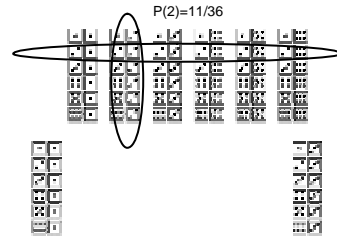


# Conditioned probability

# Intuition

- Time/Actions **change** the **sample space**.



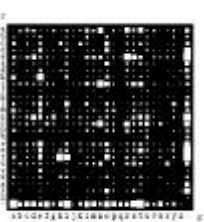
$$P(2)=11/36$$

$$P(2/\text{Second die is } 1)=1/36$$

$$P(2/\text{Second die is } 3)=1/36$$

# Intuition

- Time/Actions change the sample space.



i	$p_i$	$\omega_i$
1	0.0037	a
2	0.0028	b
3	0.0020	c
4	0.0025	d
5	0.0033	e
6	0.0173	f
7	0.0123	g
8	0.0123	h
9	0.0299	i
10	0.0096	j
11	0.0094	k
12	0.0125	l
13	0.0235	m
14	0.0296	n
15	0.0409	o
16	0.0352	p
17	0.0098	q
18	0.0296	r
19	0.0547	s
20	0.0296	t
21	0.0234	u
22	0.0069	v
23	0.0119	w
24	0.0073	x
25	0.0064	y
26	0.0007	z
27	0.0026	

$$P(k)=0.0084$$

$$P(k/y)=0$$

$$P(k/a)=0.001$$

<http://www.inference.phy.cam.ac.uk/mackay/ila/book.htm>

# Intuition

- Time/Actions change the sample space.

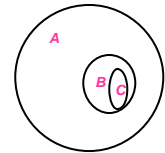
$$P(\Omega) = \text{Total Area} = 1$$

$$P(A) = \text{Area of } A$$

$$P(C) = \text{Area of } C$$

$$P(C/B) = \frac{\text{Area of } C}{\text{Area of } B}$$

$$P(B/C) = \text{Area of } C$$



# Formal definition

- Prob. of A conditioned to B is defined as

$$P(A/B) = \frac{P(A \cap B)}{P(B)}$$

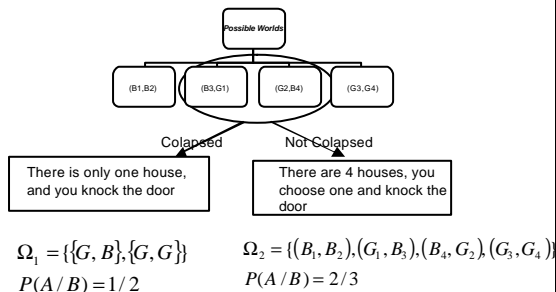
$$\text{Pr} = \frac{\text{Count of ways for a result}}{\text{Count of all possible results}}$$

# An Example: The sister problem

- You knock at the door of a family with two children, and a girl opens the door.
- Which is the probability that the other child is a boy?  $\Omega = \{(B, B), (G, B), (B, G), (G, G)\}$ 
  - A={one child is boy}
  - B={one child is girl}

$$P(A/B) = \frac{P(A \cap B)}{P(B)} = \frac{\left(\frac{2}{4}\right)}{\left(\frac{3}{4}\right)} = \frac{2}{3} \text{ Sure?}$$

## An Example: The sister problem Possible solutions



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## The sister problem Possible Worlds

### Logic:

**Kripke** In "Semantical Considerations on Modal Logic", published in 1963, Kripke responded to a difficulty with classical quantification theory. The motivation for the world-relative approach was to represent the possibility that objects in one world may fail to exist in another. If standard quantifier rules are used, however, every term must refer to something that exists in all the possible worlds. This seems incompatible with our ordinary practice of using terms to refer to things that only exist contingently.

[http://en.wikipedia.org/wiki/Possible\\_worlds](http://en.wikipedia.org/wiki/Possible_worlds)

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## An Example: The prisoner's dilemma (in prob. different from the game theory)

- Three prisoners A,B,C. One is going to be released, but they do not know the who.
- Prisoner A asks the guard the identity of one prisoner other than himself who **will not** be released.
- Guard:" your prob. of being released is now 1/3. If I tell you B, say, **will not** be released, then you would be one of only two prisoners whos fate is unknown and your probability of release would **increase** to 1/2. Since I don't want to hurt the chances of the other to be released I am not going to tell you"  
– Where is the mistake in the reasoning?

Taken from: Understanding Prob. Chance rules in everyday life. H.Tijms

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## An Example: The prisoner's dilemma (in prob. different from the game theory)

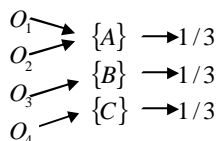
- Sample space  $\Omega = \{\{A\}, \{B\}, \{C\}\}$
- Initial Statement:  $P(\text{release A})=1/3$
- The guard seems to be saying:  
–  $P(\text{release A/guard says B released})=1/2$
- Either the voice of the guard can change probabilities or there is a different implicit sample space or independence assumption.

Taken from: Understanding Prob. Chance rules in everyday life. H.Tijms

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## An Example: The prisoner's dilemma (in prob. different from the game theory)

- If we incorporate the event  $E=\{\text{guard says B is released}\}$ , the new sampling space is:
  - $O_1=\{A, \text{guard says B is released}\}$
  - $O_2=\{A, \text{guard says C is released}\}$
  - $O_3=\{B, \text{guard says C is released}\}$
  - $O_4=\{C, \text{guard says B is released}\}$

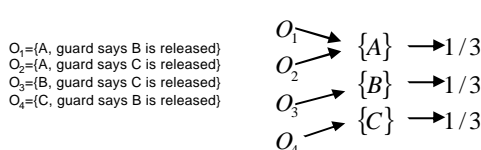


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## An Example: The prisoner's dilemma (in prob. different from the game theory)

- With:
  - $E_1=\{A \text{ is released}\}$
  - $E_2=\{\text{guard says B is released}\}$
$$P(E_1 / E_2) = \frac{P(E_1 \cap E_2)}{P(E_2)} = \frac{P(O_1)}{P(O_1) + P(O_4)} = \frac{1/6}{1/6 + 1/3} = 1/3$$



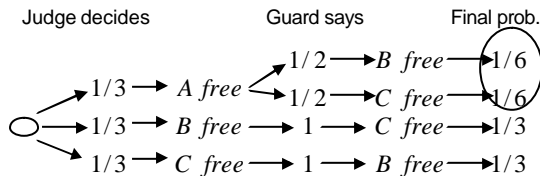
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## An Example: The prisoner's dilemma

