

### S.3 Utility and Risk

A.1. (a) Denoting  $ARA(w)$  as the absolute level of risk aversion at wealth  $w$ ,

$$ARA(w) = -\frac{u''(w)}{u'(w)} = \frac{a \exp(-aw)}{\exp(-aw)} = a$$

(b)  $\lim_{a \rightarrow 0} u'(w) = \lim_{a \rightarrow 0} \exp(-aw) = 1$

(c)  $\lim_{a \rightarrow \infty} \mathbb{E}u(\tilde{x}) = \lim_{a \rightarrow \infty} \sum_{x \geq 0} \left[ p_x \frac{1 - \exp(-ax)}{a} \right] + \sum_{x < 0} \left[ p_x \frac{1 - \exp(-ax)}{a} \right] = -\infty$   
since lottery  $x$  includes negative payoffs.

A.2. (a) Denoting the risk premium of the lottery by  $\pi$ ,

$$u(4 - \pi) = (4 - \pi)^2 = \frac{1}{2}2^2 + \frac{1}{2}6^2 = 20$$

$\therefore \pi = 4 - \sqrt{20} = -0.472$ . The risk premium is negative because the utility function is convex over the range of interest ( $u'' = 2 > 0$ ), and therefore the decision maker is willing to pay to take on a zero mean risk ('risk loving').

(b) Denoting the new risk premium of the lottery by  $\pi'$ ,

$$v(4 - \pi') = (4 - \pi')^4 = \frac{1}{2}2^4 + \frac{1}{2}6^4 = 656$$

$\therefore \pi' = 4 - \sqrt[4]{656} = -1.061$ , i.e. the risk premium decreases (or equivalently the decision maker is willing to pay a higher price to take on the zero mean risk).  $v(w) = f(u(w))$  where  $f(u) = u^2$ , a convex function ( $f''(u) > 0$ ).

A.3.  $A(w) := -\frac{u''(w)}{u'(w)} = -\frac{d}{dw} \ln u'(w)$ . Integrating over  $[z_0, z]$  gives  $\int_{z_0}^z A(w)dw := -\ln(u'(z)) + \ln(u'(z_0))$ , and so  $u'(z) = u'(z_0) \exp \left\{ -\int_{z_0}^z A(w)dw \right\}$ , which can be rewritten in the desired form.

B.1. (a)  $\mathbb{E}[\tilde{y}] = \int_0^1 \lambda y^{\lambda-1} y dy = \left[ \frac{\lambda}{\lambda+1} y^{\lambda+1} \right]_0^1 = \frac{\lambda}{\lambda+1}$   
 $\mathbb{E}[\tilde{y}^2] = \int_0^1 \lambda y^{\lambda-1} y^2 dy = \left[ \frac{\lambda}{\lambda+2} y^{\lambda+2} \right]_0^1 = \frac{\lambda}{\lambda+2}$   
 $\therefore \text{Var}[\tilde{y}] = \frac{\lambda}{\lambda+2} - \left[ \frac{\lambda}{\lambda+1} \right]^2$

(b)  $\mathbb{P}[\tilde{y} \geq 0.5] = \int_{0.5}^1 \lambda y^{\lambda-1} dy = [y^\lambda]_{0.5}^1 = 1 - 0.5^\lambda$

(c)  $u'(y) = (1-r)y^{-r}$ .  $u''(y) = -r(1-r)y^{-r-1}$  which is strictly negative for  $r \in (0, 1)$  and so the consumer is strictly risk averse.

$$\mathbb{E}u(\tilde{y}) = \int_0^1 \lambda y^{\lambda-1} y^{1-r} dy = \left[ \frac{\lambda}{\lambda-r+1} y^{\lambda-r+1} \right]_0^1 = \frac{\lambda}{\lambda-r+1}$$

