

## Problem: 6.B.2

Show that if preference relation  $\succsim$  on  $\mathcal{L}$  is represented by a utility function  $U(\cdot)$  that has the expected utility form, then  $\succsim$  satisfies the independence axiom.

*Answer*

Assume that the preference relation  $\succsim$  is represented by an v.N-M expected utility function

$$U(L) = \sum_n u_n p_n \quad \text{for every } L = (p_1, \dots, p_N) \in \mathcal{L}$$

Let  $L = (p_1, \dots, p_N) \in \mathcal{L}$ ,  $L' = (p'_1, \dots, p'_N) \in \mathcal{L}$ ,  $L'' = (p''_1, \dots, p''_N) \in \mathcal{L}$ , and  $\alpha \in (0, 1)$ .

Then  $L \succsim L'$  if and only if

$$\sum_n u_n p_n \geq \sum_n u_n p'_n$$

This inequality is equivalent to (we add a third lottery to both sides in the same proportion)

$$\alpha \left( \sum_n u_n p_n \right) + (1 - \alpha) \left( \sum_n u_n p''_n \right) \geq \alpha \left( \sum_n u_n p'_n \right) + (1 - \alpha) \left( \sum_n u_n p''_n \right)$$

This inequality holds if and only if

$$\alpha L + (1 - \alpha)L'' \succsim \alpha L' + (1 - \alpha)L''$$

Hence,  $L \succsim L'$  if and only if

$$\alpha L + (1 - \alpha)L'' \succsim \alpha L' + (1 - \alpha)L''$$

Thus the independence axiom holds.

## Problem: 6.C.4 (a, b)

Suppose that there are  $N$  risky assets whose returns  $z_n = (n = 1, \dots, N)$  per dollar invested are jointly distributed according to the distribution function  $F(z_1, \dots, z_N)$ . Assume also that all the returns are nonnegative with probability one. Consider an individual who has a continuous, increasing, and concave Bernoulli utility function  $u(\cdot)$  over  $R_+^N$ . Define the utility function  $U(\cdot)$  of this investor over  $R_+^N$ , the set of all nonnegative portfolios, by

$$U(\alpha_1, \dots, \alpha_N) = \int u(\alpha_1 z_1 + \dots + \alpha_N z_N) dF(z_1, \dots, z_N)$$

Prove that  $U(\cdot)$  is:

- a) Increasing.
- b) Concave

*Answer*

- a) Let

$$\alpha = (\alpha_1, \dots, \alpha_N) \in R_+^N$$

$$\alpha' = (\alpha'_1, \dots, \alpha'_N) \in R_+^N$$

$$\alpha \geq \alpha'$$

Then we have the following:

$$\sum_n \alpha_n z_n \geq \sum_n \alpha'_n z_n$$

for almost every realization  $(z_1, \dots, z_N)$ , because all the returns are nonnegative with probability one.

Since  $u(\cdot)$  is increasing, this implies that

$$u\left(\sum_n \alpha_n z_n\right) \geq u\left(\sum_n \alpha'_n z_n\right)$$

with probability one. Hence,

$$\int u\left(\sum_n \alpha_n z_n\right) \geq \int u\left(\sum_n \alpha'_n z_n\right)$$

That is the same as

$$U(\alpha) \geq U(\alpha')$$

**Thus,  $U$  is increasing.**

b) Let

$$\alpha = (\alpha_1, \dots, \alpha_N) \in R_+^N$$

$$\alpha' = (\alpha'_1, \dots, \alpha'_N) \in R_+^N$$

$$\lambda \in [0,1]$$

Then by concavity of  $u(\cdot)$ , we get

$$\begin{aligned} u\left(\sum_n (\lambda \alpha_n + (1-\lambda) \alpha'_n) z_n\right) &= u\left(\lambda \sum_n \alpha_n z_n + (1-\lambda) \sum_n \alpha'_n z_n\right) \\ u\left(\lambda \sum_n \alpha_n z_n + (1-\lambda) \sum_n \alpha'_n z_n\right) &\geq \lambda u\left(\sum_n \alpha_n z_n\right) + (1-\lambda) u\left(\sum_n \alpha'_n z_n\right) \end{aligned}$$

which holds almost for all realization  $(z_1, \dots, z_N)$ .

