditional probability

- ▶ We know event B occured, what is the probability of event A?
- Examples:
  - What is probability of two states going to war if they are both democracies?
  - What is the probability of a judge ruling in a pro-choice direction conditional on having daughters?
  - What is the probability that there will be a coup in a country conditional on having a presidential system?

# Conditional probability



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

# Conditional probability

- Conditioning information matters!
- ► Twins:
  - Sample space:  $\Omega = \{GG,GB,BG,BB\}.$
  - ▶  $P(BB \mid at least one boy) = P(BB \mid elder is a boy)??$

$$\begin{split} & \text{P(BB | at least one boy)} = \frac{P(BB\&(BB|BG|GB))}{P(BB|BG|GB)} = \frac{P(BB)}{P(BB|BG|GB)} = \frac{1/4}{3/4} = \frac{1}{3} \\ & \text{P(BB | elder is a boy)} = \frac{P(BB\&(BB|BG))}{P(BB|BG)} = \frac{P(BB)}{P(BB|BG)} = \frac{1/4}{1/2} = \frac{1}{2} \end{split}$$

## Conditioning info: numbers and Aggs

#### Aggies in the NFL: position groups and conferences

#### head(Ags)

```
## # A tibble: 6 x 5
##
    Player
                     Team
                                       Position Group Conference
##
    <chr>>
                     <chr>
                                       <chr>
                                                <chr> <chr>
## 1 Christian Kirk Arizona Cardinals WR.
                                                ΩF
                                                      NFC
## 2 Jake Matthews Atlanta Falcons
                                       OT
                                                OF
                                                      NFC
## 3 Otaro Alaka Baltimore Rayens LB
                                                      AFC
                                                DF
## 4 Justin Madubuike Baltimore Ravens DT
                                                      AFC
                                                DF
## 5 Tyrel Dodson
                     Buffalo Bills
                                       LB
                                                DF
                                                      AFC
## 6 Germain Ifedi
                     Chicago Bears
                                       ΠG
                                                ΟF
                                                      NFC
```

## Conditioning info: numbers and Aggs

```
# Tabulate data
t <- table(Conf = Ags$Conference, Pos.Grp = Ags$Group)
addmargins(t)</pre>
```

```
## Pos.Grp
## Conf DF OF ST Sum
## AFC 8 10 2 20
## NFC 4 12 2 18
## Sum 12 22 4 38
```

- Choose one at-random.
- ▶ What is probability of choosing Offense?

▶ 
$$P(OF) = \frac{22}{38} = 0.57$$

- What is probability of choosing Offense & NFC?
  - ▶ P(OF & NFC) =  $\frac{12}{38}$  = 0.31
- What is probability that randomly selected NFC is offense?

► 
$$P(OF \mid NFC) = \frac{P(OF \& NFC)}{P(NFC)} = \frac{12/38}{18/38} = 0.66$$

# Conditional probability in Global affairs

Military alliances: a contract





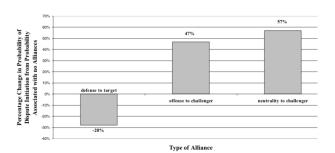




### Global military alliances

### Leeds (2003):

- Defensive cooperation.
- Offensive cooperation.
- Neutrality.
- Non-aggression.
- Consultation.



### Probability and data

## # A tibble: 6 x 24

Military Alliances (ATOP) data (1815-2018)

```
atopid member yrent moent ineffect estmode pubsecr secrart length
##
##
     <dbl>
            <dbl> <dbl> <dbl>
                                 <dbl>
                                         <dbl>
                                                 <dbl>
                                                         <dbl>
                                                                <dbl>
## 1
      3150
               58 1981
                           12
## 2
      3900
               58 1981
                            6
      4778
## 3
               58 1996
                            3
      2075
              700 1921
                            3
## 4
## 5
      2090
              700 1921
                            6
## 6
      2170
              700 1926
                            8
                                                                   36
     ... with 14 more variables: offense <dbl>, neutral <dbl>, nonagg <
## #
      consul <dbl>, active <dbl>, notaiden <dbl>, terrres <dbl>, spect
      milaid <dbl>, base <dbl>, armred <dbl>, ecaid <dbl>, StateAbb <c
## #
      StateName <chr>
## #
```

### Probability in R

► Alliance & domestic ratification

```
# Probabilities for domestic ratification
prop.table(table(Ratification = atop2$estmode))

## Ratification
## 0 1
## 0.2187919 0.7812081

# Probabilities for secret provisions
prop.table(table(publicity = atop2$pubsecr))

## publicity
## 0 1 2
## 0.92557828 0.01709688 0.05732484
```

