Homework 1 - COMM 671

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2022-09-20

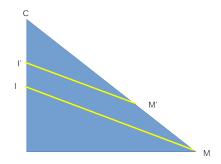
1 Show that Expected Utility Theory implies that indifference curves are parallel straight lines in the Machina triangle

I'll start by showing that EUT implies indifference curves are straight lines. Suppose we have a Machina Triangle like the one below and the agent is indifferent between points I and M. By the independence axiom, the agent must also be indifferent between N = pM + (1-p)I and J = pI + (1-p)M for any $p \in [0,1]$. But then she is indifferent between any points along the straight line connecting I and M, so the indifference curve is a straight line.

To show they're parallel, consider

$$M' = pM + (1 - p)C$$
$$I' = pI + (1 - p)C$$

Since the agent is indifferent between M and I, the independence axiom implies they must also be indifferent between M' and I'. Since we used the same value p to construct the new compound lotteries, the indifference line I'M' must be parallel to the line IM by similar triangles.



2 Roulette wheel with N outcomes

(a)
$$U(W) = \log(W), \sum_{n} a_n = 1.$$

$$U(W) = \sum_{n} p_{i} \log(a_{i}),$$

$$L = \sum_{n} p_{i} \log(a_{i}) - \lambda \left(\sum_{n} a_{i} - 1\right)$$

$$\frac{\partial L}{\partial a_{i}} = \frac{p_{i}}{a_{i}} - \lambda = 0$$

$$\implies a_{i} = \frac{p_{i}}{\lambda}$$
Set $\lambda = 1 \implies a_{i} = p_{i}$

So the prizes should be set to have the same payout as the probabilities. Containty equivalent, $\sum_{n} p_n \log(n)$

Certainty equivalent:
$$\sum_{n} p_i \log(p_i)$$

(b) $U(W) = \frac{1}{1-\gamma} W^{1-\gamma}, 0 < \gamma$

$$U(W) = \frac{1}{1 - \gamma} \left(\sum_{n} p_{i} a_{i} \right)^{1 - \gamma}$$

$$L = \frac{1}{1 - \gamma} \left(\sum_{n} p_{i} a_{i} \right)^{1 - \gamma} - \lambda \left(\sum_{n} a_{i} - 1 \right)$$

$$\frac{\partial L}{\partial a_{i}} = p_{i} (a_{i} p_{i})^{-\gamma} - \lambda = 0$$

$$\implies a_{i} = \frac{p_{i}^{1/\gamma - 1}}{\lambda^{1/\gamma}}$$

So the prizes should be set to have payouts proportional to the probabilities to the power of $1/\gamma - 1$. Certainty equivalent: $\frac{1}{1-\gamma}(\sum_n p_i^2)^{1-\gamma}$

(c)
$$U(W) = -e^{-bW}$$

$$U(W) = \sum_{n} -e^{-bp_{i}a_{i}}$$

$$L = \sum_{n} \left(-e^{-bp_{i}a_{i}}\right) - \lambda \left(\sum_{n} a_{i} - 1\right)$$

$$\frac{\partial L}{\partial a_{i}} = a_{i}p_{i}be^{-bp_{i}a_{i}} - \lambda = 0$$

$$\frac{\partial L}{\partial a_{j}} = a_{j}p_{j}be^{-bp_{j}a_{j}} - \lambda = 0$$

$$\implies \frac{a_{i}}{a_{j}} = \frac{p_{i}e^{-bp_{i}a_{i}}}{p_{j}e^{-bp_{j}a_{j}}}$$

so the prizes should be proportional to the probabilities.