$$u(\mathbb{E}(\tilde{y})) = \left( \frac{\lambda}{\lambda+1} \right)^{1-r}$$

Consider

$$\begin{aligned} \ln \left[ \frac{u(\mathbb{E}(\tilde{y}))}{\mathbb{E}(u(\tilde{y}))} \right] &= \ln \left[ \left( \frac{\lambda}{\lambda+1} \right)^{1-r} \times \frac{\lambda-r+1}{\lambda} \right] \\ &= (1-r) \ln \left[ \frac{\lambda}{\lambda+1} \right] + \ln \left[ \frac{\lambda-r+1}{\lambda} \right] \\ \ln \left[ \frac{u(\mathbb{E}(\tilde{y}))}{\mathbb{E}(u(\tilde{y}))} \right] \Big|_{r=0} &= 0 \\ \ln \left[ \frac{u(\mathbb{E}(\tilde{y}))}{\mathbb{E}(u(\tilde{y}))} \right] \Big|_{r=1} &= 0 \\ \frac{d^2}{dr^2} \ln \left[ \frac{u(\mathbb{E}(\tilde{y}))}{\mathbb{E}(u(\tilde{y}))} \right] &= -\frac{1}{(\lambda-r+1)^2} < 0 \end{aligned}$$

So  $\ln \left[ \frac{u(\mathbb{E}(\tilde{y}))}{\mathbb{E}(u(\tilde{y}))} \right]$  is strictly concave and takes the value 0 at  $r = 0$  and  $r = 1$ . We must therefore have that  $u(\mathbb{E}(\tilde{y})) > \mathbb{E}(u(\tilde{y}))$  for  $0 < r < 1$ .

B.2.  $Y_i$  are i.i.d. with mean  $\mu$  and variance  $\sigma^2$ , so  $\sum_{i=1}^n Y_i$  has mean  $n\mu$  and variance  $n\sigma^2$ .

(a) Total premiums =  $nA = n\mu + 10\sigma\sqrt{n}$ . The probability that claims exceed premiums is

$$\mathbb{P} \left( \sum Y_i > nA \right) = \mathbb{P} \left( \sum Y_i - \mathbb{E} \sum Y_i > 10\sigma\sqrt{n} \right) \leq \frac{\text{Var}(\sum Y_i)}{100\sigma^2 n} = 1/100,$$

using Chebyshev's inequality.

(b) As above, the probability that claims exceed premiums is

$$\mathbb{P} \left( \sum Y_i - \mathbb{E} \sum Y_i > 3\sigma\sqrt{n} \right).$$

By the CLT for the i.i.d. variables  $Y_i$ , this probability tends to  $\mathbb{P}(Z > 3)$  as  $n \rightarrow \infty$ , where  $Z$  has a standard normal distribution. From the table, the probability on the RHS is 0.00135, so for large  $n$  the probability that claims exceed premiums is less than 0.01 as required.

B.3. (a) Denoting the certainty equivalent and risk premium of  $x$  by  $CE_x$  and  $RP_x$  respectively,

$$u(10 - RP_x) = u(10 + CE_x) = (10 + CE_x)^{1/2} = \mathbb{E}u(10 + \tilde{x}) = \frac{1}{2}u(4) + \frac{1}{2}u(16) = 3$$

$$\therefore CE_x = 9 - 10 = -1 \text{ and } RP_x = +1.$$

(b) The Arrow-Pratt approximation of the risk premium

$$\hat{RP}_x = \frac{1}{2} \mathbb{E} \tilde{x}^2 ARA(10) = \frac{1}{2} \times 36 \times \frac{\frac{1}{4} 10^{-3/2}}{\frac{1}{2} 10^{-1/2}} = \frac{36}{40} = 0.9$$

(c)  $ARA(w) = -\frac{u'(w)}{u''(w)} = \frac{\frac{1}{4}w^{-3/2}}{\frac{1}{2}w^{-1/2}} = \frac{1}{2w}$  which is positive and decreasing with  $w$ .

$RRA(w) = -w\frac{u'(w)}{u''(w)} = \frac{1}{2}$  which is constant.

(d) Denoting the certainty equivalent and risk premium of  $x$  by  $CE_x$  and  $RP_x$  respectively,

$$\begin{aligned} v(10 - RP_x) &= v(10 + CE_x) = (10 + CE_x)^{1/4} = \mathbb{E}v(10 + \tilde{x}) \\ &= \frac{1}{2}v(4) + \frac{1}{2}v(16) = \frac{1}{\sqrt{2}} + 1 = 1.707 \text{ (4sf)} \end{aligned}$$

$\therefore CE_x = 1.707^4 - 10 = -1.507$  and  $RP_x = +1.507$ . The risk premium has increased substantially. Note that  $w^{1/4}$  is a concave transformation of  $w^{1/2}$  and so  $v$  is more risk averse than  $u$  in the sense of Arrow-Pratt.

