CS 2750 Foundations of AI
Lecture 20

Decision making in the presence of uncertainty

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Decision-making in the presence of uncertainty

- Computing the probability of some event may not be our ultimate goal
- Instead we are often interested in **making decisions about our future actions so that we satisfy some goals**
- **Example: medicine**
  - Diagnosis is typically only the first step
  - The ultimate goal is to manage the patient in the best possible way. Typically many options available:
    - Surgery, medication, collect the new info (lab test)
    - There is an **uncertainty in the outcomes** of these procedures: patient can be improve, get worse or even die as a result of different management choices.

Decision-making in the presence of uncertainty

Main issues:

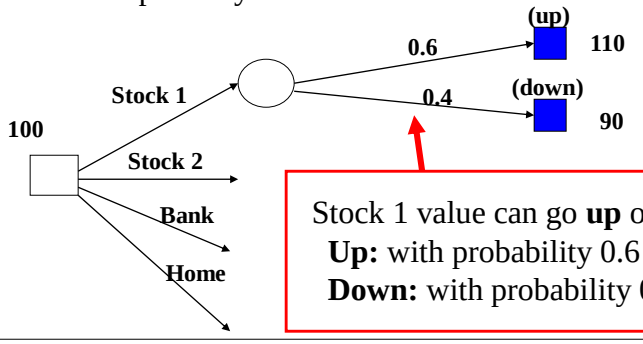
- **How to model the decision process with uncertain outcomes in the computer ?**
- **How to make decisions about actions in the presence of uncertainty?**

The field of **decision-making** studies ways of making decisions in the presence of uncertainty.

Decision making example.

Assume we want to invest $100 for 6 months

- **We have 4 choices:**
  1. Invest in Stock 1
  2. Invest in Stock 2
  3. Put money in bank
  4. Keep money at home



Stock 1 value can go **up** or **down**:
Up: with probability 0.6
Down: with probability 0.4

Decision making example.

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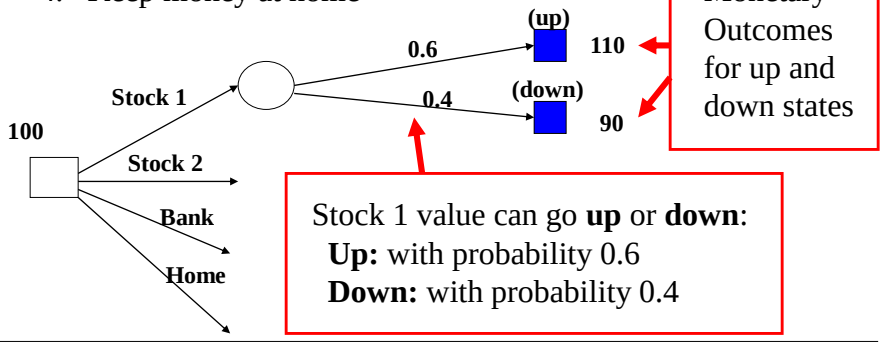
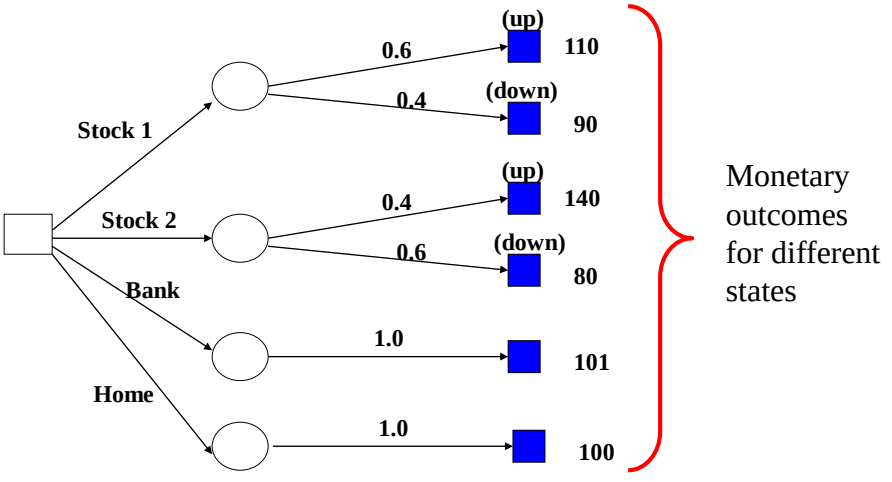


