

CS 441 Discrete Mathematics for CS  
Lecture 19

## Probabilities III

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## Probabilities

- **Experiment:**
  - a procedure that yields one of the possible outcomes
- **Sample space:** a set of possible outcomes
- **Event:** a subset of possible outcomes (E is a subset of S)
- **Assuming the outcomes are equally likely, the probability of an event E, defined by a subset of outcomes from the sample space S is**
  - $P(E) = |E| / |S|$
- The cardinality of the subset divided by the cardinality of the sample space.

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## Probabilities

### Three axioms of the probability theory:

(1) Probability of a discrete outcome  $s$  is:

- $0 \leq P(s) \leq 1$

(2) Sum of probabilities of all disjoint outcomes is  $= 1$

(3) For any two events  $E1$  and  $E2$  holds:

$$P(E1 \cup E2) = P(E1) + P(E2) - P(E1 \text{ and } E2)$$

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## Probability distribution

**Definition:** A function  $p: S \rightarrow [0,1]$  satisfying the three axioms of probability is called a **probability distribution**

**Example:** a biased coin

- Probability of head 0.6, probability of a tail 0.4
  - **Probability distribution:**
    - Head  $\rightarrow 0.6$
    - Tail  $\rightarrow 0.4$
- The sum of the probabilities sums to 1

**Note:** a **uniform distribution** is a special distribution that assigns equal probabilities to each outcome.

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## Conditional probability

**Definition:** Let E and F be two events such that  $P(F) > 0$ . The **conditional probability** of E given F

- $P(E|F) = P(E \cap F) / P(F)$

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## Conditional probability

**Corrolary:** Let E and F be two events such that  $P(F) > 0$ . Then:

- $P(E \cap F) = P(E|F) P(F)$

**Example:**

- Assume the probability of getting a flu is 0.2
- Assume the probability of having a high fever given the flu: 0.9

What is the probability of getting a flu with fever?

$$P(\text{flu and fever}) = P(\text{fever}|\text{flu})P(\text{flu}) = 0.9 \cdot 0.2 = 0.18$$

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## Bayes theorem

**Definition:** Let E and F be two events such that  $P(F) > 0$ . Then:

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

**Proof:**

$$P(E|F) = P(E \wedge F) / P(F)$$

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

Idea: Simply switch the conditioning events.

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## Bayes theorem

**Definition:** Let E and F be two events such that  $P(F) > 0$ . Then:

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

**Example:**

- Assume the probability of getting a flu is 0.2
- Assume the probability of getting a fever is 0.3
- Assume the probability of having a high fever given the flu: 0.9
- What is the probability of having a flu given the fever?
- $P(\text{flu} | \text{fever}) = P(\text{fever}|\text{flu}) P(\text{flu}) / P(\text{fever}) =$   
 $= 0.9 \times 0.2 / 0.3 = 0.18 / 0.3 = 0.3$

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## Independence

**Definition:** The two events E and F are said to be **independent** if:

- $P(E \wedge F) = P(E)P(F)$

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## Random variables

- **Definition: A random variable** is a function from the sample space of an experiment to the set of real numbers  $f: S \rightarrow \mathbb{R}$ . A random variable assigns a number to each possible outcome.
- **The distribution of a random variable X on the sample space** S is a set of pairs  $(r, p(X=r))$  for all r in S where r is the number and  $p(X=r)$  is the probability that X takes a value r.

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## Random variables

### Example:

Let  $S$  be the outcomes of a two-dice roll

Let random variable  $X$  denotes the sum of outcomes

$(1,1) \rightarrow 2$

$(1,2)$  and  $(2,1) \rightarrow 3$

$(1,3)$ ,  $(3,1)$  and  $(2,2) \rightarrow 4$

...

