

# Odds

## Definition 5.

Let

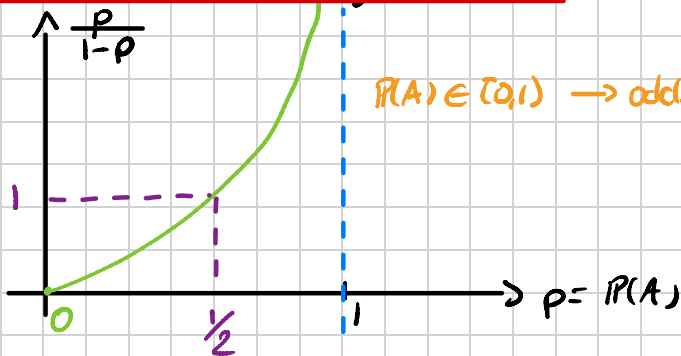
- $\mathbf{P}$  a probability on a sample space  $S$
- $A$  an event

We define the odds of  $A$  by

$$\text{odds}(A) = \frac{\mathbf{P}(A)}{\mathbf{P}(A^c)} = \frac{\mathbf{P}(A)}{1 - \mathbf{P}(A)}$$

$$\text{odds}(A) = \frac{P(A)}{1 - P(A)}$$

## Odds as a change of scale



$$P(A) \in (0,1) \rightarrow \text{odds}(A) \in (0,\infty)$$

Main use: Betting

Ex: On Wednesday  $\text{odds}(\text{Kamala}) = 1.25$

# Odds and conditioning

## Proposition 6.

**Situation:** We have

- An hypothesis  $H$ , true with probability  $\mathbf{P}(H)$
- A new evidence  $E$

**Formula:** The odds of  $H$  after evidence  $E$  are given by

$$\text{odds}(H|E) = \frac{\mathbf{P}(H|E)}{\mathbf{P}(H^c|E)} = \frac{\mathbf{P}(H)}{\mathbf{P}(H^c)} \frac{\mathbf{P}(E|H)}{\mathbf{P}(E|H^c)} = \text{odds}(H) \times \text{odds}(E|H)$$

# Proof

Inversion of conditioning: We have

$$\mathbf{P}(H|E) = \frac{\mathbf{P}(E|H)\mathbf{P}(H)}{\mathbf{P}(E)}$$

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

Conclusion:

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

# Example: coin tossing (1)

## Situation:

- Urn contains two type A coins and one type B coin.
- When a type A coin is flipped, it comes up heads with probability  $\frac{1}{4}$
- When a type B coin is flipped, it comes up heads with probability  $\frac{3}{4}$
- A coin is randomly chosen from the urn and flipped

## Question:

Given that the flip landed on heads

↪ What is the probability that it was a type A coin?

Data :  $P(H|A) = \frac{1}{4}$        $P(A) = \frac{2}{3}$   
 $P(H|B) = \frac{3}{4}$        $P(B) = \frac{1}{3}$

We wish to compute

$P(A|H)$        $<? P(A) = \frac{2}{3} ?$

Bayes

$$= \frac{P(H|A) P(A)}{P(H|A) P(A) + P(H|B) P(B)}$$
$$= \frac{\frac{1}{4} \times \frac{2}{3}}{\frac{1}{4} \times \frac{2}{3} + \frac{3}{4} \times \frac{1}{3}} = \frac{2}{5} < \frac{2}{3}$$

## Example: coin tossing (2)

Model: We set

- $A$  = type A coin flipped
- $B$  = type B coin flipped
- $H$  = Head obtained

Data:

$$\mathbf{P}(A) = \frac{2}{3}, \quad \mathbf{P}(H|A) = \frac{1}{4}, \quad \mathbf{P}(H|B) = \frac{3}{4}$$

Aim:

Compute  $\mathbf{P}(A|H)$

## Example: coin tossing (3)

Application of Proposition 6:  $\text{odds}(A)$   $\text{odds}(H|A)$

$$\text{odds}(A|H) = \frac{P(A|H)}{P(B|H)} = \frac{\overbrace{P(A)}^{\text{odds}(A)}}{\overbrace{P(H|A)}^{\text{odds}(H|A)}} \frac{P(H|B)}{P(H|A)}$$

Numerical result: We get

$$\frac{P(A|H)}{1 - P(A|H)} = \frac{P(A|H)}{P(B|H)} = \frac{2/3}{1/3} \frac{1/4}{3/4} = \frac{2}{3}$$

Therefore

*some algebra*

$$P(A|H) = \frac{2}{5}$$

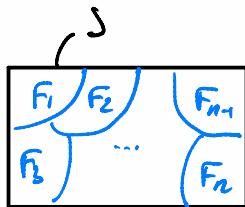


# Generalization of Proposition 3

## Proposition 7.

Let

- $\mathbf{P}$  a probability on a sample space  $S$
- $F_1, \dots, F_n$  partition of  $S$ , i.e.
  - ▶  $F_i$  mutually exclusive
  - ▶  $\bigcup_{i=1}^n F_i = S$
- $E$  another event



Then we have

$n=2$ :  $F_1, F_2 = F_1^c$  partition

$$P(E) = P(E|F_1)P(F_1) + P(E|F_2)P(F_2)$$

$$P(E) = \sum_{i=1}^n P(E|F_i) P(F_i)$$

# Generalization of Proposition 4

## Proposition 8.

Let

- $\mathbf{P}$  a probability on a sample space  $S$
- $F_1, \dots, F_n$  partition of  $S$ , i.e.
  - ▶  $F_i$  mutually exclusive
  - ▶  $\cup_{i=1}^n F_i = S$
- $E$  another event

Then we have

$$\mathbf{P}(F_j | E) = \frac{\mathbf{P}(E | F_j) \mathbf{P}(F_j)}{\sum_{i=1}^n \mathbf{P}(E | F_i) \mathbf{P}(F_i)}$$

# Example: card game (1)

## Situation:

- 3 cards identical in form (say Jack)
- Coloring of the cards on both faces:
  - ▶ 1 card RR
  - ▶ 1 card BB
  - ▶ 1 card RB
- 1 card is randomly selected, with upper side R

## Question:

What is the probability that the other side is B?

