

# Conditional probability and independence

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Probability - MA 416

Mostly taken from *A first course in probability*  
by S. Ross

# Outline

- 1 Introduction
- 2 Conditional probabilities
- 3 Bayes' formula
- 4 Independent events

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# Global objective

**Aim:** Introduce conditional probability, whose interest is twofold

- 1 Quantify the effect of a prior information on probabilities
- 2 If no prior information is available, then independence  
↔ simplification in probability computations

# Outline

1 Introduction

**2 Conditional probabilities**

3 Bayes' formula

4 Independent events

# Example of conditioning

**Dice tossing:** We consider the following situation

- We throw 2 dice
- We look for  $\mathbf{P}$ (sum of 2 faces is 9)

**Without prior information:**

$$\mathbf{P}(\text{sum of 2 faces is 9}) = \frac{1}{9}$$

**Changes with additional information:**

If we know that first face is = 4, then

↪ **how does it affect  $\mathbf{P}$ (sum of 2 faces is 9)?**

# Example of conditioning

**Probability with additional information:** If first face is = 4, then

- Only 6 possible results:

$$(4, 1), (4, 2), (4, 3), (4, 4), (4, 5), (4, 6)$$

- Among them, only (4, 5) gives sum = 9
- Probability of having sum = 9 becomes

$$p = \frac{1}{6}$$

**Conclusion:**

**We need to formalize this type of computation**

# General definition

## Definition 1.

Let

- $\mathbf{P}$  a probability on a sample space  $S$
- $E, F$  two events, such that  $\mathbf{P}(F) > 0$

Then

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



# Example: examination (1)

## Situation:

Student taking a one hour exam

**Hypothesis:** For  $x \in [0, 1]$  we have

$$\mathbf{P}(L_x) = \frac{x}{2}, \quad (1)$$

where the event  $L_x$  is defined by

$$L_x = \{\text{student finishes the exam in less than } x \text{ hour}\}$$

**Question:** Given that the student is still working after .75h  
 $\Leftrightarrow$  Find probability that the full hour is used

## Example: examination (2)

Model: We wish to find

$$\mathbf{P}(L_1^c | L_{.75}^c)$$

Computation: We have

$$\begin{aligned}\mathbf{P}(L_1^c | L_{.75}^c) &= \frac{\mathbf{P}(L_1^c L_{.75}^c)}{\mathbf{P}(L_{.75}^c)} \\ &= \frac{\mathbf{P}(L_1^c)}{\mathbf{P}(L_{.75}^c)} \\ &= \frac{1 - \mathbf{P}(L_1)}{1 - \mathbf{P}(L_{.75})}\end{aligned}$$

Conclusion: Applying (1) we get

$$\mathbf{P}(L_1^c | L_{.75}^c) = .8$$

# Simplification for uniform probabilities

General situation: We assume

- $S = \{s_1, \dots, s_N\}$  finite.
- $\mathbf{P}(\{s_i\}) = \frac{1}{N}$  for all  $1 \leq i \leq N$

Alert:

This is an important but very particular case of probability space

Conditional probabilities in this case:

Reduced sample space, i.e

Conditional on  $F$ , all outcomes in  $F$  are equally likely

# Example: family distribution (1)

## Situation:

The Popescu family has 10 kids

## Questions:

- 1 If we know that 9 kids are girls  
↪ find the probability that all 10 kids are girls

# Example: family distribution (1)

## Situation:

The Popescu family has 10 kids

## Questions:

- 1 If we know that 9 kids are girls  
↪ find the probability that all 10 kids are girls
- 2 If we know that the first 9 kids are girls  
↪ find the probability that all 10 kids are girls

## Example: family distribution (2)

Model:

- $S = \{G, B\}^{10}$
- Uniform probability: for all  $s \in S$ ,

$$\mathbf{P}(\{s\}) = \frac{1}{2^{10}} = \frac{1}{1024}$$

## Example: family distribution (3)

First conditioning: We take

$$F_1 = \{(G, \dots, G); (G, \dots, G, B); (G, \dots, G, B, G); \dots; (B, G, \dots, G)\}$$