(in prob. different from the game theory)

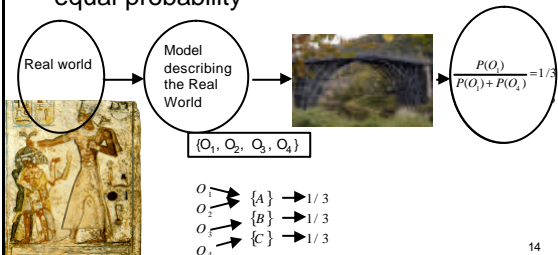
- Chance tree/Decision tree



Taken from: Understanding Prob. Chance rules in everyday life. H.Tijms 13

## Asumptions on $\{O_1, O_2, O_3, O_4\}$

- Asumption that the original events have equal probability



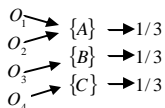
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## An Example: The prisoner's dilemma

(in prob. different from the game theory)

- What would happen if:
  - $O_1 = \{A, \text{guard says B is released}\}$
  - $O_2 = \{A, \text{guard says C is released}\}$
  - $O_3 = \{B, \text{guard says B is released}\}$
  - $O_4 = \{C, \text{guard says B is released}\}$
- $E_1 = \{A \text{ is released}\}$
- $E_2 = \{\text{guard says B is released}\}$

$$P(E_1 / E_2) = \frac{P(E_1 \cap E_2)}{P(E_2)} = \frac{P(O_1)}{P(O_1) + P(O_2) + P(O_4)} = \frac{1/6}{1/6 + 1/3 + 1/3} = 1/5$$



- Mathematical correct
- Independence assumption?

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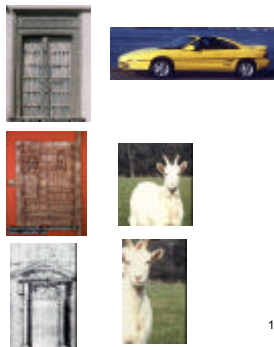
## An Example: The Monty Hall dilemma

- The contestant in a television show must choose between three doors. An expensive car is behind one door and gag prizes await behind the other two.
- The contestant must pick one of the doors randomly.
- The host opens one of the other two doors concealing one of the gag prizes.
- The contestant is asked whether he wishes to switch to the remaining door.
- *Is the prob. of winning increased by changing the chosen door?*

Taken from: Understanding Prob. Chance rules in everyday life. H.Tijms 16

## An Example: The Monty Hall dilemma

Select randomly door



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## An Example: The Monty Hall dilemma

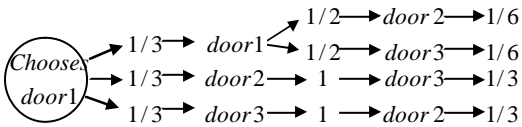
- Wrong argument:
  - It makes no difference whether the player switched doors or not. Each of the two remaining unopened doors had a  $1/2$  prob. of concealing the automobile.
  - If it is not in the opened door, it is behind either, so therefore the prob. is of  $1/2$ 
    - **Caveat** is it really random?, Note that the host has chosen with knowledge.
    - Is the the same sample space?

Taken from: Understanding Prob. Chance rules in everyday life. H.Tijms 18

## An Example: The Monty Hall dilemma

- Chance tree/Decision tree

The car is behind                      host opens                      Final prob.

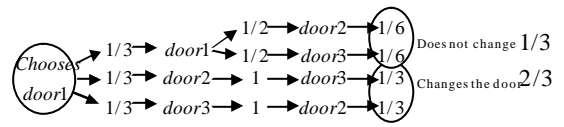


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## An Example: The Monty Hall dilemma

- Probabilities

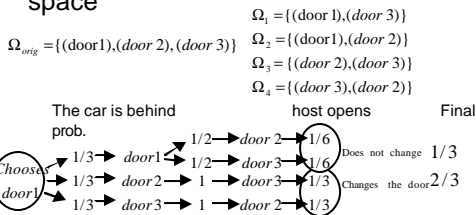
The car is behind                      host opens                      Final prob.



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## An Example: The Monty Hall dilemma

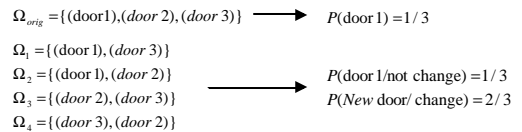
- Note that conditioning changes the sample space



Taken from: Understanding Prob. Chance rules in everyday life. H.Tijms 21

## An Example: The Monty Hall dilemma

- Note that conditioning changes the sample space



Taken from: Understanding Prob. Chance rules in everyday life. H.Tijms 22

## Computation of intersecion probability

- *Temporal structure*: B takes place and afterwards A

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

- Generalization for a sample space

$$\Omega = \{A_1, A_2, \dots, A_n\}$$

$$P(A_1 \cap A_2 \cap \dots \cap A_n)$$

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## Computation of intersecion probability

- Generalization for a sample space

$$\Omega = \{A_1, A_2, \dots, A_n\}$$

$$P(A_1 \cap A_2 \cap \dots \cap A_n) = P(A_1 \cap (A_2 \cap \dots \cap A_n)) = P(A_1)P(A_2 \cap \dots \cap A_n | A_1)$$

$$P(A \cap B) = P(A)P(B/A)$$

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## Computation of intersection probability

- Generalization for a sample space

$$\Omega = \{A_1, A_2, \dots, A_n\}$$

$$P(A_1 \cap (A_2 \cap \dots \cap A_n)) =$$

$$= P(A_1)P(A_2 \cap \dots \cap A_n / A_1)$$

$$P(A_2 \cap (A_3 \cap \dots \cap A_n) / A_1) =$$

$$= P(A_2 / A_1)P(A_3 \cap \dots \cap A_n / A_1 \cap A_2)$$

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## Computation of intersection probability

- Generalization for a sample space

$$\Omega = \{A_1, A_2, \dots, A_n\}$$

$$P(A_1 \cap A_2 \cap \dots \cap A_n) =$$

$$= P(A_1)P(A_2 / A_1)P(A_3 / A_1 \cap A_2) \dots P(A_n / A_1 \cap \dots \cap A_{n-1})$$

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## Computation of intersection probability

- Examples

- Birthday problem

Sample space  $\Omega = \left\{ \begin{array}{l} A_i / i = 1 \dots n; \\ \text{the day of the birthday of individual } i\text{-th} \\ \text{which is different from all the others.} \end{array} \right\}$

