# COMP9020 Week 9
## Term 3, 2019
## Probability

- Textbook (R & W) - Ch. 5, Sec. 5.1–5.3; Ch. 9
- Supplementary Exercises Ch. 5, 9 (R & W)

# Applications in CS

- AI
  - Machine Learning
  - Decision theory
  - Image processing
  - Speech recognition
- Algorithms
  - Algorithm analysis
  - Big Data sampling and analysis
- Security
  - Cryptography
  - Quantum computing
- Networks
  - Network traffic modelling
  - Reliability modelling

# Summary of topics

- Elementary Discrete Probability
- Exercises
- Infinite sample spaces
- Recursive probability computations
- Conditional probability

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# Elementary Probability

### Definition

**Sample space**:

$$\Omega = \{\omega_1, \ldots, \omega_n\}$$

Each point represents an outcome.

**Event**: a collection of outcomes = subset of $\Omega$

**Probability distribution**: A function $P : \text{Pow}(\Omega) \to \mathbb{R}$ such that:

- $P(\Omega) = 1$
- $E$ and $F$ disjoint events then $P(E \cup F) = P(E) + P(F)$.

### Fact

$$P(\emptyset) = 0, \quad P(E^c) = 1 - P(E)$$

# Examples

**Examples**

Tossing a coin: $\Omega = \{H, T\}$

$$P(H) = P(T) = 0.5$$

Rolling a die: $\Omega = \{1, 2, 3, 4, 5, 6\}$

$$P(1) = P(2) = P(3) = P(4) = P(5) = P(6) = \frac{1}{6}$$

# Uniform distribution

Each outcome $\omega_i$ equally likely:

$$P(\omega_1) = P(\omega_2) = \ldots = P(\omega_n) = \frac{1}{n}$$

This a called a **uniform probability distribution** over $\Omega$

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**Examples**

Tossing a coin: $\Omega = \{H, T\}$

$$P(H) = P(T) = 0.5$$

Rolling a die: $\Omega = \{1, 2, 3, 4, 5, 6\}$

$$P(1) = P(2) = P(3) = P(4) = P(5) = P(6) = \frac{1}{6}$$

# Computing Probabilities by Counting

Computing probabilities with respect to a *uniform* distribution comes down to counting the size of the event.
If $E = \{e_1, \ldots, e_k\}$ then

$$P(E) = \sum_{i=1}^{k} P(e_i) = \sum_{i=1}^{k} \frac{1}{|\Omega|} = \frac{|E|}{|\Omega|}$$

Most of the counting rules carry over to probabilities wrt. a uniform distribution.

### NB

*The expression "selected at random", when not further qualified, means:*
*"subject to / according to / . . . a* uniform *distribution."*

# Combining events

We can create complex events by combining simpler ones.
Common constructions:

- $A$ and $B$: $A \cap B$
- $A$ or $B$: $A \cup B$
- Not $A$: $\Omega \setminus A$
- $A$ followed by $B$

The first three involve events from the same set of outcomes. The last may involve events from different sets of outcomes (e.g. roll die and flip coin).

# Inclusion-exclusion rule

**Fact**

$$P(A \cup B) = P(A) + P(B) - P(A \cap B)$$

$$\begin{aligned} P(A \cup B \cup C) = {} & P(A) + P(B) + P(C) \\ & - P(A \cap B) - P(B \cap C) - P(C \cap A) \\ & + P(A \cap B \cap C) \end{aligned}$$

# Exercises

## Exercises

5.2.7 Suppose an experiment leads to events $A, B$ with probabilities $P(A) = 0.5, P(B) = 0.8, P(A \cap B) = 0.4$.
Find

- $P(B^c) =$
- $P(A \cup B) =$
- $P(A^c \cup B^c) =$

5.2.8 Given $P(A) = 0.6$, $P(B) = 0.7$, show $P(A \cap B) \geq 0.3$

# Exercises

### Exercises

5.2.7 Suppose an experiment leads to events $A, B$ with probabilities $P(A) = 0.5, P(B) = 0.8, P(A \cap B) = 0.4$. Find

- $P(B^c) = \ 1 - P(B) = 0.2$
- $P(A \cup B) = \ P(A) + P(B) - P(A \cap B) = 0.9$
- $P(A^c \cup B^c) = \ 1 - P((A^c \cup B^c)^c) = 1 - P(A \cap B) = 0.6$

5.2.8 Given $P(A) = 0.6, \ P(B) = 0.7$, show $P(A \cap B) \geq 0.3$

$$P(A \cap B) = P(A) + P(B) - P(A \cup B)$$
$$= 0.6 + 0.7 - P(A \cup B)$$
$$\geq 0.6 + 0.7 - 1 = 0.3$$

# Unifying sets of outcomes

To combine events from different sets of outcomes we unify the sample space using the **product space**: $\Omega_1 \times \Omega_2 \times \ldots \times \Omega_n$.

