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We write $\alpha L_1 + (1-\alpha)L_2$ to be the compound lottery that gives L_1 with probability α and L_2 with complementary probability. [NB Rubinstein and some other authors use the symbol \oplus , e.g., $\alpha L_1 \oplus (1-\alpha)L_2$, for such "probabilistic mixtures".]

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For every $L,L',L''\in\Delta(Z)$, such that $L'\succ L$ there exist $\overline{\alpha},\overline{\beta}\in(0,1)$ such that

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Assumption (Continuity*)

For every $L,L',L''\in\Delta(Z)$, such that $L'\succ L\succ L''$ there exist $\gamma\in(0,1)$ such that

$$L \sim \gamma L' + (1 - \gamma) L''$$

Independence Axiom

Assumption (Independence Axiom) For every $L, L', L'' \in \Delta(Z)$ and $\alpha \in (0,1], L \succsim L'$ iff $\alpha L + (1-\alpha)L'' \succsim \alpha L' + (1-\alpha)L''$

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Expected Utility

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Adding the Independence Axiom puts more structure on preferences.

Definition

The utility function $U:\Delta(Z)\to\mathbb{R}$ is a von Neumann Morgenstern (vNM) expected-utility function if there exists a vector of utilities for the prizes $(u_1,...,u_N)$ such that for every simple lottery $L=(p_1,...,p_N)\in\Delta(Z)$,

$$U(L) = p_1 u_1 + \ldots + p_N u_N.$$

The vNM utility of a lottery equals the *expected utility* of the outcomes. vNM preferences are *linear* in the prize probabilities.

Expected-Utility Theorem

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Proposition (Expected-Utility Theorem)

Preferences satisfy rationality, continuity and the Independence Axiom iff they have a von Neumann Morgenstern (vNM) expected utility representation.

Risk Aversion

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We say that a vNM decision maker is risk averse if for each lottery L, $[E[L]] \succeq L$ and for some lottery L', $[E[L']] \succ L'$.

Let's take the prize space to comprise money payments, and for simplicity $Z=\mathbb{R}.$

Proposition

A vNM DM is risk averse iff her Bernoulli utility function u is concave.

A Calibration Argument (Rabin 2000)

Assume

- 1. People are risk-averse EU maximisers
- 2. People reject small-scale gambles that real people appear to reject

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- 1. People are risk-averse EU maximisers
- 2. People reject small-scale gambles that real people appear to reject

Show that this implies implausible preferences over large-scale risk. Conclude that EU theory does not provide a satisfactory, integrated account of both small- and large-scale risk attitudes.

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Rejecting 50-50 gain \$1 lose \$ \mathcal{L} bet at w then implies that $u(w+1)-u(w)\leq 1$

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$$\Rightarrow u(w+2) - u(w) = u(w+2) - u(w+1) + u(w+1) - u(w) \\ \leq \mathcal{L} + 1.$$

Iterating,

$$u(w+k)-u(w)\leq \sum_{i=0}^{k-1}\mathcal{L}^{j}$$

Hence for any x,

$$u(x) \leq u(w) + \sum_{i=0}^{\infty} \mathcal{L}^j = u(w) + \frac{1}{1-\mathcal{L}}.$$

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Rabin (2000) makes this and stronger statements.

Alternatively, read Rabin and Thaler in Journal of Economic Perspectives (2001)

Concave Utility of Wealth Produces Second-Order Risk Aversion

- Concave utility of wealth is a theory of "second-order" risk aversion that derives from curvature of the utility function—it cannot explain "first-order" risk aversion, namely that over small stakes for plausible degrees of risk aversion.
- ► The intuition is simply that the even the most concave function is approximately linear in a small neighbourhood (that's what it means to be differentiable) and hence approximately risk neutral in that neighbourhood.
- ▶ When we estimate people's degree of risk aversion using small stakes, they appear to be much more risk averse than when we do it using large stakes.
- No concave utility of wealth function can fit people's risk attitudes towards both small and large stakes.

Evidence of First-Order Risk Aversion

Sydnor (2008) looks at American homeowner house insurance

- ➤ Typical homeowner chooses to pay \$100 to reduce deductible from \$1000 to \$500
- ► Chance of event where such coverage pays out is less than 5%, i.e., actuarially fair price of extra insurance is less than \$25

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Someone who declined the insurance and saved the \$100 premium to self insure would be on average \$2000 richer after 30 years, with only 1.6% chance of ending poorer.

Loss Aversion (Kahneman and Tversky 1979)

Kahneman and Tversky propose the reference-dependent Bernoulli utility function

$$u(r,z) = w(r) + v(z-r)$$

Bernoulli utility defined over reference point r and money outcome z: it is reference utility (w) plus a value function (v) evaluating gain or loss from the reference point that satisfies four properties:

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- 4. $\lim_{x\to 0^-} v'(x) = k \lim_{x\to 0^+} v'(x)$ for k>1: v has a kink at the reference point, where the marginal value of a loss is greater than the marginal value of a same-sized gain.

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Losses loom larger than same-sized gains—let's see how Properties 1-4 get this

P2 says if we take x = 0 then get v(y) < -v(-y), or gain from getting y above reference point is less than *loss* from getting y below reference point

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Also we can write

$$v(y) - v(x) < v(-x) - v(-y),$$

and so

$$\frac{v(y)-v(x)}{y-x}<\frac{v(-x)-v(-y)}{y-x}.$$

Letting $y \to x$ gives $v'(x) \le v'(-y) = v'(-x)$.

The value function is always steeper in the losses domain.

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This does not imply that v not differentiable (has a kink) at 0, for which we need Property 4.

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$$v(y) + v(-y) = y + \alpha(-y) = y(1 - \alpha)$$

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because $\alpha > 1$ and x < y.

$$v'(0^+) = 1 < v'(0^-) = \alpha$$
, satisfying Property 4.

The S-shape

Finally, piecewise-linear v is concave in gains and convex in losses (since linear functions are both convex and concave).

Note though that it is not just concave in the gains domain but concave overall, as the function f is concave if for any x, y and any $\beta \in [0,1]$,

$$f(\beta x + (1 - \beta)y) \ge \beta f(x) + (1 - \beta)f(y).$$

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Someone with such a v is risk averse.

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If (x,p;y,1-p) is better than fair, then a risk-averse DM with concave continuously differentiable Bernoulli utility function prefers (tx,p;ty,1-p) to [0] for positive t small enough. Intuitively, for t small, the range of outcomes in lottery shrinks to zero, as $ty-tx\to 0$. Over small stakes, the agent is approximately risk neutral and will therefore accept a better-than-fair lottery.

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This suggests how loss aversion can explain our puzzle: people are "first-order risk averse" around their reference points without being too globally risk averse.

Risk Lovingness over Losses

Prop 3 says risk-loving over losses.

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From K&T 1979

4000, .8	3000	-4000, .8	-3000
20	80	92	8
4000, .2	3000, .25	-4000, .2	-3000, .25
65	35	42	52
3000, .9	6000, .45	-3000, .9	-6000, .45
86	14	8	92
3000, .002	6000, .001	-3000,.002	-6000,.001
27	73	70	30