

Choice under Uncertainty

1. Primitives

Finite set $A = \{a_1, \dots, a_n\}$ of outcomes. A simple gamble (or lottery) is an element of $\Delta(A)$, the set of probability distributions over A . We denote the set of simple gambles by \mathcal{G}_s and a typical element of this set by $g = (p_1 \circ a_1, \dots, p_n \circ a_n)$, where (p_1, \dots, p_n) is an element of the unit simplex in \mathbb{R}^n ; i.e., $p_i \geq 0$ for $i = 1, \dots, n$ and $\sum_{i=1}^n p_i = 1$. When $p_i = 0$, we drop the term $p_i \circ a_i$ from the lottery g . The probabilities in this context are *objective*.

Not all gambles are simple. For example, if $A = \{-1, 0, 1\}$ and $\tilde{g} = (\frac{1}{2} \circ -1, \frac{1}{2} \circ 1)$, then $g = (\frac{1}{2} \circ 0, \frac{1}{2} \circ \tilde{g})$ is also a gamble. The gamble g is an example of a compound gamble. We denote the set of all compound gambles by \mathcal{G} and also use g to denote a typical element of this set. Every compound gamble g uniquely defines a probability distribution (p_1, \dots, p_n) over A . Let $g_s = (p_1 \circ a_1, \dots, p_n \circ a_n)$. We refer to g_s as the simple gamble associated to g . In the example above, $g_s = (\frac{1}{4} \circ -1, \frac{1}{2} \circ 0, \frac{1}{4} \circ 1)$.

The choice set is the set \mathcal{G} of compound gambles.

2. Preferences

As in the consumer theory, we represent preferences over gambles by a binary relation \succsim on the set \mathcal{G} . We make the following assumptions about \succsim .

1. *Completeness*: \succsim is complete.
2. *Transitivity*: \succsim is transitive.

The first two properties imply that: (i) the decision maker (agent) is able to compare any two given gambles; and (ii) the decision maker's ranking of lotteries is internally consistent. As in the consumer theory, we say that \succsim is a preference relation if it is both complete and transitive.

The preference relation \succsim on \mathcal{G} induces an order on A , that we also denote by \succsim , as follows: $a_i \succsim a_j$ if, and only if, $1 \circ a_i \succsim 1 \circ a_j$. In what follows we denote the lottery $1 \circ a_i$ by a_i , thus erasing any distinction between preferences on \mathcal{G} and preferences on A . Since \succsim

is complete, we can rank all elements of A . Assume, without loss, that $a_1 \succsim \dots \succsim a_n$.

3. Continuity: for any $g \in \mathcal{G}$, there exists $\alpha \in [0, 1]$ such that $g \sim (\alpha \circ a_1, (1 - \alpha) \circ a_n)$.

4. Monotonicity: $(\alpha \circ a_1, (1 - \alpha) \circ a_n) \succsim (\beta \circ a_1, (1 - \beta) \circ a_n)$ if, and only if, $\alpha \geq \beta$.

Note that monotonicity implies that $a_1 \succ a_n$, which rules out the case $a_1 \sim \dots \sim a_n$. Also, notice that monotonicity implies that for each gamble g the probability α in the definition of continuity is uniquely defined¹.

5. Substitution: if $g = (p_1 \circ g^1, \dots, p_k \circ g^k)$ and $h = (p_1 \circ h^1, \dots, p_k \circ h^k)$, where $g_1, \dots, h^k \in \mathcal{G}$, then $g^i \sim h^i$ for $i = 1, \dots, k$ implies that $g \sim h$.

6. Reduction to Simple Gambles: for all $g \in \mathcal{G}$, $g \sim g_s$, the simple gamble associated to g .

As in the consumer theory, the utility function $u : \mathcal{G} \rightarrow \mathbb{R}$ represents \succsim if for all $g', g \in \mathcal{G}$, $u(g') \geq u(g)$ if, and only if, $g' \succsim g$. We say that $u : \mathcal{G} \rightarrow \mathbb{R}$ representing \succsim has the *expected utility property* if for all $g \in \mathcal{G}$,

$$u(g) = \sum_{i=1}^n p_i u(a_i),$$

where $(p_1 \circ a_1, \dots, p_n \circ a_n)$ is the simple gamble induced by g . We say that $u(g)$ is the expected utility of the gamble g .