(e) The Arrow-Pratt approximation of the risk premium

$$\hat{R}P_x = \frac{1}{2}\mathbb{E}\tilde{y}^2 ARA(10) = \frac{1}{2} \times 9 \times \frac{\frac{1}{4}10^{-3/2}}{\frac{1}{2}10^{-1/2}} = \frac{9}{40} = \frac{0.9}{4}$$

The risk premium is proportional to the squared magnitude of the risk, which has decreased by a factor of 4 since the magnitude of the risk has decreased by a factor of 2.

B.4. If the decision maker has preferences that satisfy constant relative risk aversion then preferences can be represented by a utility function  $u(w) = \frac{w^{1-\gamma}}{1-\gamma}$  for some  $\gamma \in \mathbb{R}$  (or  $\ln(w)$  for the case of  $\gamma = 1$ ).

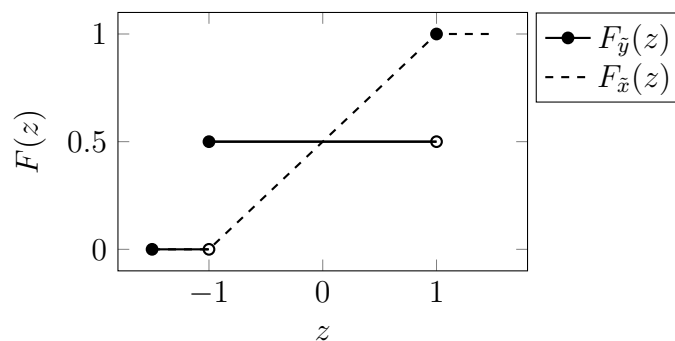
Utility of certain wealth  $w$  is  $\frac{w^{1-\gamma}}{1-\gamma}$  and expected utility of lottery  $w(1 + \tilde{x})$  is

$$\begin{aligned} \mathbb{E}[u(w(1 + \tilde{x}))] &= \mathbb{E}_{\tilde{x}} \left[ \frac{[w(1 + \tilde{x})]^{1-\gamma}}{1-\gamma} \right] \\ &= \frac{w^{1-\gamma}}{1-\gamma} \times \mathbb{E}_{\tilde{x}} [(1 + \tilde{x})^{1-\gamma}] \end{aligned}$$

and so the lottery will be (strictly) preferred to certain wealth iff  $\mathbb{E}[(1 + \tilde{x})^{1-\gamma}] \geq (>)1$ . This condition does not depend on  $w$ .

Similarly, for logarithmic utility function  $u(w) = \ln(w)$ ,  $\mathbb{E}[u(w(1 + \tilde{x}))] = \ln(w) + \mathbb{E}[\ln(1 + \tilde{x})]$  which is (strictly) greater than  $\ln(w)$  iff  $\mathbb{E}[\ln(1 + \tilde{x})] \geq (>)0$ . Again, this condition does not depend on  $w$ .

B.5. (a) See the following figure



(b) Both distributions have support on the interval  $[-1, 1]$ . We have

$$\begin{aligned}\int_{-1}^{\theta} F_{\tilde{x}}(z)dz &= \int_{-1}^{\theta} \frac{z+1}{2}dz = \left[ \frac{z^2}{4} + \frac{z}{2} \right]_{-1}^{\theta} = \frac{\theta^2}{4} + \frac{\theta}{2} + \frac{1}{4} \\ \int_{-1}^{\theta} F_{\tilde{y}}(z)dz &= \int_{-1}^{\theta} \frac{1}{2}dz = \left[ \frac{z}{2} \right]_{-1}^{\theta} = \frac{\theta}{2} + \frac{1}{2} \\ \therefore \int_{-1}^{\theta} F_{\tilde{y}}(z) - F_{\tilde{x}}(z)dz &= \frac{1-\theta^2}{4} \begin{cases} = 0 \text{ for } \theta = 1 \\ \geq 0 \text{ for } \theta \in [-1, 1) \end{cases}\end{aligned}$$

$\tilde{x}$  therefore second order stochastically dominates  $\tilde{y}$ , and so  $\tilde{y}$  is said to be riskier than  $\tilde{x}$ .