Hence,

$$\begin{aligned} u\left(\lambda \sum_n \alpha_n z_n + (1-\lambda) \sum_n \alpha'_n z_n\right) &\geq \lambda u\left(\sum_n \alpha_n z_n\right) + (1-\lambda) u\left(\sum_n \alpha'_n z_n\right) \\ \int u\left(\lambda \sum_n \alpha_n z_n + (1-\lambda) \sum_n \alpha'_n z_n\right) dF(z_1, \dots, z_N) &\geq \int \left(\lambda u\left(\sum_n \alpha_n z_n\right) + (1-\lambda) u\left(\sum_n \alpha'_n z_n\right)\right) dF(z_1, \dots, z_N) \\ U(\lambda \alpha + (1-\lambda) \alpha'_n) &\geq \int u\left(\sum_n \alpha_n z_n\right) dF(z_1, \dots, z_N) + (1-\lambda) \int u\left(\sum_n \alpha'_n z_n\right) dF(z_1, \dots, z_N) \\ U(\lambda \alpha + (1-\lambda) \alpha'_n) &\geq \lambda U(\alpha) + (1-\lambda) U(\alpha'_n) \end{aligned}$$

**Thus,  $U(\cdot)$  is concave.**

Equalities used in the part b of the problem.

$$\int u\left(\sum_n (\lambda \alpha_n + (1-\lambda) \alpha'_n) z_n\right) dF(z_1, \dots, z_N) = U(\lambda \alpha + (1-\lambda) \alpha'_n)$$

And

$$\begin{aligned} &\int \left(\lambda u\left(\sum_n \alpha_n z_n\right) + (1-\lambda) u\left(\sum_n \alpha'_n z_n\right)\right) dF(z_1, \dots, z_N) \\ &= \lambda \int u\left(\sum_n \alpha_n z_n\right) dF(z_1, \dots, z_N) + (1-\lambda) \int u\left(\sum_n \alpha'_n z_n\right) dF(z_1, \dots, z_N) \\ &= \lambda U(\alpha) + (1-\lambda) U(\alpha'_n) \end{aligned}$$

## Problem: 6.C.15 (a, b)

Assume that in the world with uncertainty, there are two assets. The first is riskless asset that pays 1 dollar. The second pays amounts  $a$  and  $b$  with probabilities of  $\pi$  and  $(1 - \pi)$ , respectively. Denote the demand for the two assets by  $(x_1, x_2)$ .

Suppose that a decision maker's preferences satisfy the axioms of expected utility theory and that he is a risk averter. The decision maker's wealth is 1, and so are the prices of the assets. Therefore, the decision maker's budget constraint is given by

$$x_1 + x_2 = 1, \quad x_1, x_2 \in [0, 1]$$

- a) Give a simple necessary condition (involving  $a$  and  $b$  only) for the demand for the riskless asset to be strictly positive.
- b) Give a simple necessary condition (involving  $a$ ,  $b$  and  $\pi$  only) for the demand for the risky asset to be strictly positive.

### Answer

Throughout the problem we assume that  $a \neq b$  because otherwise there would be no uncertainty involved in the payment of the second asset.

- a) If  $\min\{a, b\} \geq 1$ , the risky asset pays at least as high a return as the riskless asset at both states, and strictly higher return at one of them. Then all the wealth is invested to the risky asset.

Thus,  $\min\{a, b\} < 1$  is a necessary condition for the demand for the riskless asset to be strictly positive.

- b) If  $[\pi a + (1 - \pi)b] \leq 1$ , then the expected return does not exceed the payments of the riskless asset and hence the risk-averse decision maker does not demand the risky asset at all.

Since, we care about the utilities of returns, not the returns itself, we have the following.

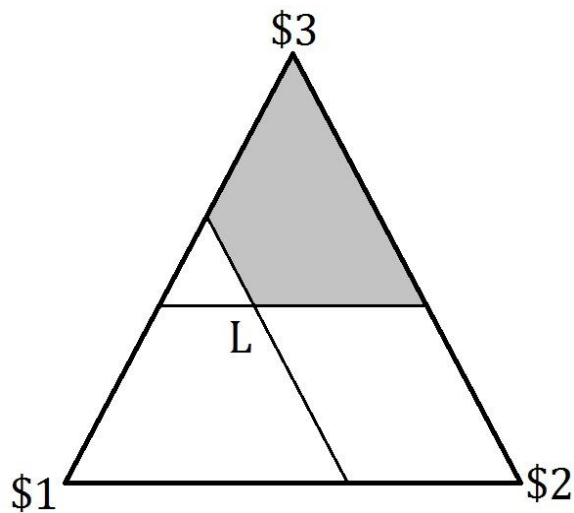
$\pi u(a) + (1 - \pi)u(b) > u(1)$  is a necessary condition for the demand for the risky asset to be strictly positive.