Theorem 1. *Suppose \succsim on G satisfies properties (axioms) 1 to 6 above. Then there exists a utility function u representing \succsim that has the expected utility property.*

Proof: For each $g \in \mathcal{G}$, let $u(g) \in [0, 1]$ be such that $g \sim (u(g) \circ a_1, (1 - u(g)) \circ a_n)$. This number exists by continuity. Moreover, by monotonicity, $u(g)$ is uniquely defined for every gamble g . This defines a utility function $u : \mathcal{G} \rightarrow \mathbb{R}$.

We claim that u represents \succsim . Indeed, by transitivity, $g' \succsim g$ implies that

$$(u(g') \circ a_1, (1 - u(g')) \circ a_n) \succsim (u(g) \circ a_1, (1 - u(g)) \circ a_n), \quad (1)$$

which implies that $u(g') \geq u(g)$ by monotonicity. Now observe, by monotonicity, that $u(g') \geq u(g)$ implies that 1 holds. Thus, by transitivity, $g' \succsim g$, which proves the claim.

¹Why is that?

We now prove that u has the expected utility property. By reduction to simple gambles, we only need to prove this for simple gambles. Let then $g = (p_1 \circ a_1, \dots, p_n \circ a_n)$ and notice that for each $i \in \{1, \dots, n\}$, $a_i \sim (u(a_i) \circ a_1, (1 - u(a_i)) \circ a_n) \equiv q^i$. By substitution, we then have that $g' = (p_1 \circ q^1, \dots, p_n \circ q^n) \sim g$. It is easy to see that

$$g'_s = \left(\left(\sum_{i=1}^n p_i u(a_i) \right) \circ a_1, \left(1 - \sum_{i=1}^n p_i u(a_i) \right) \circ a_n \right).$$

Thus, by reduction and transitivity,

$$g \sim (u(g) \circ a_1, (1 - u(g)) \circ a_n) \sim g'_s \sim \left(\left(\sum_{i=1}^n p_i u(a_i) \right) \circ a_1, \left(1 - \sum_{i=1}^n p_i u(a_i) \right) \circ a_n \right).$$

Since $u(g)$ is uniquely defined, it must be that $u(g) = \sum_{i=1}^n p_i u(a_i)$, the desired result. \square

It is immediate from the proof above that the existence of an utility function u representing \succsim follows from properties 2 to 4. Since a binary relation is a preference relation if it can be represented by an utility function, we then have that the completeness of \succsim follows from transitivity, continuity, and monotonicity. The fact that u has the expected utility property is a consequence of properties 5 and 6 (together with transitivity).

A decision maker with preferences \succsim over compound gambles is an expected utility maximizer if these preferences are represented by a utility function with the expected utility property and the decision maker chooses the lottery g that maximizes his expected utility. From now on we work with expected utility maximizers (despite the example below).

Example (Allais Paradox): Suppose that $A = \{2.5, 0.5, 0\}$, where each unit is measured in millions of euros. Consider now the following lotteries:

$$\begin{aligned} g_1 &= 0.5, \\ g_2 &= \left(\frac{1}{10} \circ 2.5, \frac{89}{100} \circ 0.5, \frac{1}{100} \circ 0 \right), \\ g_3 &= \left(\frac{11}{100} \circ 0.5, \frac{89}{100} \circ 0 \right), \\ g_4 &= \left(\frac{1}{10} \circ 2.5, \frac{9}{10} \circ 0 \right). \end{aligned}$$

Most people (when asked) prefer g_1 over g_2 and g_4 over g_3 . This is not compatible with expected utility maximization. Indeed, if \succsim is represented by an utility function u with the expected utility property, then $g_1 \succ g_2$ if, and only if,

$$u(0.5) > \frac{1}{10}u(2.5) + \frac{89}{100}u(0.5) + \frac{1}{100}u(0),$$

which is equivalent to

$$\frac{11}{100}u(0.5) + \frac{89}{100}u(0) > \frac{1}{10}u(2.5) + \frac{9}{10}u(0).$$

This, however, implies that $g_3 \succ g_4$. Most people would agree that g_3 preferred to g_4 is insane. Since we know that expected utility follows from properties 5 and 6, the relevant question is whether one takes these properties as having normative content or a positive content. Most people would go with the second alternative.