(c) For each outcome  $x$  of  $\tilde{x}$  add a zero mean noise  $\tilde{\varepsilon}_x := \{1-x, \frac{1+x}{2}; -(1+x), \frac{1-x}{2}\}$ . Conditional on any  $x$ , there are two potential outcomes,  $x+1-x = +1$  and  $x-(1+x) = -1$ , and  $\mathbb{E}[\tilde{\varepsilon}_x] = 0$ . By symmetry, and over all possible outcomes of  $x$ , these two outcomes of  $\tilde{x} + \tilde{\varepsilon}$  have equal probability of  $\frac{1}{2}$ , where conditional on  $\tilde{x}$ ,  $\tilde{\varepsilon}$  has distribution  $\tilde{\varepsilon}_x$ , and so  $\tilde{x} + \tilde{\varepsilon} \sim (-1, \frac{1}{2}; +1, \frac{1}{2})$ , which is the distribution of  $\tilde{y}$ .

B.6. Let  $\tilde{A} = (80, \frac{1}{4}; 100, \frac{1}{4}; 120, \frac{1}{4}; 140, \frac{1}{4})$ ,  $\tilde{B} = (90, \frac{1}{2}; 130, \frac{1}{2})$  and  $\tilde{\varepsilon} = (+10, \frac{1}{2}; -10, \frac{1}{2})$ , and  $\tilde{B}$  and  $\tilde{\varepsilon}$  independent. Then  $\tilde{B} + \tilde{\varepsilon} \sim \tilde{A}$ . Conditional on each outcome of  $\tilde{B}$ ,  $\tilde{\varepsilon}$  has zero mean. Therefore by the Rothschild-Stiglitz Theorem, Project B SSD Project A.

B.7. (a) Since  $g(x)$  is a probability density function with support  $[0, 1]$ , we must have

$$\int_0^1 g(x)dx = 1 \Rightarrow c \int_0^1 x^2 - x + \frac{1}{4}dx = c \left[ \frac{x^3}{3} - \frac{x^2}{2} + \frac{x}{4} \right]_0^1 = \frac{c}{12} = 1$$

and so  $c = 12$ .

$$\int_0^1 x g(x)dx = 12 \int_0^1 x^3 - x^2 + \frac{x}{4} dx = 12 \left[ \frac{x^4}{4} - \frac{x^3}{3} + \frac{x^2}{8} \right]_0^1 = 12 \left( \frac{6}{24} - \frac{8}{24} + \frac{3}{24} \right) = \frac{1}{2}$$

This shows that the mean of  $L_b$  is equal to the mean of  $L_a$ .

Defining the interval  $I := \left[ \frac{1}{2} - \sqrt{1/12}, \frac{1}{2} + \sqrt{1/12} \right]$  and denoting the pdf of  $L_a$  as  $f$ , we therefore have  $g(x) \geq f(x)$  for  $x \in [0, 1] \setminus I$  (the set  $[0, 1]$  excluding elements in  $I$ ), and  $g(x) \leq f(x)$  for  $x \in I$ . Distribution  $g(x)$  can be obtained from distribution  $f(x)$  by moving some probability mass from the interval  $I$  to outside of  $I$  without changing the mean, and so  $L_b$  is a mean-preserving spread of  $L_a$ .

(b) No. The cdf of lottery  $L_a$  is  $F(x) = x$ ,  $x \in [0, 1]$  and the cdf of lottery  $L_b$  is

$$G(x) = \int_0^x g(z)dz = 12 \left[ \frac{z^3}{3} - \frac{z^2}{2} + \frac{z}{4} \right]$$

$G(1/4) = 0.4375 > 0.25 = F(1/4)$  and  $G(3/4) = 0.5625 < 0.75 = F(3/4)$ . Therefore neither  $G(x) \geq F(x) \forall x \in [0, 1]$  nor  $G(x) \leq F(x) \forall x \in [0, 1]$ .