## Problem 6.D.1

The purpose of this exercise is to prove Proposition 4.D.1 in a two-dimensional probability simplex. Suppose that there are three monetary outcomes: 1 dollar, 2 dollars, and 3 dollars. Consider the probability simplex of Figure 6.B.1(b).

- For a given lottery  $L$  over these outcomes, determine the region of the probability simplex in which lie the lotteries whose distributions first-order stochastically dominate the distribution of  $L$ .
- Given a lottery  $L$ , determine the region of the probability simplex in which lie the lotteries  $L'$  such that  $F(x) \leq G(x)$  for every  $x$ , where  $F(\cdot)$  is the distribution of  $L'$  and  $G(\cdot)$  is the distribution of  $L$ . [Notice that we get the same region as in **(a)**].

Answer



6.D.2 Prove that if  $F(\cdot)$  first-order stochastically dominates  $G(\cdot)$ , then the mean of  $x$  under  $F(\cdot)$ ,  $\int x dF(x)$ , exceeds that under  $G(\cdot)$ ,  $\int x dG(x)$ . Also provide an example where  $\int x dF(x) > \int x dG(x)$ , but  $F(\cdot)$  does not first-order stochastically dominate  $G(\cdot)$ .

6.D.2 [First printing errata: The phrase "the mean of  $x$  under  $F(\cdot)$ ,  $\int x dF(x)$ , exceeds that under  $G(\cdot)$ ,  $\int x dG(x)$ " should be "the mean of  $x$  under  $G(\cdot)$ ,  $\int x dG(x)$ , cannot exceed that under  $F(\cdot)$ ,  $\int x dF(x)$ ". That is, the equality of the two means should be allowed.] For the first assertion, simply put  $u(x) = x$  and apply Definition 6.D.1. As for the second, let  $p \in (0, 1/2)$  and consider the following two distributions:

$$F(z) = \begin{cases} 0 & \text{if } z < 0, \\ p & \text{if } 0 \leq z < 2, \\ 1 & \text{if } 2 \leq z, \end{cases}$$

$$G(z) = \begin{cases} 0 & \text{if } z < 1, \\ 1 & \text{if } 1 \leq z. \end{cases}$$

Then  $F(1/2) = p > 0 = G(1/2)$  and  $\int x dF(x) = 2(1 - p) > 1 = \int x dG(x)$ . Hence  $F(\cdot)$  does not first-order stochastically dominate  $G(\cdot)$ , but the mean of  $F(\cdot)$  is larger than that of  $G(\cdot)$ .

Definitions: First Order Stochastic Dominance (FOSD):

$F(\cdot)$  FOSD  $G(\cdot)$  if for every nondecreasing  $u: \mathbb{R} \rightarrow \mathbb{R}$ :

$$\int u(x) dF(x) \geq \int u(x) dG(x)$$

or  $F(x) \leq G(x)$  for every  $x$ .

Second Order Stochastic Dominance (SOSD):

For  $F(\cdot)$  and  $G(\cdot)$  with the same mean,  $F(\cdot)$  SOSD  $G(\cdot)$  (is less risky than) if for every nondecreasing concave function  $u: \mathbb{R}_+ \rightarrow \mathbb{R}$ :

$$\int u(x) dF(x) \geq \int u(x) dG(x)$$

Prove: if  $F(\cdot)$  FOSD  $G(\cdot)$  then mean of  $x$  under  $F(\cdot)$  exceeds that under  $G(\cdot)$ :

$$\int x \, dF(x) \geq \int x \, dG(x)$$

give an example where  $\int x \, dF(x) \geq \int x \, dG(x)$  but  $F(\cdot)$  does not FOSD  $G(\cdot)$ .

Answer: From FOSD, if  $F(\cdot)$  FOSD  $G(\cdot)$ :

$$\int u(x) \, dF(x) \geq \int u(x) \, dG(x)$$

for every nondecreasing  $u(x)$ .

Let  $u(x) = x$  (so that  $u$  remains nondecreasing).