Theorem 2. *Suppose that u represents \succsim and has the expected utility property. Then v with the expected utility property represents \succsim if, and only if, there exists $\alpha \in \mathbb{R}$ and $\beta > 0$ such that $v(g) = \alpha + \beta u(g)$ for all $g \in \mathcal{G}$.*

Proof: Sufficiency is obvious. Indeed, if v is such that $v(g) = \alpha + \beta u(g)$ for every gamble g with $\alpha \in \mathbb{R}$ and $\beta > 0$, then v has the expected utility property.

In order to prove necessity we just need to consider simple gambles. For this, let $A = \{a_1, \dots, a_n\}$, with $a_1 \succsim \dots \succsim a_n$, and consider the simple gamble $g = (p_1 \circ a_1, \dots, p_n \circ a_n)$. Moreover, assume that $a_1 \succ a_n$, for otherwise necessity is immediate².

Since u represent \succsim , we have that $u(a_1) \geq \dots \geq u(a_n)$. So, for each $i \in \{1, \dots, n\}$, there exists $\alpha_i \in [0, 1]$ such that $u(a_i) = \alpha_i u(a_1) + (1 - \alpha_i)u(a_n)$; $\alpha_i > 0$ since $a_1 \succ a_n$. Since u has the expected utility property, $\alpha_i u(a_1) + (1 - \alpha_i)u(a_n) = u(g)$, where $g = (\alpha_i \circ a_1, (1 - \alpha_i) \circ a_n)$. Since u represents \succsim , we then have that

$$a_i \sim (\alpha_i \circ a_1, (1 - \alpha_i) \circ a_n). \tag{2}$$

²You should check this.

Now observe that v also has the expected utility property and represents \succsim . Thus, by 2, $v(a_i) = \alpha_i v(a_1) + (1 - \alpha_i)v(a_n)$. Hence, for all $i \in \{1, \dots, n\}$ such that $a_i \succ a_n$, we have that

$$\frac{u(a_1) - u(a_i)}{u(a_i) - u(a_n)} = \frac{1 - \alpha_i}{\alpha_i} = \frac{v(a_1) - v(a_i)}{v(a_i) - v(a_n)},$$

which implies that

$$[u(a_1) - u(a_i)][v(a_i) - v(a_n)] = [v(a_1) - v(a_i)][u(a_i) - u(a_n)]. \quad (3)$$

Since 3 is also satisfied when $a_i \sim a_n$ and $u(a_1) > u(a_n)$, we then have (omitting the algebra) that

$$\begin{aligned} v(a_i) &= \frac{u(a_1)v(a_n) - v(a_1)u(a_n)}{u(a_1) - u(a_n)} + \frac{v(a_1) - v(a_n)}{u(a_1) - u(a_n)}u(a_i) \\ &= \alpha + \beta u(a_i), \end{aligned}$$

where $\beta > 0$ since $v(a_1) > v(a_n)$.

Therefore, since both v and u have the expected utility property,

$$v(g) = \sum_{i=1}^n p_i v(a_i) = \sum_{i=1}^n p_i (\alpha + \beta u(a_i)) = \alpha + \beta \sum_{i=1}^n p_i u(a_i) = \alpha + \beta u(g).$$

□

The restriction to finite outcome spaces is done for simplicity. We can extend Theorem 1 to the case where A is infinite, at the cost of some technicalities. In most applications, the case of interest is when $A = \mathbb{R}_+$ and an element of A represents wealth. In this case, we can represent a simple gamble on A by a cdf F^3 . For any simple gamble F , the expected utility of F is

$$u(F) = \int_0^\infty u(w) dF(w).$$

We refer to the function $u : \mathbb{R}_+ \rightarrow \mathbb{R}$ such that $u(w)$ is the agent's payoff from the lottery F that gives w with probability one as the agent's utility function over wealth. Monotonicity implies that u is strictly increasing.

³Recall that a cdf in \mathbb{R}_+ is an increasing and right-continuous function $F : \mathbb{R}_+ \rightarrow [0, 1]$ such that $F(0) = 0$ and $F(+\infty) = 1$.

3. Attitudes to Risk

Here and in the next two sections, assume that $A = \mathbb{R}_+$. Consider an individual with utility over wealth u . Now let g be a gamble (over final wealth) and denote its expected value by $\mathbb{E}[g]$. We say the individual is: (i) risk-neutral at g if $u(g) = u(\mathbb{E}[g])$; (ii) risk-averse at g if $u(g) < u(\mathbb{E}[g])$; (iii) risk-loving at g if $u(g) > u(\mathbb{E}[g])$.