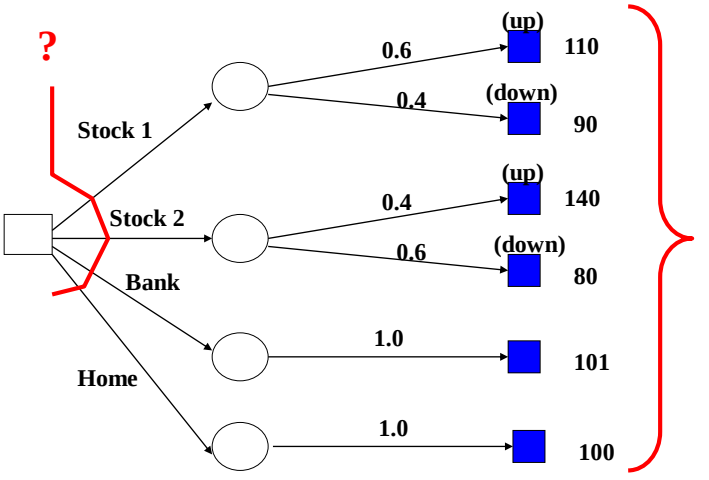
Decision making example.

Investing of $100 for 6 months



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. But how?

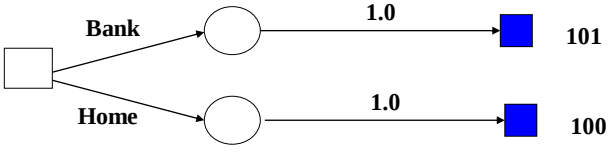


Monetary outcomes for different scenarios

Decision making example.

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

The result is guaranteed – the outcome is deterministic

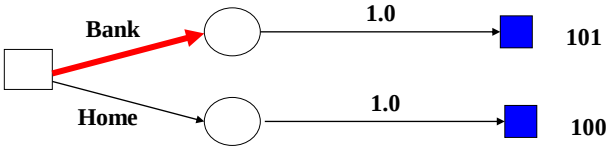


What is the rational choice assuming our goal is to make money?

Decision making. Deterministic outcome.

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

These choices are deterministic.



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

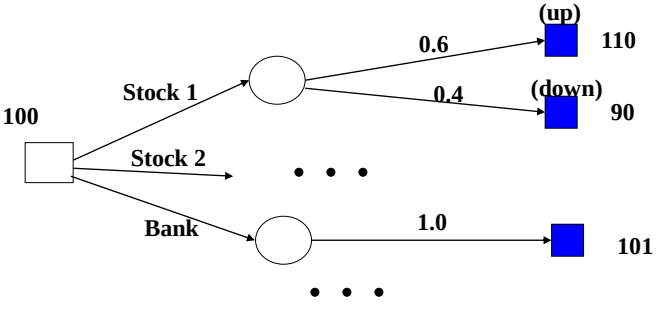
Answer: Put money into the bank. The choice is always strictly better in terms of the outcome

But what to do if we have uncertain outcomes?

Decision making. Stochastic outcome

- **How to quantify the goodness of the stochastic outcome?**

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



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Decision making. Stochastic outcome

- **How to quantify the goodness of the stochastic outcome?**

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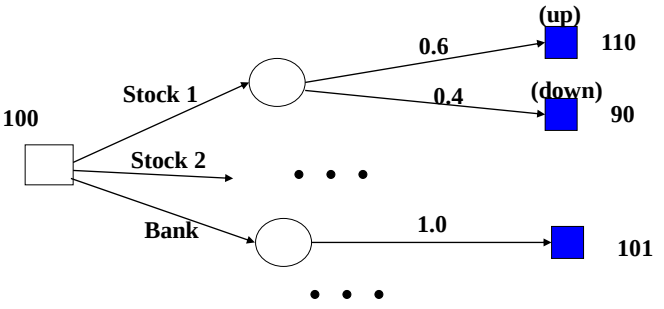


Idea: **Use the expected value of the outcome**

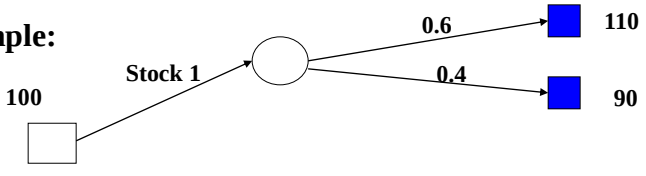
Expected value

- Let X be a random variable representing the monetary outcome with a discrete set of values Ω_X .
- **Expected value** of X is:

$$E(X) = \sum_{x \in \Omega_X} xP(X = x)$$

Intuition: **Expected value** summarizes all stochastic outcomes into a single quantity.

- **Example:**



- What is the expected value of the outcome of Stock 1 option?

