
Collaborative transmission in wireless sensor networks

Introduction to probability theory

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Overview and Structure

- Introduction to context aware computing
- Wireless sensor networks
- Wireless communications
- Basics of probability theory
- Randomised search approaches
- Cooperative transmission schemes
- Distributed adaptive beamforming
 - Feedback based approaches
 - Asymptotic bounds on the synchronisation time
 - Alternative algorithmic approaches
 - Alternative Optimisation environments

Overview and Structure

- Introduction to context aware computing
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- **Basics of probability theory**
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Outline

Basics of probability theory

- 1 Introduction
- 2 Notation
- 3 Calculation with probabilities
- 4 The Markov inequality
- 5 The Chernoff bound

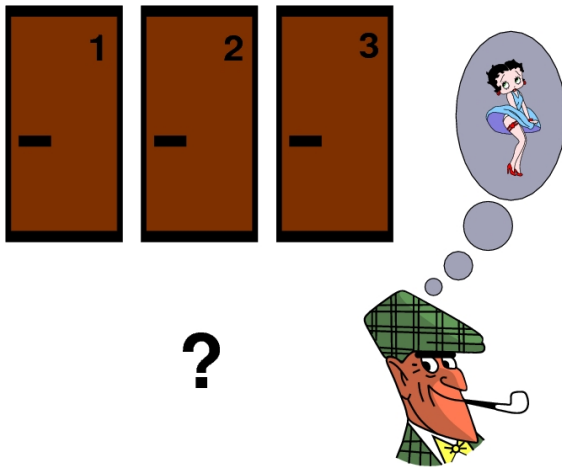
Probability in everyday life

We are confronted with Probability constantly:

- Weather forecasts
- Quiz shows
- ...

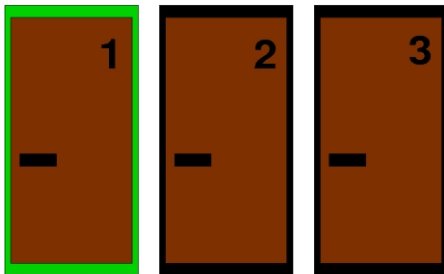
Example

The treasure behind the doors



Example

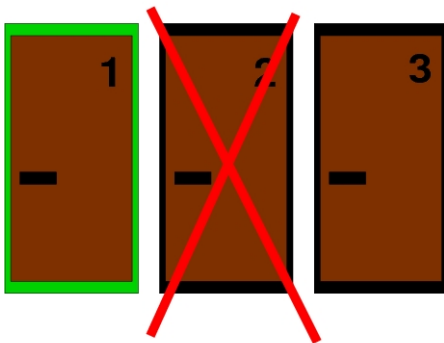
The treasure behind the doors



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Example

The treasure behind the doors



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Example

The treasure behind the doors

- What shall the candidate do?
 - Alter his decision?
 - Retain his decision?
 - Does it make a difference?

Example

The treasure behind the doors

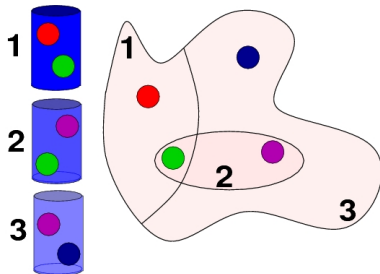
- What shall the candidate do?
 - Alter his decision?
 - Retain his decision?
 - Does it make a difference?
- We will consider the solution to this Problem in some minutes

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Notation

Experiments, Events and sample points



- The results of experiments or observations are called events.
- Events are sets of sample points.
- The sample space is the set of all possible events.

Notation

Experiments, Events and sample points



- What is the sample space for the experiment of tossing a coin two times?

Example

Sample spaces

- Three distinct balls (a,b,c) are to be placed in three distinct bins.

	1	2	3	4	5	6	7	8	9	10	11	12	13
1	abc			ab	ab	c		c		ac	ac	b	
2		abc		c		ab	ab		c	b		ac	ac
3			abc		c		c	ab	ab		b		b
14	15	16	17	18	19	20	21	22	23	24	25	26	27
b		bc	bc	a		a		a	a	b	b	c	c
	b	a		bc	bc		a	b	c	a	c	a	b
ac	ac		a		a	bc	bc	c	b	c	a	b	a

Example

Sample spaces

- Suppose that the three balls are not distinguishable.