We say the individual is

(i) *risk-neutral* if $u(g) = u(\mathbb{E}[g])$ for all gambles g ;

(ii) *risk-averse* if $u(g) \leq u(\mathbb{E}[g])$ for all gambles g and $u(g) < u(\mathbb{E}[g])$ for at least one non-degenerate gamble⁴ g ;

(iii) *risk-loving* if $u(g) \geq u(\mathbb{E}[g])$ for all gambles g and $u(g) > u(\mathbb{E}[g])$ for at least one non-degenerate gamble g .

It is easy to see that if u is linear (affine) in wealth, i.e., if $u(w) = \alpha + \beta w$ with $\alpha \in \mathbb{R}$ and $\beta > 0$, then the individual is risk-neutral. Now, for each $w, w' \in A$ and $t \in (0, 1)$, let $g = (t \circ w, (1 - t) \circ w')$. If the individual is risk-averse at g , then

$$u(tw + (1 - t)w') \geq tu(w) + (1 - t)u(w').$$

Thus, if the individual is risk-averse, his utility over wealth is concave and non-linear. It follows immediately from this that if the individual is risk-loving, his utility over wealth is convex and non-linear. Since a function can be concave and convex at the same time only if it is linear, we then have that if the individual is risk-neutral, his utility over wealth is linear. In other words, the individual is risk-neutral if, and only if, u is linear.

The next result is useful.

Lemma 1 (Jensen's Inequality). *Suppose $u : \mathbb{R}_+ \rightarrow \mathbb{R}$ is concave. Then, for every cdf F ,*

$$\int_0^\infty u(w)dF(w) \leq u\left(\int_0^\infty wdF(w)\right) = u(\mathbb{E}[F]).$$

⁴A gamble is non-degenerate if it assigns positive probability to some non-empty interval in \mathbb{R}_+ .

Notice that Jensen's inequality reduces to

$$u(g) = \sum_{i=1}^n p_i u(g_i) \leq u\left(\sum_{i=1}^n p_i w_i\right) = u(\mathbb{E}[g]) \quad (4)$$

when F is the cdf associated to the simple gamble $g = (p_1 \circ w_1, \dots, p_n \circ w_n)$.

Proof of Jensen's: We establish Jensen's inequality in the case where F assigns positive probability only to a finite subset of \mathbb{R}_+ , in which case it can be represented by a simple gamble $g = (p_1 \circ w_1, \dots, p_n \circ w_n)$. Since u is concave, (4) is satisfied when $n = 2$. Suppose then, by induction, that (4) is satisfied when $n = K$, with $K \geq 2$. If $p_{K+1} \in \{0, 1\}$, (4) is satisfied when $n = K + 1$. So assume that $p_{K+1} \notin \{0, 1\}$ and observe that

$$\sum_{i=1}^{K+1} p_i w_i = \sum_{i=1}^K p_i \bar{w} + p_{K+1} w_{K+1} = (1 - p_{K+1})\bar{w} + p_{K+1} w_{K+1},$$

where $\bar{w} = \sum_{i=1}^K p_i w_i / \sum_{i=1}^K p_i$; notice that \bar{w} is well-defined since $\sum_{i=1}^K p_i = 1 - p_{K+1} > 0$.

Now observe that

$$\begin{aligned} u\left(\sum_{i=1}^{K+1} p_i w_i\right) &\geq (1 - p_{K+1})u(\bar{w}) + p_{K+1}u(w_{K+1}) \\ &\geq \sum_{i=1}^K p_i w_i + p_{K+1} w_{K+1}, \end{aligned}$$

where the first inequality follows from the concavity of u and the second inequality follows from the induction hypothesis (why?). Hence, (4) holds when $n = K + 1$. By induction, (4) holds for all $n \geq 2$. \square

It is easy to see, by Jensen's inequality, that if u is concave, then $u(g) \leq u(\mathbb{E}[g])$ for every gamble g . Now observe that if u is non-linear, there exists $w, w' \in \mathbb{R}_+$ with $w \neq w'$ and $t \in (0, 1)$ such that $u(tw + (1-t)w') > tu(w) + (1-t)u(w')$. Let then $g = (t \circ w, (1-t) \circ w')$; notice that g is a non-degenerate gamble. By construction, $u(g) < u(\mathbb{E}[g])$. Thus, if u is concave and non-linear, the individual is risk-averse. Therefore, from the discussion before Jensen's inequality, the individual is risk-averse if, and only if, his utility over wealth is concave and non-linear. It is the immediate to see from this that the individual is risk-loving if, and only if, his utility over wealth is convex and non-linear.

