# **Expected Utility Over Money and Risk Aversion**

Econ 3030

Fall 2025

#### Lecture 11

#### Outline

- Expected Utility Of Wealth
- Betting & Insurance
- Risk Aversion
- Certainty Equivalent
- Risk Premium

### **Probability Distribution On Wealth**

- Many applications of expected utility consider preferences on probability distributions of wealth (a continuous variable).
- A probability distribution is characterized by its cumulative distribution function.

#### **Definition**

A cumulative distribution function (cdf)  $F : \mathbb{R} \to [0,1]$  satisfies:

- $x \ge y$  implies  $F(x) \ge F(y)$  (nondecreasing);
- $\lim_{y \downarrow x} F(y) = F(x)$  (right continuous);
- $\lim_{x\to-\infty} F(x) = 0$  and  $\lim_{x\to\infty} F(x) = 1$ .
- <sup>a</sup>Recall that, if it exists,  $\lim_{y\downarrow x} f(y) = \lim_{n\to\infty} f(x+\frac{1}{n})$ .

## **Probability Distribution On Wealth**

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#### **Notation**

- $\mu_F$  denotes the expected value of F, i.e.  $\mu_F = \int x \, dF(x)$ .
- $\delta_x$  is the degenerate distribution function at x; i.e.  $\delta_x$  yields x with certainty:

$$\delta_x(z) = \begin{cases} 0 & \text{if } z < x \\ 1 & \text{if } z \ge x \end{cases}.$$

<sup>&</sup>lt;sup>a</sup>Recall that, if it exists,  $\lim_{y\downarrow x} f(y) = \lim_{n\to\infty} f(x+\frac{1}{n})$ .

## **Expected Utility Of Wealth**

 As usual, the space of all distribution functions is convex and one can define preferences on it.

The utility index  $v: \mathbb{R} \to \mathbb{R}$  is defined over wealth (NOTE: values can be negative).

The expected utility is the integral of v with respect to F

$$\int v(x)dF(x) = \int vdF$$

• If F is differentiable, the expectation is computed using the density f = F':  $\int v(x)dF = \int v(x)f(x)dx$ .

### von Neumann and Morgenstern Expected Utility

Under some axioms, there exists a utility function U on distributions defined as  $U(F) = \int v(x) dF$ , for some continuous index  $v : \mathbb{R} \to \mathbb{R}$  over wealth, such that

$$F \succsim G \iff \int v(x) dF \ge \int v(x) dG$$

- Axioms are not important (need a stronger continuity assumption).
- ullet We always think of v as a weakly increasing function (more wealth cannot be bad).

## **Simple Probability**

- A simple probability distribution  $\pi$  on  $X \subset \mathbb{R}$  is specified by
  - a finite subset of X called the support and denoted  $supp(\pi)$ , and
  - for each  $x \in X$ ,  $\pi(x) > 0$  with  $\sum_{x \in supp(\pi)} \pi(x) = 1$
- If we restrict attention to simple probability distributions, then even if X is infinite, only elements with strictly positive probability count.
- The utility index  $v: X \to \mathbb{R}$  is defined over wealth (can be negative).
- The expected utility is the expected value of  $\nu$  with respect to  $\pi$

$$\sum_{x \in supp(\pi)} \pi(x) v(x)$$

- One can write more money is better as: for each  $x,y \in X$  such that x > y then  $\delta_x \succ \delta_y$ .
- We can use this setting to think about many applied choice under uncertainty problems like betting and insurance.

# Betting

#### A Gamble

Suppose an individual is offered the following bet:

win ax with probability p

lose x with probability 1-p

• The expected value of this bet is

$$pax + (1 - p)(-x) = [pa + (1 - p)(-1)]x$$

## Definition

A bet is actuarially fair if it has expected value equal to zero (i.e.  $a = \frac{1-p}{p}$ ); it is better than fair if the expected value is positive and worse than fair if it is negative.

