#### **Topic 1: Decision Theory**



### **Outcomes & Objectives**

- Master the ability to study/model and analyze agent choices (lotteries) using *expected utility theory*.
  - Demonstrate why people do not maximize expected rewards.
  - Characterize agent's choice preferences using utilities, and outcome beliefs using subjective probabilities.
  - Develop an intuitive axiomatic framework in which agent picks choices to maximize his/her expected utility.
- Illustrate the limitations of EUT and formulate models to better accomodate various deviating behaviors.
  - Relate axiom violations to other well-known normative models.
    Identify certain deviations from experiments and associate them with descriptive models.
- Devise and become proficient in a decision model based on domination, when agents cannot evaluate beliefs.

### Philosophy of Decision Theory

- ► Agent: One decision maker (or a team of multiple decision makers working in tandem) in a given system.
- ► Agent Rationality: The philosophy (principle) used by agents to make decisions.

How does an agent make decisions under uncertainty? Can we model agent rationality mathematically?

#### Two fundamental approaches:

- Normative Models: Identification of optimal decision outcomes – Prescriptive in nature
- Descriptive Models: Describe observed behaviors using consistent rules/models.

### **Modeling Choice Uncertainty**

- $\mathcal{X}$ : (Discrete) Set of all possible prizes
- ► ∆(X): Set of all possible randomizations over choices, called a probability *simplex*



However, these choice probabilities are conditional to the information available at the agent.

### Modeling Choice Uncertainty (cont...)

- Say, you have two route choices: A and B.
- $\blacktriangleright$  Normally, decide A or B based on preference evaluation.

What if there was an unexpected accident in route A?

- Outcomes rely on the state of the choice experiment.
- $\Omega$ : Set of all possible states.

#### Definition

A lottery is any probability distribution  $f : \Omega \to \Delta(\mathcal{X})$ that specifies a non-negative number f(x|t) for every prize  $x \in \mathcal{X}$  and every state  $t \in \Omega$  such that  $\sum_{x \in \mathcal{X}} f(x|t) =$ 1 holds true for every state  $t \in \Omega$ .

•  $\mathcal{L}$ : Set of all such lotteries in the choice experiment.

#### **Randomization of Lotteries**

Given two lotteries  $f,g\in\mathcal{L}$ , and a number  $\alpha\in[0,1]$ , the lottery  $\alpha f+(1-\alpha)g$  denotes a lottery in  $\mathcal{L}$  such that

$$(\alpha f + (1 - \alpha)g)(x|t) = \alpha f(x|t) + (1 - \alpha)g(x|t)$$

for all  $x \in \mathcal{X}$  and  $t \in \Omega$ .



**Example:**  $\alpha^{p_2}$  is the probability with which an accident can take place in route A.

#### **Preference Relations**

Consider any two lotteries  $f, g \in \mathcal{L}$  (finite, countably infinite, or uncountable) be given. Assume that an event  $E \in \mathcal{E}$  has been observed that reveals state information.

**•** Binary Relation:  $f \succ_E g$ .

Example: f is greater than g, f is more tasty than g

• Negation: 
$$\neg(f \succ_E g)$$
.

Example: f is not greater than g, f is not as tasty as g

► Indifference:  $f \sim_E g$ .

Example: f is equal to g, f is similar (incomparable) in taste to g

• Weak Relation:  $f \succeq_E g$ .

Example: f is greater than or equal to g, f is at least as tasty as gSid Nadendla (CS 5408: Game Theory for Computing) 7

#### St. Petersburg Paradox

(Invoked in Blaise Pascal's Wager, published in Pensèes in 1670)

Consider the following choice experiment...



**Final outcome:** Accumulate all the rewards obtained over all time instances of the experiment.

Can you formally state this experiment as a set of lotteries?

# If you were to choose the length of play beforehand, how long would you play this game?

### St. Petersburg Paradox (cont...)

(Invoked in Blaise Pascal's Wager, published in Pensèes in 1670)



In other words, we do not maximize expected rewards!

### St. Petersburg Paradox (cont...)

(Daniel Bernoulli in Commentaries of the Imperial Academy of Science of Saint Petersburg in 1738)









T = 1Outcome: Win 2, if heads in T = 1Win 0, otherwise.

T = 2Outcome: Win 4, if heads in T = 1, 2Win 0, otherwise.

T = 3T = 4Outcome: Outcome: Win 8, if heads in T = 1, 2, 3 Win 16, if heads in T = 1, 2, 3, 4Win 0, otherwise. Win 0, otherwise.