Example: Consider an individual with utility over wealth u , where u is strictly increasing and strictly concave; thus, the individual is risk-averse. Let $w_0 > 0$ be the individual's initial wealth. There is a probability $\alpha \in (0, 1)$ that he loses an amount $0 < L < w_0$ of his wealth. The individual can insure himself against this loss. The price of one unit of insurance is ρ . We assume that insurance is actuarially fair, i.e., $\rho = \alpha$.

The individual's expected payoff from contracting an amount x of insurance is

$$\begin{aligned} v(x) &= \alpha u(w_0 - L + x - \rho x) + (1 - \alpha)u(w_0 - \rho x) \\ &= \alpha u(w_0 - L + x - \alpha x) + (1 - \alpha)u(w_0 - \alpha x). \end{aligned}$$

Indeed, with probability $1 - \alpha$ he suffers no loss, and so his final wealth is $w_0 - \alpha x$. With probability α he suffers a loss, and so his final wealth is $w_0 - L + x - \alpha x$; the individual pays for insurance regardless of whether he suffers a loss or not. The problem of the individual is to choose the amount of insurance $x \in [0, w_0/\alpha]$ that maximizes $v(x)$. It is easy to show that v is strictly concave, so that the solution to the individual's problem is characterized by its first-order condition. In particular, if there exists $x^* \in (0, w_0/\alpha)$ such that $v'(x^*) = 0$, then x^* is the solution. Notice that

$$v'(x) = (1 - \alpha)\alpha u'(w_0 - L + (1 - \alpha)x) - \alpha(1 - \alpha)u'(w_0 - \alpha x).$$

Since u' is strictly decreasing, $v'(x^*) = 0$ if, and only if,

$$w_0 - L + (1 - \alpha)x^* = w_0 - \alpha x^* \Leftrightarrow x^* = L.$$

Thus, it is optimal for the individual to insure himself completely against the loss.

4. Risk Aversion

Measuring the Risk of Gambles

Consider an individual with a concave utility function over wealth. For any lottery g , the certainty equivalent CE of g is the amount of wealth such that $u(g) = u(CE)$. Since u is strictly increasing and concave, Jensen's inequality implies that $CE \leq \mathbb{E}[g]$ for every gamble g . The risk premium of g is the value P such that $u(g) = u(\mathbb{E}[g] - P)$; i.e., $P = \mathbb{E}[g] - CE$.

The risk premium of a gamble measures how much the (risk-averse) individual is willing to pay in order to avoid the risk associated with the gamble.

The Arrow-Pratt Measure of Absolute Risk Aversion

Suppose that u is twice differentiable with $u' > 0$; since u is concave, $u'' \leq 0$. The Arrow-Pratt measure of absolute risk aversion is

$$R_a(w) = -\frac{u''(w)}{u'(w)}.$$

Consider two individuals, 1 and 2, with utilities over wealth u and v , respectively, and suppose that for all $w \geq 0$,

$$R_a^1(w) = -\frac{u''(w)}{u'(w)} > R_a^2(w) = -\frac{v''(w)}{v'(w)}. \quad (5)$$

For simplicity assume that $v(w) \geq 0$ for all $w \geq 0$. Hence, we can define $h : [0, +\infty) \rightarrow \mathbb{R}$ to be such that $h(x) = u(v^{-1}(x))$. Notice that

$$h'(x) = \frac{u'(v^{-1}(x))}{v'(v^{-1}(x))} > 0$$

and that

$$\begin{aligned} h''(x) &\propto u''(v^{-1}(x)) - v''(v^{-1}(x)) \frac{u'(v^{-1}(x))}{v'(v^{-1}(x))} \\ &\propto -R_a^1(w) + R_a^2(w) < 0 \end{aligned}$$

by 5. Therefore, h is strictly increasing and strictly concave.