C.1. We are required to calculate the risk premium  $\pi(w_0, u, \tilde{y})$  when  $u(w) = -\exp\{-Aw\}$  and  $\tilde{y} \sim N(0, \sigma^2)$ . We have

$$\begin{aligned}
u(w_0 - \pi) &= -\exp\{-A(w_0 - \pi)\} \\
&= -\int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left\{-\frac{y^2}{2\sigma^2}\right\} \exp\{-A(w_0 + y)\} dy \\
&= -\int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left\{-\frac{y^2 + 2\sigma^2 A(w_0 + y)}{2\sigma^2}\right\} dy \\
&= -\int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left\{-\frac{(y + \sigma^2 A)^2}{2\sigma^2}\right\} \exp\left\{-Aw_0 + \frac{1}{2}\sigma^2 A^2\right\} dy \\
&= -\exp\left\{-Aw_0 + \frac{1}{2}\sigma^2 A^2\right\} \\
\therefore \pi &= \frac{1}{2}\sigma^2 A
\end{aligned}$$

Since  $\tilde{y}$  has zero mean,  $\mathbb{E}[y^2] = \sigma^2$ , and since  $u$  satisfies CARA,  $A(w_0) = A$ , and so we have the required result.

C.2 Let  $\tilde{y}$  denote the distribution of final wealth under the perfect diversification strategy. Now under any alternative feasible strategy  $A = (\alpha_1, \dots, \alpha_n)$  we have

$$\sum_{i=1}^n \alpha_i \tilde{x}_i = \tilde{y} + \sum_{i=1}^n \left(\alpha_i - \frac{1}{n}\right) \tilde{x}_i$$

From the Rothschild-Stiglitz theorem we are done if we can show that  $\sum_{i=1}^n \left(\alpha_i - \frac{1}{n}\right) \tilde{x}_i$  has a zero mean conditional on  $\tilde{y}$ . Now, by joint independence of  $\tilde{x}_i$  and symmetry of  $\tilde{y}$ ,  $\mathbb{E}[\tilde{x}_i | \tilde{y}]$  is independent of  $i$ . We may denote it  $k$ . Then, we have

$$\begin{aligned}
\mathbb{E}\left[\sum_{i=1}^n \left(\alpha_i - \frac{1}{n}\right) \tilde{x}_i | \tilde{y}\right] &= \sum_{i=1}^n \left(\alpha_i - \frac{1}{n}\right) \mathbb{E}[\tilde{x}_i | \tilde{y}] \\
&= k \sum_{i=1}^n \left(\alpha_i - \frac{1}{n}\right) = 0
\end{aligned}$$

This proves that the alternative strategy is second order stochastically dominated by the perfect diversification strategy.

C.3. (a)  $u$  is piecewise linear with gradient  $u'(z) = 1$  for  $z < z_0$  and  $u'(z) = a < 1$  for  $z > z_0$ . We must show that for any  $z, z' \in \mathbb{R}$ ,  $\lambda \in [0, 1]$ ,

$$\lambda u(z) + (1 - \lambda)u(z') \leq u(\lambda z + (1 - \lambda)z').$$

If  $z, z' \geq z_0$  or  $z, z' \leq z_0$  then this weak inequality is satisfied with equality since  $u$  is linear between  $z$  and  $z'$ . If  $z \leq z_0$  and  $z' \geq z_0$  then from the definition of  $u$ ,

$$\lambda u(z) + (1 - \lambda)u(z') = \lambda z + (1 - \lambda)az' + (1 - \lambda)(1 - a)z_0$$

There are two cases. For  $\lambda z + (1 - \lambda)z' \leq z_0$ ,

$$\begin{aligned} u(\lambda z + (1 - \lambda)z') &= \lambda z + (1 - \lambda)z' \\ \therefore \lambda u(z) + (1 - \lambda)u(z') - u(\lambda z + (1 - \lambda)z') &= (1 - a)(1 - \lambda)(z_0 - z') \leq 0 \end{aligned}$$

since  $z' \geq z_0$  by assumption, and in the second case,  $\lambda z + (1 - \lambda)z' \geq z_0$ ,

$$\begin{aligned} u(\lambda z + (1 - \lambda)z') &= (1 - a)z_0 + a(\lambda z + (1 - \lambda)z') \\ \therefore \lambda u(z) + (1 - \lambda)u(z') - u(\lambda z + (1 - \lambda)z') &= (1 - a)\lambda(z - z_0) \leq 0 \end{aligned}$$

since  $z \leq z_0$  by assumption.