#### **Decreasing Marginal Utilities:**

Expected utility = 
$$\ln 2 \times \frac{1}{2} + \ln 4 \times \frac{1}{4} + \cdots$$
  
=  $\left[\frac{1}{2} + \frac{2}{4} + \frac{3}{8} + \cdots + \frac{k}{2^k} + \cdots\right] \ln 2$   
<  $\infty$ .

However, logarithmic distortion of rewards does not characterize diverse choice preferences across different agents.

### **Ordinal Utility**

#### Definition

Ordinal utility is any deterministic function  $u : \mathcal{X} \times \Omega \to \mathbb{R}$ such that, for any  $E \subseteq \Omega$ ,

 $x_1 \succeq_E x_2 \iff u(x_1, t) \ge u(x_2, t) \text{ for all } x_1, x_2 \in \mathcal{X}, t \in E.$ 

**Example:** Consider an agent who is presented with four choice outcomes,  $\mathcal{X} = \{a, b, c, d\}$  and  $\Omega$  contains only one state. Let the agent's preference ordering be  $a \succ b \succ c \succ d$ .

Then, how can we assign real numbers to outcomes so as to reflect the above preference ordering?

Assignments: Uncountably infinite possibilities...

▶ a:4,b:3,c:2,d:1, a:100,b:50,c:10,d:0

Such assignments do not capture the degree of agent's preferences. **Does such utility functions exist in all choice experiments?** Sid Nadendla (CS 5408: Game Theory for Computing) 11

#### **Expected Utility**

#### Definition

Given any conditional distribution p, utility function u, lottery  $f \in \mathcal{L}$  and any event  $E \in \mathcal{E}$ , the expected utility of the prize determined by f is given by

$$\mathbb{E}_p(u(f)|E) = \sum_{t \in E} p(t|E) \sum_{x \in \mathcal{X}} u(x,t) f(x|t).$$

#### **Expected Utility: Example**

Suppose a commuter has two route choices A and B. Route A is a local road with utility  $u_A = 0.3$ . Route B is a highway route, which has a utility

 $u_B = \begin{cases} 1, & \text{if Route } A \text{ is normal} \\ 0.5, & \text{if Route } A \text{ is under construction} \end{cases}$ 

if there is no accident, and

 $u_B = \begin{cases} 0.25, & \text{if Route } A \text{ is normal} \\ 0.1, & \text{if Route } A \text{ is under construction} \end{cases}$ 

if there is an accident. Let Route B be under construction with probability 0.4. If the commuter takes Route B with probability 0.75, then the expected utilities are given by

 $\begin{array}{lll} EU_{NoAccident} & = & 0.6 \times (1 \times 0.75 + 0.3 \times 0.25) + 0.4 \times (0.5 \times 0.75 + 0.3 \times 0.25) \\ & = & 0.675 \\ \\ EU_{Accident} & = & 0.6 \times (0.25 \times 0.75 + 0.3 \times 0.25) + 0.4 \times (0.1 \times 0.75 + 0.3 \times 0.25) \\ & = & 0.2175 \end{array}$ 

#### **Axioms of Decision Theory**

Basic properties that a rational agent's preferences may satisfy:

- 1. Completeness: Either  $f \succ_E g$ , or  $g \succ_E f$ . or  $f \sim_E g$
- 2. Transitivity: If  $f \succeq_E g$  and  $g \succeq_E h$ , then  $f \succeq_E h$ .
- 3. **Relevance:** If  $f(\cdot|t) = g(\cdot|t)$  for all  $t \in E$ , then  $f \sim_E g$ .
- 4. Monotonicity: If  $f \succeq_E g$  and  $0 \le \beta \le \alpha \le 1$ , then  $\alpha f + (1 - \alpha)g \succeq_E \beta f + (1 - \beta)g.$
- 5. Continuity: If  $f \succeq_E g$  and  $g \succeq_E h$ , then there exists  $\alpha_g \in [0,1]$  such that

