

# The Course So Far

- **Traditional AI:** Deterministic single agent domains
  - Atomic agent: uninformed, informed, local
  - Specific KR languages
    - Constraint Satisfaction
    - Logic and Satisfiability
- **Traditional AI:** Deterministic Adversarial domains
  - Atomic agent: minimax
  - Knowledge: utility function, game specific heuristics

# Rest of the Course

- **Modern AI:** Uncertainty
- Uncertainty comes in many forms
  - uncertainty due to another agent's policy
  - uncertainty in outcome of my own action
  - uncertainty in my knowledge of the world

# Rest of the Course

- **Modern AI: Uncertainty**
- Uncertainty comes in many forms
  - uncertainty due to another agent's policy
  - **uncertainty in outcome of my own action**
  - uncertainty in my knowledge of the world

# Fundamentals of Decision Theory

## Chapter 16

**Mausam**

(Based on slides of someone from NPS,  
Maria Fasli)

# Decision Theory

- **“an analytic and systematic approach to the study of decision making”**

## Good decisions:

- based on reasoning
- consider all available data and possible alternatives
- employ a quantitative approach

## Bad decisions:

- not based on reasoning
- do not consider all available data and possible alternatives
- do not employ a quantitative approach

- **A good decision may occasionally result in an unexpected outcome; it is still a good decision if made properly**
- **A bad decision may occasionally result in a good outcome if you are lucky; it is still a bad decision**

# Steps in Decision Theory

1. List the possible alternatives (actions/decisions)
2. Identify the possible outcomes
3. List the payoff or profit or reward
4. Select one of the decision theory models
5. Apply the model and make your decision

# Example

## The Thompson Lumber Company

- Problem.
  - The Thompson Lumber Co. must decide whether or not to expand its product line by manufacturing and marketing a new product, backyard storage sheds
- Step 1: List the possible alternatives
  - alternative:* “a course of action or strategy that may be chosen by the decision maker”
  - (1) Construct a large plant to manufacture the sheds
  - (2) Construct a small plant
  - (3) Do nothing

# The Thompson Lumber Company

- Step 2: Identify the states of nature
  - (1) The market for storage sheds could be favorable
    - high demand
  - (2) The market for storage sheds could be unfavorable
    - low demand

*state of nature*: “an outcome over which the decision maker has little or no control”  
e.g., lottery, coin-toss, whether it will rain today



# The Thompson Lumber Company

- Step 3: List the possible rewards
  - A reward for all possible combinations of alternatives and states of nature
  - *Conditional values*: “reward depends upon the alternative and the state of nature”
    - with a favorable market:
      - a large plant produces a net profit of \$200,000
      - a small plant produces a net profit of \$100,000
      - no plant produces a net profit of \$0
    - with an unfavorable market:
      - a large plant produces a net loss of \$180,000
      - a small plant produces a net loss of \$20,000
      - no plant produces a net profit of \$0

# Reward tables

- A means of organizing a decision situation, including the rewards from different situations given the possible states of nature

<b>Actions</b>	<b>States of Nature</b>	
	<b>a</b>	<b>b</b>
<b>1</b>	<b>Reward 1a</b>	<b>Reward 1b</b>
<b>2</b>	<b>Reward 2a</b>	<b>Reward 2b</b>

# The Thompson Lumber Company

	<b>States of Nature</b>	
<b>Actions</b>		

# The Thompson Lumber Company

	<b>States of Nature</b>	
<b>Actions</b>	<b>Favorable Market</b>	<b>Unfavorable Market</b>
<b>Large plant</b>	<b>\$200,000</b>	<b>-\$180,000</b>
<b>Small plant</b>	<b>\$100,000</b>	<b>-\$20,000</b>
<b>No plant</b>	<b>\$0</b>	<b>\$0</b>

# The Thompson Lumber Company

- Steps 4/5: Select an appropriate model and apply it
  - Model selection depends on the operating environment and degree of uncertainty

# Future Uncertainty

- Nondeterministic
- Probabilistic

# Non-deterministic Uncertainty

	<b>States of Nature</b>	
<b>Actions</b>	<b>Favorable Market</b>	<b>Unfavorable Market</b>
<b>Large plant</b>	<b>\$200,000</b>	<b>-\$180,000</b>
<b>Small plant</b>	<b>\$100,000</b>	<b>-\$20,000</b>
<b>No plant</b>	<b>\$0</b>	<b>\$0</b>

- What should we do?

