

# Intersection and conditioning

## Situation:

- Urn with 8 Red and 4 White balls
- Draw 2 balls without replacement

## Question: Let

- $R_1$  = 1st ball drawn is red
- $R_2$  = 2nd ball drawn is red

Then find  $\mathbf{P}(R_1 R_2)$

# Intersection and conditioning (2)

Recall:

- Urn with 8 Red and 4 White balls
- Draw 2 balls without replacement

Computation: We have

$$\mathbf{P}(R_1 R_2) = \mathbf{P}(R_1) \mathbf{P}(R_2 | R_1)$$

Thus

$$\mathbf{P}(R_1 R_2) = \frac{8}{12} \frac{7}{11} = \frac{14}{33} \simeq .42$$

8R, 4W

## Intersection of 3 events

$R_1 = R$  for 1<sup>st</sup> draw

$R_2 = R$  for 2<sup>nd</sup> draw

$W_3 = W$  for 3<sup>rd</sup> draw

Then

$$P(W_3 \cap R_2 \cap R_1)$$

$$= P(R_1) P(R_2 | R_1) P(W_3 | R_1 \cap R_2)$$

$$= \frac{8}{12} \times \frac{7}{11} \times \frac{4}{10} \approx .17$$

# The multiplication rule

For  $n=3$ :  $P(E_1, E_2, E_3) = P(E_1) P(E_2 | E_1) P(E_3 | E_1, E_2)$

## Proposition 2.

Let

- $\mathbf{P}$  a probability on a sample space  $S$
- $E_1, \dots, E_n$   $n$  events

Then

$$\mathbf{P}(E_1 \cdots E_n) = \mathbf{P}(E_1) \prod_{k=1}^{n-1} \mathbf{P}(E_{k+1} | E_1 \cdots E_k) \quad (2)$$

# Proof

Expression for the rhs of (2):

$$\mathbf{P}(E_1) \frac{\mathbf{P}(E_1 E_2)}{\mathbf{P}(E_1)} \frac{\mathbf{P}(E_1 E_2 E_3)}{\mathbf{P}(E_1 E_2)} \cdots \frac{\mathbf{P}(E_1 \cdots E_{n-1} E_n)}{\mathbf{P}(E_1 \cdots E_{n-1})}$$

Conclusion:

By telescopic simplification

# Example: deck of cards (1)

## Situation:

- Ordinary deck of 52 cards
- Division into 4 piles of 13 cards

## Question: If

$$E = \{\text{each pile has one ace}\},$$

compute  $\mathbf{P}(E)$

## Example: deck of cards (2)

Model: Set

$E_1 = \{\text{the ace of S is in any one of the piles}\}$

$E_2 = \{\text{the ace of S and the ace of H are in different piles}\}$

$E_3 = \{\text{the aces of S, H \& D are all in different piles}\}$

$E_4 = \{\text{all 4 aces are in different piles}\}$

We wish to compute

$$\mathbf{P}(E_1 E_2 E_3 E_4)$$

## Example: deck of cards (3)

Applying the multiplication rule: write

$$\mathbf{P}(E_1 E_2 E_3 E_4) = \mathbf{P}(E_1) \mathbf{P}(E_2 | E_1) \mathbf{P}(E_3 | E_1 E_2) \mathbf{P}(E_4 | E_1 E_2 E_3)$$

Computation of  $\mathbf{P}(E_1)$ : Trivially

$$\mathbf{P}(E_1) = 1$$

Computation of  $\mathbf{P}(E_2 | E_1)$ : Given  $E_1$ ,

- Reduced space is  
    {51 labels given to all cards except for ace S}
- $\mathbf{P}(E_2 | E_1) = \frac{51-12}{51} = \frac{39}{51}$



## Example: deck of cards (4)

Other conditioned probabilities:

$$\begin{aligned}\mathbf{P}(E_3 | E_1 E_2) &= \frac{50 - 24}{50} = \frac{26}{50}, \\ \mathbf{P}(E_4 | E_1 E_2 E_3) &= \frac{49 - 36}{49} = \frac{13}{49}\end{aligned}$$

**Conclusion:** We get

$$\begin{aligned}\mathbf{P}(E) &= \mathbf{P}(E_1) \mathbf{P}(E_2 | E_1) \mathbf{P}(E_3 | E_1 E_2) \mathbf{P}(E_4 | E_1 E_2 E_3) \\ &= \frac{39 \cdot 26 \cdot 13}{51 \cdot 50 \cdot 49} \simeq .105\end{aligned}$$

# Outline

- 1 Introduction
- 2 Conditional probabilities
- 3 Bayes' formula**
- 4 Independent events
- 5 Conditional probability as a probability

# Thomas Bayes

## Some facts about Bayes:

- England, 1701-1760
- Presbyterian minister
- Philosopher and statistician
- Wrote 2 books in entire life
- Bayes formula unpublished



# Decomposition of $\mathbf{P}(E)$

## Proposition 3.

Let

- $\mathbf{P}$  a probability on a sample space  $S$
- $E, F$  two events with  $0 < \mathbf{P}(F) < 1$

Then

$$\mathbf{P}(E) = \mathbf{P}(E|F)\mathbf{P}(F) + \mathbf{P}(E|F^c)\mathbf{P}(F^c)$$

Situation : we know

$P(E|F)$  → we also need  
 $P(E|F^c), P(F)$

Can we compute

$P(F|E)$  ?

