

Stat 134 Lec 2

Last time

(OR)

Addition rule

if A, B mutually exclusive sets \rightarrow

$$P(A \text{ or } B) = P(A) + P(B).$$

(OR)

Inclusion-exclusion

$$P(A \text{ or } B) = P(A) + P(B) - P(AB)$$

Today

① sec 1.3 Distributions

② sec 1.4 Conditional Probability

③ sec 1.5 Bayes' rule

Please load b-courses on your device

Some easy facts about sets :

$$A_1 \cup A_2 \cup A_3 = (A_1 \cup A_2) \cup A_3 \Rightarrow \bigcup_{i=1}^{m+1} A_i = \left(\bigcup_{i=1}^m A_i \right) \cup A_{m+1}$$

$$A_1 A_3 \cup A_2 A_3 = (A_1 \cup A_2) A_3 \Rightarrow \bigcup_{i=1}^m (A_i A_{m+1}) = \left(\bigcup_{i=1}^m A_i \right) A_{m+1}$$

Sec 1.3 Distribution

Proof by induction

see #12 p31
 \Leftarrow Let A_1, \dots, A_{n+1} be events.

$$\text{Show } P\left(\bigcup_{i=1}^{n+1} A_i\right) = P\left(\bigcup_{i=1}^n A_i\right) + P(A_{n+1}) - P\left(\bigcup_{i=1}^n A_i; A_{n+1}\right)$$

Pf/ we prove by induction.

(1) base case \Rightarrow true, $P(A_1 \cup A_2) = P(A_1) + P(A_2) - P(A_1 A_2)$
 by inclusion exclusion rule.

(2) Assume true for union of n events and
 show true for union of $n+1$ events :

$$\begin{aligned} P\left(\bigcup_{i=1}^{n+1} A_i\right) &\leq P\left(\bigcup_{i=1}^n A_i \cup A_{n+1}\right) && P((A_i) A_{n+1}) \\ &= P\left(\bigcup_{i=1}^n A_i\right) + P(A_{n+1}) - P\left(\bigcup_{i=1}^n A_i; A_{n+1}\right) \\ &\quad \text{by inclusion exclusion rule.} \end{aligned}$$

By the principle of mathematical induction
 we have shown the claim for all n .

□

Named distribution

Can repeat
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Uniform distribution on a finite set  $\{x_1, \dots, x_n\}$ :

Imagine you have numbers  $x_1, \dots, x_n$  in a hat.

Let  $X$  be a random draw of one of these numbers.

$$P(X = x_i) = \frac{1}{n} \text{ for all } i$$

We say that  $X$  has uniform distribution on  $\{x_1, \dots, x_n\}$ .

☞ Suppose a word is randomly picked from this sentence. What is the distribution of the length of the word picked.

answ       $\text{unif}(\{7, 1, 4, 2, 8, 6, 4, 4, 8\})$

# Conc test pt 1

A deck of cards is shuffled. What is the chance that the top card is the **king** of spades **or** the bottom card is the **king** of spades

- a  $\frac{1}{52} \times \frac{1}{51}$
- b  $\frac{1}{52} + \frac{1}{51}$
- c  $\frac{1}{52} + \frac{1}{52} - \frac{1}{52} \times \frac{1}{51}$

(d) none of the above

King of spades and  
King of spades are  
mutually exclusive so  
have inclusion exclusion  
rule.

$$\frac{1}{52} + \frac{1}{52}$$

unconditional probability

## Concert test pt 2

A deck of cards is shuffled. What is the chance that the top card is the **king** of spades **or** the bottom card is the **queen** of spades

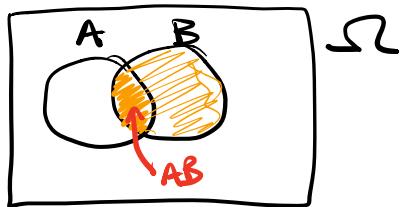
- a  $\frac{1}{52} \times \frac{1}{51}$
- b  $\frac{1}{52} + \frac{1}{51}$
- c  $\frac{1}{52} + \frac{1}{52} - \frac{1}{52} \times \frac{1}{51}$
- d none of the above

King of spades and queen of spades aren't mutually exclusive so have inclusion exclusion rule.

$$\begin{aligned} P(\text{King of spades and queen of spades}) \\ = \frac{1}{52} \cdot \frac{1}{51} \\ \text{by multiplication rule} \end{aligned}$$

## Sec 1.4 Conditional Probability and Independence

Let  $A, B$  be subsets of  $\Omega$  (i.e events).



Baye's rule says  $P(A|B) = \frac{P(AB)}{P(B)}$  given

$$\Leftrightarrow P(AB) = P(A|B)P(B)$$

*A and B*

multiplication rule,  
(A AND)

We say  $A$  and  $B$  are independent if

$$P(A|B) = P(A)$$

or equivalently if  $P(AB) = P(A)P(B)$

ex  $A =$  last card is queen of spades  
 $B =$  1<sup>st</sup> card is king of spades  
 $A$  and  $B$  are dependent

$$P(AB) = P(B)P(A|B) = \left[ \frac{1}{52} \cdot \frac{1}{51} \right]$$

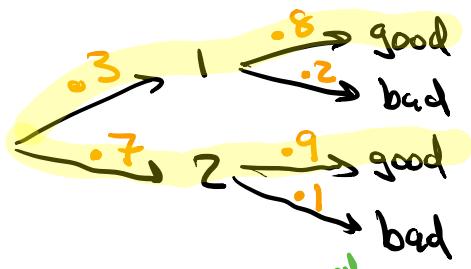
### Sec 1.5 Baye's rule

Ex A factory produces 2 models of cell phones.

Given  $P(1) = .3$

$$P(\text{good} | 1) = .8$$

$$P(\text{good} | 2) = .9$$



Find  $P(1, \text{good})$  *and*  $= (.3)(.8) = .24$

$$P(\text{good}) = (.3)(.8) + (.7)(.9) = .87$$

$$P(1 | \text{good}) = \frac{P(1, \text{good})}{P(\text{good})} = \frac{.24}{.87} = .28$$

### Note

Another way to write Bayes rule

$$P(1 | \text{good}) = \frac{P(1, \text{good})}{P(\text{good})} = \frac{P(\text{good}|1)P(1)}{P(\text{good}|1)P(1) + P(\text{good}|2)P(2)}$$