### Probability in R

- ► Alliance → commitment.
- ▶ US guarantee military assistance?

```
# Subset data (tidyverse): US alliances only
atop.us <- atop2 %>%
  filter(member == 2)

# Probability of military commitment
prop.table(table(atop.us$defense))
```

Conditional probability

```
## Types of military aid given that alliance has defensive provision
prop.table(table(atop2$milaid[atop2$defense == 1]))
##
```

```
## 0 1 2 3 4
## 0.81632653 0.03755102 0.01551020 0.11183673 0.01877551
```

### Probability in R

- Joint probability tables
- Marginal probabilities → sum of rows/columns

```
# Defense and Offense provisions
j1 <- prop.table(table(def = atop2$defense, off = atop2$offense))</pre>
addmargins(j1)
        off
## def
                                       Sum
   0 0.56569709 0.01738549 0.58308258
         0.33132732 0.08559010 0.41691742
    Sum 0.89702441 0.10297559 1.00000000
# Offensive and secret provisions
j2 <- prop.table(table(secret = atop2$secrart, off = atop2$offense))</pre>
addmargins(j2)
##
         off
## secret
##
          0.849480389 0.076097888 0.925578277
          0.003687563 0.000000000 0.003687563
##
         0.003687563 0.000000000 0.003687563
##
          0.004022796 0.001005699 0.005028495
##
         0.001005699 0.000000000 0.001005699
         0.000000000 0.001005699 0.001005699
##
          0.002681864 0.000000000 0.002681864
##
          0.034193765 0.023131076 0.057324841
##
      Sum 0.898759638 0.101240362 1.000000000
```

### Independence

- Events are not related.
- Knowing the A occurred does not affect the probability of B occurring.
- Marginal probability of B (knowing A occurred) remains P(B).
- ► Formally:
  - ► P(A&B) = P(A) \* P(B)
  - P(A|B) = P(A)
  - P(B|A) = P(B)

### Independence in ATOP data

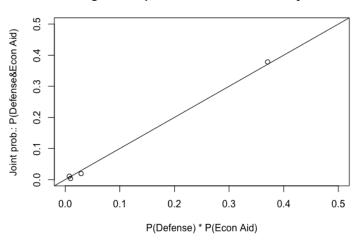
▶ Defense treaties & Economic aid: related?

```
# Marginal probability: levels of economic aid
prop.table(table(EconAid = atop2$ecaid))
## EconAid
##
## 0.88870037 0.01798439 0.02341364 0.06990159
# Marginal probability: defense alliance
prop.table(table(Defense = atop2$defense))
## Defense
##
## 0.5830826 0.4169174
# Joint probability: defense and econ aid
prop.table(table(Defense = atop2$defense, EconAid = atop2$ecaid))
##
          EconAid
## Defense
         0 0.510349508 0.007125891 0.019681032 0.050559891
##
         1 0.378350865 0.010858500 0.003732609 0.019341703
##
```

### Plotting independence

Defense treaties & Economic aid

#### Checking for independence of events: Military Alliances



### Independence

- ▶ Throw conditional probability into the mix.
- ► The Monty Hall problem:



## Bayesian probability

- ▶ The subjective side of probability estimates.
- How prior knowledge and new evidence shape our behavior?
- Bayes rule: mathematical solution to update our beliefs.

$$P(A|B) = \frac{P(B|A)*P(A)}{P(B)} = \frac{P(B|A)*P(A)}{P(B|A)*P(A) + P(B|A^C)*P(A^C)}$$

- P(A): prior probability.
- Event B occur.
- ► P(A|B) = posterior probability

## Bayes in real life

- ▶ Where is the phone?
- ► Health diagnosis.
- Monetary policy.
- ▶ Insurance premiums and hazard events.

### Bayes and the British code breakers

#### Alan Touring and Enigma Machine





- Near-infinite potential code translations.
- ightharpoonup Solutions ightarrow previous encrypted messages.
- $\blacktriangleright$  U-Boats  $\rightarrow$  weather and shipping phrases.

## Wrapping up week 9

### Summary:

- Probability: tool to measure uncertainty in events.
- ▶ What is it good for?
- Conditional probability: importance of information.
- Independence of events.
- Bayesian reasoning.

#### Final project

### Final research project

- Policy brief document: FP situation or policy.
- ▶ Define the issue/situation with evidence: capture the audience!
- Use data analysis to support evaluation and/or recommendations.
- ▶ Data: use **at least one** of 6 proposed data sets.
- Writing and development:
  - ▶ 5Ps (people, purpose, problem, product and process).
  - BLUF (Bottom-Line-Up-Front).
  - More on course website.
- Draft version for feedback by November 23, 2021.