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**Example**

Flipping a coin and rolling a die:

$$\Omega_1 = \{\text{heads}, \text{tails}\} \qquad \Omega_2 = \{1, 2, 3, 4, 5, 6\}$$

$$\Omega = \Omega_1 \times \Omega_2 = \{(\text{heads}, 1), (\text{heads}, 2), \ldots\}$$

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**NB**

*This approach can also be used to model sequences of outcomes.*

# Events in the product space

Events are lifted into the product space by restricting the appropriate co-ordinate. E.g. $A \subseteq \Omega_1$ translates to $A' = A \times \Omega_2 \times \ldots \times \Omega_n$.

### Example

Coin shows heads and die shows an even number:

$$\Omega_1 = \{\text{heads, tails}\} \qquad A = \{\text{heads}\}$$
$$\Omega_2 = \{1, 2, 3, 4, 5, 6\} \qquad B = \{2, 4, 6\}$$

$$\Omega = \Omega_1 \times \Omega_2 = \{(\text{heads}, 1), (\text{heads}, 2), \ldots\}$$
$$A' = A \times \Omega_2 \qquad B' = \Omega_1 \times B$$

"$A$ and $B$" or "$A$ followed by $B$" corresponds to:
$$A' \cap B' = (A \times \Omega_2) \cap (\Omega_1 \times B) = A \times B$$

# Probability in the product space

**NB**

*Cannot assume that $P(A \times B) = P(A)P(B)$*

**Example**

Toss two coins.

- $A$: First coin shows heads
- $B$: Both coins show tails

$\Omega_1 = \{H, T\}$ $\qquad \Omega_2 = \{HH, HT, TH, TT\}$

$A = \{H\}$ $\qquad A' = \{(H, HH), (H, HT), (H, TH), (H, TT)\}$

$B = \{TT\}$ $\qquad B' = \{(H, TT), (T, TT)\}$

$A' \cap B' = A \times B = \{(H, TT)\}$

$$P(A) = \frac{1}{2} \quad P(B) = \frac{1}{4} \quad P(A' \cap B') = 0$$

# Independence: Intuition

Given probability distributions on the component spaces, there is a natural probability distribution on the product space:

$$P(E_1 \times E_2 \times \ldots \times E_n) = P_1(E_1).P_2(E_2).\ldots.P_n(E_n)$$

Intuitively, the probability of an event in one dimension is not affected by the outcomes in the other dimensions.

# Independence

Intuitively, events are *independent* if the outcomes in one do not affect the outcomes in the other. More generally, we define independence on events of the same sample space.

### Definition

$A$ and $B$ are **(stochastically) independent** (notation: $A \perp B$) if
$$P(A \cap B) = P(A) \cdot P(B)$$

### NB

*Intuitive notion of independence corresponds to the stochatic independence of the "lifted" events $A'$ and $B'$*

# Probability of Sequential Independent Outcomes

## Example

Team $A$ has probability $p = 0.5$ of winning a game against $B$.
What is the probability $P_p$ of $A$ winning a best-of-seven match if
(a) $A$ already won the first game?
(b) $A$ already won the first two games?
(c) $A$ already won two out of the first three games?

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(a) Sample space $S$ — 6-sequences, formed from wins (W) and losses (L)

$$|S| = 2^6 = 64$$

Favourable sequences $F$ — those with three to six W

$$|F| = \binom{6}{3} + \binom{6}{4} + \binom{6}{5} + \binom{6}{6} = 20 + 15 + 6 + 1 = 42$$

Therefore $P_{0.5} = \frac{42}{64} \approx 66\%$

## Probability of Sequential Independent Outcomes

**Example (cont'd)**

(b) Sample space $S$ — 5-sequences of W and L

$$|S| = 2^5 = 32$$

Favourable sequences $F$ — those with two to five W

$$|F| = \binom{5}{2} + \binom{5}{3} + \binom{5}{4} + \binom{5}{5} = 10 + 10 + 5 + 1 = 26$$

Therefore $P_{0.5} = \frac{26}{32} \approx 81\%$

(c)

$$|S| = 2^4 = 16$$

$$|F| = \binom{4}{2} + \binom{4}{3} + \binom{4}{4} = 6 + 4 + 1 = 11$$

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# Examples

### Example

A four-digit number $n$ is selected at random (i.e. randomly from $[1000 \ldots 9999]$). Find the probability $p$ that $n$ has each of 0, 1, 2 among its digits.