Then plugging into the above inequality we obtain:

$$\int x \, dF(x) \geq \int x \, dG(x)$$

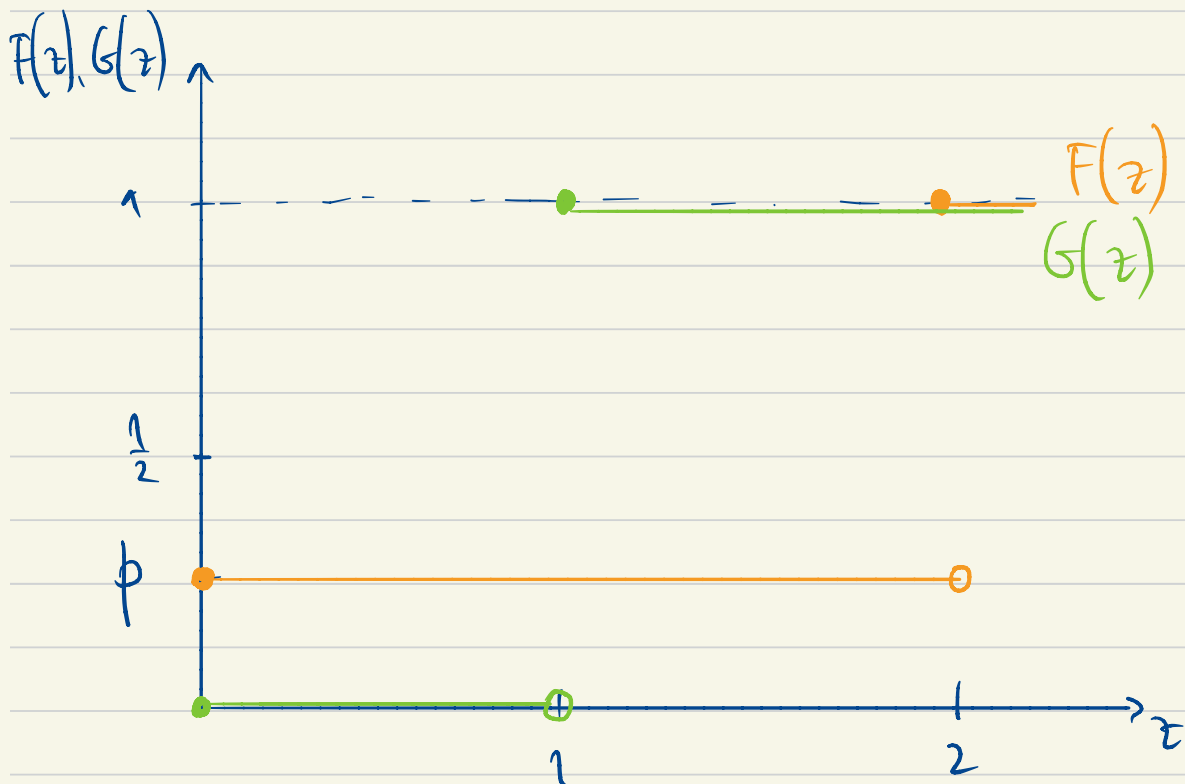
and so the mean of  $x$  under  $F(\cdot)$  exceed (or is equal) to that under  $G(\cdot)$ .

Now, for an example where  $\int x dF(x) > \int x dG(x)$  but  $F(\cdot)$  does not FOSD  $G(\cdot)$ .

Let  $p \in (0, \frac{1}{2})$  and consider the (cumulative) distributions:

$$F(z) = \begin{cases} 0 & \text{if } z < 0 \\ p & \text{if } 0 \leq z < 2 \\ 1 & \text{if } z \geq 2 \end{cases}$$

$$G(z) = \begin{cases} 0 & \text{if } z < 1 \\ 1 & \text{if } z \geq 1 \end{cases}$$





Graphically,  $F(z)$  FOSD  $G(z)$  if  $F(z)$  lies below  $G(z)$  (or at the same level, but never above) for all possible values of  $z$ . (higher probs of bigger payoff)

Note that this does not happen for  $z \in [0, 1]$ .

- $\int x dF(x) = 2(1-p)$

- $\int x dG(x) = 1$

Thus,  $\int x dF(x) > \int x dG(x)$  for  $0 < p < \frac{1}{2}$ .

But, for example,  $z = \frac{1}{2}$ :

- $F\left(\frac{1}{2}\right) = p$

- $G\left(\frac{1}{2}\right) = 0$

So  $F\left(\frac{1}{2}\right) > G\left(\frac{1}{2}\right)$  (will also hold for  $0 \leq z < 1$ )

Thus  $F(\cdot)$  does not FOSD  $G(\cdot)$ .

hence the mean of  $F(\cdot)$  is larger than that of  $G(\cdot)$  but  $F(\cdot)$  does not FOSD  $G(\cdot)$ .

In conclusion: if  $F(\cdot)$  FOSD  $G(\cdot) \rightarrow \int x dF(x) > \int x dG(x)$   
but  $\int x dF(x) > \int x dG(x) \not\Rightarrow F(\cdot)$  FOSD  $G(\cdot)$ .