### Distribution of $X$ :

- $2 \rightarrow 1/36$ ,
- $3 \rightarrow 2/36$ ,
- $4 \rightarrow 3/36$  ...
- $12 \rightarrow 1/36$

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## Probabilities

- Assume a repeated coin flip
- $P(\text{head}) = 0.6$  and the probability of a tail is 0.4. Each coin flip is independent of the previous.
- What is the probability of seeing:
  - HHHHH - 5 heads in a row
- $P(\text{HHHHH}) = 0.6^5 =$ 
  - Assume the outcome is HHTTT
- $P(\text{HHTTT}) = 0.6 * 0.6 * 0.4^3 = 0.6^2 * 0.4^3$ 
  - Assume the outcome is TTHHT
- $P(\text{TTHHT}) = 0.4^2 * 0.6^2 * 0.4 = 0.6^2 * 0.4^3$
- What is the probability of seeing three tails and two heads?
- The number of two-head-three tail combinations =  $C(5,2)$
- $P(\text{two-heads-three tails}) = C(5,2) * 0.6^2 * 0.4^3$

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## Probabilities

- Assume a variant of a repeated coin flip problem
- The space of possible outcomes is the count of occurrences of heads in 5 coin flips. For example:
- TTTTT yields outcome 0
- HTTTT or TTHTT yields 1
- HTHHT yields 3 ...
- What is the probability of an outcome 0?
- $P(\text{outcome}=0) = 0.6^0 * 0.4^5$
- $P(\text{outcome}=1) = C(5,1) 0.6^1 * 0.4^4$
- $P(\text{outcome}=2) = C(5,2) 0.6^2 * 0.4^3$
- $P(\text{outcome}=3) = C(5,3) 0.6^3 * 0.4^2$
- ...

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## Expected value and variance

**Definition:** The expected value of the random variable  $X(s)$  on the sample space is equal to:

$$E(X) = \sum_{s \in S} p(s) X(s)$$

**Example:** roll of a dice

- Outcomes: 1 2 3 4 5 6
- Expected value:

$$E(X) = 1*1/6 + 2*1/6 + 3*1/6 + 4*1/6 + 5*1/6 + 6*1/6 = 7/2 = 3.5$$

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## Expected value

### Example:

Flip a fair coin 3 times. The outcome of the trial  $X$  is the number of heads. What is the expected value of the trial?

### Answer:

Possible outcomes:

$= \{HHH \ HHT \ HTH \ THH \ HTT \ THT \ TTH \ TTT\}$   
3      2      2      2      1      1      1      0

$$E(X) = 1/8 (3 + 3 \cdot 2 + 3 \cdot 1 + 0) = 12/8 = 3/2$$

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## Expected value

- **Theorem:** If  $X_i$   $i=1,2,3, n$  with  $n$  being a positive integer, define random variables on  $S$ , and  $a$  and  $b$  are real numbers then:
  - $E(X_1 + X_2 + \dots + X_n) = E(X_1) + E(X_2) + \dots + E(X_n)$
  - $E(aX + b) = aE(X) + b$

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## Expected value

### Example:

- Roll a pair of dices. What is the expected value of the sum of outcomes?

- **Approach 1:**

- Outcomes: (1,1) (1,2) (1,3) .... (6,1)... (6,6)  
                  2      3      4              7      12