Let $q = 1 - p$ be the complementary probability and define

$A_i = \{n : \text{no digit } i\}, A_{ij} = \{n : \text{no digits } i, j\}, A_{ijk} = \{n : \text{no } i, j, k\}$

Then define
$T = A_0 \cup A_1 \cup A_2 = \{n : \text{ missing at least one of } 0, 1, 2\}$
$S = (A_0 \cup A_1 \cup A_2)^c = \{n : \text{ containing each of } 0, 1, 2\}$

# Examples

**Example (cont'd)**

Once we find the cardinality of $T$, the solution is

$$q = \frac{|T|}{9000}, \ p = 1 - q$$

To find $|A_i|, |A_{ij}|, |A_{ijk}|$ we reflect on how many choices are available for the first digit, for the second etc. A special case is the leading digit, which must be $1, \ldots, 9$

# Examples

### Example (cont'd)

$$|A_0| = 9^4, \quad |A_1| = |A_2| = 8 \cdot 9^3$$
$$|A_{01}| = |A_{02}| = 8^4, \quad |A_{12}| = 7 \cdot 8^3$$
$$|A_{012}| = 7^4$$

$$\begin{aligned}
|T| &= |A_0 \cup A_1 \cup A_2| \\
&= |A_0| + |A_1| + |A_2| - |A_0 \cap A_1| - |A_0 \cap A_2| - |A_1 \cap A_2| \\
&\quad + |A_0 \cap A_1 \cap A_2| \\
&= 9^4 + 2 \cdot 8 \cdot 9^3 - 2 \cdot 8^4 - 7 \cdot 8^3 + 7^4 \\
&= 25 \cdot 9^3 - 23 \cdot 8^3 + 7^4 = 8850
\end{aligned}$$

$$q = \frac{8850}{9000}, \quad p = 1 - q \quad \approx 0.01667$$

# Examples

### Example

Previous example generalised: Probability of an $r$-digit number having all of 0,1,2,3 among its digits.

We use the previous notation: $A_i$ — set of numbers *n missing* digit $i$, and similarly for all $A_{ij...}$

We aim to find the size of $T = A_0 \cup A_1 \cup A_2 \cup A_3$, and then to compute $|S| = 9 \cdot 10^{r-1} - |T|$.

$$
\begin{aligned}
|A_0 \cup A_1 \cup A_2 \cup A_3| = {} & \text{sum of } |A_i| \\
& - \text{sum of } |A_i \cap A_j| \\
& + \text{sum of } |A_i \cap A_j \cap A_k| \\
& - \text{sum of } |A_i \cap A_j \cap A_k \cap A_l|
\end{aligned}
$$

# Examples

### Examples

5.6.38 (Supp) Of 100 problems, 75 are 'easy' and 40 'important'.
(b) $n$ problems chosen randomly. What is the probability that all $n$ are important?

$$p = \frac{\binom{40}{n}}{\binom{100}{n}} = \frac{40 \cdot 39 \cdots (41 - n)}{100 \cdot 99 \cdots (101 - n)}$$

5.2.3 A 4-letter word is selected at random from $\Sigma^4$, where $\Sigma = \{a, b, c, d, e\}$. What is the probability that
(a) the letters in the word are all distinct?
(b) there are no vowels ("a", "e") in the word?
(c) the word begins with a vowel?

# Examples

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(a) $|E| = \Pi(5, 4), \quad P(E) = \frac{5 \cdot 4 \cdot 3 \cdot 2}{5^4} = \frac{120}{625} \approx 19\%$
(b) $|E| = 3^4, \quad P(E) = \frac{81}{625} \approx 13\%$
(c) $|E| = 2 \cdot 5^3, \quad P(E) = \frac{2}{5}$

# Exercise

**Exercises**

5.2.11 Two dice, a red die and a black die, are rolled.
What is the probability that
(a) the sum of the values is even?

(b) the number on the red die is bigger than on the black die?

(c) the number on the black die is twice the one on the red die?

5.2.12 (a) the maximum of the numbers is 4?
(b) their minimum is 4?