$$g \sim_E \alpha_g f + (1 - \alpha_g)h.$$

### Axioms of Decision Theory (cont...)

6. Objective Substitution: If  $f_1 \succeq_E g_1$  and  $f_2 \succeq_E g_2$  and  $\alpha \in [0, 1]$ , then

$$\alpha f_1 + (1-\alpha)f_2 \succeq_E \alpha g_1 + (1-\alpha)g_2.$$

- 7. Subjective Substitution: If  $f \succeq_{E_1} g$  and  $f \succeq_{E_2} g$  and  $E_1 \cap E_2 = \emptyset$ , then  $f \succeq_{E_1 \cup E_2} g$ .
- Interest: For every t ∈ Ω, there exists at least one pair of prizes x<sub>1</sub>, x<sub>2</sub> ∈ X such that x<sub>1</sub> ≻<sub>{t}</sub> x<sub>2</sub>.
- 9. State Neutrality: For any two states  $s, t \in \Omega$ , if  $f(\cdot|s) = f(\cdot|t), g(\cdot|s) = g(\cdot|t)$  and  $f \succeq_{\{s\}} g$ , then  $f \succeq_{\{t\}} g$ .

### **Expected Utility Maximization (EUM)**

Credit: Von Neumann and Morgenstern, 1947

#### Theorem 1

Axioms 1-8 are jointly satisfied if and only if there exists a utility function  $u:\mathcal{X}\times\Omega\to[0,1]$  and a conditional probability function  $p:\mathcal{E}\to\Delta(\Omega)$  such that

•  $f \succeq_E g$  if and only if  $\mathbb{E}_p(u(f)|E) \ge \mathbb{E}_p(u(g)|E)$  for all  $f, g \in \mathcal{L}$  and for all  $E \in \mathcal{E}$ .

If there are more than two lotteries, by transitivity, the most preferred lottery also has the largest expected utility!

### State-Independent Utility Maximization

Credit: Von Neumann and Morgenstern, 1947

**State-Independent Utility:** u(x,t) = U(x), for all t, x. (inspired from *State Neutrality* axiom)

#### Corollary 1

Axioms 1-9 are satisfied if and only if there exists a state-independent utility function  $u: \mathcal{X} \times \Omega \to [0,1]$  and a conditional probability function  $p: \mathcal{E} \to \Delta(\Omega)$  such that

•  $f \succeq_E g$  if and only if  $\mathbb{E}_p(u(f)|E) \ge \mathbb{E}_p(u(g)|E)$  for all  $f, g \in \mathcal{L}$  and for all  $E \in \mathcal{E}$ .

#### Given that utility functions exist under Axioms 1-9, how can we construct<sup>1</sup> them from agents' revelations?

<sup>1</sup>This is beyond the scope of this course. However, interested students may refer to Revealed Preference Theory and Afriat's Theorem.

#### Example

Suppose an agent wants to buy a used 4-volume boxed set of *The Art of Computer Programming* by Don Knuth. Assume that the item arrives in any of the following conditions: *Very Good*, *Good* and *Acceptable*. Following are the two marketplaces available to the agent:

- Market 1: Good with probability 0.3, or Very Good with probability 0.7.
- Market 2: Acceptable with probability 0.3, or Good with probability 0.2, or Very Good with probability 0.5.

Let the utilities be  $u_A = 100$ ,  $u_G = 200$  and  $u_{VG} = 300$ . Then,

$$\begin{split} EU(\mathsf{Market 1}) &= 0.3 \times 200 + 0.7 \times 300 = 270 \\ EU(\mathsf{Market 2}) &= 0.3 \times 100 + 0.2 \times 200 + 0.5 \times 300 = 220 \end{split}$$

#### **Prescription:** Market $1 \succ$ Market 2

#### **Affine Transformation**

Credit: Von Neumann and Morgenstern, 1947

#### Theorem 2

Let  $E \in \mathcal{E}$  be any given subjective event. Suppose the agent's preferences satisfy Axioms 1-9, and let u:  $\mathcal{X} \times \Omega \rightarrow [0,1]$  and  $p : \mathcal{E} \rightarrow \Delta(\Omega)$  denote the stateindependent utility function and conditional probability function as stated in Corollary 1. Let v be a stateindependent utility function and q be a conditional probability function, which represent the preference ordering  $\succeq_E$ . Then, there exists numbers a > 0 and b such that

$$v(x) = au(x) + b$$
, for all  $x \in \mathcal{X}$ .

### **Types of Utility Functions**

- ► Risk Aversion: A concave function of monetary value (wealth), i.e., u(λx + (1 − λ)y) > λu(x) + (1 − λ)u(y) ∀ x, y ∈ X
- ► Risk Seeking: A convex function of monetary value (wealth), i.e., u(λx + (1 − λ)y) < λu(x) + (1 − λ)u(y)</p>
- ▶ Risk Neutral: An affine function of monetary value (wealth).