**Expected value:**  $1/36 (2*1 + \dots) = 7$

- **Approach 2 (theorem):**

- $E(X_1 + X_2) = E(X_1) + E(X_2)$
- $E(X_1) = 7/2$  and  $E(X_2) = 7/2$
- $E(X_1 + X_2) = 7$

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## Expected value

### Investment problem:

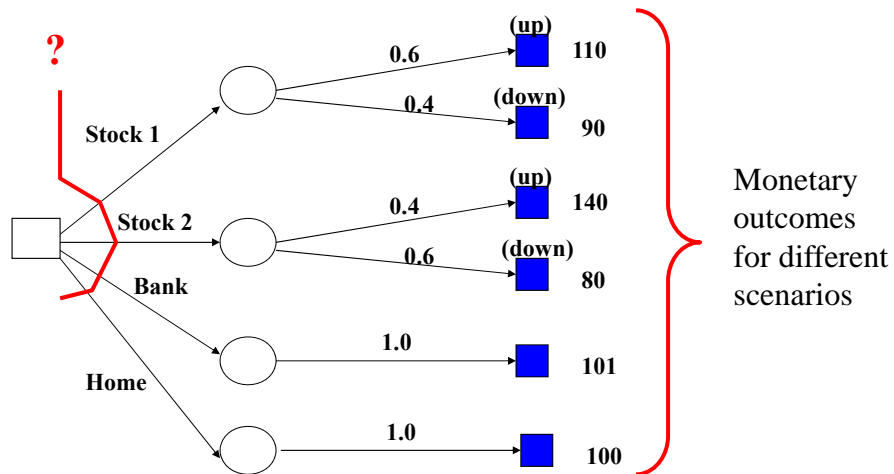
- You have 100 dollars and can invest into a stock. The returns are volatile and you may get either \$120 with probability of 0.4, or \$90 with probability 0.6.
- **What is the expected value of your investment?**
- $E(X) = 0.4*120 + 0.6*90 = 48 + 54 = 102$
- **Is it OK to invest?**

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## Decision making example

We need to make a choice whether to invest in Stock 1 or 2, put money into bank or keep them at home.

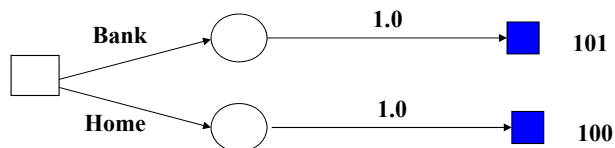


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## Decision making example.

Assume the simplified problem with the Bank and Home choices only.

The result is guaranteed – the outcome is deterministic



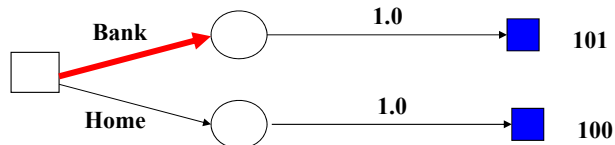
What is the rational choice assuming our goal is to make money?

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## Decision making. Deterministic outcome.

Assume the simplified problem with the Bank and Home choices only.

These choices are deterministic.



Our goal is to make money. What is the rational choice?

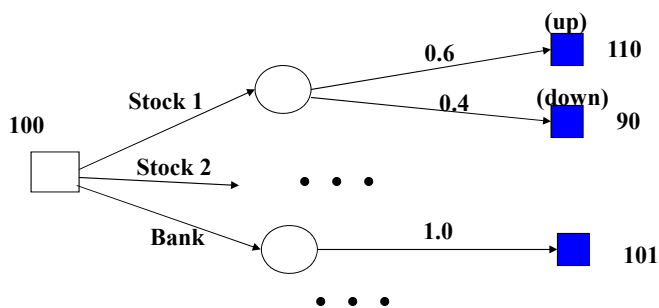
**Answer:** Put money into the bank. The choice is always strictly better in terms of the outcome

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## Decision making

- How to quantify the goodness of the stochastic outcome?

We want to compare it to deterministic and other stochastic outcomes.