# Bayes' formula

## Proposition 4.

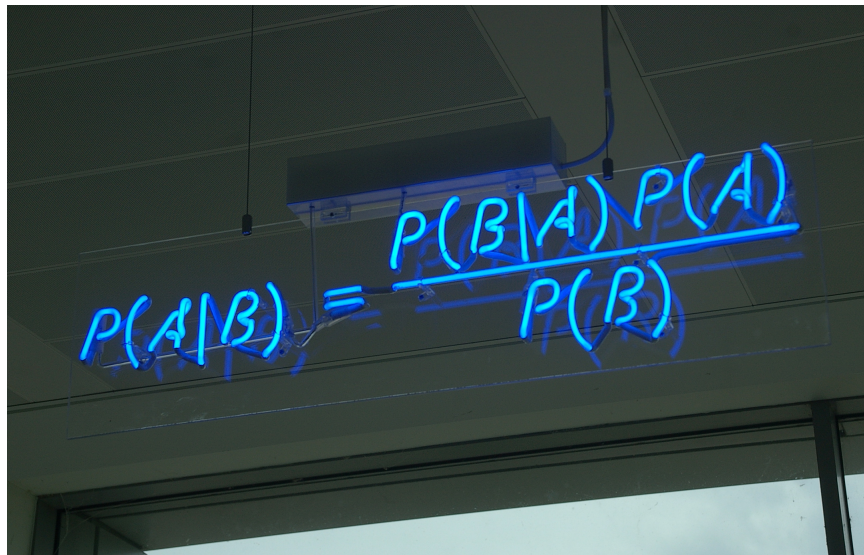
Let

- $\mathbf{P}$  a probability on a sample space  $S$
- $E, F$  two events with  $0 < \mathbf{P}(F) < 1$

Then

$$\mathbf{P}(F|E) = \frac{\mathbf{P}(E|F)\mathbf{P}(F)}{\mathbf{P}(E|F)\mathbf{P}(F) + \mathbf{P}(E|F^c)\mathbf{P}(F^c)}$$

# Iconic Bayes (offices of HP Autonomy)


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

# Example: insurance company (1)

## Situation:

- Two classes of people:  
those who are accident prone and those who are not.
- Accident prone: probability .4 of accident in a one-year period
- Not accident prone: probab .2 of accident in a one-year period
- 30% of population is accident prone

## Question:

Probability that a new policyholder will have an accident within a year of purchasing a policy?



$A =$  accident prone     $A_1 =$  accident in 1 year  
Insurance example

Sample space: for our experiment

$$S = \{ (A, A_1), (A, A_1^c), (A^c, A_1), (A^c, A_1^c) \}$$

Probab.: defined through conditioning.

$A$ : accident prone       $A_1$ : accident in 1 year

Data:  $P(A_1 | A) = .4$

$$P(A_1 | A^c) = .2$$

$$P(A) = .3$$

$$P(A^c) = .7$$

We compute

$$P(A_1) = P(A_1 | A) P(A) + P(A_1 | A^c) P(A^c)$$

$$= .4 \times .3 + .2 \times .7$$

$$= 26\%$$

## Example: insurance company (2)

Model: Define

- $A_1$  = Policy holder has an accident in 1 year
- $A$  = Accident prone

Then

- $S = \{(A_1, A); (A_1^c, A); (A_1, A^c); (A_1^c, A^c)\}$
- Probability: given indirectly by conditioning

Aim:

Compute  $\mathbf{P}(A_1)$

## Example: insurance company (3)

Given data:

$$\mathbf{P}(A_1|A) = .4, \quad \mathbf{P}(A_1|A^c) = .2, \quad \mathbf{P}(A) = .3$$

Application of Proposition 3:

$$\mathbf{P}(A_1) = \mathbf{P}(A_1|A) \mathbf{P}(A) + \mathbf{P}(A_1|A^c) \mathbf{P}(A^c)$$

We get

$$\mathbf{P}(A_1) = 0.4 \times 0.3 + 0.2 \times 0.7 = 26\%$$

# Example: swine flu (1)

## Situation:

We assume that 20% of a pork population has swine flu.

A test made by a lab gives the following results:

- Among 50 tested porks with flu, 2 are not detected
- Among 30 tested porks without flu, 1 is declared sick

## Question:

Probability that a pork is healthy while his test is positive?

*= probab. to kill the whole population  
without proper justification*

Events:  $F = \text{"Flu"}$      $T = \text{"Test positive"}$

Data:  $P(F) = .2$      $P(F^c) = .8$

$$P(T^c | F) = 2/50 = 1/25$$

$$P(T | F^c) = 1/30$$

We compute

Test is not very good!

$$P(F^c | T)$$

Bayes

$$= \frac{P(T | F^c) P(F^c)}{P(T | F^c) P(F^c) + P(T | F) P(F)}$$

$$= \frac{1/30 \times .8}{1/30 \times .8 + (1 - 1/25) \times .2} = 12\%$$

## Example: swine flu (2)

**Model:** We set  $F = \text{"Flu"}$ ,  $T = \text{"Positive test"}$

We have

$$\mathbf{P}(F) = \frac{1}{5}, \quad \mathbf{P}(T^c | F) = \frac{1}{25}, \quad \mathbf{P}(T | F^c) = \frac{1}{30}$$

**Aim:**

Compute  $\mathbf{P}(F^c | T)$

## Example: swine flu (3)

Application of Proposition 4:

$$\begin{aligned}\mathbf{P}(F^c | T) &= \frac{\mathbf{P}(T | F^c) \mathbf{P}(F^c)}{\mathbf{P}(T | F^c) \mathbf{P}(F^c) + \mathbf{P}(T | F) \mathbf{P}(F)} \\ &= \frac{\mathbf{P}(T | F^c) \mathbf{P}(F^c)}{\mathbf{P}(T | F^c) \mathbf{P}(F^c) + [1 - \mathbf{P}(T^c | F)] \mathbf{P}(F)} \\ &= 0.12\end{aligned}$$

Conclusion:

12% chance of killing swines without proper justification