Reduced sample space:

Each outcome in  $F_1$  has probability  $\frac{1}{11}$

Conditional probability:

$$\mathbf{P}(\{(G, \dots, G)\} | F_1) = \frac{1}{11}$$

## Example: family distribution (4)

Second conditioning: We take

$$F_2 = \{(G, \dots, G); (G, \dots, G, B)\}$$

Reduced sample space:

Each outcome in  $F_2$  has probability  $\frac{1}{2}$

Conditional probability:

$$\mathbf{P}(\{(G, \dots, G)\} | F_2) = \frac{1}{2}$$



# Example: bridge game (1)

## Bridge game:

- 4 players, E, W, N, S
- 52 cards dealt out equally to players

**Conditioning:** We condition on the set

$$F = \{N + S \text{ have a total of 8 spades}\}$$

**Question:** Conditioned on  $F$ ,  
Probability that E has 3 of the remaining 5 spades

## Example: bridge game (2)

Model: We take

$$S = \{\text{Divisions of 52 cards in 4 groups}\}$$

and we have

- Uniform probability on  $S$
- $|S| = \binom{52}{13,13,13,13} \simeq 5.36 \cdot 10^{28}$

Reduced sample space: Conditioned on  $F$ ,

$$\tilde{S} = \{\text{Combinations of 13 cards among 26 cards with 5 spades}\}$$

and  $|\tilde{S}| = 10,400,600$

## Example: bridge game (3)

Conditional probability:

$$\mathbf{P}(\text{E has 3 of the remaining 5 spades} | F) = \frac{\binom{5}{3} \binom{21}{10}}{\binom{26}{13}} \simeq .339$$

# 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

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

# The multiplication rule

## 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

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

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# 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

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

# Bayes' formula

## Proposition 4.

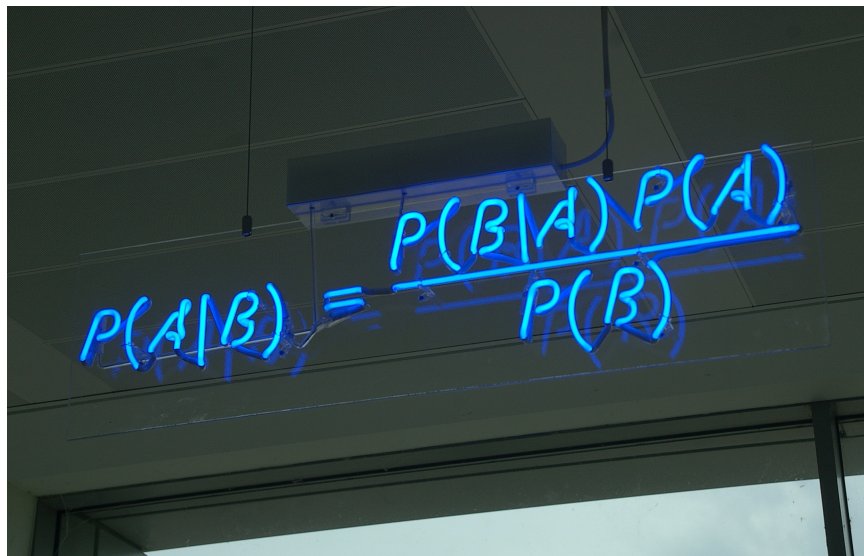
Let

- $\mathbf{P}$  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)





# 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?

## 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?

## 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

# Henri Poincaré

## Some facts about Poincaré:

- Born in **Nancy**, 1854-1912
- Cousin of Raymond Poincaré  
↔ French president during WW1
- Mathematician and engineer
- Numerous contributions in
  - ▶ Celestial mechanics
  - ▶ Relativity
  - ▶ Gravitational waves
  - ▶ Topology
  - ▶ Differential equations



Graph. Fern. Havard.

