All the axioms that Savage and von Neumann-Morgenstern use seem so reasonable.

Savage views his axioms as characterizing rationality Is that reasonable?

They certainly don't always characterize how people act

Recall the Allais paradox:

The set of prizes is $X = \{\$0, \$1, 000, 000, \$5, 000, 000\}.$

▶ Which probability do you prefer: $p_1 = (0.00, 1.00, 0.00)$ or $p_2 = (0.01, 0.89, 0.10)$?

Allais Paradox

The set of prizes is $X = \{\$0, \$1, 000, 000, \$5, 000, 000\}.$

- ► Which probability do you prefer: p₁ = (0.00, 1.00, 0.00) or p₂ = (0.01, 0.89, 0.10)?
- ▶ Which probability do you prefer: $p_3 = (0.90, 0.00, 0.10)$ or $p_4 = (0.89, 0.11, 0.00)$?

Many subjects report: $p_1 \succ p_2$ and $p_3 \succ p_4$

Inconsistent with Maximizing Expected Utility

Suppose (u_0, u_1, u_5) represents \succ . Then $p_1 \succ p_2$ implies

$$\begin{split} u_1 > .01 u_0 + .89 u_1 + .1 u_5 \\ .11 u_1 - .01 u_0 > .1 u_5 \\ .11 u_1 + .89 u_0 > .1 u_5 + .9 u_0. \end{split}$$

So $p_4 \succ p_3$. Which axiom is violated? Independence: $a \succ b$ iff $\alpha a + (1 - \alpha)c \succ \alpha b + (1 - \alpha)c$.

homework – explain exactly how.

We considered the *Ellsberg paradox* in the first class:

There is one urn with 90 balls: 30 of these balls are red (R) and the rest are either blue (B) or yellow (Y). Consider the following two choice situations:

- I: *a*. Win \$100 if a ball drawn from the urn is R and nothing otherwise.
 - a'. Win \$100 if a ball drawn from the urn is B and nothing otherwise.

Ellsberg Paradox

There is one urn with with 90 balls: 30 of these balls are red (R) and the rest are either blue (B) or yellow (Y). Consider the following two choice situations:

- I: *a*. Win \$100 if a ball drawn from the urn is R and nothing otherwise.
 - a'. Win \$100 if a ball drawn from the urn is B and nothing otherwise.
- II: b. Win \$100 if a ball drawn from the urn is R or Y and nothing otherwise.
 - b'. Win \$100 if a ball drawn from the urn is B or Y and nothing otherwise.

Inconsistent with SEU

Suppose a decision maker's preferences are such that $a \succ a'$ and $b' \succ b.$

If there are subjective probabilities then the first choice implies that the probability of a red ball is greater than the probability of a blue ball and the second choice implies the reverse.

Which of Savage's axioms is violated?

▶ Independence: Remember that an act is a function from states to outcomes. Let $T \subseteq S$ be a subset of states. Then

$$f_Tg \succeq f'_Tg$$
 iff $f_Th \succeq f'_Th$.

Homework: prove that the standard choices in the Ellsberg paradox violate this.

These examples suggest that maximizing expected utility is not obviously always the "right" thing to do.

- But if we don't do that, what should we do?
 - ▶ We've already seen some alternatives (regret, maximin, ...)
 - Let's go back to one of them: maxmin expected utility

Maxmin Expected Utility Rule

Suppose that the decision maker's uncertainty can be represented by a set ${\mathcal P}$ of probabilities . Let

$$\underline{E}_{\mathcal{P}}(u_a) = \inf_{\Pr \in \mathcal{P}} \{ E_{\Pr}(u_a) : \Pr \in \mathcal{P} \}$$

Recall the maximin expected utility rule: (covered earlier in the course):

•
$$a >_{\mathcal{P}}^{1} a'$$
 iff $\underline{E}_{\mathcal{P}}(u_a) > \underline{E}_{\mathcal{P}}(u_{a'})$

This is like maximin:

Optimizing the worst-case expectation

This could explain the Ellsberg Paradox:

• Let $\mathcal{P} = \{(1/3, p_B, p_Y) : 0 \le p_B \le 2/3\}$

Gilboa and Schmeidler axiomatized the maxmin expected utility rule

- It does not satisfy independence
- Gilboa and Schmeidler replaced independence by a weaker axiom.