Probability of Event =  $\{A_1 \cap A_2 \cap \dots \cap A_n\}$

↓

Probability of the simultaneous occurrence of the atomic events  $A_i \quad i = 1, \dots, n$

$P(A_2 / A_1) \rightarrow$  Probability of the birthday in day

$A_2$  when the day  $A_1$  is occupied

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## Computation of intersection probability

- Examples

- Birthday problem

$$P(A_1 \cap A_2 \cap \dots \cap A_n) =$$

$$= P(A_1)P(A_2 / A_1)P(A_3 / A_1 \cap A_2) \dots P(A_n / A_1 \cap \dots \cap A_{n-1})$$

$$Pr = \left( \frac{N}{N} \right) \left( \frac{N-1}{N} \right) \left( \frac{N-2}{N} \right) \dots \left( \frac{N-n+1}{N} \right)$$

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## Computation of intersection probability

- Examples: Grammars for computing the probability of a correct sentence

$$P(w_1 \cap w_2 \cap \dots \cap w_n) =$$

$$= P(w_1)P(w_2 / w_1)P(w_3 / w_1 \cap w_2) \dots P(w_n / w_1 \cap \dots \cap w_{n-1})$$

$$= P(w_1)P(w_2 / w_1)P(w_3 / w_1 \cap w_2)P(w_4 / w_2 \cap w_3) \dots P(w_n / w_{n-2} \cap w_{n-1})$$

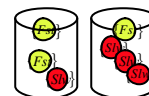


To be, or not to be—that is the question;  
Whether 'tis nobler in the mind to suffer  
The slings and arrows of outrageous  
fortune,  
Or to take arms against a sea of troubles,  
And by opposing end them? To die, to  
sleep  
Microsoft® Encarta®

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## Example

- We can connect to two servers:
  - S1 has **2 high** speed links and **1 slow** link
  - S2 has **1 high** speed link and **3 slow** links
- We select one server at random
- Which is the probability of getting a slow link?



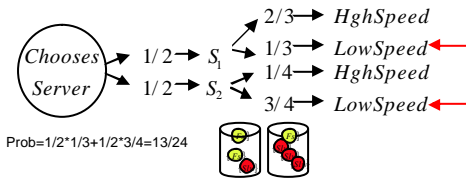
30

## Example

- Sample space:

$$\Omega = \{S_1Fst, S_1Slw, S_2Fst, S_2Slw\}$$

– **Event:** Select Server and then type of link



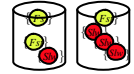
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## Example

- Probabilities:

$$P(S_1) = P(S_2) = 1/2$$

$$P(Slw / S_1) = \frac{1}{3} = \frac{P(Slw \cap S_1)}{P(S_1)}$$



$$P(Slw / S_2) = \frac{3}{4} = \frac{P(Slw \cap S_2)}{P(S_2)}$$

$$P(Slw) = P(S_2)P(Slw / S_2) + P(S_1)P(Slw / S_1)$$

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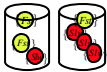
## Example

- Probabilities: Another way

$$P(Slw) = P((Slw \cap S_1) \cup (Slw \cap S_2))$$

$$P(Slw) = P(Slw \cap S_1) + P(Slw \cap S_2)$$

$$P(Slw) = P(S_2)P(Slw / S_2) + P(S_1)P(Slw / S_1)$$

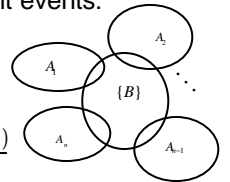


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## Properties

- If we have a set of disjoint events:

$$S = \{A_1, A_2, \dots, A_n\}$$



$$\begin{aligned} P(A_1 \cup A_2 \cup \dots \cup A_n / B) &= \\ &= \frac{P((A_1 \cap B) \cup (A_2 \cap B) \cup \dots \cup (A_n \cap B))}{P(B)} \\ &= \frac{P(A_1 \cap B) + P(A_2 \cap B) + \dots + P(A_n \cap B)}{P(B)} \end{aligned}$$

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## Properties

- The universe conditioned to a given event has prob. one:

$$P(\Omega / B) = \frac{P(\Omega \cap B)}{P(B)} = 1$$

- Both prob. spaces will have the same properties

$$(\Omega, P(\cdot / B)) \quad (B, P(\cdot / B))$$

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## Properties

- Total probabilities:

– Given a set of events  $S = \{B_1, B_2, \dots, B_n\}$  pairwise disjoint or mutually disjoint<sup>1</sup>, such that

$$B_1 \cup B_2 \cup \dots \cup B_n = \Omega$$

– We have:

$$P(A) = P(B_1)P(A / B_1) + P(B_2)P(A / B_2) + \dots + P(B_n)P(A / B_n)$$

<sup>1</sup> - Mutually Disjoint: for all (i, j) such that  $i \neq j$   
 $B_i \cap B_j = \emptyset$

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## Properties

- Total probabilities:

– Proof:

$$A = A \cap \Omega = A \cap (B_1 \cup B_2 \cup \dots \cup B_n) =$$

$$= (A \cap B_1) \cup (A \cap B_2) \cup \dots \cup (A \cap B_n)$$

$$P(A) = P(A \cap B_1) \cup P(A \cap B_2) \cup \dots \cup P(A \cap B_n)$$

$$P(A) = P(A \cap B_1) + P(A \cap B_2) + \dots + P(A \cap B_n)$$

but  $P(A \cap B_i) = P(B_i)P(A / B_i)$

$$P(A) = P(B_1)P(A / B_1) + P(B_2)P(A / B_2) + \dots + P(B_n)P(A / B_n)$$

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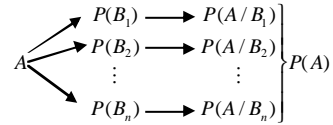
## Properties

- Total probabilities:

– Temporal/Spatial interpretation

$$P(A) = P(B_1)P(A / B_1) + P(B_2)P(A / B_2) + \dots + P(B_n)P(A / B_n)$$

– When  $B_i$  happens, the odds of A change



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## Example

- A Clinical analysis is used for the diagnosis of three illness,  $B_1, B_2, B_3$ .
- The proportion of people with a given illness is: 3%, 2%, 1%
- The analysis gives
 

$B_1 \rightarrow$	85%
positive result for: $B_2 \rightarrow$	92%
$B_3 \rightarrow$	78%
$B_0 \rightarrow$	0.5%
- Compute the prob. of a positive.