- (b) Denoting the cdf of  $\tilde{x}$  as  $F(x)$ , where  $\mathbb{E}[\tilde{x}] = 0$ , the risk premium  $\pi(z_0, u, k\tilde{x})$  is nonnegative since  $u$  is concave, and therefore satisfies

$$\begin{aligned} u(z_0 - \pi) &= z_0 - \pi = \mathbb{E}[u(z_0 + k\tilde{x})] \\ &= \int_{-\infty}^0 (z_0 + kx)dF(x) + \int_0^{\infty} ((1 - a)z_0 + a(z_0 + kx))dF(x) \\ &= z_0 + k \left[ \int_{-\infty}^0 x dF(x) + a \int_0^{\infty} x dF(x) \right] \\ &= z_0 + k \left[ \mathbb{E}[\tilde{x}] - (1 - a) \int_0^{\infty} x dF(x) \right] \\ &= z_0 - k(1 - a) \int_0^{\infty} x dF(x) \\ \therefore \pi(z_0, u, k\tilde{x}) &= k(1 - a) \int_0^{\infty} x dF(x) \end{aligned}$$

For  $\tilde{x}$  non-degenerate,  $\mathbb{E}[\tilde{x}] = 0$  implies  $\int_0^{\infty} x dF(x)$  is positive and finite and so  $\pi(z_0, u, k\tilde{x})$  is linear in  $k$  and  $\lim_{k \rightarrow 0} \pi(z_0, u, k\tilde{x}) = 0$ .

- (c) Consider

$$\phi(z) := \begin{cases} z & \text{if } z \leq z_0 \\ z_0 + b(z - z_0) & \text{if } z > z_0 \end{cases}.$$

where  $0 < b < 1$ . We therefore have

$$\phi(u(z)) := \begin{cases} z & \text{if } z \leq z_0 \\ z_0 + ab(z - z_0) & \text{if } z > z_0 \end{cases}.$$

$\phi$  is an increasing concave transformation from the first part of this exercise.

Now consider a decrease in parameter  $a$  of function  $u$  to  $a'$ . The new utility function is the same as  $\phi \circ u$  where  $\phi$  is defined with parameter  $b = a'/a \in [0, 1]$ . A reduction in  $a$  to  $a'$  therefore increases the degree of risk aversion because it can result from transforming  $u$  with an increasing, concave function  $\phi$ .

- (d) We prove by finding a pure risk with zero risk at wealth  $w_0$  and strictly positive risk premium at wealth  $w'_0 > w_0$ . This is sufficient to show that  $u$  does not exhibit decreasing absolute risk aversion.

Let the pure risk to consider be the lottery  $\tilde{x} := (+x, \frac{1}{2}; -x, \frac{1}{2})$  and let  $w_0, w_0 + x \leq z_0$ . Then the risk premium is zero at wealth level  $w_0$  since from the definition of the risk premium  $w_0 - \pi(w_0, u, \tilde{x}) = 0.5(w_0 - x) + 0.5(w_0 + x) = w_0$ . Now consider a level of wealth  $w'_0 > w_0$  such that  $z_0 - x < w'_0 < z_0$ . The risk premium  $\pi(w'_0, u, \tilde{x})$  is defined by

$$\begin{aligned} 0.5(w'_0 - x) + 0.5((1 - a)z_0 + a(w'_0 + x)) &= w'_0 - \pi(w'_0, u, \tilde{x}) \\ \therefore \pi(w'_0, u, \tilde{x}) &= 0.5(1 - a)(w'_0 + x - z_0) > 0 \end{aligned}$$

since  $w'_0 + x > z_0$  by construction and  $a \in (0, 1)$