# Exercise

### Exercises

5.2.11 Two dice, a red die and a black die, are rolled.
What is the probability that
(a) the sum of the values is even?
$P\big(R + B \in \{2, 4, \ldots, 12\}\big) = \frac{18}{36} = \frac{1}{2}$

(b) the number on the red die is bigger than on the black die?
$P(R > B) = P(R < B)$; also $P(R = B) = \frac{1}{6}$
Therefore $P(R < B) = \frac{1}{2}(1 - P(R = B)) = \frac{5}{12}$

(c) the number on the black die is twice the one on the red die?
$P(R = 2 \cdot B) = P(\{(2, 1), (4, 2), (6, 3)\}) = \frac{3}{36} = \frac{1}{12}$

5.2.12 (a) the maximum of the numbers is 4?  $P(E_1) = \frac{7}{36}$
(b) their minimum is 4?  $P(E_2) = \frac{5}{36}$
Check:
$P(E_1 \cup E_2) = \frac{7}{36} + \frac{5}{36} - P(E_1 \cap E_2) = \frac{7+5-1}{36} = \frac{11}{36}$

# Exercise

**Exercises**

5.2.5 An urn contains 3 red and 4 black balls. 3 balls are removed without replacement. What are the probabilities that
(a) all 3 are red
(b) all 3 are black
(c) one is red, two are black

# Exercise

### Exercises

5.2.5 An urn contains 3 red and 4 black balls. 3 balls are removed without replacement. What are the probabilities that
(a) all 3 are red
(b) all 3 are black
(c) one is red, two are black

All probabilities are computed using the same sample space: all possible ways to draw three balls without replacement.

The size of the sample space is $\dfrac{7 \cdot 6 \cdot 5}{3!} = 35$

(a) $E =$ All balls are red: 1 combination

(b) $E =$ All balls are black: $\binom{4}{3} = 4$ combinations

(c) $E =$ One red and two black: $\binom{3}{1} \cdot \binom{4}{2} = 18$ combinations

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- Elementary Discrete Probability
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# Infinite sample spaces

Probability distributions generalize to infinite sample spaces **with some provisos**.

- In continuous spaces (e.g. $\mathbb{R}$):
  - Probability distributions are *measures*;
  - Sums are *integrals*;
  - Non-zero probabilities apply to *ranges*;
  - Probability of a single event is $0$. Note: Probability $0$ is not the same as impossible.
- In discrete spaces (e.g. $\mathbb{N}$):
  - Probability $0$ is the same as impossible.
  - No uniform distribution!
  - Non-uniform distributions exist, e.g. $P(0) = 1$, $P(n) = 0$ for $n > 0$; or $P(0) = 0$, $P(n) = \frac{1}{2^n}$ for $n > 0$.
  - May consider limiting probabilities if that makes sense.

# Asymptotic Estimate of Relative Probabilities

### Example

Event $A \overset{\text{def}}{=}$ one die rolled $n$ times and you obtain two 6's

Event $B \overset{\text{def}}{=}$ $n$ dice rolled simultaneously and you obtain one 6

$$P(A) = \frac{\binom{n}{2} \cdot 5^{n-2}}{6^n} \quad P(B) = \frac{\binom{n}{1} \cdot 5^{n-1}}{6^n}$$

Therefore $\frac{P(A)}{P(B)} = \frac{\binom{n}{2}}{\binom{n}{1}} \cdot \frac{1}{5} = \frac{n(n-1)}{2} \cdot \frac{1}{5n} = \frac{n-1}{10} \in \Theta(n)$

| $n$ | 1 | 2 | 3 | 4 | ... | 11 | ... | 20 | ... |
|---|---|---|---|---|---|---|---|---|---|
| $P(A)$ | 0 | $\frac{1}{36}$ | $\frac{5}{72}$ | $\frac{25}{216}$ | ... | 0.296 | ... | 0.198 | ... |
| $P(B)$ | $\frac{1}{6}$ | $\frac{10}{36}$ | $\frac{25}{72}$ | $\frac{125}{324}$ | ... | 0.296 | ... | 0.104 | ... |

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# Use of Recursion in Probability Computations

### Question

*Given $n$ tosses of a coin, what is the probability of two HEADS in a row? Compute for $n = 5, 10, 20, \ldots$*

Approaches:

- I. Write down all possibilities — 32 for $n = 5$, 1024 for $n = 10, \ldots$

- II. Write a program; running time $O(2^n)$ — why?