# Maximax Criterion

“Go for the Gold”

- Select the decision that results in the maximum of the maximum rewards
- A very optimistic decision criterion
  - Decision maker assumes that the most favorable state of nature for each action will occur
- Most risk prone agent



# Maximax

Decision	States of Nature		Maximum in Row
	Favorable	Unfavorable	
Large plant	\$200,000	-\$180,000	<b>\$200,000</b>
Small plant	\$100,000	-\$20,000	\$100,000
No plant	\$0	\$0	\$0

- **Thompson Lumber Co. assumes that the most favorable state of nature occurs for each decision alternative**
- **Select the maximum reward for each decision**
  - **All three maximums occur if a favorable economy prevails (a tie in case of no plant)**
- **Select the maximum of the maximums**
  - **Maximum is \$200,000; corresponding decision is to build the large plant**
  - **Potential loss of \$180,000 is completely ignored**

# Maximin Criterion

“Best of the Worst”

- Select the decision that results in the maximum of the minimum rewards
- A very pessimistic decision criterion
  - Decision maker assumes that the minimum reward occurs for each decision alternative
  - Select the maximum of these minimum rewards
- Most risk averse agent

# Maximin

Decision	States of Nature		Minimum in Row
	Favorable	Unfavorable	
Large plant	\$200,000	-\$180,000	-\$180,000
Small plant	\$100,000	-\$20,000	-\$20,000
No plant	\$0	\$0	<b>\$0</b>

- **Thompson Lumber Co. assumes that the least favorable state of nature occurs for each decision alternative**
- **Select the minimum reward for each decision**
  - **All three minimums occur if an unfavorable economy prevails (a tie in case of no plant)**
- **Select the maximum of the minimums**
  - **Maximum is \$0; corresponding decision is to do nothing**
  - **A conservative decision; largest possible gain, \$0, is much less than maximax**

# Equal Likelihood Criterion

- Assumes that all states of nature are equally likely to occur
  - Maximax criterion assumed the most favorable state of nature occurs for each decision
  - Maximin criterion assumed the least favorable state of nature occurs for each decision
- Calculate the *average reward* for each alternative and select the alternative with the maximum number
  - Average reward: the sum of all rewards divided by the number of states of nature
- Select the decision that gives the highest average reward

# Equal Likelihood

Decision	States of Nature		Row
	Favorable	Unfavorable	Average
Large plant	\$200,000	-\$180,000	\$10,000
Small plant	\$100,000	-\$20,000	<b>\$40,000</b>
No plant	\$0	\$0	\$0

## Row Averages

$$\text{Large Plant} = \frac{\$200,000 - \$180,000}{2} = \$10,000$$

$$\text{Small Plant} = \frac{\$100,000 - \$20,000}{2} = \$40,000$$

$$\text{Do Nothing} = \frac{\$0 + \$0}{2} = \$0$$

- Select the decision with the highest weighted value
  - Maximum is \$40,000; corresponding decision is to build the small plant

# Criterion of Realism

- Also known as the weighted average or Hurwicz criterion
  - A compromise between an optimistic and pessimistic decision
- A coefficient of realism,  $\alpha$ , is selected by the decision maker to indicate optimism or pessimism about the future

$$0 \leq \alpha \leq 1$$

When  $\alpha$  is close to 1, the decision maker is optimistic.

When  $\alpha$  is close to 0, the decision maker is pessimistic.

- Criterion of realism =  $\alpha(\text{row maximum}) + (1-\alpha)(\text{row minimum})$ 
  - A weighted average where maximum and minimum rewards are weighted by  $\alpha$  and  $(1 - \alpha)$  respectively

# Criterion of Realism

- Assume a coefficient of realism equal to 0.8

Decision	States of Nature		Criterion of Realism
	Favorable	Unfavorable	
Large plant	\$200,000	-\$180,000	<b>\$124,000</b>
Small plant	\$100,000	-\$20,000	\$76,000
No plant	\$0	\$0	\$0

## Weighted Averages

$$\text{Large Plant} = (0.8)(\$200,000) + (0.2)(-\$180,000) = \$124,000$$

$$\text{Small Plant} = (0.8)(\$100,000) + (0.2)(-\$20,000) = \$76,000$$

$$\text{Do Nothing} = (0.8)(\$0) + (0.2)(\$0) = \$0$$

Select the decision with the highest weighted value

**Maximum is \$124,000; corresponding decision is to build the large plant**

# Minimax Regret

- Regret/Opportunity Loss: “the difference between the optimal reward and the actual reward received”
- Choose the alternative that minimizes the maximum regret associated with each alternative
  - Start by determining the maximum regret for each alternative
  - Pick the alternative with the minimum number