A handwritten signature of Henri Poincaré in dark ink, written in a cursive style.

# An example by Poincaré (1)

## Situation:

- We are on a train
- Someone gets on the train and proposes to play a card game
- The unknown person wins

## Question:

Probability that this person has cheated?



# An example by Poincaré (2)

**Model:** We set

- $p$  = probability to win without cheating
- $q$  = probability that the unknown person has cheated
- $W$  = "The unknown person wins"
- $C$  = "The unknown person has cheated"

**Hypothesis on probabilities:** We assume

$$\mathbf{P}(W | C^c) = p, \quad \mathbf{P}(W | C) = 1, \quad \mathbf{P}(C) = q$$

**Aim:**

Compute  $\mathbf{P}(C | W)$

# An example by Poincaré (3)

Application of Proposition 4:

$$\begin{aligned}\mathbf{P}(C | W) &= \frac{\mathbf{P}(W | C) \mathbf{P}(C)}{\mathbf{P}(W | C) \mathbf{P}(C) + \mathbf{P}(W | C^c) \mathbf{P}(C^c)} \\ &= \frac{q}{q + p(1 - q)}\end{aligned}$$

Remarks:

(1) We have  $\mathbf{P}(C | W) \geq q = \mathbf{P}(C)$ .

↪ the unknown's win increases his probability to cheat

(2) We have

$$\lim_{p \rightarrow 0} \mathbf{P}(C | W) = 1$$

## Definition 5.

Let

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

We define the odds of  $A$  by

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

# 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

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

# 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?

## 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:

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

Numerical result: We get

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

Therefore

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



# 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
  - ▶  $\cup_{i=1}^n F_i = S$
- $E$  another event

Then we have

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

# 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  $j = 1, 2, 3$ ?

## 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:

$$\mathbf{P}(F_1|A) = \frac{0.7 \times 0.2}{0.41} = \frac{14}{41} \simeq 34\%$$

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# 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  $E \perp\!\!\!\perp F$

## Some remarks

**Interpretation:** If  $E \perp\!\!\!\perp F$ , then

$$\mathbf{P}(E|F) = \mathbf{P}(E),$$

that is the knowledge of  $F$  does not affect  $\mathbf{P}(E)$

**Warning:** Independent  $\neq$  mutually exclusive!

Specifically

$$A, B \text{ mutually exclusive} \Rightarrow \mathbf{P}(AB) = 0$$

$$A, B \text{ independent} \Rightarrow \mathbf{P}(AB) = \mathbf{P}(A)\mathbf{P}(B)$$

Therefore  $A$  et  $B$  both independent and mutually exclusive

$\Leftrightarrow$  we have either  $\mathbf{P}(A) = 0$  or  $\mathbf{P}(B) = 0$

# Example: dice tossing (1)

**Experiment:** We throw two dice

**Sample space:**

- $S = \{1, \dots, 6\}^2$
- $\mathbf{P}(\{(s_1, s_2)\}) = \frac{1}{36}$  for all  $(s_1, s_2) \in S$

**Events:** We consider

$$A = \text{"1}^{\text{st}} \text{ outcome is 1"}, \quad B = \text{"2}^{\text{nd}} \text{ outcome is 4"}$$

**Question:**

Do we have  $A \perp\!\!\!\perp B$ ?

## Example: dice tossing (2)

Description of  $A$  and  $B$ :

$$A = \{1\} \times \{1, \dots, 6\}, \quad \text{and} \quad B = \{1, \dots, 6\} \times \{4\}.$$

Probabilities for  $A$  and  $B$ : We have

$$\mathbf{P}(A) = \frac{|A|}{36} = \frac{1}{6}, \quad \mathbf{P}(B) = \frac{|B|}{36} = \frac{1}{6}$$

Description of  $AB$ : We have  $AB = \{(1, 4)\}$ . Thus

$$\mathbf{P}(AB) = \frac{1}{36} = \mathbf{P}(A) \mathbf{P}(B)$$

Conclusion:  $A$  and  $B$  are **independent**

# Example: tossing $n$ coins (1)

Experiment:

Tossing a coin  $n$  times

Events: We consider

$A =$  "At most one Head"

$B =$  "At least one Head and one Tail"

Question:

Are there values of  $n$  such that  $A \perp\!\!\!\perp B$ ?

