Session 9: The expected utility framework

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Questions

- How do humans make decisions when faced with uncertainty?
- How can decision theory be used to solve problems of portfolio choice?

Decision making under uncertainty

- Ordinary' utility theory deals with problems like apples and oranges: Look for tangency of the budget constraint w.r.t. indifference curves.
- What is a comparable technology for dealing with uncertainty?

Historical introduction

First attempts

- One plausible theory: "Humans behave asif they maximise E(x)".
- It appears reasonable to think that when faced with decisions, humans compute E(x) and choose the option with the highest E(x). For example, the NPV-based method of choosing between alternative cashflows.
- This proves to be an incomplete solution.

The St. Petersburg paradox

- You pay a fixed fee to enter a game.
- A coin will be tossed until a head appears.
- You win Rs.1 if the head is on the 1st toss; Rs.2 if on the 2nd, Rs.4 if on the 3rd toss, etc.
- How much would you be willing to pay to enter the game?

(Posed by Daniel Bernoulli, 1738).

Analysis

 \blacksquare Pr(the first head appears on the kth toss) is:

$$p_k = \frac{1}{2^k}$$

- Pr(you win more than Rs.1024) is less than 0.001.
- BUT the expected winning is infinite!

$$E = \sum_{k=1}^{\infty} p_k 2^{k-1} = \sum_{k=1}^{\infty} \frac{1}{2} = \infty$$

- **The sum diverges to \infty.**
- No matter how much you pay to enter (e.g. Rs.100,000), you come out ahead on expectation of present the part of the th

The paradox

- You or I might feel like paying Rs.5 for the lottery.
- But it's expected value is infinity.
- How do we reconcile this?

Expected utility hypothesis

- Theory:
 - "Humans behave asif they maximise E(u(x))".
- There is a fair supply of anomalies and paradoxes, but this remains our benchmark hypothesis.

John von Neumann and Oskar Morgenstern, 1946.

Characteristics of utility functions

Simple utility functions

Exponential	$U(x) = -e^{-ax}$	a > 0
Logarithmic	$U(x) = \log(x)$	
Power	$U(x) = bx^b$	$b \le 1, b \ne 0$. If $b = 1$, it's riskneutral.
Quadratic	$U(x) = x - bx^2$	b > 0. Is increasing only on $x < 1/(2b)$.

Equivalent utility functions

- Two utility functions are equivalent if they yield identical rankings in x.
- Monotonic transforms do not matter. Example:
 - $U(x) = \log(x)$ versus
 - $\overline{U(x)} = a \log(x) + \log c$ is just a monotonic transform.
 - Hence, $V(x) = \log(cx^a)$ is equivalent to $U(x) = \log(x)$.
- Sometimes, it's convenient to force a monotonic transform upon a U(x) of interest, in order to make it more convenient.

Expected utility hypothesis

Calculating expected utility

- When the choice variable x is constant, then E(U(x)) = U(x).
- When the choice variable x is a random variable, then E(U(x)) is driven by the PDF of x.
- If x has k outcomes, each with probability p_k , then

$$E(U(x)) = \sum_{1}^{k} p_i U(x_i)$$

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- Say, $U(x) = 10 + 2x 0.1x^2$
- x has the following PDF:

X	p(x)
-1	0.3
0.5	0.5
1	0.2

• What is E(U(x))?

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U(x) has the following PDF:

X	p(x)	U(x)
-1.0	0.3	7.90
0.5	0.5	10.98
1.0	0.2	11.90

E(U(x)) = 0.3*7.9+10.98*0.5+11.90*0.2 = 10.42

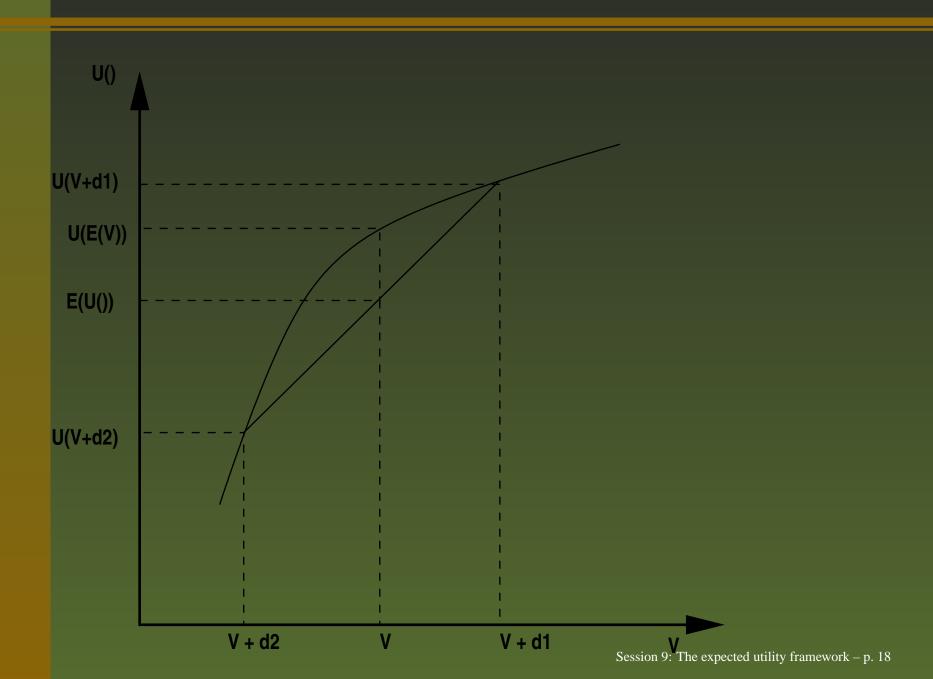
Risk aversion

Definition: A utility function is *risk averse* on [a, b] if it is concave on [a, b]. If U is concave everywhere, it is risk averse. U is concave if for all $0 \le \alpha \le 1$ and on any x, y in [a, b]:

$$U(\alpha x + (1 - \alpha)y) \ge \alpha U(x) + (1 - \alpha)U(y)$$

- Risk aversion: when expected utility across all possibilities is lower than utility of the expectation of all possibilities.
- Greater curvature is greater risk aversion; the straight line utility function is risk—neutral.

Concave utility functions



Certainty equivalence

The certainty equivalent C of a random lottery x is:

$$U(C) = E(U(x))$$

- Under a risk neutral utility function, C = E(x);
- Under a risk averse utility function, C < E(x); The greater the risk aversion, the greater the distance between C and E(x).

NOTE: U() has no units, but C can be nicely interpreted.

Example: U(x) = a + bx

If $x \sim N(\mu_x, \sigma_x^2)$, then

- $E(x) = \mu_x$
- $U(E(x)) = a + b\mu_x$
- $E(U(x)) = E(a + bx) = a + b\mu_x$

$$E(U(x)) = E(x).$$

Here the choice result is the same as if the individual was maximising E(x).

Therefore, a person with this utility function is risk-neutral.

Example: $U(x) = a + bx - cx^2$

If
$$x \sim N(\mu_x, \sigma_x^2)$$
, then

- $U(E(x)) = a + b\mu_x c\mu_x^2$
- $E(U(x)) = E(a+bx-cx^2) = a+b\mu_x-c(\sigma_x^2+\mu_x^2)$
- $E(U(x)) \neq U(E(x)).$ In fact, U(E(x)) > E(U(x)).

A person with this utility function is risk-averse.

person

There is a significant literature on eliciting the risk aversion of a person.

- Ask the user to assign certainty equivalents to a series of lotteries. In principle, this can non-parametrically trace out the entire utility function.
- Choose a parametric utility function, in which case we are down to the easier job of just choosing the parameter values. Once again, the user can be asked to choose between a few lotteries.