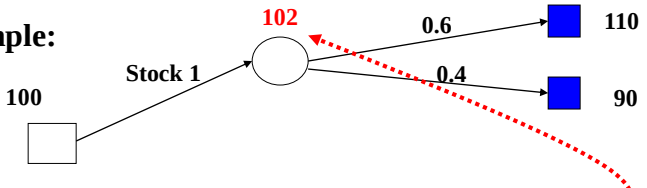
Expected value

- Let X be a random variable representing the monetary outcome with a discrete set of values Ω_X .
- **Expected value** of X is:

$$E(X) = \sum_{x \in \Omega_X} xP(X = x)$$

  - **Expected value** summarizes all stochastic outcomes into a single quantity

- **Example:**

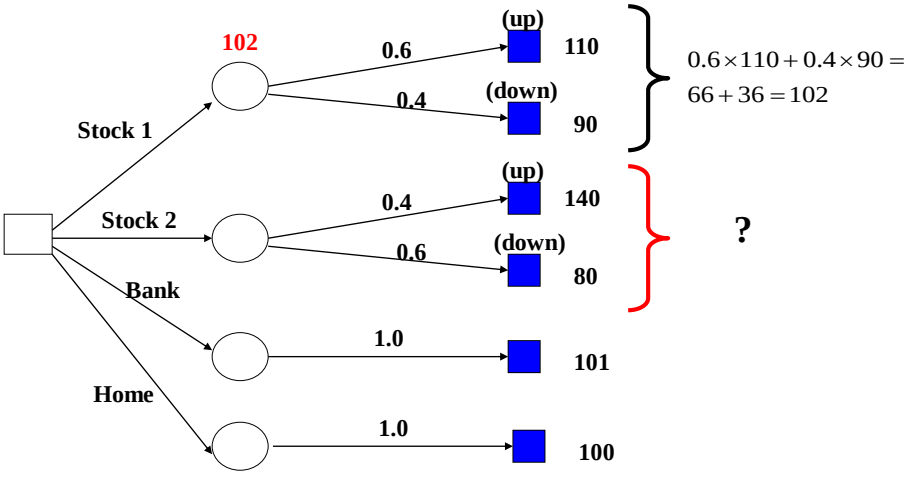


Expected value for the outcome of the Stock 1 option is:

$$0.6 \times 110 + 0.4 \times 90 = 66 + 36 = 102$$

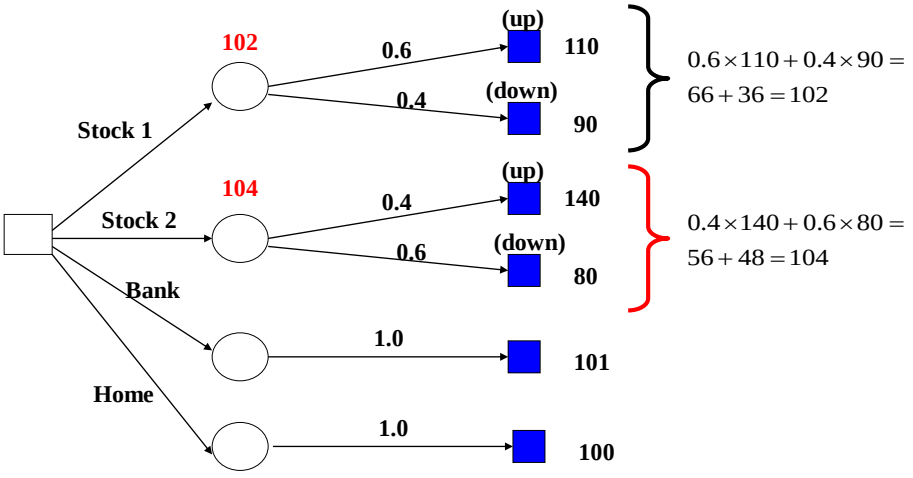
Expected values

Investing $100 for 6 months



Expected values

Investing $100 for 6 months



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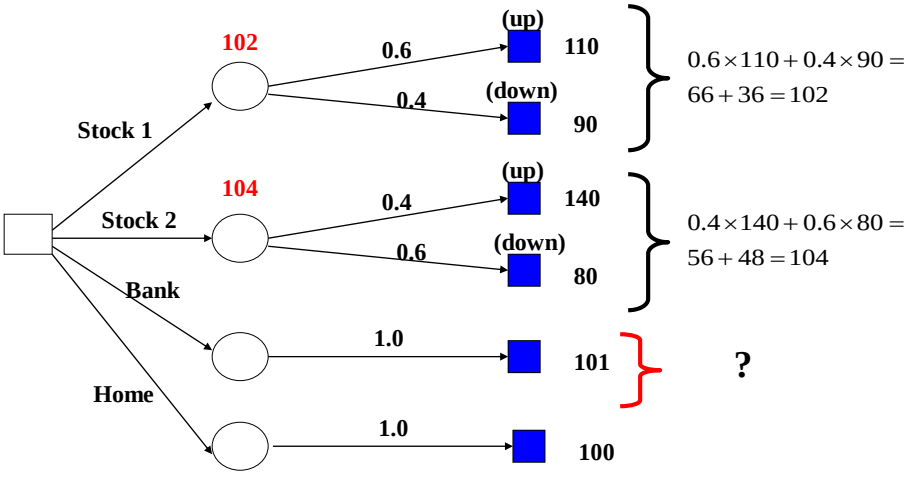
Expected values

