

AMS 7 - Lecture 4.26.18



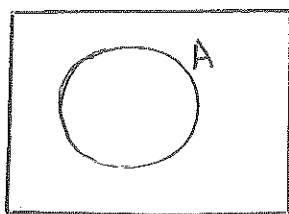
▷ This Time: Conditional Probability

▷ Next Time: Prob. Models for sums

* Read DD ch. 1-3 (A) & DD ch. 1-9 (B)

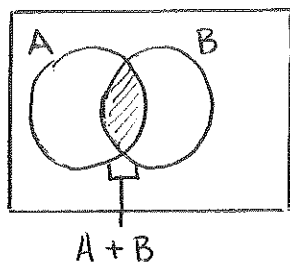
Today → LN pg. 95-118

Next Time → LN pg. 119-126



• What's the probability of A?

$$P(A) = \frac{\text{circle A}}{\text{square}} \leftarrow 100\%$$



• What's the probability of A given B?

$$P(A|B) = \frac{\text{shaded overlap}}{\text{circle B}}$$

← overlap of A & B

"given" ←

• DEFINITION (A. de Moivre) - 1705

$$\hookrightarrow P(A|B) = \begin{cases} P(A \text{ and } B) / P(B) & \text{if } P(B) > 0 \\ \text{undefined} & \end{cases}$$

R-37 $P(A|B) = \frac{P(A \text{ and } B)}{P(B)} \rightarrow P(B) \cdot P(A|B) = P(A \text{ and } B)$

Prob. Chain rule

$$\hookrightarrow P(B|A) = \frac{P(A \text{ and } B)}{P(A)} \rightarrow P(A) \cdot P(B|A) = P(A \text{ and } B)$$

* $P(y_1 = 2 \text{ and } y_2 = 2) \stackrel{\text{SRS}}{=} P(y_1 = 2) \cdot P(y_2 = 2 | y_1 = 2)$

$$\frac{1}{3} \cdot 0 = 0 \checkmark$$



$$* P(y_1 = 2 \text{ and } y_2 = 2) \stackrel{\text{IID}}{=} P(y_1 = 2) \cdot P(y_2 = 2 | y_1 = 2) \\ = P(y_1 = 2) \cdot P(y_2 = 2)$$

• DEFINITION

↳ A & B are independent if and only if information about A does not change chances of B & vice versa: $P(A \& B) \stackrel{\text{indep}}{=} P(A) \cdot P(B)$

* Tay - Sachs Case Study

$$\rightarrow P(\text{1 or more T-S babies in family of 5, both parents carriers}) \\ = 1 - P(0 \text{ T-S babies}) \\ = 1 - P(\text{1st baby NOT T-S} \text{ AND } \text{2nd baby NOT T-S} \text{ AND } \dots \text{ AND } \text{5th baby NOT T-S})$$

∴ by independence:

$$= 1 - P(\text{NOT T-S on 1st}) \cdot P(\text{NOT T-S on 2nd}) \cdot \dots \cdot P(\text{NOT T-S on 5th}) \\ = 1 - (1 - \frac{1}{4}) \cdot (1 - \frac{1}{4}) \cdot \dots \cdot (1 - \frac{1}{4}) \\ = 1 - (1 - \frac{1}{4})^5 = 0.76 = 76\%$$

2-37 * UCLA Marijuana Case Study

Y	X
0	1
1	0
1	0
⋮	⋮
⋮	⋮
⋮	⋮
1	1

MLP
n=106

Y = outcome (yes or no)
X = Predictor (Female or male)

SORT

Female (1), Male (0)
Yes (1), No (0)

Y	X	
0	0	} 5
0	0	
0	0	
⋮	⋮	} 20
0	1	
0	1	
1	0	} 52
1	0	
1	0	
⋮	⋮	} 29
1	1	
1	1	

MLP

▷ 2x2 contingency table
→ categorical data analysis

	Yes	No	
Female	29	20	29+20=49
Male	52	5	52+5=57
	81	25	n=106



MIDDLE TABLE

$$P(DP | VW) = \frac{30}{214} \doteq 14.0\%$$

white
victim

$$P(DP | DW \text{ AND } VW) = \frac{19}{151} \doteq 12.6\%$$

$$P(DP | DB \text{ AND } VW) = \frac{11}{63} \doteq 17.5\%$$

BOTTOM TABLE

$$P(DP | VB) = \frac{6}{112} \doteq 5.4\%$$

black
victim

$$P(DP | VB \text{ AND } DW) = \frac{0}{9} \doteq 0\%$$

$$P(DP | VB, DB) = \frac{6}{103} \doteq 5.8\%$$

the direction of relationship
between X & Y changes
when Z is accounted for
*Simpson's Paradox
(1950)

▷ Why did the Simpson's Paradox arise?

↳ Victim usually knows killer

↳ White people mostly hang around white people & black
with black, therefore, whites mostly kills whites &
blacks kill blacks but if victim is white, much more
likely to get death penalty, therefore, it will look like
white defendants get DP more than "they really are"

Probability Models for sums & Means

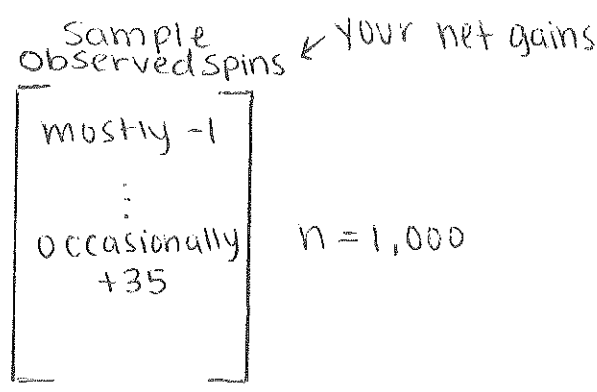
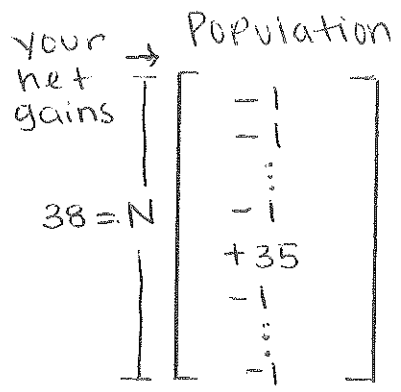
*Roulette (38 possibilities) ELM ✓ & >> IID <<

$$P(\text{Win on a single play, single \#}) = \frac{1}{38} \doteq 2.5\%$$

$$\text{Split} = \frac{2}{38} = 5\%$$

R-52

Possible outcome on a single spin



Single # 6

$$\text{mean } \mu = \frac{37(-1) + (+35)}{38}$$

↑

$$\text{"mean"} = \frac{-2}{38} = -\$0.05$$

↳ negative

σ = sigma

* on average, I expect to win $\mu = -\$0.05$ on each \$1.00 bet for a single #