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## Example

- We define prob. of a positive as  $P(A)$

$$P(A) = P(B_1)P(A / B_1) + P(B_2)P(A / B_2) + P(B_3)P(A / B_3) + P(B_0)P(A / B_0)$$

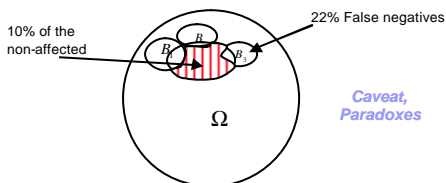
$$= 0.03 * 0.85 + 0.02 * 0.92 + 0.01 * 0.78 + 0.94 * 0.005 = 0.0564$$

- **5,6% instead of 6% not bad** ☺
- What would happen with  $P(A / B_0) = 0.1$  ?

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## Example

- What would happen with  $P(A / B_0) = 0.1$  ?
- $$P(A) = P(B_1)P(A / B_1) + P(B_2)P(A / B_2) + P(B_3)P(A / B_3) + P(B_0)P(A / B_0)$$
- $$= 0.03 * 0.85 + 0.02 * 0.92 + 0.01 * 0.78 + 0.94 * 0.1 = 0.146$$
- Why 14.6%? **Only 6% should give positive!!!!**



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## Bayes's formula

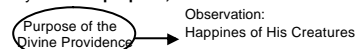


- Intuitively not clear (?)

$$P(B_i / A) = \frac{P(B_i)P(A / B_i)}{P(B_1)P(A / B_1) + P(B_2)P(A / B_2) + \dots + P(B_n)P(A / B_n)}$$

- Context:

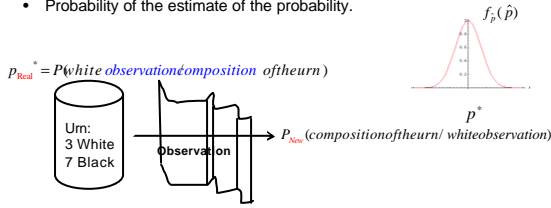
- Solution to a problem of "inverse probability" was presented in the *Essay Towards Solving a Problem in the Doctrine of Chances*
- Published *Divine Benevolence, or an Attempt to Prove That the Principal End of the Divine Providence and Government is the Happiness of His Creatures* (?)
- **End** is synonym of **purpose, aim**.



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## Kinds of probability

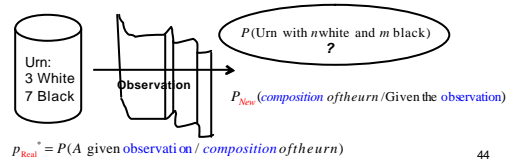
- Probability of an observation
- **Probability of the cause of the observation**
- Probability of the estimate of the probability.



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## Bayes Formula

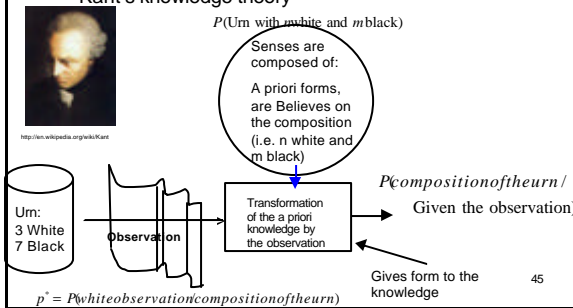
- Motivation:
  - We would like the **probability of the causes** that generate the observations.
  - **Update** our knowledge after the observations.



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## Bayes Formula

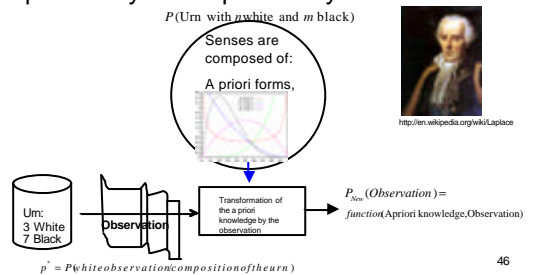
- Context of the Bayes formula:
  - Kant's knowledge theory



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## Bayes Formula

- Laplace proposes taking into account the probability of the probability:



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## Bayes' formula

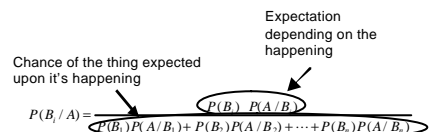
- Bayes' formula:
  - Given a set of events  $S = \{B_1, B_2, \dots, B_n\}$  pairwise disjoint or mutually disjoint<sup>1</sup>, such that  $B_1 \cup B_2 \cup \dots \cup B_n = \Omega$
  - We have:  $P(B_i / A) = \frac{P(B_i \cap A)}{P(A)} = \frac{P(B_i)P(A / B_i)}{P(A)}$

$$P(B_i / A) = \frac{P(B_i)P(A / B_i)}{P(B_1)P(A / B_1) + P(B_2)P(A / B_2) + \dots + P(B_n)P(A / B_n)}$$

1 - Mutually Disjoint : for all (i, j) such that  $i, j \neq j$   
 $B_i \cap B_j = \emptyset$

## Bayes's formula

- Context:
  - Bayes defines *probability* as follows :
    - The probability of any event is the ratio between the **value at which an expectation** depending on the happening of the event ought to be computed, and the **chance of the thing expected** upon it's happening



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## Bayes's formula



- Expectation
  - **1. anticipation of something happening:** a confident belief or strong hope that a particular event will happen
  - **2. notion of something:** a mental image of something expected, often compared to its reality (*often used in the plural*)
  - **3. expected standard:** a standard of conduct or performance expected by or of somebody (*often used in the plural*)
  - Microsoft® Encarta® Reference Library 2003. © 1993-2002 Microsoft Corporation. All rights reserved.

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## Bayes's formula and *Utility function*



- Idea of utility
    - Expectation is a subjective concept.
    - Instead of the probabilistic expectation, we could think in the **'moral expectation'**
- Value → Chance of the thing × Moral expectation depending on the happening

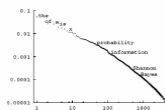
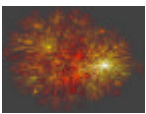
$$Value(B_i / A) \propto P(B_i) \text{ Utility}(A / B_i)$$

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## Bayes's formula and *Utility function*



- Usefulness of the Idea of utility
  - People behave as if they were maximizing an Utility function, i.e. a **'moral expectation'**
  - Objects generated by the independent and free actions can be understood by means of the utility theory.
  - Example:  $Value(B_i / A) \propto P(B_i) \text{ Utility}(A / B_i)$



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## Application to the clinical analysis problem

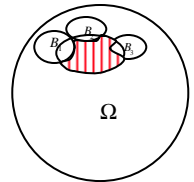
- Probability of having the illness given a positive result (**A**).

$$P(B_1 / A) = \frac{P(B_1)P(A / B_1)}{P(A)} = \frac{0.03 * 0.85}{0.0564} = 0.452$$

$$P(B_2 / A) = \frac{P(B_2)P(A / B_2)}{P(A)} = \frac{0.02 * 0.92}{0.0564} = 0.326$$

$$P(B_3 / A) = \frac{P(B_3)P(A / B_3)}{P(A)} = \frac{0.01 * 0.78}{0.0564} = 0.138$$

$$P(B_0 / A) = \frac{P(B_0)P(A / B_0)}{P(A)} = \frac{0.94 * 0.05}{0.0564} = 0.083$$



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## Application to the clinical analysis problem

- Probability of having the illness given a positive result (**A**).