- III. Inter-relate the numbers of relevant possibilities

$N_n \stackrel{\text{def}}{=}$ No. of sequences of $n$ tosses *without* $\ldots$ HH $\ldots$ pattern
Initial values:
$N_0 = 1$, $N_1 = 2$, $N_2 = 3$ (all except "HH")
$N_3 = 5$ (why?) $\quad N_4 = 8$ (why?)

**Answer**

*We can summarise all possible outcomes in a* **recursive tree**

$N_n$:

First toss

T     H

$N_{n-1}$

Second toss

T     H

$N_{n-2}$

•
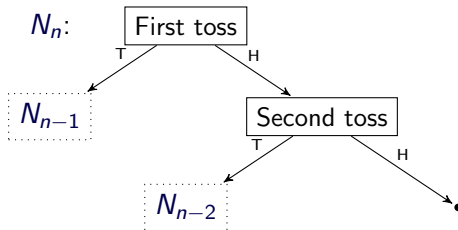
$N_n = N_{n-1} + N_{n-2}$ — Fibonacci recurrence: $N_n = \text{FIB}(n+1)$

$N_n \approx \frac{1}{\sqrt{5}} \left( \frac{\sqrt{5}+1}{2} \right)^{n+1} \approx 0.72 \cdot (1.6)^n$

$p_n = \frac{2^n - \text{FIB}(n+1)}{2^n} \approx 1 - 0.72 \cdot (0.8)^n$

# Example

**Question**

*Given $n$ tosses, what is the probability $q_n$ of at least one HHH?*

$q_0 = q_1 = q_2 = 0; q_3 = \frac{1}{8}$
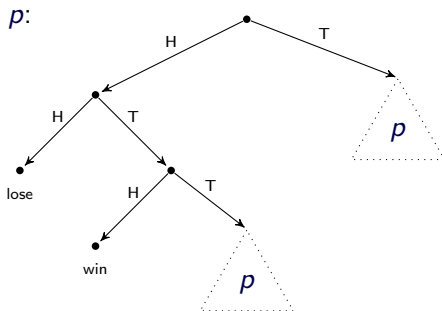
Then recursive computation:

$$
\begin{aligned}
q_n =& \frac{1}{2}q_{n-1} && \text{(initial: T)} \\
&+ \frac{1}{4}q_{n-2} && \text{(initial: HT)} \\
&+ \frac{1}{8}q_{n-3} && \text{(initial: HHT)} \\
&+ \frac{1}{8} && \text{(start with: HHH)}
\end{aligned}
$$

# Example

**Question**

*A coin is tossed 'indefinitely'. Which pattern is more likely (and by how much) to appear first, HTH or HHT?*
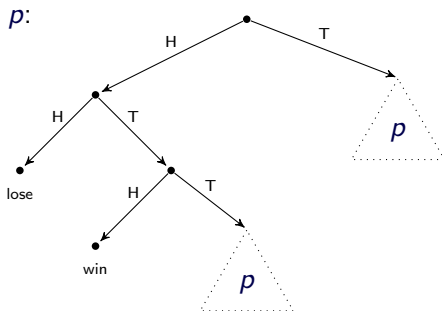
let $p = P(HTH \text{ first})$

# Example

**Question**

*A coin is tossed 'indefinitely'. Which pattern is more likely (and by how much) to appear first, HTH or HHT?*

let $p = P(HTH \text{ first})$



$$p = \frac{1}{8} + \frac{1}{8}p + \frac{1}{2}p \;\rightarrow\; \frac{3}{8}p = \frac{1}{8} \;\rightarrow\; p = \frac{1}{3}$$
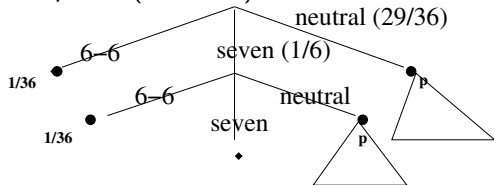
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# Example

### Question

*Two dice are rolled repeatedly. What is the probability that '6–6' will occur before two consecutive (back-to-back) 'totals seven'?*

### NB

*The probability of either occurring at a given roll is the same:* $\frac{1}{36}$.
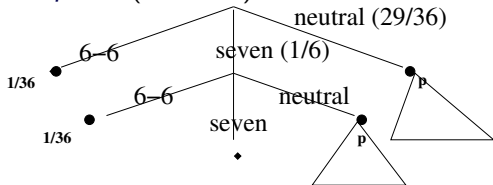
Let $p = P(6\text{–}6 \text{ first})$

# Example

### Question

*Two dice are rolled repeatedly. What is the probability that '6–6' will occur before two consecutive (back-to-back) 'totals seven'?*