## Example: card game (2)

**Model:** We define the events

- RR: chosen card is all red
- BB: chosen card is all black
- RB: chosen card is red and black
- R: upturned side of chosen card is red

**Aim:**

Compute  $\mathbf{P}(RB| R)$

## Example: card game (3)

Application of Proposition 8:

$$\begin{aligned} \mathbf{P}(RB|R) \\ = \frac{\mathbf{P}(R|RB)\mathbf{P}(RB)}{\mathbf{P}(R|RR)\mathbf{P}(RR) + \mathbf{P}(R|RB)\mathbf{P}(RB) + \mathbf{P}(R|BB)\mathbf{P}(BB)} \end{aligned}$$

Numerical values:

$$\mathbf{P}(RB|R) = \frac{\frac{1}{2} \times \frac{1}{3}}{1 \times \frac{1}{3} + \frac{1}{2} \times \frac{1}{3} + 0 \times \frac{1}{3}} = \frac{1}{3}$$

# Example: disposable flashlights

## Situation:

- Bin containing 3 different types of disposable flashlights
- Proba that a type 1 flashlight will give over 100 hours of use is .7
- Corresponding probabilities for types 2 & 3: .4 and .3
- 20% of the flashlights are type 1, 30% are type 2, and 50% are type 3

## Questions:

- 1 What is the probability that a randomly chosen flashlight will give more than 100 hours of use?
- 2 Given that a flashlight lasted over 100 hours, what is the conditional probability that it was a type  $j$  flashlight, for  $j = 1, 2, 3$ ?

A = "more than 100k of use"     $F_j$  = "Type j chosen"

<u>Data</u>	$P(A   F_1) = .7$	$P(F_1) = .2$
	$P(A   F_2) = .4$	$P(F_2) = .3$
	$P(A   F_3) = .3$	$P(F_3) = .5$

We wish to compute

$$P(A)$$

Bayes!

$$= P(A | F_1) P(F_1) + P(A | F_2) P(F_2) + P(A | F_3) P(F_3)$$

$$= .7 \times .2 + .4 \times .3 + .3 \times .5 = 41\%$$

$$\begin{array}{ll}
 P(A|F_1) = .7 & P(F_1) = .2 \\
 P(A|F_2) = .4 & P(F_2) = .3 \\
 P(A|F_3) = .3 & P(F_3) = .5
 \end{array}$$

We wish to compute

$$P(F_1|A) > P(F_1) = .2 ?$$

Bayes 2

$$\begin{aligned}
 &= \text{computed} \leftarrow \frac{P(A|F_1) P(F_1)}{P(A|F_1) P(F_1) + P(A|F_2) P(F_2) + P(A|F_3) P(F_3)} \\
 &= \frac{P(A|F_1) P(F_1)}{P(A)} \\
 &= \frac{.7 \times .2}{.41} = .34 > .2
 \end{aligned}$$



## Example: disposable flashlights (2)

**Model:** We define the events

- $A$ : flashlight chosen gives more than 100h of use
- $F_j$ : type  $j$  is chosen

**Aim 1:**

Compute  $\mathbf{P}(A)$

## Example: disposable flashlights (3)

Application of Proposition 7:

$$\mathbf{P}(A) = \sum_{j=1}^3 \mathbf{P}(A|F_j) \mathbf{P}(F_j)$$

Numerical values:

$$\mathbf{P}(A) = 0.7 \times 0.2 + 0.4 \times 0.3 + 0.3 \times 0.5 = .41$$

## Example: disposable flashlights (4)

Aim 2:

Compute  $\mathbf{P}(F_1|A)$

Application of Proposition 8:

$$\mathbf{P}(F_1|A) = \frac{\mathbf{P}(A|F_1)\mathbf{P}(F_1)}{\mathbf{P}(A)}$$

Numerical value:

$$\mathbf{P}(F_1|A) = \frac{0.7 \times 0.2}{0.41} = \frac{14}{41} \simeq 41\%$$

# Outline

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

# Definition of independence

## Definition 9.

Let

- $\mathbf{P}$  a probability on a sample space  $S$
- $E, F$  two events

Then  $E$  and  $F$  are independent if

$$\mathbf{P}(EF) = \mathbf{P}(E)\mathbf{P}(F)$$

Notation:

$E$  and  $F$  independent denoted by  $E \perp\!\!\!\perp F$

Def  $E \perp F$  if  $P(E \cap F) = P(E)P(F)$

Rmk 1 If  $E \perp F$ , we also have

$$P(E|F) = P(E)$$

Rmk 2

$E \perp F$  is very different from  $E \cap F = \emptyset$