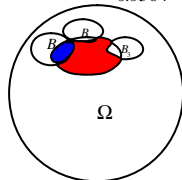
• Meaning:  $\frac{P(B_1 \cap A)}{P(A)} = \frac{0.03 * 0.85}{0.0564} = \frac{0.0255}{0.0564}$

$$P(B_1 / A) \rightarrow 45\%$$

$$P(B_2 / A) \rightarrow 32\%$$

$$P(B_3 / A) \rightarrow 13\%$$

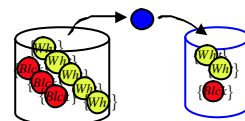
$$P(B_0 / A) \rightarrow 8.3\% \quad \text{False positive}$$



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## Bayes and chance trees

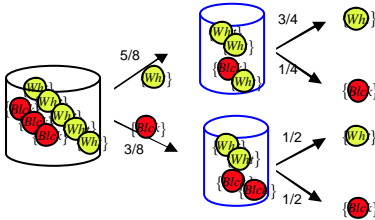
- Urn has 5 white balls and 3 black. One ball is taken randomly and introduced in another with 2 white and 1 black.
- A ball **white** is taken from the second urn.
- What is the probability that the first ball was black?.



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## Bayes and chance trees

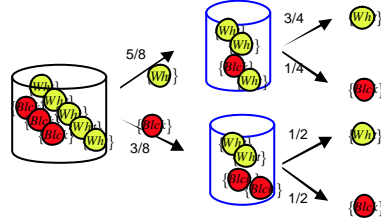
- Urn has 5 white balls and 3 black. One ball is taken randomly and introduced in another with 2 white and 1 black.
- A ball **white** is taken from the second urn.
- What is the **probability that the first ball was black?**



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## Bayes and chance trees

- What is the **probability that the first ball was black?**



$$P(B_1 / W_2) = \frac{P(B_1)P(W_2 / B_1)}{P(B_1)P(W_2 / B_1) + P(W_1)P(W_2 / W_1)} = \frac{3/8 * 1/2}{3/8 * 1/2 + 5/8 * 3/4}$$

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## $P(A/B) \neq P(B/A)$

- Speaks spanish/spanish citizenship

$$P(\text{speaks spanish} / \text{spanish citizenship}) = \frac{40 \text{million}}{40 \text{million}} = 1$$

$$P(\text{spanish citizenship} / \text{speaks spanish}) = \frac{40 \text{million}}{500 \text{million}} = \frac{4}{5}$$

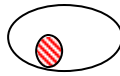
- Kasparov/winning in chess

$$P(\text{Kasparov winning in chess}) = \frac{1}{1000 \text{million}} \approx 0$$

$$P(\text{winning in chess} / \text{Kasparov}) = 1$$

- A more difficult one: OJ Simpson:

- P(being killed / a beating husband)
- P(a beating husband / being killed)



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## Bayes and "the prosecutor's fallacy".

- The prosecutor's fallacy

- The prosecutor's fallacy is the assertion that, because the story before the court is highly improbable, the defendant's innocence is equally improbable.

- OJ Simpson:

- Prob. Having a beating husband **given** that a woman has been killed -> 1/10.000

- Many more causes of death: accidents, age, illness, husband, etc

- Prob. of being killed **given** a beating husband -> 1/100

- If we exclude other causes, the conclusion

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## $P(\text{cancer/positive test}) \neq P(\text{positive test/cancer})$

- Problem of concept:
  - P(cancer/positive) -> Attribute of the patient
  - P(positive/cancer) -> Attribute of the test.
- What we want to know: (either)
  - **Effectiveness** of the treatment given how the patients fare
  - **Diagnosis** of the patient given the result of the test
- **Diagnosis** problem: P(cancer/positive test)
  - We know:
    - Clinical study: P(positive test/cancer)=0.9
    - Statistics of the population: P(cancer)=10/1000
  - Note: could be the people that go the hospital

Taken from: Understanding Prob. Chance rules in everyday life. H. Tijms

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## $P(\text{cancer/positive test}) \neq P(\text{positive test/cancer})$

- **Diagnosis** problem: P(cancer/positive)

- We know:

- Clinical study: P(positive test/cancer)=0.9
- Statistics of the population: P(cancer)=10/1000

We define  $\rightarrow \begin{cases} C: \text{Patient is ill of cancer} \\ P: \text{Positive result in the test} \end{cases}$

$$P(C/P) = \frac{P(C)P(P/C)}{P(C)P(P/C) + P(\bar{C})P(P/\bar{C})} = \frac{\frac{10}{1000} \cdot 0.90}{\frac{10}{1000} \cdot 0.90 + \frac{990}{1000} \cdot 0.10} = 0.0833$$

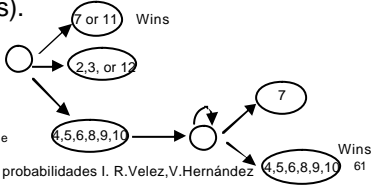
Note:

- 90% vs. 8.3%
- 1% is increased to 8,3%

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## Game of Craps

- Dice game. The player throws two dice; if the sum is 7 or 11 wins, if it is 2,3, or 12 loses. If it has another result, continues until a 7 (loses) or the result of the first throw (wins).



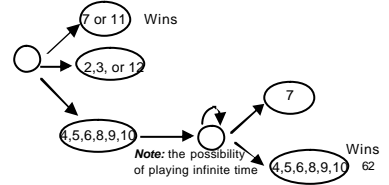
Note: the possibility of playing infinite time

Taken from: Cálculo de probabilidades I. R.Velez,V.Hernández

61

## Game of Craps

- Justification of 28 slides for a problem:
  - Indefinite duration
  - Geometric series
  - Conditioned probability and Bayes in different ways.