### NB

*The probability of either occurring at a given roll is the same: $\frac{1}{36}$.*

Let $p = P(\text{6–6 first})$



$$p = \frac{1}{36} + \frac{1}{6} \cdot \frac{1}{36} + \frac{1}{6} \cdot \frac{29}{36}p + \frac{29}{36}p \;\rightarrow\; 216p = 7 + 203p \;\rightarrow\; p = \frac{7}{13}$$

### NB

*The majority of problems in probability and statistics do not have such elegant solutions. Hence the use of computers for either precise calculations or approximate simulations is mandatory. However, it is the use of recursion that simplifies such computing or, quite often, makes it possible in the first place.*

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# Conditional Probability

### Definition

**Conditional** probability of $E$ **given** $S$:

$$P(E|S) = \frac{P(E \cap S)}{P(S)}, \quad E, S \subseteq \Omega$$

It is defined only when $P(S) \neq 0$

### NB

$P(A|B)$ and $P(B|A)$ are, in general, not related — one of these values predicts, by itself, essentially nothing about the other. The only exception, applicable when $P(A), P(B) \neq 0$, is that $P(A|B) = 0$ iff $P(B|A) = 0$ iff $P(A \cap B) = 0$.

If $P$ is the uniform distribution over a finite set $\Omega$, then

$$P(E|S) = \frac{\frac{|E \cap S|}{|\Omega|}}{\frac{|S|}{|\Omega|}} = \frac{|E \cap S|}{|S|}$$

This observation can help in calculations...

### Example

$\boxed{9.1.6}$ A coin is tossed four times. What is the probability of
(a) two consecutive HEADS
(b) two consecutive HEADS *given* that $\geq 2$ tosses are HEADS

| | | | | | | | | |
|---|---|---|---|---|---|---|---|---|
| T | T | T | T | | H | T | T | T |
| T | T | T | H | | H | T | T | H |
| T | T | H | T | | H | T | H | T |
| T | T | H | H | | H | T | H | H |
| T | H | T | T | | H | H | T | T |
| T | H | T | H | | H | H | T | H |
| T | H | H | T | | H | H | H | T |
| T | H | H | H | | H | H | H | H |

(a) $\frac{8}{16}$  (b) $\frac{8}{11}$

# Some General Rules

**Fact**

- $A \subseteq B \rightarrow P(A|B) \geq P(A)$
- $A \subseteq B \rightarrow P(B|A) = 1$
- $P(A \cap B|B) = P(A|B)$
- $P(\emptyset|A) = 0$ for $A \neq \emptyset$
- $P(A|\Omega) = P(A)$
- $P(A^c|B) = 1 - P(A|B)$

**NB**

- $P(A|B)$ and $P(A|B^c)$ are not related
- $P(A|B), P(B|A), P(A^c|B^c), P(B^c|A^c)$ are not related

**Example**

Two dice are rolled and the outcomes recorded as $b$ for the black die, $r$ for the red die and $s = b + r$ for their total.

Define the events $B = \{b \geq 3\}$, $R = \{r \geq 3\}$, $S = \{s \geq 6\}$.

$P(S|B) = \frac{4+5+6+6}{24} = \frac{21}{24} = \frac{7}{8} = 87.5\%$

$P(B|S) = \frac{4+5+6+6}{26} = \frac{21}{26} = 80.8\%$

The (common) numerator $4 + 5 + 6 + 6 = 21$ represents the size of the $B \cap S$ — the common part of $B$ and $S$, that is, the number of rolls where $b \geq 3$ and $s \geq 6$. It is obtained by considering the different cases: $b = 3$ and $s \geq 6$, then $b = 4$ and $s \geq 6$ etc.

The denominators are $|B| = 24$ and $|S| = 26$

# Example

## Example (cont'd)

Recall: $B = \{b \geq 3\}$, $R = \{r \geq 3\}$, $S = \{s \geq 6\}$

$P(B) = P(R) = 2/3 = 66.7\%$

$P(S) = \frac{5+6+5+4+3+2+1}{36} = \frac{26}{36} = 72.22\%$

$P(S|B \cup R) = \frac{2+3+4+5+6+6}{32} = \frac{26}{32} = 81.25\%$

The set $B \cup R$ represents the event '$b$ or $r$'.

It comprises all the rolls except for those with *both* the red and the black die coming up either 1 or 2.