Now let F be a cdf and define \hat{w}_1 and \hat{w}_2 be such that

$$\begin{aligned} u(\hat{w}_1) &= \int_0^\infty u(w) dF(w), \\ v(\hat{w}_2) &= \int_0^\infty v(w) dF(w); \end{aligned}$$

\hat{w}_1 is the certainty equivalent for 1 of the gamble defined by F and \hat{w}_2 is the certainty equivalent for 2 of the same gamble. Since $h(v(w)) = u(w)$,

$$u(\hat{w}_1) = \int_0^\infty h(v(w)) dF(w) < h\left(\int_0^\infty v(w) dF(w)\right) = h(v(\hat{w}_2)) = u(\hat{w}_2),$$

where the inequality follows from Jensen's inequality. Thus, since u is strictly increasing, $\widehat{w}_2 > \widehat{w}_1$. Consequently, for every gamble g , the risk premium of g is greater for 1 than for 2. In other words, 1 is more risk-averse than 2. In particular, if 1 and 2 have the same wealth, then 2 accepts any gamble that 1 accepts, and 1 rejects any gamble than 2 rejects.

Example (DARA): We say an individual has decreasing absolute risk-aversion (DARA) if his utility over wealth is such that $R_a(w)$ is strictly decreasing in w .

Consider an individual with DARA who must decide how much of his initial wealth $w_0 > 0$ he allocates to a risky asset. The asset has n possible rates of return, r_1 to r_n , with probability of r_i equal to $p_i \in (0, 1)$.

Let u be the individual's utility over wealth and assume $u'' < 0$. Now let β be the amount of wealth he allocates to the risky asset. The individual's final wealth is then $w_0 - \beta + (1 + r_i)\beta = w_0 + \beta r_i$ with probability p_i . Thus, the problem of the individual is to choose $\beta \in [0, w_0]$ that maximizes

$$v(\beta) = \sum_{i=1}^n p_i u(w_0 + \beta r_i).$$

It is easy to see that $u'' < 0$ implies that $v'' < 0$. So, the solution to the individual's problem is unique and is characterized by its first-order condition. Denote by β^* the optimal choice of β . Notice that

$$v'(\beta) = \sum_{i=1}^n p_i r_i u'(w_0 + \beta r_i).$$

Thus,

$$v'(0) = \sum_{i=1}^n p_i r_i u'(w_0) = u'(w_0) \mathbb{E}[r],$$

where $\mathbb{E}[r]$ is the expected return of the asset. Hence, $\mathbb{E}[r] > 0$ is sufficient for $\beta^* > 0$; in fact, $\mathbb{E}[r] > 0$ is also necessary for $\beta^* > 0$ (prove this). Now observe that

$$v'(w_0) = \sum_{i=1}^n p_i r_i u'((1 + r_i)w_0),$$

and that a necessary and sufficient condition for $\beta^* < w_0$ is that $v'(w_0) < 0$. A necessary condition for $v'(w) < 0$ is that $r_i < 0$ for at least one i . In what follows, assume that β^* is interior.

By assumption, β^* is the unique solution to $v'(\beta^*) = 0$. This equation defines β^* implicitly as a function of w (and of the rates of return r_i as well). By the Implicit Function Theorem, we have that β^* is differentiable and that

$$\frac{d\beta^*}{dw} = \frac{-\sum_{i=1}^n p_i r_i u''(w + \beta^* r_i)}{\sum_{i=1}^n p_i r_i^2 u''(w + \beta^* r_i)}.$$

The denominator of $d\beta^*/dw$ is negative. Thus $d\beta^*/dw > 0$, i.e., the risky asset is a normal good, if the numerator is negative as well. Now notice that

$$-\sum_{i=1}^n p_i r_i u''(w + \beta^* r_i) = \sum_{i=1}^n p_i u'(w + \beta^* r_i) R_a(w + \beta^* r_i) r_i.$$

Since $R_a(w)$ is strictly decreasing in w , $r_i > 0$ implies that $R_a(w) r_i > R_a(w + \beta^* r_i) r_i$ and $r_i < 0$ implies that $R_a(w) r_i > R_a(w + \beta^* r_i) r_i$. Thus, since at least one r_i is different from zero (otherwise $\mathbb{E}[r] = 0$), we have that

$$-\sum_{i=1}^n p_i r_i u''(w + \beta^* r_i) < \sum_{i=1}^n p_i r_i u'(w + \beta^* r_i) R_a(w) = 0,$$

by the first-order condition. Hence, DARA implies the risky asset is normal.