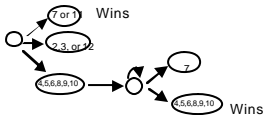


Note: the possibility of playing infinite time

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## Game of Craps

- Compute:
  - Probability of winning
  - Probability of getting a 5 in the first throw if it is known that the player has won.
    - Note: inverse probability, which is the cause.
  - The probability that the player wins if there have been k throws.

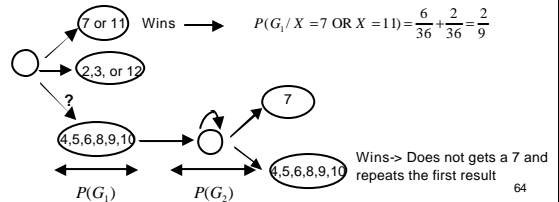


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## Game of Craps

- Probability of winning :

2	3	4	5	6	7	8	9	10	11	12
1/36	2/36	3/36	4/36	5/36	6/36	5/36	4/36	3/36	2/36	1/36



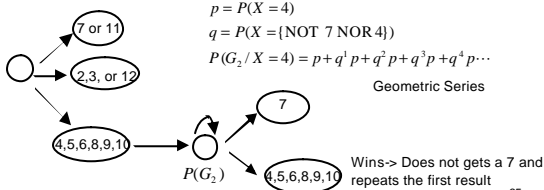
$$P(G_1 / X = 7 \text{ OR } X = 11) = \frac{6}{36} + \frac{2}{36} = \frac{2}{9}$$

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## Game of Craps

- Probability of winning :

2	3	4	5	6	7	8	9	10	11	12
1/36	2/36	3/36	4/36	5/36	6/36	5/36	4/36	3/36	2/36	1/36



$$p = P(X = 4)$$

$$q = P(X = \{\text{NOT } 7 \text{ NOR } 4\})$$

$$P(G_2 / X = 4) = p + q^1 p + q^2 p + q^3 p + q^4 p \dots$$

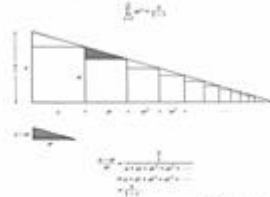
Geometric Series

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## Game of Craps

- How do we sum an infinite geometric series?  $S = p + q^1 p + q^2 p + q^3 p + q^4 p \dots = \frac{p}{1-q}$

Real world World



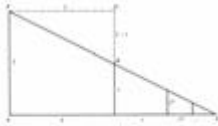
1. de Matemática  
Universidade de São Paulo

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## Game of Craps

- How do we sum an infinite geometric series?  $S = p + q^1 p + q^2 p + q^3 p + q^4 p \dots = \frac{p}{1-q}$

Proof Without Words



$p = q^1 + q^2 + q^3 + \dots$

Michael J. Steele  
New York University  
Department of Mathematics  
New York, NY 10003

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## Game of Craps

- How do we sum a **finite** geometric series?  $S = p + q^1 p + q^2 p + q^3 p \dots + q^{N-1} p = p \frac{1-q^N}{1-q}$

Proof without Words



$p = q^1 + q^2 + q^3 + \dots + q^{N-1}$

J. H. Stone  
University of Cape Town

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## Game of Craps

- How do we sum an **finite** geometric series?

- Analytical Solution

$$S = 1 + q^1 + q^2 + q^3 \dots + q^{N-1}$$

$$S = 1 + q^1 + q^2 + q^3 \dots + q^{N-1} + (q^N - q^N)$$

$$S = 1 + q^1 (1 + q^1 + q^2 + q^3 \dots + q^{N-1}) - q^N$$

$$S = 1 + qS - q^N$$

$$S = \frac{1 - q^N}{1 - q}$$

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## Game of Craps

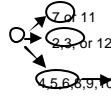
- We can sum  $P(G_2 / X = 4)$

2	3	4	5	6	7	8	9	10	11	12
1/36	2/36	3/36	4/36	5/36	6/36	5/36	4/36	3/36	2/36	1/36

$$p = P(X = 4) = \frac{3}{36}$$

$$q = P(X = \{\text{NOT } 7 \text{ NOR } 4\}) = \frac{27}{36}$$

$$P(G_2 / X = 4) = p \frac{1 - q^N}{1 - q} = \frac{3}{36} \frac{1 - \frac{27}{36}}{1 - \frac{27}{36}} = \frac{1}{3}$$



Wins -> Does not get a 7 and repeats the first result

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## Game of Craps

- The other cases

2	3	4	5	6	7	8	9	10	11	12
1/36	2/36	3/36	4/36	5/36	6/36	5/36	4/36	3/36	2/36	1/36

$$p = P(X = 5) = \frac{4}{36}$$

$$q = P(X = \{\text{NOT } 7 \text{ NOR } 5\}) = \frac{26}{36}$$

$$P(G_2 / X = 5) = p \frac{1 - q^N}{1 - q} = \frac{4}{36} \frac{1 - \frac{26}{36}}{1 - \frac{26}{36}} = \frac{2}{5}$$

$$p = P(X = 6) = \frac{5}{36}$$

$$q = P(X = \{\text{NOT } 7 \text{ NOR } 6\}) = \frac{25}{36}$$

$$P(G_2 / X = 6) = p \frac{1 - q^N}{1 - q} = \frac{5}{36} \frac{1 - \frac{25}{36}}{1 - \frac{25}{36}} = \frac{5}{11}$$

Note that by symmetry:

$$P(G_2 / X = 4) = P(G_2 / X = 10)$$

$$P(G_2 / X = 5) = P(G_2 / X = 9)$$

$$P(G_2 / X = 6) = P(G_2 / X = 8)$$

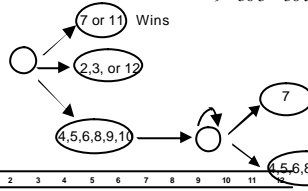
71

## Game of Craps

- Finally the probability of winning is:

$$P(G) = P(G_1 / X = 7 \text{ OR } 11) + P(G_1 / X = 4 \text{ OR } 5 \text{ OR } 6 \text{ OR } 8 \text{ OR } 9 \text{ OR } 10)$$

$$P(G) = \frac{2}{9} + \frac{3}{36} \frac{1}{3} + \frac{4}{36} \frac{2}{5} + \frac{5}{36} \frac{5}{11} + \frac{4}{36} \frac{2}{5} + \frac{3}{36} \frac{1}{3} \cong 0.493$$