#### How does she evaluate this bet? Use the expected utility model to find out

• If vNM index is  $v(\cdot)$  and initial wealth is w, expected utility is:

```
probability of winning v\left(w+ax
ight) probability of losing v\left(w-x
ight) utility of wealth if win v\left(w-x
ight) utility of wealth if lose
```

## **Betting and Expected Utility**

How much does she want of this bet? Answer by finding the optimal x.

win ax with probability p

lose x with probability 1-p

The consumer solves

$$\max_{x} pv(w + ax) + (1 - p)v(w - x)$$

The FOC is

$$\mathit{pav}'\left(w+\mathit{ax}\right)=\left(1-\mathit{p}\right)\mathit{v}'\left(w-\mathit{x}\right)$$

rearranging

$$\frac{pa}{(1-p)} = \frac{v'(w-x)}{v'(w+ax)}$$

- If the bet is fair, the left hand side is 1. Therefore, at an optimum, the right hand side must also be 1.
- If the vNM utility function is strictly increasing and strictly concave (v' > 0 and v'' < 0), the only way a fair bet can satisfy this FOC is to solve

$$w + ax = w - x$$

which implies x = 0.

- She will take no part of a fair bet.
- What happens with a better than fair bet?

#### Insurance

#### **An Insurance Problem**

• An individual faces a potential "accident":

the loss is L with probability  $\pi$ 

nothing happens with probability  $1-\pi$ 

### **Definition**

An insurance contract establishes an initial premium P and then reimburses an amount Z if and only if the loss occurs.

#### **Definition**

Insurance is actuarially fair when its expected cost is zero; it is less than fair when its expected cost is positive.

• The expected cost (to the individual) of an insurance contract is

$$P - \left[\pi(-Z) + (1-\pi) \quad (0)\right] = P - \pi Z$$

Fair insurance means

$$P = \pi Z$$

## **Insurance and Expected Utility**

#### An Insurance Problem

An individual with current wealth W and utility function  $v(\cdot)$  faces a potential accident:

lose L with probability  $\pi$  or lose zero with probability  $1-\pi$ 

If she buys insurance, her expected utility is

$$\pi v(\underbrace{W-L-P+Z}) + (1-\pi) v(\underbrace{W-P}_{\text{wealth if no loss}})$$

• For example, if the loss is fully reimbursed (Z = L), this becomes

$$\pi v (W - P) + (1 - \pi) v (W - P) = v (W - P)$$

Will she buy any insurance? Yes if

$$\underbrace{\pi v \left(W - L - P + Z\right) + \left(1 - \pi\right) v \left(W - P\right)}_{\text{expected utility with insurance}} \ge \underbrace{\pi v \left(W - L\right) + \left(1 - \pi\right) v \left(W\right)}_{\text{expected utility without insurance}}$$

• Find how much coverage she wants (if any) by finding the optimal Z.

• The answer depends on the premium set by the insurance company P (which could depend on Z) as well as the curvature of the utility function v.

#### **Curvature of the Utility Function**

The answers to the previous problems depend on the curvature of the utility function v.

• The curvature of *v* captures important characteristics of preferences in many applied situations.

#### **Risk Aversion**

#### **Definitions**

The preference relation  $\succsim$  is

- risk averse if, for all cumulative distribution functions F,  $\delta_{\mu_F} \gtrsim F$ .
- risk loving if, for all cumulative distribution functions F,  $F \succsim \delta_{\mu_F}$ .
- risk neutral if it is both risk averse and risk loving:  $\delta_{\mu_{\rm E}} \sim F$ .
- DM is risk averse if she always prefers the expected value  $\mu_F$  for sure to the uncertain distribution F.
- This definition does not depend on the expected utility representation (or any other).

#### Remark

Risk attitudes are defined directly from preferences.

## Risk Aversion: An example

#### **Exercise**

Let  $\succeq$  be a preference relation on the space of all cumulative distribution functions represented by the following utility function:

$$U(F) = \begin{cases} x & \text{if } F = \delta_x \text{ for some } x \in \mathbf{R} \\ 0 & \text{otherwise} \end{cases}$$

- - False: If  $\mu_F < 0$ , then  $F \succ \mu_F$ .

## **Certainty Equivalent**

#### Definition

Given a strictly increasing and continuous vNM index v over wealth, the certainty equivalent (CE) of F, denoted c(F, v), is defined by

$$v(c(F,v)) = \int v(\cdot) dF.$$

- The certainty equivalent of F is the amount of wealth  $c(\cdot)$  such that  $c(\cdot) \sim F$ .
  - DM is indifferent between a distribution and the certainty equivalent of that distribution.
  - The certainty equivalent is constructed to satisfy this indifference.
- One can compare two lotteries by comparing their certainty equivalents.
- Unlike risk aversion, the certainty equivalent definition assumes a given preference representation (needs some utility function that represents preferences).
- The value of the certainty equivalent is related to risk aversion.