## Example: tossing $n$ coins (2)

Model: We take

- $S = \{h, t\}^n$
- $\mathbf{P}(\{s\}) = \frac{1}{2^n}$  for all  $s \in S$

Description of  $A$  and  $B$ :

$$A = \{(t, \dots, t), (h, t, \dots, t), (t, h, t, \dots, t), (t, \dots, t, h)\}$$
$$B = \{(h, \dots, h), (t, \dots, t)\}^c$$



## Example: tossing $n$ coins (3)

Computing probabilities for  $A$  and  $B$ : We have

$$\mathbf{P}(A) = \frac{|A|}{2^n} = \frac{n+1}{2^n}$$

$$\mathbf{P}(B) = 1 - \mathbf{P}(B^c) = 1 - \frac{1}{2^{n-1}}$$

Description of  $AB$  and

$$AB = A \setminus \{(t, \dots, t)\} \quad \Rightarrow \quad \mathbf{P}(AB) = \frac{n}{2^n}$$

## Example: tossing $n$ coins (4)

Checking independence: We have  $A \perp\!\!\!\perp B$  iff

$$\frac{n+1}{2^n} \left(1 - \frac{1}{2^{n-1}}\right) = \frac{n}{2^n} \iff n - 2^{n-1} + 1 = 0$$

Conclusion: One can check that

$$x \mapsto x - 2^{x-1} + 1$$

vanishes for  $x = 3$  only on  $\mathbb{R}_+$ . Thus

We have  $A \perp\!\!\!\perp B$  iff  $n = 3$

# Independence and complements

## Proposition 10.

Let

- $\mathbf{P}$  a probability on a sample space  $S$
- $E, F$  two events
- We assume that  $E \perp\!\!\!\perp F$

Then

$$E \perp\!\!\!\perp F^c, \quad E^c \perp\!\!\!\perp F, \quad E^c \perp\!\!\!\perp F^c$$

# Proof

Decomposition of  $\mathbf{P}(E)$ : Write

$$\begin{aligned}\mathbf{P}(E) &= \mathbf{P}(E F) + \mathbf{P}(E F^c) \\ &= \mathbf{P}(E) \mathbf{P}(F) + \mathbf{P}(E F^c)\end{aligned}$$

Expression for  $\mathbf{P}(E F^c)$ : From the previous expression we have

$$\begin{aligned}\mathbf{P}(E F^c) &= \mathbf{P}(E) - \mathbf{P}(E) \mathbf{P}(F) \\ &= \mathbf{P}(E) (1 - \mathbf{P}(F)) \\ &= \mathbf{P}(E) \mathbf{P}(F^c)\end{aligned}$$

Conclusion:

$$E \perp\!\!\!\perp F^c$$

# Counterexample: independence of 3 events (1)

## Warning:

In certain situations we have  $A, B, C$  pairwise independent, however

$$\mathbf{P}(A \cap B \cap C) \neq \mathbf{P}(A) \mathbf{P}(B) \mathbf{P}(C)$$

**Example:** tossing two dice

- $S = \{1, \dots, 6\}^2$
- $\mathbf{P}(\{(s_1, s_2)\}) = \frac{1}{36}$  for all  $(s_1, s_2) \in S$

**Events:** Define

$A =$  "even number for the 1<sup>st</sup> outcome"

$B =$  "odd number for the 2<sup>nd</sup> outcome"

$C =$  "same parity for the two outcomes"