Wins -> Does not get a 7 and repeats the first result

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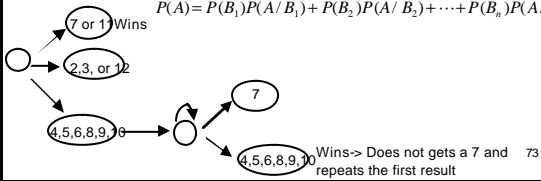
## Game of Craps

- Finally the probability of winning is:
  - Note that we have used the total probability

$$P(G) = \frac{2}{9} + \frac{3}{363} + \frac{4}{365} + \frac{5}{3611} + \frac{4}{365} + \frac{3}{363} = 0.493$$

Total Probability

$$P(A) = P(B_1)P(A/B_1) + P(B_2)P(A/B_2) + \dots + P(B_n)P(A/B_n)$$

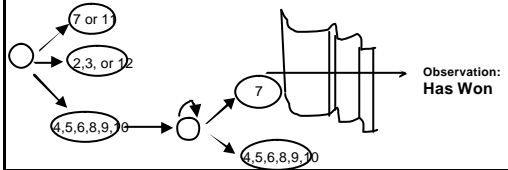


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## Game of Craps

- Probability of getting a 5 in the first throw *if* it is **known** that the player has **won**.
  - Note: inverse probability, which is the cause?

$$P(\{X=5\}/G) = \frac{P(\{X=5\})P(G/\{X=5\})}{P(G)} = \frac{\frac{4}{36} \cdot \frac{2}{5}}{0.493} = \frac{11}{122}$$



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## Game of Craps

- The probability that the player wins if there have been  $k$  throws.
- The condition is that the player wins and at least there have been  $N > k$  rounds:  $P(G/\{N > k\})$ 
  - Note: that the temporal structure depends on getting in the first throw one element of the set = {4, 5, 6, 8, 9, 10}

$$P(G/\{N > k\}) \longrightarrow \begin{cases} P(N > k / X = 4) \\ P(N > k / X = 5) \\ P(N > k / X = 6) \\ P(N > k / X = 8) \\ P(N > k / X = 9) \\ P(N > k / X = 10) \end{cases}$$

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## Game of Craps

- The probability that the player wins if there have been  $k$  throws.
- Case:  $P(N > k / X = 4)$ 
  - Note: the condition for playing  $N > k$  rounds given that the first sum was 4 is:

$$\{G \text{ et } 4\} \text{ AND } \underbrace{\{\text{NOT } 7 \text{ NOR } 4\} \text{ AND } \dots \text{ AND } \{\text{NOT } 7 \text{ NOR } 4\}}_{k-1 \text{ times}} \text{ AND } \{\text{anything else}\}$$

$$P(N > k / X = 4) = \frac{P(X = 4)P(X = \{\text{NOT } 7 \text{ NOR } 4\})^{k-1}P(\text{anything else})}{P(X = 4)}$$

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## Game of Craps

- The probability that the player wins if there have been  $k$  throws.

$$P(N > k / X = 4) = \frac{P(X = 4)P(X = \{\text{NOT } 7 \text{ NOR } 4\})^{k-1}P(\text{anything else})}{P(X = 4)}$$

$$P(X = \{\text{NOT } 7 \text{ NOR } 4\}) = \frac{27}{36}$$

$$P(N > k / X = 4) = \left(\frac{27}{36}\right)^{k-1} = P(N > k / X = 10)$$

2	3	4	5	6	7	8	9	10	11	12
1/36	2/36	3/36	4/36	5/36	6/36	5/36	4/36	3/36	2/36	1/36

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## Game of Craps

- The probability that the player wins if there have been  $k$  throws.
  - For the rest of the set = {4, 5, 6, 8, 9, 10}

$$P(N > k / X = 4) = \left(\frac{27}{36}\right)^{k-1} = P(N > k / X = 10)$$

$$P(N > k / X = 5) = \left(\frac{26}{36}\right)^{k-1} = P(N > k / X = 9)$$

$$P(N > k / X = 6) = \left(\frac{25}{36}\right)^{k-1} = P(N > k / X = 8)$$

2	3	4	5	6	7	8	9	10	11	12
1/36	2/36	3/36	4/36	5/36	6/36	5/36	4/36	3/36	2/36	1/36

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## Game of Craps

- The probability that the player wins *if there have been k* throws.
  - We will compute the probability of  $N > k$  which we will need for **conditioning**.

$$P(N > k) = (P(X=4)P(N > k | X=4) + P(X=5)P(N > k | X=5) + P(X=6)P(N > k | X=6)) * 2$$

$$P(N > k) = \frac{6}{36} \left(\frac{27}{36}\right)^{k-1} + \frac{8}{36} \left(\frac{26}{36}\right)^{k-1} + \frac{10}{36} \left(\frac{25}{36}\right)^{k-1}$$

Symetry of the prob. of the set={4,5,6,8,9,10}

2	3	4	5	6	7	8	9	10	11	12
1/36	2/36	3/36	4/36	5/36	6/36	5/36	4/36	3/36	2/36	1/36

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## Game of Craps

- The probability that the player wins *if there have been k* throws.

$$\{N > k\} \cap G \text{ given that } \{X=4\} \cup \{X=10\}$$

For  $i = k + 1$  to infinite happens either of the following events

$$\{Get a 4 \text{ OR } 10\} \text{ AND } \{ \text{NOT } 7 \text{ NOR } (4, 10) \} \text{ AND } \dots \text{ AND } \{ \text{NOT } 7 \text{ NOR } (4, 10) \} \text{ AND } \{win\}$$

$i - 2$  times  
 $i \geq k + 1$

$$P(\{N > k\} \cap G / \{X=4\} \cup \{X=10\}) = \sum_{i=k+1}^{\infty} \frac{6}{36} \left(\frac{27}{36}\right)^{i-2}$$

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## Game of Craps

- The probability that the player wins *if there have been k* throws.