Event	1	2	3	4	5	6	7	8	9	10
Bin										
1	***			**	**	*		*		*
2		***		*		**	**		*	*
3			***		*		*	**	**	*

Example

Sample spaces

- Indistinguishable balls and indistinguishable bins

	Event	1	2	3
Bin				
1		***	**	*
2			*	*
3				*

Notation

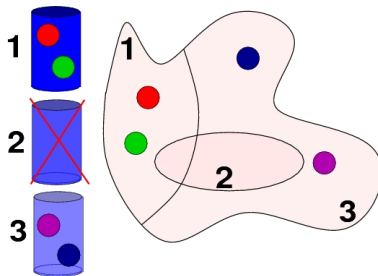
Probability space

Probability space

A probability space (Π, P) consists of a sample space Π and a probability measure $P : \Pi \rightarrow [0, 1]$. This function satisfies the following conditions

- For each subset $\Pi' \subseteq \Pi$, $0 \leq P(\Pi') \leq 1$
- $P(\Pi) = 1$
- For each $\Pi' \subseteq \Pi$, $P(\Pi') = \sum_{\chi \in \Pi'} P(\chi)$

Impossible events



Impossible event

With $\chi = \{\}$ we denote the fact that event χ contains no sample points. It is impossible to observe event χ as an outcome of the experiment.

Probability of events

Probability of events

Given a sample space Π and an event $\chi \in \Pi$, the occurrence probability $P(\chi)$ of event χ is the sum probability of all sample points from χ :

$$P(\chi) = \sum_{x \in \chi} P(x). \quad (1)$$

Statistical independence

Independence

A collection of events χ_i that form the sample space Π is independent if for all subsets $\Pi' \subseteq \Pi$

$$P\left(\bigcap_{\chi_i \in \Pi'} \chi_i\right) = \prod_{\chi_i \in \Pi'} P(\chi_i). \quad (2)$$

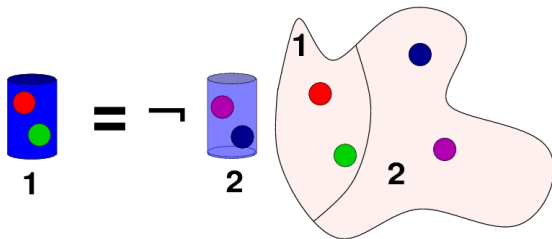
- Statistical independence is required for many useful results in probability theory.
- Be careful to apply such results not in cases where independence between sample points is not provided.

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Calculation with probabilities

Negation of events



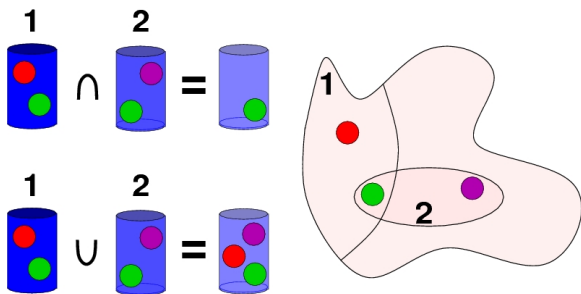
For every event χ there is an event $\neg\chi$ that is defined as ' χ does not occur '.

Negation of events

The event consisting of all sample points x with $x \notin \chi$ is the complementary event (or negation) of χ and is denoted by $\neg\chi$.

Calculation with probabilities

Subsuming events

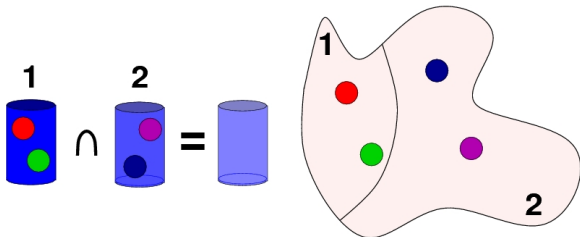


$$\chi_1 \cap \chi_2 = \{x | x \in \chi_1 \wedge x \in \chi_2\} \quad (3)$$

$$\chi_1 \cup \chi_2 = \{x | x \in \chi_1 \vee x \in \chi_2\} \quad (4)$$

Calculation with probabilities

Mutual exclusive events



Mutual exclusive events

When the events χ_1 and χ_2 have no sample point x in common, the event $\chi_1 \cap \chi_2$ is impossible: $\chi_1 \cap \chi_2 = \{\}$.