Investing $100 for 6 months



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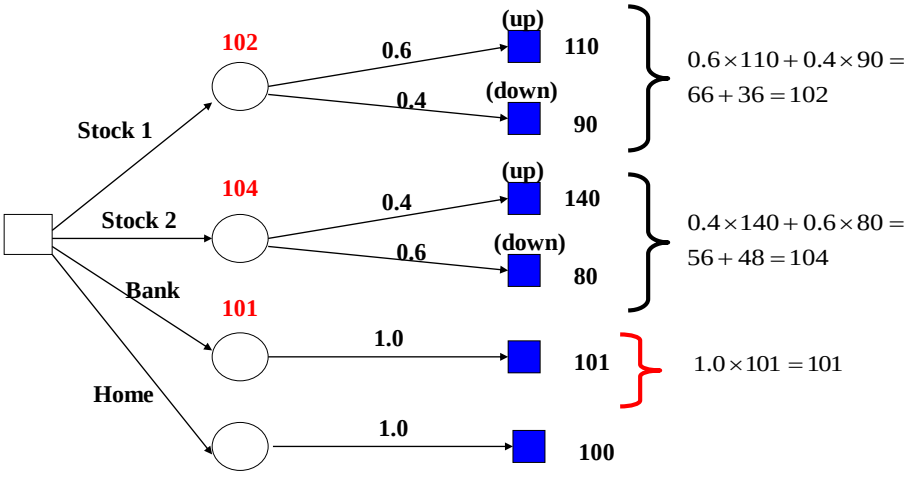
Expected values

Investing $100 for 6 months



Expected values

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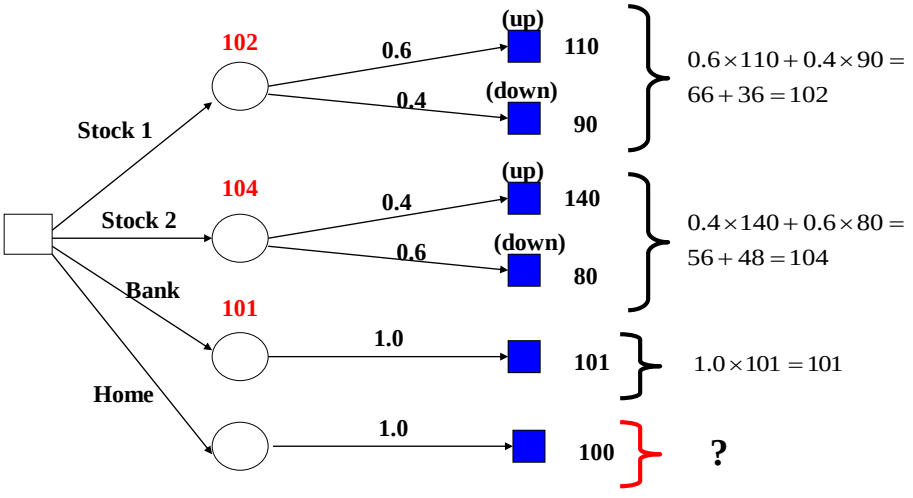
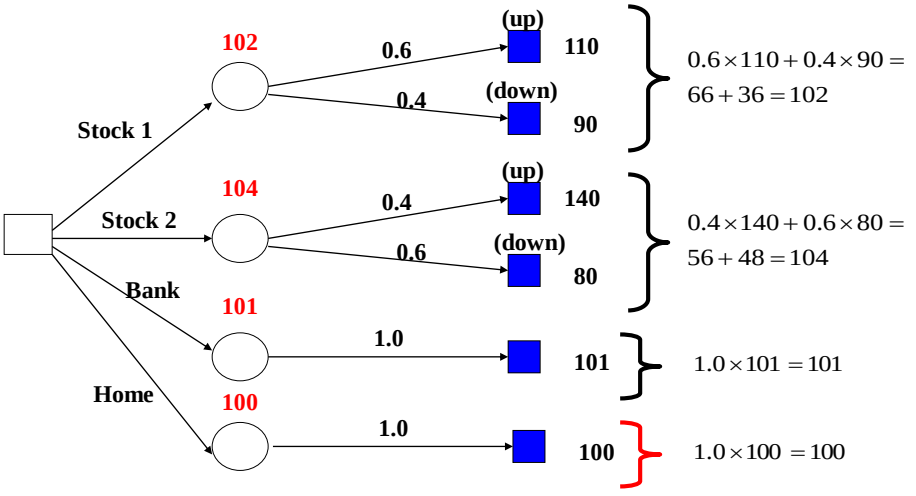