- ▶ Suppose that there are four states, w, x, y, z.
- f is the act where you get 1 in state w, 0 in all other states
- f' is the act where you get 1 in state x, 0 in all other states
- g is the act where you get 1 in state y, 0 in all other states
- h is the act where you get 1 in state z, 0 in all other states

• Let $A = \{w, x\}$.

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- Let $A = \{w, x\}$.

| act | w | x | y | z |
|----------|---|---|---|---|
| $f_A g$ | 1 | 0 | 1 | 0 |
| $f'_A g$ | 0 | 1 | 1 | 0 |
| $f_A h$ | 1 | 0 | 0 | 1 |
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Now suppose that $Pr(\{w, y\}) = Pr(\{x, z\}) = 1/2$, but you don't know the probability of individual states. It seems reasonable that

• $f_Ag \succ f'_Ag$ (getting 1 with probability 1/2 is better than getting 1 with some unknown probability)

•
$$f'_A h \succ f_A h$$
 (same logic).

• Suppose that there are four states, w, x, y, z.

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- $f_Ag \succ f'_Ag$ (getting 1 with probability 1/2 is better than getting 1 with some unknown probability)
- $f'_A h \succ f_A h$ (same logic). Independence fails!

Comonotonic independence

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Acts f and g are $\ensuremath{\textit{comonotonic}}$ if there do not exist states s and t such that

 $f(s) \succ f(t) \text{ and } g(t) \succ g(s)$

- ▶ f and g are comonotonic if you can't be happier to be in state s than state t when doing f and be happier to be in state t than state s when doing g.
- If h is a constant act, then f and h are comonotonic for all acts f (since we never have h(s) ≻ h(t)).

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- If h is a constant act, then f and h are comonotonic for all acts f (since we never have h(s) ≻ h(t)).
- ► [Comonotonic Independence:] If f and h and g and h are both comonotonic, then $f \succ g$ iff for all $\alpha \in (0, 1]$, $\alpha f + (1 \alpha)h \succ \alpha g + (1 \alpha)h$.

Gilboa and Schmeidler proved a representation theorem for their axiomatization:

- An agent's preference order obeys the Gilboa-Schmeidler axioms iff he has a utility function (unique up to affine transformations) and a set *P* of probability measures such that a₁ ≻ a₂ iff a₁ >¹_P a₂.
 - If the agent's preference order obeys the axioms iff he is acting like a maxmin expected utility maximizer
 - \mathcal{P} is not unique, but its convex closure is.
 - ► The *convex closure* of *P* is the smallest closed convex set containing *P*.
 - If $p, p' \in P$, then $\alpha p + (1 \alpha)p' \in \mathcal{P}$
 - $\blacktriangleright \ \mathcal{P}$ also contains its limit points
 - ▶ if (u, \mathcal{P}) and (u, \mathcal{P}') both represent the axioms, then the convex closure of \mathcal{P} = the convex closure of \mathcal{P}'

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 - $\blacktriangleright \ \mathcal{P}$ also contains its limit points
 - if (u, \mathcal{P}) and (u, \mathcal{P}') both represent the axioms, then the convex closure of $\mathcal{P} =$ the convex closure of \mathcal{P}'

While maxmin expected utility captures some aspects of human behavior, but certainly doesn't capture all its complexities.

Cumulative Prospect Theory

Cumulative prospect theory (CPT) is a model of decision making due to Tversky and Kahneman (1992)

- An improvement over *prospect theory*, their earlier approach
- CPT was a significant part of why Kahneman won the Nobel Prize (Tversky would have won it too had he not died)
- It is largely viewed as the best descriptive theory of decision making that we have
 - It's used by practitioners

Key insights of CPT:

- People care more about losses than gains
- People tend to overweight extreme unlikely events and underweight "average" events.

In the utility functions considered in homework and in class up to now, we took the utility of (final) wealth

- CPT considers a reference point, and takes utility of wealth relative to that reference point
 - Above the reference point is a gain; below it is a loss.