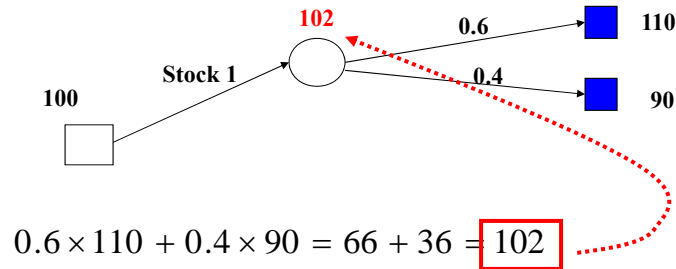


Idea: Use the expected value of the outcome

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## Expected value

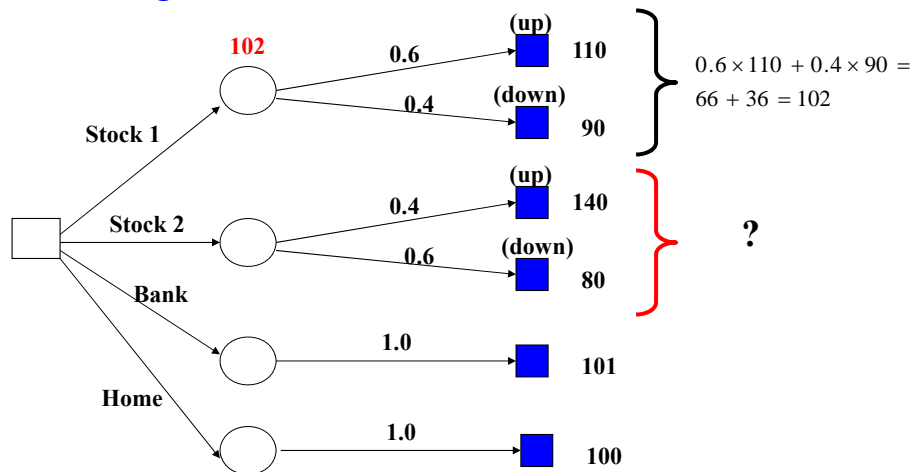
- **Expected value** summarizes all stochastic outcomes into a single quantity
  - Expected value for the outcome of the Stock 1 option is:



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## Expected values

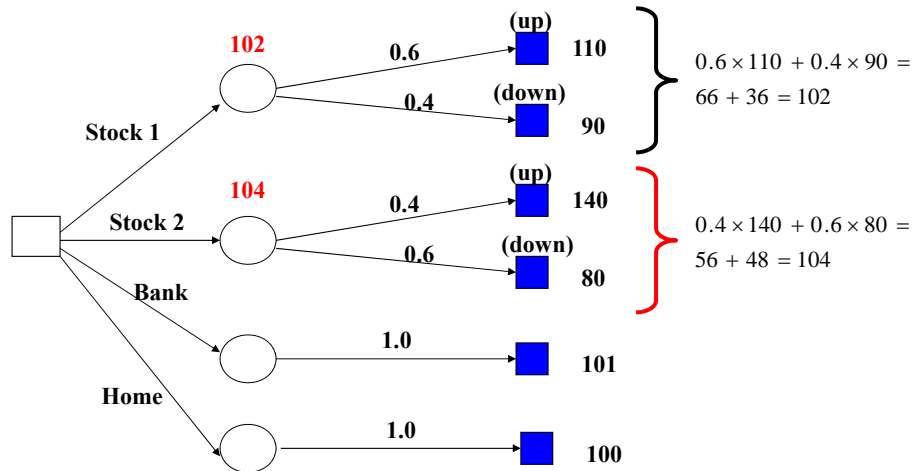
Investing \$100 for 6 months



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## Expected values

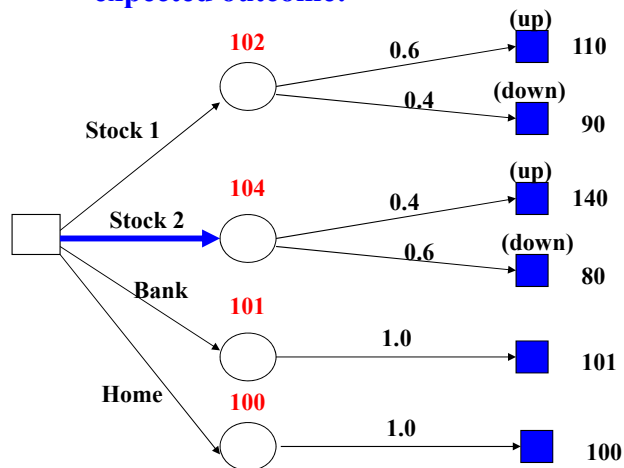
Investing \$100 for 6 months



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## Selection based on expected values

The optimal action is the option that maximizes the expected outcome:



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