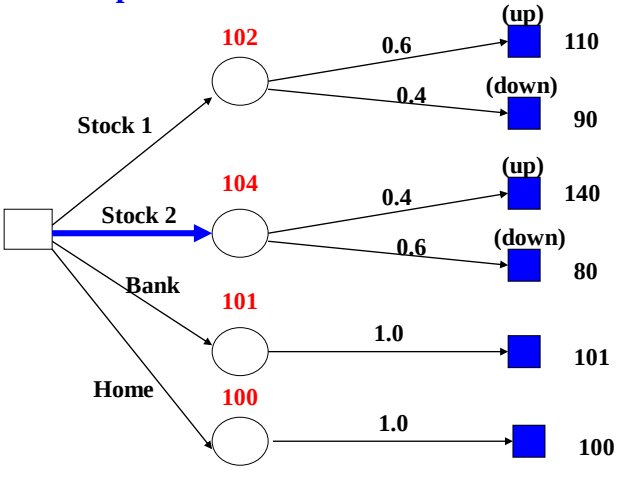
Expected values

Investing $100 for 6 months



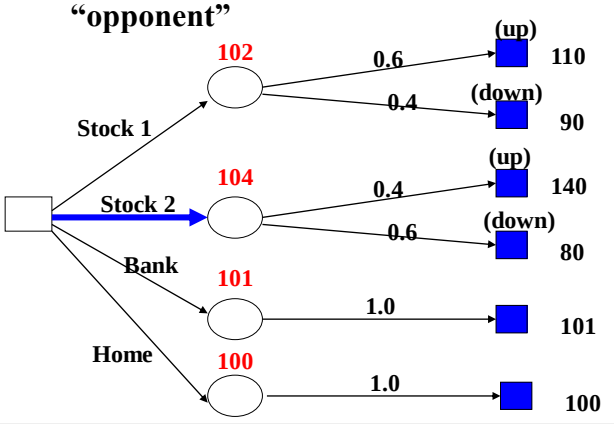
Selection based on expected values

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



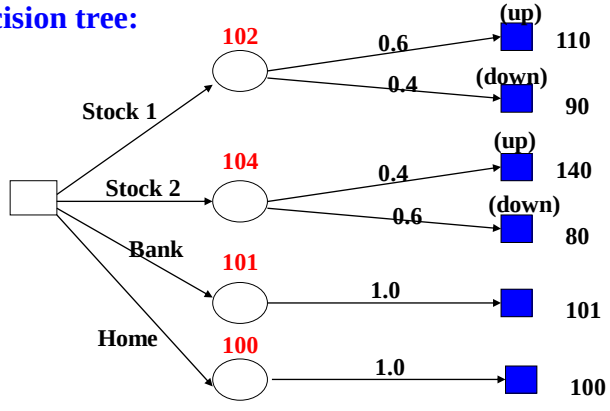
Relation to the game search

- **Game search: minimax algorithm**
  - **considers the rational opponent and its best move**
- **Decision making: maximizes the expectation**
  - **play against the nature – a stochastic non-malicious "opponent"**



(Stochastic) Decision tree

- **Decision tree:**



- **decision node**
- **chance node**
- **outcome (value) node**

Sequential (multi-step) problems

The decision tree can be build to capture multi-step decision problems:

- Choose an action
- Observe the stochastic outcome
- And repeat

How to make decisions for multi-step problems?

- Start from the leaves of the decision tree (outcome nodes)
- Compute expectations at chance nodes
- Maximize at the decision nodes

Algorithm is sometimes called **expectimax**

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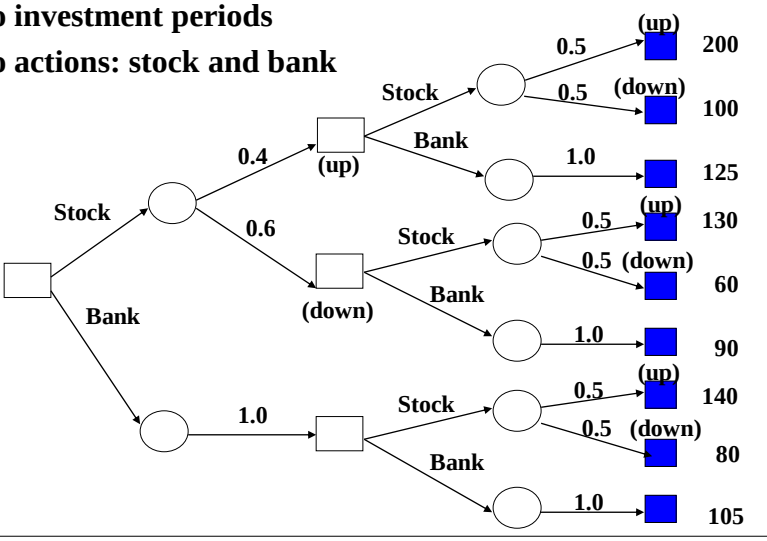
Multi-step problem example