## Counterexample: independence of 3 events (2)

Description of  $A, B, C$ :

$$A = \{2, 4, 6\} \times \{1, \dots, 6\}$$

$$B = \{1, \dots, 6\} \times \{1, 3, 5\}$$

$$C = (\{2, 4, 6\} \times \{2, 4, 6\}) \cup (\{1, 3, 5\} \times \{1, 3, 5\})$$

Pairwise independence: we find

$$A \perp\!\!\!\perp B, A \perp\!\!\!\perp C \text{ and } B \perp\!\!\!\perp C$$

Independence of the 3 events: We have  $A \cap B \cap C = \emptyset$ . Thus

$$0 = \mathbf{P}(A \cap B \cap C) \neq \mathbf{P}(A) \mathbf{P}(B) \mathbf{P}(C) = \frac{1}{8}$$

# Independence of 3 events

## Definition 11.

Let

- $\mathbf{P}$  a probability on a sample space  $S$
- 3 events  $A_1, A_2, A_3$

We say that  $A_1, A_2, A_3$  are independent if

$$\mathbf{P}(A_1 A_2) = \mathbf{P}(A_1) \mathbf{P}(A_2), \quad \mathbf{P}(A_1 A_3) = \mathbf{P}(A_1) \mathbf{P}(A_3)$$

$$\mathbf{P}(A_2 A_3) = \mathbf{P}(A_2) \mathbf{P}(A_3)$$

and

$$\mathbf{P}(A_1 A_2 A_3) = \mathbf{P}(A_1) \mathbf{P}(A_2) \mathbf{P}(A_3)$$

# Independence of $n$ events

## Definition 12.

Let

- $\mathbf{P}$  a probability on a sample space  $S$
- $n$  events  $A_1, A_2, \dots, A_n$

We say that  $A_1, A_2, \dots, A_n$  are independent if for all  $2 \leq r \leq n$  and  $j_1 < \dots < j_r$  we have

$$\mathbf{P}(A_{j_1} A_{j_2} \dots A_{j_r}) = \mathbf{P}(A_{j_1}) \mathbf{P}(A_{j_2}) \dots \mathbf{P}(A_{j_r})$$



# Independence of an $\infty$ number of events

## Definition 13.

Let

- $\mathbf{P}$  a probability on a sample space  $S$
- A sequence of events  $\{A_i; i \geq 1\}$

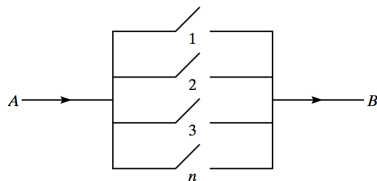
We say that the  $A_i$ 's are independent if for all  $2 \leq r < \infty$  and  $j_1 < \dots < j_r$  we have

$$\mathbf{P}(A_{j_1} A_{j_2} \dots A_{j_r}) = \mathbf{P}(A_{j_1}) \mathbf{P}(A_{j_2}) \dots \mathbf{P}(A_{j_r})$$

# Example: parallel system (1)

## Situation:

- Parallel system with  $n$  components
- All components are independent
- Probability that  $i$ -th component works:  $p_i$



## Question:

Probability that the system functions

## Example: parallel system (2)

Model: We take

- $S = \{0, 1\}^n$
- Probability  $\mathbf{P}$  on  $S$  defined by

$$\mathbf{P}(\{(s_1, \dots, s_n)\}) = \prod_{i=1}^n p_i^{s_i} (1 - p_i)^{1-s_i}$$

Events:

$A =$  "System functions" ,  $A_i =$  " $i$ -th component functions"

Facts about  $A_i$ 's:

The events  $A_i$  are independent and  $\mathbf{P}(A_i) = p_i$

## Example: parallel system (3)

Computations for  $\mathbf{P}(A^c)$ :

$$\begin{aligned}\mathbf{P}(A^c) &= \mathbf{P}\left(\bigcap_{i=1}^n A_i^c\right) \\ &= \prod_{i=1}^n \mathbf{P}(A_i^c) \\ &= \prod_{i=1}^n (1 - p_i)\end{aligned}$$