# Regret Table

- If I knew the future, how much I'd regret my decision...
- Regret for any state of nature is calculated by subtracting each outcome in the column from the best outcome in the same column

# Minimax Regret

Decision	States of Nature				
	Favorable		Unfavorable		Row
	Payoff	Regret	Payoff	Regret	Maximum
Large plant	\$200,000	\$0	-\$180,000	\$180,000	\$180,000
Small plant	\$100,000	\$100,000	-\$20,000	\$20,000	<b>\$100,000</b>
No plant	\$0	\$200,000	\$0	\$0	\$200,000
Best payoff	\$200,000		\$0		

- Select the alternative with the lowest maximum regret

**Minimum is \$100,000; corresponding decision is to build a small plant**

# Summary of Results

<b>Criterion</b>	<b>Decision</b>
<b>Maximax</b>	<b>Build a large plant</b>
<b>Maximin</b>	<b>Do nothing</b>
<b>Equal likelihood</b>	<b>Build a small plant</b>
<b>Realism</b>	<b>Build a large plant</b>
<b>Minimax regret</b>	<b>Build a small plant</b>

# Future Uncertainty

- Non deterministic
- Probabilistic

# Probabilistic Uncertainty

- Decision makers know the probability of occurrence for each possible outcome
  - Attempt to maximize the expected reward
- Criteria for decision models in this environment:
  - Maximization of expected reward
  - Minimization of expected regret
    - Minimize expected regret = maximizing expected reward!

# Expected Reward (Q)

- called Expected Monetary Value (EMV) in DT literature
- “the probability weighted sum of possible rewards for each alternative”
  - Requires a reward table with conditional rewards and probability assessments for all states of nature

$$\begin{aligned} Q(\text{action } a) = & \text{(reward of 1st state of nature)} \\ & X \text{ (probability of 1st state of nature)} \\ & + \text{(reward of 2nd state of nature)} \\ & X \text{ (probability of 2nd state of nature)} \\ & + \dots + \text{(reward of last state of nature)} \\ & X \text{ (probability of last state of nature)} \end{aligned}$$

# The Thompson Lumber Company

- Suppose that the probability of a favorable market is exactly the same as the probability of an unfavorable market. Which alternative would give the greatest Q?

	States of Nature		EMV
	Favorable Mkt p = 0.5	Unfavorable Mkt p = 0.5	
Decision			
Large plant	\$200,000	-\$180,000	\$10,000
Small plant	\$100,000	-\$20,000	<b>\$40,000</b>
No plant	\$0	\$0	\$0

$$Q(\text{large plant}) = (0.5)(\$200,000) + (0.5)(-\$180,000) = \$10,000$$

$$Q(\text{small plant}) = (0.5)(\$100,000) + (0.5)(-\$20,000) = \$40,000$$

$$Q(\text{no plant}) = (0.5)(\$0) + (0.5)(\$0) = \$0$$

**Build the small plant**

# Expected Value of Perfect Information (EVPI)

- It may be possible to purchase additional information about future events and thus make a better decision
  - Thompson Lumber Co. could hire an economist to analyze the economy in order to more accurately determine which economic condition will occur in the future
    - How valuable would this information be?



# EVPI Computation

- Look first at the decisions under each state of nature
  - If information was available that perfectly predicted which state of nature was going to occur, the best decision for that state of nature could be made
    - *expected value with perfect information* (EV w/ PI): “the expected or average return if we have perfect information before a decision has to be made”

# EVPI Computation

- Perfect information changes environment from decision making under risk to decision making with certainty
  - Build the large plant if you know for sure that a favorable market will prevail
  - Do nothing if you know for sure that an unfavorable market will prevail

Decision	States of Nature	
	Favorable $p = 0.5$	Unfavorable $p = 0.5$
Large plant	<b>\$200,000</b>	-\$180,000
Small plant	\$100,000	-\$20,000
No plant	\$0	<b>\$0</b>

# EVPI Computation

- Even though perfect information enables Thompson Lumber Co. to make the correct investment decision, each state of nature occurs only a certain portion of the time
  - A favorable market occurs 50% of the time and an unfavorable market occurs 50% of the time
  - EV w/ PI calculated by choosing the best alternative for each state of nature and multiplying its reward times the probability of occurrence of the state of nature