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### Limitations of EUM: Preference Intensity

- $\begin{array}{l} \blacktriangleright \quad x_1 \succ x_2 \succ x_3 \succ x_4 \text{ and} \\ u(x_1) u(x_2) > u(x_3) u(x_4) \\ \not \Longrightarrow \quad \text{change from } x_2 \text{ to } x_1 \\ \text{is more preferred than change} \\ \text{from } x_4 \text{ to } x_3. \end{array}$ 
  - Ordinal utility does not measure both intensity and direction of preferences.
  - Need for cardinal utility to capture the strength of preferences.
  - Note: The notion of cardinal utility has a different meaning in measurement theory in psychology, which is irrelevant to our discussion.
  - Captures information framing effects

#### Approaching efficiency

Improvements to vehicles' fuel consumption at the low end of the scale have a greater effect than those to already more efficient vehicles.



#### Limitations of EUM: Other Inconsistencies

People's preferences does not necessarily satisfy Axioms 1-9.

- ► Bounded Rationality: Decisions under Limited Time/Memory/Attention ⇒ Satisficing.
- Behavioral Complexities: Loss Aversion, Probability Weighting, Framing Effects and Preference Reversals, Anchoring Bias, Confirmation Bias, Polarization...
- Prosociality: Social Reputation/Pride

#### **Allias Paradox**

Credit: Maurice Allias, 1953

#### **Experiment 1 Experiment 2** Lottery 1A Lottery 1B Lottery 2A Lottery 2B Winnings Winnings Winnings Chance Winnings Chance Chance Chance STATITON DOLLARS THE INTITLED STATUES OF AMERICA 89% 89% THE UNTITLED STATUES OF AMER 90% THE UNTITLED STATUES OF AMERIC 1% 100% CALIFORNIO CONTRACTO 10% 10%

#### Pick one lottery from each experiment!

## Allias Paradox (cont...)

Credit: Maurice Allias, 1953

Experiment 1			Experiment 2				
Lottery 1A		Lottery 1B		Lottery 2A		Lottery 2B	
Winnings	Chance	Winnings	Chance	Winnings	Chance	Winnings	Chance
	100%		89%		89%		90%
			1%				
			10%		11%		10%

#### Usually $1A \succ 1B$ and $2B \succ 2A$ – inconsistent with EUM!

#### **Prospect Theory – A Descriptive Model**

Credit: Daniel Kahneman and Amos Tversky, 1979

Assuming that the choice experiment has only one state,



## Variants<sup>2</sup> of Expected Utility Theory

		TABLE 1
		NINE VARIANTS OF THE
		Expected Utility Model
1.	$\Sigma p_i x_i$	Expected Monetary Value
2.	$\sum p_i v(x_i)$	Bernoullian Expected Utility (1738)
3.	$\sum p_i u(x_i)$	von Neumann-Morgenstern Expected Utility (1947)
4.	$\sum f(p_i)x_i$	Certainty Equivalence Theory (Schneeweiss,
	•	1974; Handa, 1977; de Finetti, 1937)
5.	$\sum f(p_i)v(x_i)$	Subjective Expected Utility (Edwards, 1955)
6.	$\sum f(p_i)u(x_i)$	Subjective Expected Utility (Ramsey, 1931;
	•	Savage, 1954; Quiggin, 1980)
7.	$\sum w(p_i)x_i$	Weighted Monetary Value
8.	$\sum w(p_i)v(x_i)$	Prospect Theory (Kahneman and Tversky, 1979)
9.	$\sum w(p_i)u(x_i)$	Subjectively Weighted Utility (Uday Karmarkar, 1978)

Note: v(x) denotes an interval scaled utility measure constructed under certainty; u(x) denotes one constructed via lotteries.

#### Several other models have been proposed since 1982...

<sup>&</sup>lt;sup>2</sup>Credit: P. J. H. Schoemaker, "The Expected Utility Model: Its Variants, Purposes, Evidence and Limitations", J. Economic Literature, vol. 20, no. 2, pp. 529-563, June 1982. Sid Nadendla (CS 5408: Game Theory for Computing)

### **Decisions Under Ignorance**

Sometimes...

- Difficult to assess subjective probabilities
  - Unknown environments
  - Limited computational capabilities
  - ► Limited time/memory...
- ► Easy to eliminate some choices the *dominated* ones!

How can we identify and eliminate the dominated choices?