Conclusion:

$$\mathbf{P}(A) = 1 - \prod_{i=1}^n (1 - p_i)$$

# Example: rolling dice (1)

## Experiment:

- Roll a pair of dice
- Outcome: sum of faces

## Event: We define

- $E =$  "5 appears before 7"

## Question:

Compute  $\mathbf{P}(E)$

## Example: rolling dice (2)

Family of events: For  $n \geq 1$  set

$E_n$  = no 5 or 7 on first  $n - 1$  trials, then 5 on  $n$ -th trial

Relation between  $E_n$  and  $E$ : We have

$$E = 5 \text{ appears before } 7 = \bigcup_{n \geq 1} E_n$$

## Example: rolling dice (3)

Computation for  $\mathbf{P}(E_n)$ : by independence

$$\mathbf{P}(E_n) = \left(1 - \frac{10}{36}\right)^{n-1} \frac{4}{36} = \left(\frac{13}{18}\right)^{n-1} \frac{1}{9}$$

Computation for  $\mathbf{P}(E)$ :

$$\mathbf{P}(E) = \sum_{n=1}^{\infty} \mathbf{P}(E_n) = \frac{1}{9} \frac{1}{1 - \frac{13}{18}}$$

Thus

$$\mathbf{P}(E) = \frac{2}{5}$$

# Same example with conditioning (1)

New events: We set

- $E =$  "5 appears before 7"
- $F_5 =$  "1st trial gives 5"
- $F_7 =$  "1st trial gives 7"
- $H =$  "1st trial gives an outcome  $\neq 5,7$ "



## Same example with conditioning (2)

Conditional probabilities:

$$\mathbf{P}(E|F_5) = 1, \quad \mathbf{P}(E|F_7) = 0, \quad \mathbf{P}(E|H) = \mathbf{P}(E)$$

Justification:  $E \perp\!\!\!\perp H$  since

$$E \cap H = H \cap \{\text{Event which depends on } i\text{-th trials with } i \geq 2\}$$

## Same example with conditioning (3)

Applying Proposition 7:

$$\mathbf{P}(E) = \mathbf{P}(E|F_5) \mathbf{P}(F_5) + \mathbf{P}(E|F_7) \mathbf{P}(F_7) + \mathbf{P}(E|H) \mathbf{P}(H) \quad (3)$$

Computation: We get

$$\mathbf{P}(E) = \frac{1}{9} + \frac{13}{18} \mathbf{P}(E),$$

and thus

$$\mathbf{P}(E) = \frac{2}{5}$$

# Problem of the points

## Experiment:

- Independent trials
- For each trial, success with probability  $p$

## Question:

What is the probability that  $n$  successes occur before  $m$  failures?

# Pascal's solution

Notation: set

$$A_{n,m} = \text{"}n \text{ successes occur before } m \text{ failures"}, \quad P_{n,m} = \mathbf{P}(A_{n,m})$$

Conditioning on 1st trial: Like in (3) we get

$$P_{n,m} = pP_{n-1,m} + (1-p)P_{n,m-1} \quad (4)$$

Initial conditions:

$$P_{n,0} = p^n, \quad P_{0,m} = (1-p)^m \quad (5)$$

Strategy:

Solve difference equation (4) with initial condition (5)

# Fermat's solution

Expression for  $A_{n,m}$ : Write

$A_{n,m}$  = "at least  $n$  successes in  $m + n - 1$  trials"

Thus  $A_{n,m} = \cup_{k=n}^{m+n-1} E_{k,m,n}$  with

$E_{k,m,n}$  = "exactly  $k$  successes in  $m + n - 1$  trials"

Expression for  $P_{n,m}$ : We get

$$P_{n,m} = \sum_{k=n}^{m+n-1} \binom{m+n-1}{k} p^k (1-p)^{m+n-1-k}$$