# EVPI Computation

**EV w/ PI = (best reward for 1st state of nature)  
X (probability of 1st state of nature)  
+ (best reward for 2nd state of nature)  
X (probability of 2nd state of nature)**

**EV w/ PI = (\$200,000)(0.5) + (\$0)(0.5) = \$100,000**

Decision	States of Nature	
	Favorable p = 0.5	Unfavorable p = 0.5
Large plant	\$200,000	-\$180,000
Small plant	\$100,000	-\$20,000
No plant	\$0	\$0

# EVPI Computation

- Thompson Lumber Co. would be foolish to pay more for this information than the extra profit that would be gained from having it
  - *EVPI*: “the maximum amount a decision maker would pay for additional information resulting in a decision better than one made *without perfect information*”
    - EVPI is the expected outcome with perfect information minus the expected outcome without perfect information

$$EVPI = EV \text{ w/ PI} - Q$$

$$EVPI = \$100,000 - \$40,000 = \$60,000$$

# Using EVPI

- EVPI of \$60,000 is the maximum amount that Thompson Lumber Co. should pay to purchase perfect information from a source such as an economist
  - “Perfect” information is extremely rare
  - An investor typically would be willing to pay some amount less than \$60,000, depending on how reliable the information is perceived to be

# Is Expected Value sufficient?

- Lottery 1
  - returns \$0 always
- Lottery 2
  - return \$100 and -\$100 with prob 0.5
- Which is better?

# Is Expected Value sufficient?

- Lottery 1
  - returns \$100 always
- Lottery 2
  - return \$10000 (prob 0.01) and \$0 with prob 0.99
- Which is better?
  - depends



# Is Expected Value sufficient?

- Lottery 1
  - returns \$3125 always
- Lottery 2
  - return \$4000 (prob 0.75) and -\$500 with prob 0.25
- Which is better?

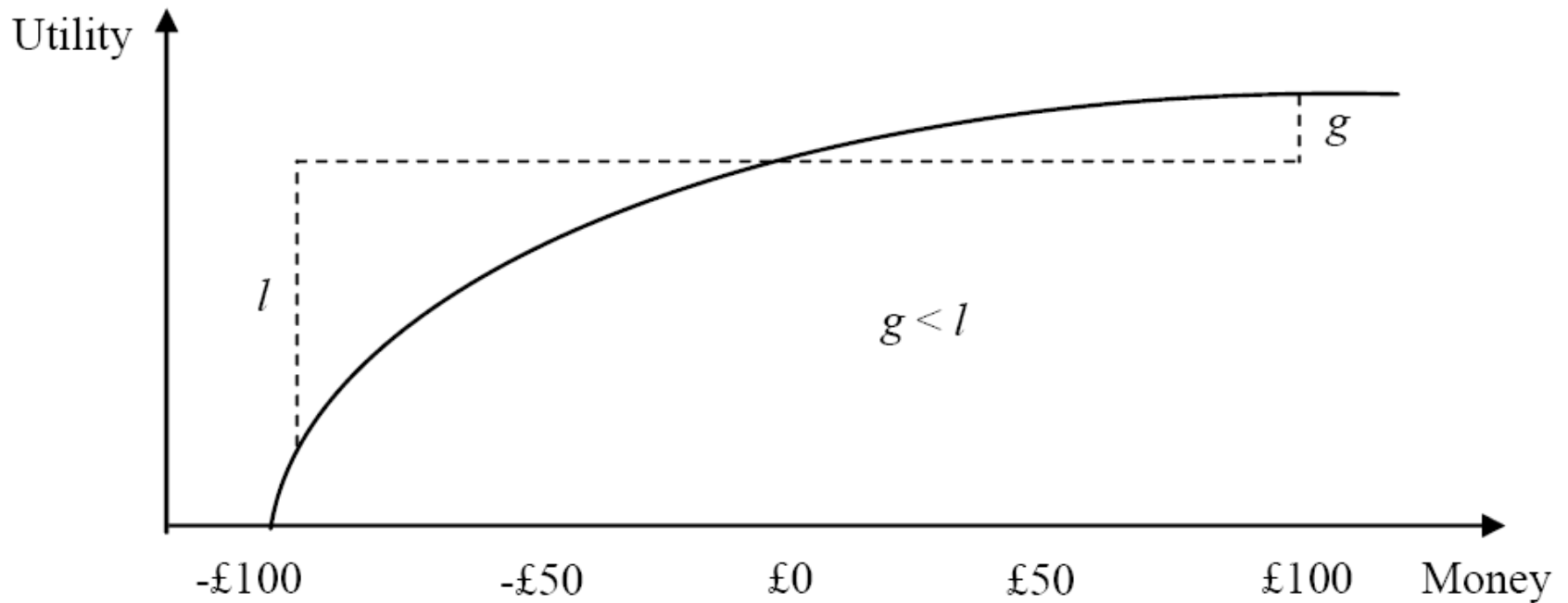
# Is Expected Value sufficient?