$P(S|B \cap R) = 1 = 100\%$ — because $S \supseteq B \cap R$

# Exercise

9.1.9 Consider three red and eight black marbles; draw two without replacement. We write $b_1$ — Black on the first draw, $b_2$ — Black on the second draw, $r_1$ — Red on first draw, $r_2$ — Red on second draw
Find the probabilities
(a) both Red:

# Exercise

**Exercise**

9.1.9 Consider three red and eight black marbles; draw two
without replacement. We write $b_1$ — Black on the first draw,
$b_2$ — Black on the second draw, $r_1$ — Red on first draw,
$r_2$ — Red on second draw
Find the probabilities
(a) both Red:

$$P(r_1 \wedge r_2) = P(r_1)P(r_2|r_1) = \frac{3}{11} \cdot \frac{2}{10} = \frac{3}{55}$$

Equivalently:
$|\text{two-samples}| = \binom{11}{2} = 55$; $|\text{Red two-samples}| = \binom{3}{2} = 3$
$P(\cdot) = \frac{\binom{3}{2}}{\binom{11}{2}} = \frac{3}{55}$

(b) both Black:

(c) one Red, one Black:

$$P(r_1 \wedge b_2) + P(b_1 \wedge r_2) = \frac{3 \cdot 8}{\binom{11}{2}} \quad \text{— why?}$$

**Exercise**

(b) both Black:
$$P(b_1 \wedge b_2) = P(b_1)P(b_2|b_1) = \frac{8}{11} \cdot \frac{7}{10} = \frac{28}{55} = \frac{\binom{8}{2}}{\binom{11}{2}}$$

(c) one Red, one Black:
$$P(r_1 \wedge b_2) + P(b_1 \wedge r_2) = \frac{3 \cdot 8}{\binom{11}{2}} \quad \text{— why?}$$

By textbook (the 'hard way')
$$P(r_1 \wedge b_2) + P(b_1 \wedge r_2) = \frac{3}{11} \cdot \frac{8}{10} + \frac{8}{11} \cdot \frac{3}{10}$$

or
$$P(\cdot) = 1 - P(r_1 \wedge r_2) - P(b_1 \wedge b_2) = \frac{55 - 3 - 28}{55}$$

# Exercise

## Exercise

9.1.12 What is the probability of a flush given that all five cards in a Poker hand are red?

# Exercise

# Exercise

### Exercise

9.1.22 Prove the following:
If $P(A|B) > P(A)$ ("positive correlation") then $P(B|A) > P(B)$

# Exercise

### Exercise

9.1.22 Prove the following:
If $P(A|B) > P(A)$ ("positive correlation") then $P(B|A) > P(B)$

$P(A|B) > P(A)$

$\rightarrow P(A \cap B) > P(A)P(B)$

$\rightarrow \frac{P(A \cap B)}{P(A)} > P(B)$

$\rightarrow P(B|A) > P(B)$

# Stochastic Independence

### Definition

$A$ and $B$ are **(stochastically) independent** (notation: $A \perp B$) if
$P(A \cap B) = P(A) \cdot P(B)$

If $P(A) \neq 0$ and $P(B) \neq 0$, all of the following are *equivalent* definitions:

- $P(A \cap B) = P(A)P(B)$
- $P(A|B) = P(A)$
- $P(B|A) = P(B)$
- $P(A^c|B) = P(A^c)$ or $P(A|B^c) = P(A)$ or $P(A^c|B^c) = P(A^c)$

The last one claims that

$$A \perp B \leftrightarrow A^c \perp B \leftrightarrow A \perp B^c \leftrightarrow A^c \perp B^c$$

Basic non-independent sets of events

- $A \subseteq B$
- $A \cap B = \emptyset$
- Any pair of one-point events $\{x\}, \{y\}$:
  either $x = y$ and $P(x|y) = 1$
  or $x \neq y$ and $P(x|y) = 0$

Basic non-independent sets of events

- $A \subseteq B$
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  either $x = y$ and $P(x|y) = 1$
  or $x \neq y$ and $P(x|y) = 0$

Independence of $A_1, \ldots, A_n$

$$P(A_{i_1} \cap A_{i_2} \cap \ldots \cap A_{i_k}) = P(A_{i_1}) \cdot P(A_{i_2}) \cdots P(A_{i_k})$$

for all possible collections $A_{i_1}, A_{i_2}, \ldots, A_{i_k}$.
This is often called (for emphasis) a *full* independence

Pairwise independence is a *weaker* concept.