The events χ_1 and χ_2 are mutually exclusive.

Calculation with probabilities

Combining probabilities

- To compute the probability $P(\chi_1 \cup \chi_2)$ that either χ_1 or χ_2 or both occur we add the occurrence probabilities

$$P(\chi_1 \cup \chi_2) \leq P(\chi_1) + P(\chi_2) \quad (5)$$

Calculation with probabilities

Combining probabilities

- To compute the probability $P(\chi_1 \cup \chi_2)$ that either χ_1 or χ_2 or both occur we add the occurrence probabilities

$$P(\chi_1 \cup \chi_2) \leq P(\chi_1) + P(\chi_2) \quad (5)$$

- The ' \leq '-relation is correct since sample points might be contained in both events:

$$P(\chi_1 \cup \chi_2) = P(\chi_1) + P(\chi_2) - P(\chi_1 \cap \chi_2). \quad (6)$$

Example

Coin tosses







Question

What is the probability that in two coin tosses either head occurs first or tail occurs second ?




Example

Coin tosses

Events	coin tosses	probability
head - head		$\frac{1}{2} \cdot \frac{1}{2} = \frac{1}{4}$
head - tail		$\frac{1}{2} \cdot \frac{1}{2} = \frac{1}{4}$
tail - head		$\frac{1}{2} \cdot \frac{1}{2} = \frac{1}{4}$
tail - tail		$\frac{1}{2} \cdot \frac{1}{2} = \frac{1}{4}$

Example

Coin tosses

Events	coin tosses	probability	sum probability
head - head		$\frac{1}{2} \cdot \frac{1}{2} = \frac{1}{4}$	
head - tail		$\frac{1}{2} \cdot \frac{1}{2} = \frac{1}{4}$	
tail - tail		$\frac{1}{2} \cdot \frac{1}{2} = \frac{1}{4}$	

Calculation with probabilities

Conditional probability

Conditional probability

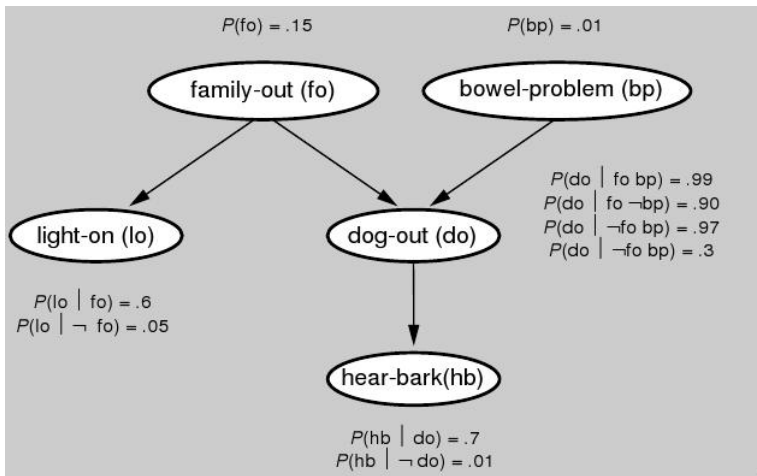
The conditional probability of two events χ_1 and χ_2 with $P(\chi_2) > 0$ is denoted by $P(\chi_1|\chi_2)$ and is calculated by

$$\frac{P(\chi_1 \cap \chi_2)}{P(\chi_2)} \quad (7)$$

$P(\chi_1|\chi_2)$ describes the probability that event χ_1 occurs in the presence of event χ_2 .

Example

Conditional probability



Calculation with probabilities

Bayes Rule

With rewriting and some simple algebra we obtain the Bayes rule:

Bayes Rule

$$P(\chi_1|\chi_2) = \frac{P(\chi_2|\chi_1) \cdot P(\chi_1)}{\sum_i P(\chi_2|\chi_i) \cdot P(\chi_i)}. \quad (8)$$

- This equation is useful in many statistical applications.
- With Bayes rule we can calculate $P(\chi_1|\chi_2)$ provided that we know $P(\chi_2|\chi_1)$ and $P(\chi_1)$.