- Lottery 1
  - returns \$0 always
- Lottery 2
  - return \$1,000,000 (prob 0.5) and -\$1,000,000 with prob 0.5
- Which is better?

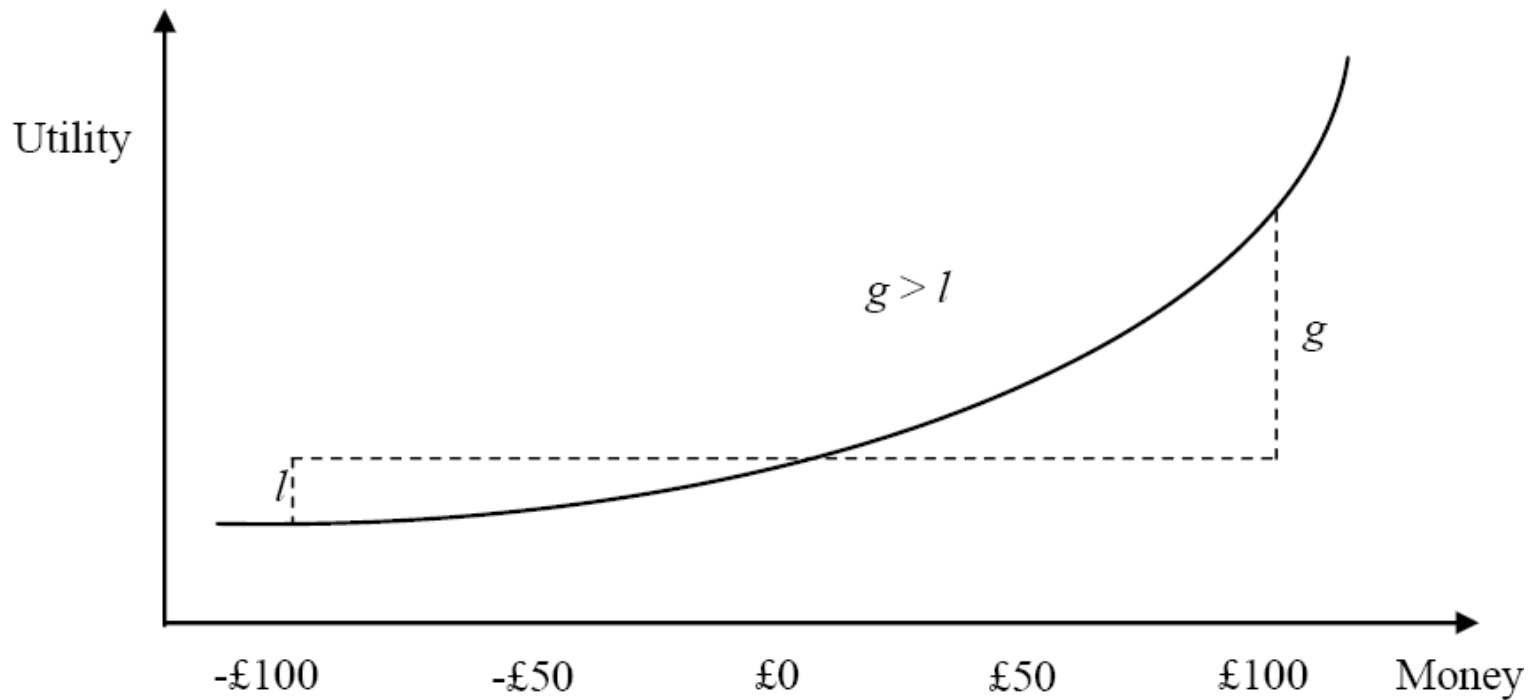
# Utility Theory

- Adds a layer of utility over rewards
- Risk averse
  - |Utility| of high negative money is much MORE than utility of high positive money
- Risk prone
  - Reverse
- Use expected utility criteria...

# Utility function of risk-averse agent



# Utility function of a risk-prone agent



# Utility function of a risk-neutral agent