Assume:

- **Two investment periods**
- **Two actions: stock and bank**



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Multi-step problem example

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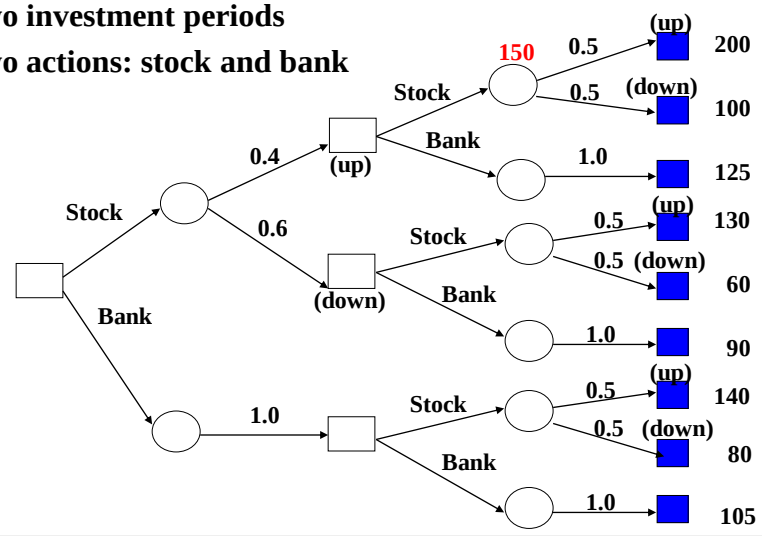

Multi-step problem example

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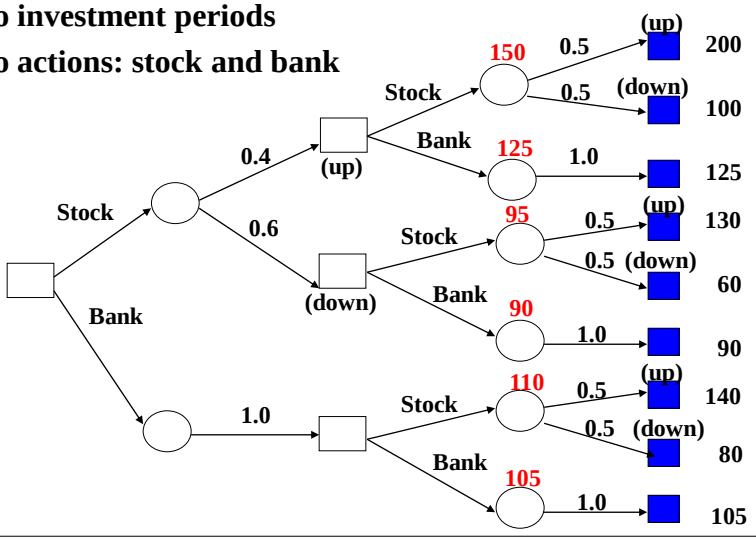

Multi-step problem example

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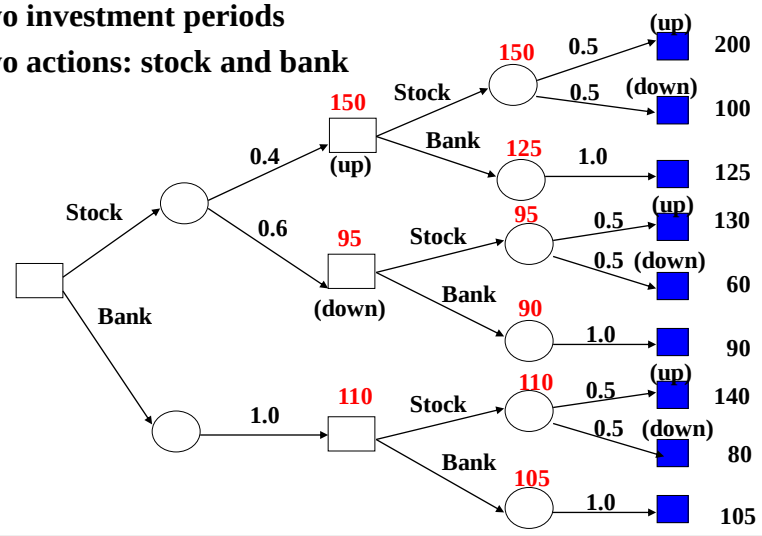