Calculation with probabilities

Expectation

Expectation

The expectation of an event χ is defined as

$$E[\chi] = \sum_{x \in \mathbb{R}} x \cdot P(\chi = x) \quad (9)$$

Example

Expectation

Example

Consider the event χ of throwing a dice. The Sample space is given by $S_\chi = \{1, 2, 3, 4, 5, 6\}$.

What is the expectation of this event?

Example

Expectation

Example

Consider the event χ of throwing a dice. The Sample space is given by $S_\chi = \{1, 2, 3, 4, 5, 6\}$.

What is the expectation of this event?

- The expectation of this event is

$$E[\chi] = \frac{1}{6} \cdot (1 + 2 + 3 + 4 + 5 + 6) = 3.5 \quad (10)$$

Calculation with probabilities

Calculation with expectations

Linearity of expectation

For any two random variables χ_1 and χ_2 ,

$$E[\chi_1 + \chi_2] = E[\chi_1] + E[\chi_2]. \quad (11)$$

Multiplying expectations

For an independent random variables χ_1 and χ_2 ,

$$E[\chi_1 \cdot \chi_2] = E[\chi_1] \cdot E[\chi_2]. \quad (12)$$

Calculation with probabilities

Law of large numbers

Law of large numbers

Let $\{\bar{\chi}\}$ be a sequence of mutually independent random variables with a common distribution. If the expectation $E[\bar{\chi}]$ exists, then for every $\varepsilon > 0$ and $n \rightarrow \infty$

$$P \left\{ \left| \frac{\chi_1 + \dots + \chi_n}{n} - E[\bar{\chi}] \right| > \varepsilon \right\} \rightarrow 0 \quad (13)$$

- Probability that the average S_n/n will differ from expectation by less than ε tends to one.

Calculation with probabilities

Variance

Variance

The variance of a random variable χ is defined as

$$\text{var}[\chi] = E[(\chi - E[\chi])^2]. \quad (14)$$

Calculation with probabilities

Calculation with variance

Add variances

For any independent random variables χ_1 and χ_2

$$\text{var}[\chi_1 + \chi_2] = \text{var}[\chi_1] + \text{var}[\chi_2]. \quad (15)$$

Multiplying variances

For any random variable χ and any $c \in \mathbb{R}$,

$$\text{var}[c\chi] = c^2 \text{var}[\chi]. \quad (16)$$

The Markov inequality

Estimate the deviation of an event from its expectation

Markov inequality

Let (Π, P) be a probability space and $x : \Pi \rightarrow \mathbb{R}^+$ a non-negative random variable. For $t \in \mathbb{R}^*$ the following inequality holds:

$$P(x \geq t \cdot E[x]) \leq \frac{1}{t} \quad (17)$$

The Chernoff bound

Estimate the deviation of an event from its expectation

Chernoff bound

Let (Π, P) be a probability space and $x_1, x_2, \dots, x_n : \Pi \rightarrow \{0, 1\}$ independent random variables with $0 < P(x_i = 1) < 1$ for all $i \in \{1, 2, \dots, n\}$. For $X := \sum_{1 \leq i \leq n} x_i$ and $\delta > 0$ the following inequality holds:

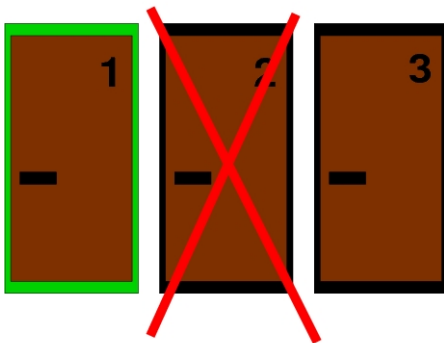
$$P(X < (1 + \delta)E[X]) < \left(\frac{e^\delta}{(1 + \delta)^{1+\delta}} \right)^{E[X]} \quad (18)$$

and for all δ with $0 < \delta < 1$

$$P(X < (1 - \delta)E[X]) < e^{-\frac{E[X]\delta^2}{2}} \quad (19)$$

Example

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