Using expected utility hypothesis

Choosing between uncertain alternatives

- Say, θ influences the pdf of a random outcome. For example, for a binomial distribution, $\theta = p$, the probability of success.
- The typical optimisation problem is that a person chooses a parameter θ .
- **How** should the optimal value, θ^* , be chosen?
- When faced with choices θ_1 and θ_2 , the person picks θ_1 iff $EU(\theta_1) > EU(\theta_2)$.
- Therefore, the choice is made as:

$$\theta^* = \arg\max E(U(x(\theta)))$$

- An individual has the utility function U(x) = 10 + 2.5x
- $x_1 \sim N(5.5, 4.5)$
- $x_2 \sim N(4.5, 3.5)$
- Which of x_1, x_2 would the individual choose?

 $\sim N(5.5, 4.5)$

$$E(U(x_1)) = 10 + 2.5\mu_{x_1} = 10 + 2.5 * 5.5 = 23.75$$

 $x_2 \sim N(4.5, 3.5)$

$$E(U(x_2)) = 10 + 2.5\mu_{x_2} = 10 + 2.5 * 4.5 = 21.25$$

Since $E(U(x_1) > E(U(x_2))$, the individual would choose x_1 .

- Another individual has the utility function $U(x) = 10 + 2.5x 0.5x^2$
- $x_1 \sim N(5.5, 4.5)$
- $x_2 \sim N(4.5, 3.5)$
- Which of x_1, x_2 would the individual choose?

 $x_1 \sim N(5.5, 4.5), E(U(x_1))$

$$10 + 2.5\mu_{x_1} - 0.5(\sigma_{x_1}^2 + \mu_{x_1}^2)$$
$$10 + 2.5 * 5.5 - 0.5(4.5 + 5.5^2) = 6.38$$

 $x_2 \sim N(4.5, 3.5), E(U(x_2))$

$$10 + 2.5\mu_{x_2} - 0.5(\sigma_{x_2}^2 + \mu_{x_2}^2)$$
$$10 + 2.5 * 4.5 - 0.5(3.5 + 4.5^2) = 9.38$$

Since $E(U(x_2) > E(U(x_1))$, this individual would choose x_2 .

Non-corner solutions

- In the previous two examples, we forced the two individuals to choose either one or the other.

 These are called **corner solutions** to the optimisation problem.
- What if the two could choose a linear combination of the two choices, ie $\lambda x_1 + (1 \lambda)x_2$ where $0 > \lambda > 1$?
- Assume that the covariance between $x_1, x_2 = 0$.

example of a non-corner solution and risk-neutrality: $\lambda = 0.5$

For the risk neutral individual, $E(U(0.5*x_1 + 0.5*x_2))$

$$= 10 + 2.5(0.5*5.5 + 0.5*4.5) = 10 + 2.5*5.0 = 22.5$$

- This is much less than the original solution of choosing x_1 , where $E(U(x_1) = 23.75)$ This person would choose x_1 above any linear combination with x_2 .
- Observation: risk-neutral individuals prefer corner solutions!

Example of a non-corner solution and risk-aversion: $\lambda = 0.5$

For the risk averse individual,

$$E(U(0.5 * x_1 + 0.5 * x_2))$$
= $10 + 2.5 * 5.0 - 0.2(\sigma_{0.5x_1+0.5x_2}^2)$
= $10 + 2.5 * 5.0 - 0.5 * \frac{3.5 + (4.5 * 4.5) + 4.5 + (5.5 * 5.5)}{4}$
= 15.19

- This is much more than the original solution of choosing x_2 , where $E(U(x_2) = 9.38$ This person would choose this linear combination above the corner solution of only x_2 !
- Observation: risk-averse individuals prefer non-corner solutions!

What is the optimal combination for a risk-averse individual?

In a world with

- Several opportunities, x, with uncertain outcomes where
- **Each** x has a different PDF $f(\theta)$,
- What is the optimal choice of the combination of x for the individual to maximise E(U(x))?

We are back to the original question posed in the last class – the Markowitz problem!