One can similarly construct a set of $n$ events where any $k$ of them are independent, while any $k+1$ are dependent (for $k < n$). Independence of events, even just pairwise independence, can greatly simplify computations and reasoning in AI applications. It is common for many expert systems to make an approximating assumption of independence, even if it is not completely satisfied.



$$P(\text{sense}_t \,|\, \text{loc}_t, \text{sense}_{t-1}, \text{loc}_{t-1}, \ldots) = P(\text{sense}_t \,|\, \text{loc}_t)$$

# Exercise

### Exercise

9.1.7 Suppose that an experiment leads to events $A$, $B$ and $C$ with $P(A) = 0.3$, $P(B) = 0.4$ and $P(A \cap B) = 0.1$

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

(b) $P(A^c) =$

(c) Is $A \perp B$ ?

(d) Is $A^c \perp B$ ?

# Exercise

$\boxed{9.1.7}$ Suppose that an experiment leads to events $A$, $B$ and $C$ with $P(A) = 0.3$, $P(B) = 0.4$ and $P(A \cap B) = 0.1$

(a) $P(A|B) = \frac{P(A \cap B)}{P(B)} = \frac{1}{4}$

(b) $P(A^c) = 1 - P(A) = 0.7$

(c) Is $A \perp B$?   No.  $P(A) \cdot P(B) = 0.12 \neq P(A \cap B)$

(d) Is $A^c \perp B$?   No, as can be seen from (c).

Note:   $P(A^c \cap B) = P(B) - P(A \cap B) = 0.4 - 0.1 = 0.3$
         $P(A^c) \cdot P(B) = 0.7 \cdot 0.4 = 0.28$

# Exercise

### Exercise

9.1.8 Given $A \perp B$, $P(A) = 0.4$, $P(B) = 0.6$

$P(A|B) =$
$P(A \cup B) =$
$P(A^c \cap B) =$

# Exercise

**Exercise**

$\boxed{9.1.8}$ Given $A \perp B$, $P(A) = 0.4$, $P(B) = 0.6$

$P(A|B) = P(A) = 0.4$
$P(A \cup B) = P(A) + P(B) - P(A)P(B) = 0.76$
$P(A^c \cap B) = P(A^c)P(B) = 0.36$

# Exercise

**Exercise**

9.1.25 Does $A \perp B \perp C$ imply $(A \cap B) \perp (A \cap C)$ ?

# Exercise

**Exercise**

9.1.25 Does $A \perp B \perp C$ imply $(A \cap B) \perp (A \cap C)$ ?

No; this is almost never the case.
If somehow $(A \cap B) \perp (A \cap C)$ then it would give

$$P(A \cap B \cap C) = P(A \cap B \cap A \cap C) = P(A \cap B) \cdot P(A \cap C)$$

As $A$ is independent of $B$ and of $C$ it would suggest

$$P(A \cap B \cap C) \stackrel{?}{=} P(A) \cdot P(B) \cdot P(A) \cdot P(C)$$

instead of the correct

$$P(A \cap B \cap C) = P(A) \cdot P(B) \cdot P(C)$$

## Supplementary Exercise

### Exercises

9.5.5 (Supp) We are given two events with $P(A) = \frac{1}{4}$, $P(B) = \frac{1}{3}$.
True, false or could be either?
(a) $P(A \cap B) = \frac{1}{12}$
(b) $P(A \cup B) = \frac{7}{12}$
(c) $P(B|A) = \frac{P(B)}{P(A)}$
(d) $P(A|B) \geq P(A)$
(e) $P(A^c) = \frac{3}{4}$
(f) $P(A) = P(B)P(A|B) + P(B^c)P(A|B^c)$

# Supplementary Exercise

## Exercises

$\boxed{9.5.5}$ (Supp) We are given two events with $P(A) = \frac{1}{4}$, $P(B) = \frac{1}{3}$.
True, false or could be either?

(a) $P(A \cap B) = \frac{1}{12}$ — possible; it holds when $A \perp B$

(b) $P(A \cup B) = \frac{7}{12}$ — possible; it holds when $A, B$ are disjoint

(c) $P(B|A) = \frac{P(B)}{P(A)}$ — false; correct is: $P(B|A) = \frac{P(B \cap A)}{P(A)}$

(d) $P(A|B) \geq P(A)$ — possible (it means that $B$ "supports" $A$)

(e) $P(A^c) = \frac{3}{4}$ — true, since $P(A^c) = 1 - P(A)$

(f) $P(A) = P(B)P(A|B) + P(B^c)P(A|B^c)$ — true

(also known as *total probability*)