Certainty equivalent: $\sum_{n}^{\infty} -e^{-bp_i^2}$

3 Two lotteries

(a) $U(W) = -\frac{1}{W}$

Expected utility from lottery 1:

$$E[U(W)] = 0.5U(10) + 0.5U(50) = 0.5\left(-\frac{1}{10}\right) + 0.5\left(-\frac{1}{50}\right) = -0.06$$

Expected utility from lottery 2:

$$E[U(W)] = U(X) = -\frac{1}{X} = -0.06 \implies X = \frac{50}{3}$$

(b) $U(W) = \log(W)$

Expected utility from lottery 1:

$$E[U(W)] = 0.5U(10) + 0.5U(50) = 0.5 \log 10 + 0.5 \log 50 \approx 3.107$$

Expected utility from lottery 2:

$$E[U(W)] = U(X) = \log X = 1.349 \implies X = e^{3.107} \approx 22.35$$

(c) $U(W) = \frac{1}{1-\gamma}W^{1-\gamma}, \gamma = 0.25; 0.75$ Expected utility from lottery 1:

$$E[U(W)] = 0.5U(10) + 0.5U(50) = 0.5 \frac{1}{1 - \gamma} 10^{1 - \gamma} + 0.5 \frac{1}{1 - \gamma} 50^{1 - \gamma} \approx 16.28, \gamma = 0.25; 8.87, \gamma = 0.75$$

Expected utility from lottery 2:

$$E[U(W)] = U(X) = \frac{1}{1 - \gamma} X^{1 - \gamma} = EU(\text{lottery 1}) \implies X = 28.12, \gamma = 0.25; 24.23, \gamma = 0.75$$

Increasing γ means that the certainty equivalent decreases which implies that higher γ is equivalent to higher risk aversion.

4 Proof of Theorem

Theorem: Suppose that utility functions u_1 and u_2 are twice continuously differentiable with continuous second derivatives and strictly increasing. The following are equivalent: 1) $A_1(y) \geq A_2(y)$ for every y. 2) $\pi_1(y,\tilde{z}) \geq \pi_2(y,\tilde{z})$ for every y and every random variable \tilde{z} . 3) u_1 is a concave transformation of u_2 ; that is, $u_1 = f \circ u_2$ for f concave and strictly increasing. 4) u_1 is at least as risk averse as u_2 .

 $1 \Longrightarrow 3$ Suppose $A_1(y) \ge A_2(y)$ for every y. Suppose $u_1 = f \circ u_2 = u_1(u_2^{-1}(t))$. Then $\frac{d}{dt}u_1(u_2^{-1}(t)) = \frac{u_1'(u_2^{-1}(t))}{u_2'(u_2^{-1}(t))}$ which is decreasing (concave) iff $\log u_1'(x)/u_2'(x)$ is, which follows from $A_1(y) \ge A_2(y)$.

$$3 \implies 2$$

Suppose $u_1 = f \circ u_2$ for f concave and strictly increasing. $\pi_1(y,\tilde{z}) = y + E(\tilde{z}) - u_i^{-1}(E[u_i(y+\tilde{z})])$. Then

$$\pi_1(y,\tilde{z}) - \pi_2(y,\tilde{z}) = u_2^{-1}(E[u_2(y+\tilde{z})]) - u_1^{-1}(E[u_1(y+\tilde{z})])$$

= $u_2^{-1}(E[u_2(y+\tilde{z})]) - u_1^{-1}(E[u_1(u_2^{-1}(u_2(y+\tilde{z})))])$

Since $u_1(u_2^{-1}(\cdot))$ is concave, $E[u_1(u_2^{-1}(u_2(y+\tilde{z})))] \le u_1(u_2^{-1}(E[u_2(y+\tilde{z})]))$ so $\pi_1(y,\tilde{z}) \ge \pi_2(y,\tilde{z})$. $2 \implies 1$ (actually not $1 \implies \text{not } 2$)

If 1 does not hold, then 1 holds with A_1 and A_2 interchanged. Then 3 and 2 also hold with A_1 and A_2 interchanged, and therefore do not hold in the correct ordering. This completes the proof.

5 Mean variance does not satisfy independence

Suppose I have two lotteries L_1 and L_2 . L_1 has high mean but also high variance, while L_2 has high mean and smaller variance, so L_2 is preferred to L_1 . Then consider L_3 which has a high variance and is negatively correlated with L_1 but positively correlated with L_2 . Then the combination lottery of L_1 and L_3 will be preferred to the combination of L_2 and L_3 , which violates the independence axiom.