The DM is choosing among prospects (i.e., acts) mapping a state space S to a finite outcome space

$$X = \{x_{-m} \dots, x_{-1}, x_0, \dots, x_n\}.$$

- ▶ The x_i 's are ordered by goodness: $x_j \succ x_i$ if j > i
- x_0 is the reference point

•
$$v(x_0) = 0$$
; $v(x_i) > 0$ if $i > 0$; $v(x_i) < 0$ if $i < 0$.

For prospect f, A_i^f is the set of states s such that $f(s) = x_i$.

$$\blacktriangleright V(f) = \sum_{i=-m}^{n} \pi_i^f v(x_i).$$

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- ► Like expected utility, but π^f_i is not the probability of getting outcome i with act f
- ► A *capacity* W is a generalization of a probability measure:

•
$$W(\emptyset) = 0; W(S) = 1; W(A) \le W(B)$$
 if $A \subseteq B$.

- $\pi_n^f = W(A_n^f); \ \pi_{-m}^f = W(A_{-m}^f)$ $\pi_i^f = W(A_i^f \cup \ldots \cup A_n^f) W(A_{i+1}^f \cup \ldots \cup A_n^f) \ \text{if} \ 0 \le i \le n-1$
- $\pi_i^f = W(A_{-m}^f \cup \ldots \cup A_i^f) W(A_{-m}^f \cup \ldots \cup A_{i-1}^f)$ if $-m < i \le 0$
 - \blacktriangleright W lets us treat losses differently from gains
 - W lets us overweight extreme unlikely events $(A_{-m}^f \text{ and } A_n^f)$
 - if W is a probability, then π^f_i is the probability of getting outcome x_i with f.

Framing Effects—Kahneman and Tversky

A disease is expected to kill 600 people. Two alternative programs have been proposed:

- Program A: 200 people will be saved
- Program B:
 - ▶ probability 1/3: 600 people will be saved
 - probability 2/3: no one will be saved

Which program would you favor?

Framing Effects—Kahneman and Tversky

A disease is expected to kill 600 people. Two alternative programs have been proposed:

- Program C: 400 people will die
- Program D:
 - ▶ probability 1/3: no one will die
 - ▶ probability 2/3: 600 will die

Which program would you favor?

Framing Effects—Kahneman and Tversky

Kahneman and Tversky found:

- ▶ 72% chose A over B.
- ▶ 22% chose C over D.

But if 200 people will be saved out of 600 is the same to the decision-maker as 400 people will die out of 600, and so on, then A and C are identical and so are B and D.

 losses (death) are weighed more heavily than gains (people staying alive)

Conjunction Fallacy or Failure of Extensionality

Tom is a rancher from Montana.

Which bet would you prefer?

- ▶ Win \$10 if Tom drives either a Ford or a Chevy, otherwise win nothing
- ► Win \$10 if Tom drives either a Chevy truck or Ford truck, otherwise win nothing

Kahneman and Tversky experiment:

Linda is 31 years old, single, outspoken, and very bright. She majored in philosophy. As a student, she was deeply concerned with issues of discrimination and social justice, and also participated in anti-nuclear demonstrations. Which is more probable?

Linda is a bank teller.

Linda is a bank teller and is active in the feminist movement.
85% of subjects chose the second option.

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- Linda is a bank teller.
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Another systematic error: Ignoring priors

A cab was involved in a hit and run accident last night. Two cab companies, Green and Blue, operate in the city. You know:

- A witness identified the cab as Blue.
- ▶ Witnesses are pretty reliable: Tests have shown that in similar cirumstances witnesses correctly identify each of the two cabs 80% of the time and misidentify them 20% of the time.
- $\blacktriangleright~85\%$ of the cabs in the city are Green the rest are Blue. What is the probability that the cab involved in the accident was Blue?

The correct answer requires Bayes rule:

$$Pr(B|idB) = \frac{Pr(idB|B)Pr(B)}{Pr(idB)}$$
$$= \frac{(.8)(.15)}{(.8)(.15) + (.2)(.85)}$$
$$= .41$$

Computational limitations

People use "fast and frugal" heuristics [Gigerenzer]

- simple decision rules for making decisions that often work surprisingly well
- Which has a larger population: Detroit or Milwaukee?
 - Europeans do better on this question than Americans!

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Claim: People are rational, but the "mistakes" that we're seeing are the outcome of computational limitations [Wilson; Halpern/Pass/Seeman]