#### Dominance

Let the agent have

- state-dependent utility function  $u: \mathcal{X} \times \Omega \to \mathbb{R}$ ,
- a subjective probability p(t) in a state  $t \in \Omega$ .

Let the agent does not make random decisions, i.e. lotteries are **deterministic**!

#### Definition

The choice  $y \in \mathcal{X}$  is dominating in a deterministic experiment only if

$$\sum_{t \in \Omega} p(t)u(y,t) \ge \sum_{t \in \Omega} p(t)u(x,t)$$

for all  $x \in \mathcal{X}$ .

**Note:** Dominance may hold true only in some  $p(t) \in \Delta(\Omega)$ .

## Dominance (cont...)

For what distributions of p can a given choice  $y \in \mathcal{X}$  be optimal for an agent with a state-dependent utility function u?

#### Theorem

Given  $u : \mathcal{X} \times \Omega \to \mathbb{R}$  and given  $y \in \mathcal{X}$ , the set of all  $p \in \Delta(\Omega)$  such that y is optimal is convex.

**Example:** Suppose  $\mathcal{X} = \{\alpha, \beta, \gamma\}$ ,  $\Omega = \{t_1, t_2\}$  and the corresponding utilities are as follows.

Decision	State $t_1$	State $t_2$
$\alpha$	8	1
eta	5	3
$\gamma$	4	7

### Dominance (cont...)

Let  $p(t_1) = p$ . Then, we have  $p(t_2) = 1 - p$ .

• The decision  $\alpha$  is optimal if and only if

$$\begin{split} &8p+1[1-p] \geq 5p+3(1-p),\\ &8p+1[1-p] \geq 4p+7(1-p). \end{split}$$

In other words,  $p \ge 0.6$ .

• The decision  $\beta$  is optimal if and only if

$$\begin{array}{l} 5p + 3(1-p) \geq 8p + 1[1-p], \\ 5p + 3(1-p) \geq 4p + 7(1-p). \end{array}$$

But, this is an empty set!

So,  $\beta$  can never be optimal for any set of beliefs! This is called a *strongly dominated* choice.

#### A Caveat...

Just because  $\alpha$  is the optimal choice in state  $t_1$  and  $\gamma$  is the optimal choice is state  $t_2$ , we **cannot** claim that  $\beta$  (an intermediate choice) is dominated!

**Example:** Consider the earlier example with the following utility table.

Decision	State $t_1$	State $t_2$
$\alpha$	8	1
eta	6	3
$\gamma$	4	7

Now, the decision  $\beta$  is optimal whenever  $5/7 \le p \le 1/3$ .

#### **Dominance and Lotteries**

## Definition A choice $y \in \mathcal{X}$ is strongly dominated by a lottery $f \in \Delta(\mathcal{X})$ if $\sum_{x \in \mathcal{X}} f(x|t)u(x,t) > u(y,t)$ for all $t \in \Omega$ .

\_\_\_\_

#### Theorem

Given  $u: \mathcal{X} \times \Omega \to \mathbb{R}$  and any choice  $y \in \mathcal{X}$ , there exists a lottery  $f \in \Delta(\mathcal{X})$  such that y is strongly dominated by f, if and only if there does not exist any probability distribution  $p \in \Delta(\Omega)$  such that y is optimal in a deterministic experiment.

### Weak Dominance

#### Definition

A choice  $y \in \mathcal{X}$  is weakly dominated by a lottery  $f \in \Delta(\mathcal{X})$  if

$$\sum_{x \in \mathcal{X}} f(x|t)u(x,t) \ge u(y,t)$$

for all  $t \in \Omega$ , and there exists at least one state in  $\Omega$  such that the above inequality is strict.

**Example:** Let  $\mathcal{X} = \{\alpha, \beta\}$ ,  $\Omega = \{t_1, t_2\}$  and utilities as given below.

Decision	State $t_1$	State $t_2$
$\alpha$	5	3
$\beta$	5	1

Here,  $\beta$  is weakly dominated by  $\alpha$  (due to the case where p = 1.). Sid Nadendla (CS 5408: Game Theory for Computing) 33

### Summary

- ► St. Petersburg Paradox: Why do we need utility functions?
- Preference Axioms: How does ideal agent's preferences look like?
- Expected Utility Maximization: Ideal agents maximize expected utilities.
- *Limitations:* People are not ideal agents.
- Allais Paradox and Prospect Theory: One example of a descriptive model.
- *Domination:* Decision making under ignorance