#### **Risk Premium**

#### Definition

Given a strictly increasing and continuous vNM index v over wealth, the risk premium of F, denoted r(F, v) is defined by

$$r(F, v) = \mu_F - c(F, v).$$

- This measures the difference between the expected value of a particular distribution and its certainty equivalent.
- The definition of risk premium also assumes a given preference representation.
- The risk premium is also related to risk aversion.

## Risk Aversion, Certainty Equivalent, and Risk Premium

 If preferences satisfy the vNM axioms, risk aversion is characterized by concavity of the utility index and a non-negative risk-premium.

## **Proposition**

Suppose  $\succeq$  has an expected utility representation and v is the corresponding von Neumann and Morgestern utility index over money. The following are equivalent:

- $\bullet$   $\succeq$  is risk averse;
- v is concave;
- $r(F,v) \geq 0;$ 
  - The proof uses Jensen's inequality.

## Jensen's Inequality

**Reminder**: a function  $f : \mathbb{R} \to \mathbb{R}$  is concave if for any  $x, y \in \mathbb{R}$  and any  $\alpha \in [0, 1]$   $f(\alpha x + (1 - \alpha)y) \ge \alpha f(x) + (1 - \alpha)f(y)$ 

## Jensen's inequality

A function g is concave if and only if for all distributions F

$$g\left(\int xdF\right)\geq\int g\left(x\right)dF$$

 $g(\mathsf{E}(X)) \geq \mathsf{E}(g(X))$ 

This says

- Consequences of Jensen's inequality
- Hence,  $v(\cdot)$  is concave if and only if for all distributions F

$$\underbrace{v\left(\int dF\right)}_{\text{utility of the expected value of }F} \geq \underbrace{\int vdF}_{\text{expected utility of }F}$$

# Risk Aversion, CE, and Risk Premium

We prove  $(1) \Rightarrow (2) \Rightarrow (3) \Rightarrow (1)$ . Start with  $(1) \Rightarrow (2)$ .

## Proof.

 $\succsim$  is risk averse, hence  $\delta_{\mu_F} \succsim F$  for all  $F \in \Delta \mathbb{R}$ .

- For any  $x, y \in \mathbb{R}$  and  $\alpha \in [0, 1]$ , let the discrete random variable X be such that
  - $P(X = x) = \alpha$  and  $P(X = y) = 1 \alpha$ . Let  $F_{x,y}^{\alpha}$  be the associated cumulative distribution.
- By risk aversion we have:

$$v(\mu_{F_{x,y}^{\alpha}}) \geq \int v(z)dF_{x,y}^{\alpha}(z)$$

$$\Rightarrow$$

$$v(\alpha x + (1 - \alpha)y) \geq \sum_{z} v(z)P(X = z) = \alpha v(x) + (1 - \alpha)v(y)$$

<del>\_</del>. .



# Risk Aversion, CE, and Risk Premium

Now prove that  $(2) \Rightarrow (3)$ 

## Proof.

Let v be concave, and X be a random variable with cdf F.

By Jensen's inequality:

$$v(\mathsf{E}(X)) > \mathsf{E}(v(X))$$

or

$$v(\mu_F) \ge \int v(x)dF(x) = v(c(F, v))$$

- Since v is an increasing function, we have
- $\mu_{\mathsf{F}} > c(\mathsf{F},\mathsf{v})$
- Thus

$$\mu_{\mathsf{F}} - c(\mathsf{F}, \mathsf{v}) = r(\mathsf{F}, \mathsf{v}) > 0$$

# Risk Aversion, CE, and Risk Premium

$$\succeq \text{ is risk averse} \Leftrightarrow \underbrace{v \text{ is concave}}_{(2)} \Leftrightarrow \underbrace{r(F,v) \geq 0}_{(3)}$$

 $(3) \Rightarrow (1)$ 

# Proof.

Let  $r(F, v) \ge 0$  for all cdfs F.

Then we have

$$\mu_F \geq c(F, v)$$

which in turn implies that

$$v(\mu_F) \ge v(c(F, v)) = \int v(x)dF(x)$$

• Hence  $\delta_{\mu_F} \succsim F$  for all F; therefore  $\succsim$  is risk averse.

We have shown that  $(1) \Rightarrow (2) \Rightarrow (3) \Rightarrow (1)$ , thus the proof is complete.

#### **Next Class**

- Relative Risk Aversion
- Stochastic Dominance
- Random Consumption Bundles