Multi-step problem example

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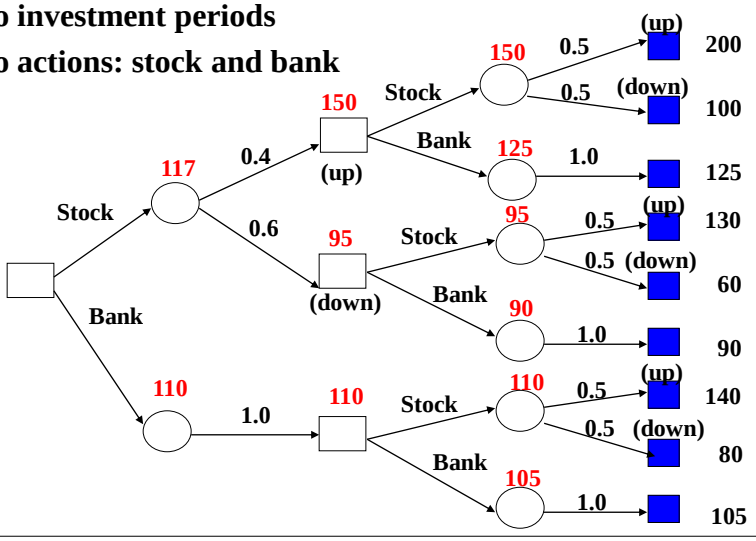

Multi-step problem example

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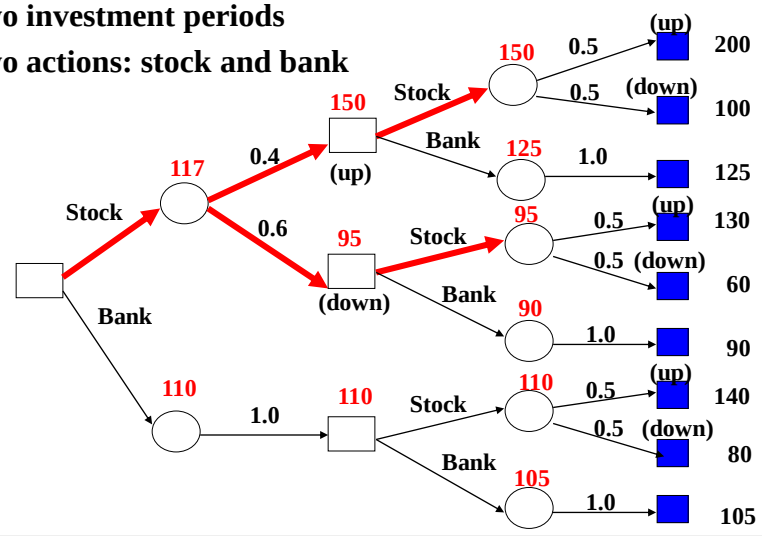
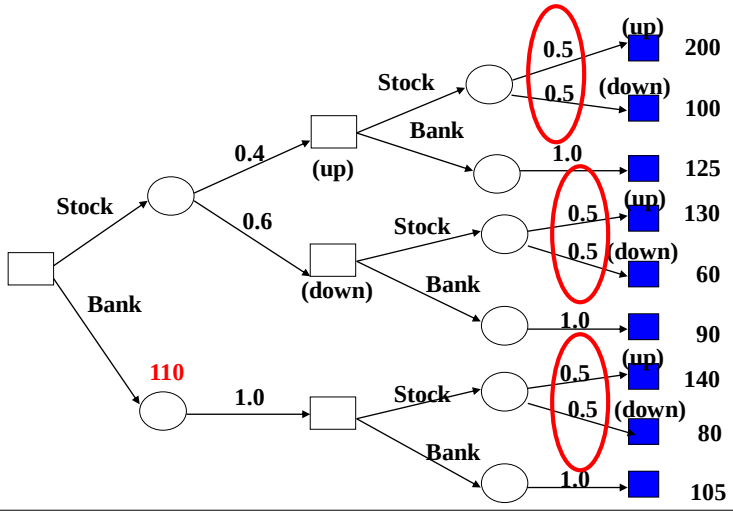


Multi-step problems. Conditioning.