5. Stochastic Dominance

First, let us understand better what the integral

$$\int_0^\infty u(w) dF(w) \tag{6}$$

means. If the cdf F has a density f , i.e., if there exists $f : \mathbb{R}_+ \rightarrow \mathbb{R}_+$ such that

$$F(w) = \int_0^w f(s) ds$$

for all $w \geq 0$, then

$$\int_0^\infty u(w) dF(w) = \int_0^\infty u(w) f(w) dw.$$

If F is the cdf associated with the simple lottery $(p_1 \circ w_1, \dots, p_n \circ w_n)$, then

$$\int_0^\infty u(w) dF(w) = \sum_{i=1}^n p_i u(w_i) = \sum_{i=1}^n [F(w_i) - F(w_{i-1})] u(w_i),$$

where we adopt the convention that $F(w_0) = 0$.

The integral (6) is well-defined for any cdf F , whether it has a density or not. It has the property that if $u = \mathbb{I}_{(a,b]}$, the characteristic function of the interval $(a, b]$, then

$$\int_0^\infty u(w)dF(w) = F(b) - F(a) = \Pr\{a < w \leq b\}. \quad (7)$$

In fact, property (7) completely determines the integral (6) in the following sense. For any two cdfs F and G , write $F \equiv G$ if $F(b) - F(a) = G(b) - G(a)$ for all $0 \leq a < b < \infty$. Then, $F \equiv G$ implies that

$$\int_0^\infty u(w)dF(w) = \int_0^\infty u(w)dG(w)$$

for every (integrable) function $u : \mathbb{R}_+ \rightarrow \mathbb{R}$.

In what follows we refer to the probability distribution on \mathbb{R}_+ induced by the cdf F as (the distribution) F . The integral (6) is the expected value of the random variable u given the distribution F .

First-Order Stochastic Dominance

The distribution F first-order stochastically dominates the distribution G , $F \succsim_{fosd} G$, if for all increasing $u : \mathbb{R}_+ \rightarrow \mathbb{R}$, we have that

$$\int_0^\infty u(w)dF(w) \geq \int_0^\infty u(w)dG(w).$$

Then, given two (simple) lotteries F and G , any expected utility maximizer prefers F over G if $F \succsim_{fosd} G$. It is possible to show that if $F \sim_{fosd} G$, then $F \equiv G$.

Lemma 2. $F \succsim_{fosd} G$ if, and only if, $F(w) \leq G(w)$ for all $w \geq 0$.

Proof: *Necessity.* For each $w \geq 0$, let $u = \mathbb{I}_{(w,+\infty)}$. Then u is increasing, and so

$$1 - F(w) = \int_0^\infty u(w)dF(w) \geq \int_0^\infty u(w)dG(w) = 1 - G(w),$$

which implies the desired result.

Sufficiency. We prove sufficiency for discrete cdfs. Let F be the cdf associated to the gamble $(p_1 \circ w_1, \dots, p_n \circ w_n)$ and G be the cdf associated to the gamble $(q_1 \circ w'_1, \dots, q_m \circ w'_m)$.

Without loss, we can assume that $m = n$ and $w_i = w'_i$ for all $i \in \{1, \dots, n\}$. Moreover, let $w_1 \leq \dots \leq w_n$.

Now observe that if H is the cdf associated to $(r_1 \circ w_1, \dots, r_n \circ w_n)$, then

$$\begin{aligned} \sum_{i=1}^n r_i u(w_i) &= \sum_{i=1}^n [H(w_i) - H(w_{i-1})] u(w_i) \\ &= u(w_n) + \sum_{i=1}^{n-1} H(w_i) u(w_i) - \sum_{i=2}^n H(w_{i-1}) u(w_i) \\ &= u(w_n) - \sum_{i=1}^{n-1} H(w_i) [u(w_{i+1}) - u(w_i)]. \end{aligned}$$

Hence,

$$\sum_{i=1}^n p_i u(w_i) - \sum_{i=1}^n q_i u(w_i) = \sum_{i=1}^{n-1} \underbrace{[G(w_i) - F(w_i)]}_{\geq 0} \underbrace{[u(w_{i+1}) - u(w_i)]}_{\geq 0} \geq 0.$$

□

An immediate corollary of Lemma 2 is that $F \succsim_{f\text{osd}} G$ implies that $\mathbb{E}[F] \geq \mathbb{E}[G]$. The converse is not true.