For  $i = k + 1$  to infinite happens either of the following events

$$\{Get a 4 \text{ OR } 10\} \text{ AND } \{ \text{NOT } 7 \text{ NOR } (4, 10) \} \text{ AND } \dots \text{ AND } \{ \text{NOT } 7 \text{ NOR } (4, 10) \} \text{ AND } \{win\}$$

$i - 2$  times  
 $i \geq k + 1$

$$P(\{N > k\} \cap G / \{X=4\} \cup \{X=10\}) = \sum_{i=k+1}^{\infty} \frac{6}{36} \left(\frac{27}{36}\right)^{i-2}$$

$$\frac{3}{36} + \frac{3}{36} = \frac{6}{36}$$

GeometriSeries

2	3	4	5	6	7	8	9	10	11	12
1/36	2/36	3/36	4/36	5/36	6/36	5/36	4/36	3/36	2/36	1/36

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## Game of Craps

- Sum of the geometric series

$$\frac{6}{36} \sum_{i=k+1}^{\infty} \left(\frac{27}{36}\right)^{i-2} = \frac{6}{36} \sum_{i=k+1}^{\infty} \left(\frac{27}{36}\right)^{i+k-1} = \frac{6}{36} \left(\frac{27}{36}\right)^{k-1} \sum_{i=0}^{\infty} \left(\frac{27}{36}\right)^i$$

$$= \frac{6}{36} \left(\frac{27}{36}\right)^{k-1} \frac{1}{1 - \frac{27}{36}} = \frac{6}{36} \left(\frac{27}{36}\right)^{k-1} \frac{36}{36-27} = \frac{1}{3} \left(\frac{27}{36}\right)^{k-1}$$

Criterion for selecting the change of variables:

Transform the original sum into the known one  $S = p \sum_{n=0}^{\infty} q^n = \frac{p}{1-q}$

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## Game of Craps

- The probability that the player wins *if there have been k* throws.

$$P(\{N > k\} \cap G / \{X=4\} \cup \{X=10\}) = \sum_{i=k+1}^{\infty} \frac{3}{36} \left(\frac{27}{36}\right)^{i-2} = \frac{1}{3} \left(\frac{27}{36}\right)^{k-1}$$

$$P(\{N > k\} \cap G / \{X=5\} \cup \{X=9\}) = \sum_{i=k+1}^{\infty} \frac{4}{36} \left(\frac{26}{36}\right)^{i-2} = \frac{4}{10} \left(\frac{26}{36}\right)^{k-1}$$

$$P(\{N > k\} \cap G / \{X=6\} \cup \{X=8\}) = \sum_{i=k+1}^{\infty} \frac{5}{36} \left(\frac{25}{36}\right)^{i-2} = \frac{5}{11} \left(\frac{25}{36}\right)^{k-1}$$

2	3	4	5	6	7	8	9	10	11	12
1/36	2/36	3/36	4/36	5/36	6/36	5/36	4/36	3/36	2/36	1/36

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## Game of Craps

- The probability that the player wins *if there have been k* throws.

$$P(\{N > k\} \cap G) = \frac{6}{36} \frac{1}{3} \left(\frac{27}{36}\right)^{k-1} + \frac{8}{3610} \left(\frac{26}{36}\right)^{k-1} + \frac{10}{3611} \left(\frac{25}{36}\right)^{k-1}$$

Totalprobabilij

$$P(A) = P(B_1)P(A/B_1) + P(B_2)P(A/B_2) + P(B_3)P(A/B_3)$$

$$P(\{N > k\} \cap G / \{X=4\} \cup \{X=10\})$$

$$P(\{N > k\} \cap G / \{X=5\} \cup \{X=9\})$$

$$P(\{N > k\} \cap G / \{X=6\} \cup \{X=8\})$$

2	3	4	5	6	7	8	9	10	11	12
1/36	2/36	3/36	4/36	5/36	6/36	5/36	4/36	3/36	2/36	1/36

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# Game of Craps

- The probability that the player wins **if there have been k** throws.
  - Final Probability

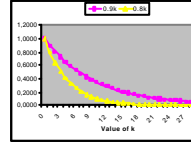
$$P(G/\{N > k\}) = \frac{P(\{N > k\} \cap G)}{P(N > k)} = \frac{\frac{6}{36} \left(\frac{27}{36}\right)^{k-1} + \frac{8}{36} \left(\frac{26}{36}\right)^{k-1} + \frac{10}{36} \left(\frac{25}{36}\right)^{k-1}}{\frac{6}{36} \left(\frac{27}{36}\right)^{k-1} + \frac{8}{36} \left(\frac{26}{36}\right)^{k-1} + \frac{10}{36} \left(\frac{25}{36}\right)^{k-1}}$$

# Game of Craps

- The probability that the player wins **if there have been k** throws.
  - What happens when k->infinite

Value of k	d.p	d.p
0	1.0000	1.0000
1	0.9028	0.8028
2	0.8150	0.6620
3	0.7299	0.5120
4	0.6561	0.4096
5	0.5926	0.3277
6	0.5314	0.2621
7	0.4783	0.2097
8	0.4305	0.1678
9	0.3874	0.1342
10	0.3487	0.1074
11	0.3138	0.0859

$$P(G/\{N > k\}) \cong \frac{\frac{6}{36} \left(\frac{27}{36}\right)^{k-1} + e}{\frac{6}{36} \left(\frac{27}{36}\right)^{k-1} + d} \cong \frac{1}{3}$$



# Bayes and "the prosecutor's fallacy".

- The prosecutor's fallacy
  - The prosecutor's fallacy is the assertion that, because the story before the court is highly improbable, the defendant's innocence is equally improbable.
- OJ Simpson:
  - The chance that a random sample of DNA would match that of O.J. Simpson was put at one in 4m. Long odds: but, as Johnnie Cochran, Mr Simpson's counsel, explained to the jury, there are 20m people in the Los Angeles area. Mr Simpson was therefore one of several people whose blood might be matched to the scene and he could not be guilty beyond reasonable doubt.

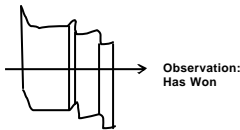
$$P(\text{Guilty} / \text{DNA Matches}) = \frac{P(\text{Guilty})P(\text{DNA Matches} / \text{Guilty})}{P(\text{Guilty})P(\text{DNA Matches} / \text{Guilty}) + P(\text{Not Guilty})P(\text{DNA Matches} / \text{Not Guilty})}$$

$$P(\text{DNA Matches} / \text{Guilty}) = \frac{P(\text{DNA Matches} / \text{DNA Matches} / \text{Guilty})}{P(\text{DNA Matches} / \text{DNA Matches} / \text{Guilty}) + P(\text{DNA Matches} / \text{DNA don't Match} / \text{Guilty})}$$

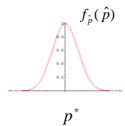
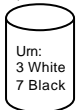
REVISSAR!!!!

What is the chance of your being guilty?  
Financial Times 19 June 2003

- Probability of getting a 5 in the first throw if it is known that the player has won.
  - Note: inverse probability, which is the cause.



$$P_{\text{Real}}^* = P(\text{White observation} / \text{composition of the urn})$$



$$P_{\text{New}}^* (\text{composition of the urn} / \text{white observation})$$