- Notice that the probability of stock going up and down in the 2nd step is independent of the 1st step (=0.5)

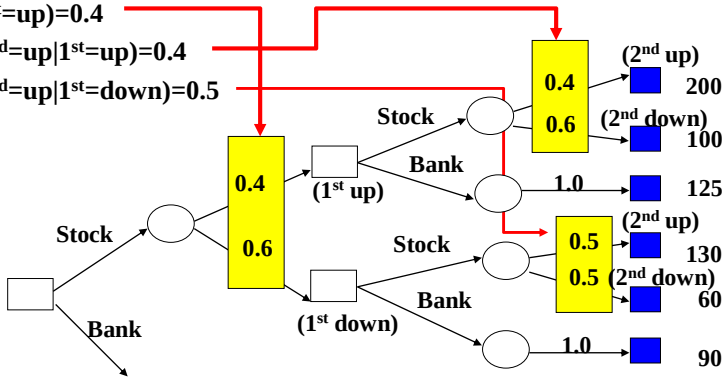


Conditioning in the decision tree

- **But this may not hold in general. In decision trees:**
  - **Later outcomes can be conditioned on the earlier stochastic outcomes and actions**

Example: stock movement probabilities. Assume:

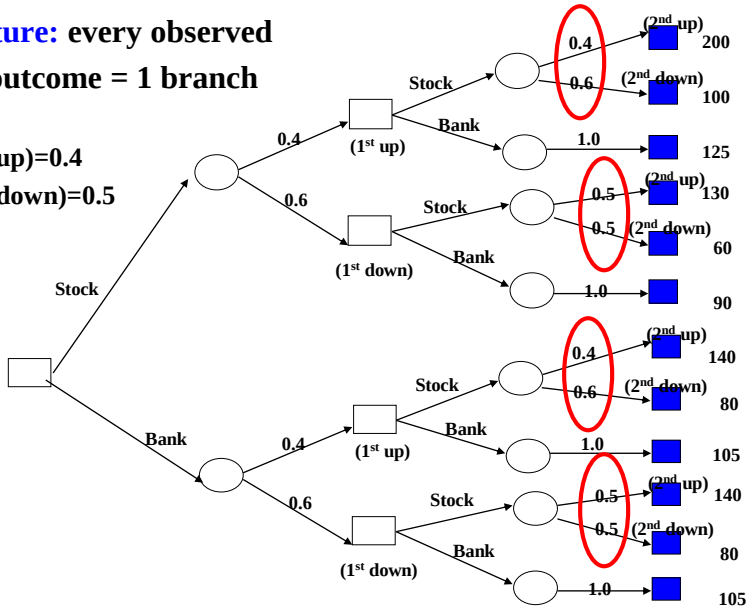
P(1st=up)=0.4
P(2nd=up|1st=up)=0.4
P(2nd=up|1st=down)=0.5



Multi-step problems. Conditioning.

Tree Structure: **every observed stochastic outcome = 1 branch**

P(1st=up)=0.4
P(2nd=up|1st=up)=0.4
P(2nd=up|1st=down)=0.5



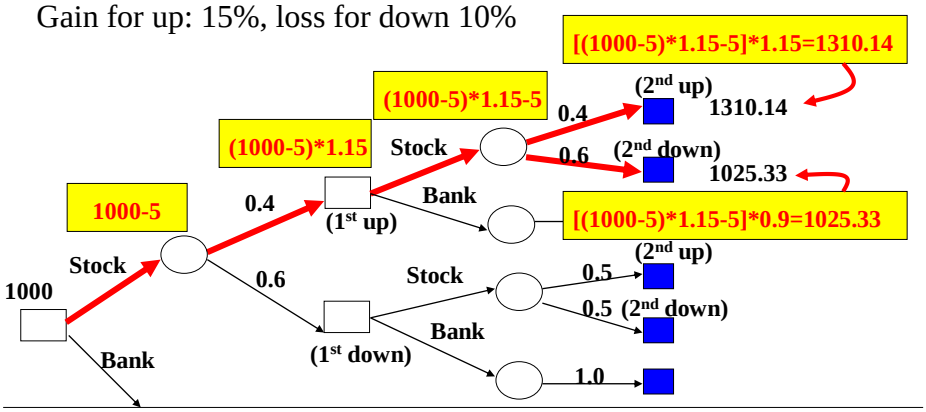
Trajectory payoffs

- **Outcome values at leaf nodes (e.g. monetary values)**
  - **Rewards and costs for the path trajectory**

Example: stock fees and gains. **Assume:**

Fee per period: $5 paid at the beginning

Gain for up: 15%, loss for down 10%



Constructing a decision tree

- **The decision tree is rarely given to you directly.**
  - Part of the problem is to construct the tree.

Example: stocks, bonds, bank for k periods

Stock:

- Probability of stocks going up in the first period: 0.3
- Probability of stocks going up in subsequent periods:
  - P(kth step=Up| (k -1)th step =Up)=0.4
  - P(kth step =Up| (k -1)th step=Down)=0.5
- Return if stock goes up: 15 % if down: 10%
- Fixed fee per investment period: $5

Bonds:

- Probability of value up: 0.5, down: 0.5
- Return if bond value is going up: 7%, if down: 3%
- Fee per investment period: $2

Bank:

- Guaranteed return of 3% per period, no fee