Example (Upward Probabilistic Shift): Consider the following two-stage lottery. First, draw w according to the cdf G . Once w is drawn, draw z according to the cdf H_w with $H_w(0) = 0$. The final wealth is $w' = w + z$. Denote the distribution implied by the two-stage lottery by F . The cdf of F is

$$F(w') = Pr\{w + z \leq w'\} = Pr\{z \leq w' - w\} = \int_0^\infty H_w(w' - w) dG(w)$$

and we say that F is an upward probabilistic shift of G . Since $H_w(z) = 0$ if $z \leq 0$ and $H_w(z) \leq 1$ for all z , we have that

$$F(w') \leq \int_0^{w'} H_w(w' - w) dG(w) \leq \int_0^{w'} dG(w) = G(w').$$

Thus, $F \succsim_{f\text{osd}} G$. It is possible to show (but quite difficult) that if $F \succsim_{f\text{osd}} G$, then F is an upward probabilistic shift of G .

Second-Order Stochastic Dominance

The distribution F second-order stochastically dominates G , $F \succsim_{sosl} G$, if for all concave $u : \mathbb{R}_+ \rightarrow \mathbb{R}$,

$$\int_0^\infty u(w)dF(w) \geq \int_0^\infty u(w)dG(w).$$

Then, given two (simple) lotteries F and G , any risk-averse expected utility maximizer weakly prefers F over G if $F \succsim_{sosl} G$. Notice that if $F \succsim_{sosl} G$, then $Var(F) \leq Var(G)$. The converse is not true, though⁵.

Example (Mean-Preserving Spread): Consider the following two-stage lottery. First, draw w according to G . Once w is drawn, obtain z from H_w (defined on \mathbb{R}) such that $\mathbb{E}[H_w] = 0$ and $H_w(-w) = 0$ for all w . The final wealth is $w' = w + z$. Denote the distribution implied by the two-stage lottery by F . The cdf of F is

$$F(w') = Pr\{w + z \leq w'\} = Pr\{z \leq w' - w\} = \int_0^\infty H_w(w' - w)dG(w).$$

We say that F is a mean-preserving spread of G .

In what follows, assume G has a density and H_w has a density for all w , and denote the densities of G and H_w by g and h_w , respectively. Then

$$\begin{aligned} F(w') &= \int_0^\infty H_w(w' - w)g(w)dw \\ &= \int_0^\infty \left(\int_{-w}^{w'-w} h_w(s)ds \right) g(w)dw \\ &= \int_{-w}^{w'-w} \left(\int_0^\infty h_w(s)g(w)dw \right) ds \\ &= \int_0^{w'} \left(\int_0^\infty h_w(s - w)g(w)dw \right) ds \end{aligned}$$

where the second equality follows from the fact that $H_w(-w) = 0$, the third equality follows from Fubini's theorem, and the last equality follows from the change of variable $s \mapsto s - w$.

Thus, F has a density, which is given by

$$f(s) = \int_0^\infty h_w(s - w)g(w)dw.$$

⁵See the papers by Rothschild and Stiglitz for a more detailed explanation of why one shouldn't compare the risk of two distributions by their variances.

Hence,

$$\begin{aligned}
\int_0^\infty u(w')dF(w') &= \int_0^\infty u(w')f(w')dw' \\
&= \int_0^\infty \left(\int_0^\infty h_w(w' - w)g(w)dw \right) u(w')dw' \\
&= \int_0^\infty \left(\int_0^\infty u(w')h_w(w' - w)dw' \right) g(w)dw \\
&= \int_0^\infty \left(\int_{-w}^\infty u(w + z)h_w(z)dz \right) g(w)dw \\
&= \int_0^\infty \left(\int_{-w}^\infty u(w + z)dH_w(z) \right) dG(w),
\end{aligned}$$

where the third equality follows from Fubini and the fourth equality follows from the change of variable $w' \mapsto z + w$.

First, notice that

$$\int_0^\infty w'dF(w') = \int_0^\infty \left(\int_{-w}^\infty (w + z)dH_w(z) \right) dG(w) = \int_0^\infty wdG(w).$$

Thus (confirming intuition), $\mathbb{E}[F] = \mathbb{E}[G]$. Now observe that if u is concave, then Jensen's inequality implies that

$$\begin{aligned}
\int_0^\infty \left(\int_{-w}^\infty u(w + z)dH_w(z) \right) dG(w) &\leq \int_0^\infty u \left(\int_{-w}^\infty (w + z)dH_w(z) \right) dG(w) \\
&= \int_0^\infty u(w)dG(w).
\end{aligned}$$

Thus, $G \succsim_{sosl} F$. The converse is also true (but also quite difficult to prove): if $G \succsim_{sosl} F$, then F is a mean-preserving spread of G .