Theory of Risky Choice

Joseph Tao-yi Wang 2019/10/9

(Lecture 10, Micro Theory I)

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Theory of Risky Choice

- We analyzed preferences, utility and choices
- Apply them to study risk and uncertainty – Preference for probabilities
 - Expected Utility
- Discuss Experimental Anomalies
- 1. Allais paradox and Ellsberg paradox
- 2. Bayes' Rule paradoxes: Soft vs. Hard prob., Game show paradox (Monty Hall problem)
- 3. Rabin paradox

States and Probabilities

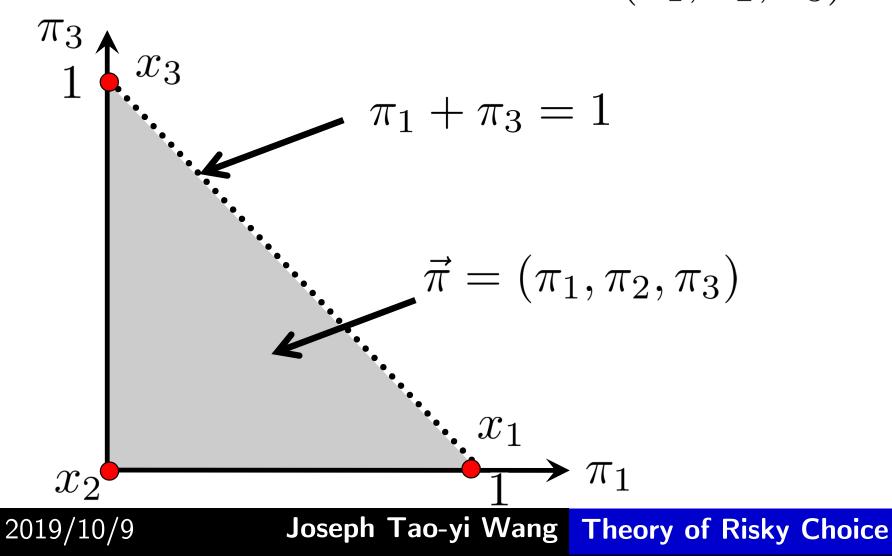
- Consequence x_s happens in state $s = 1, \cdots, S$
- Assign (subjective) probability π_s to state s
- A prospect $(\vec{\pi}; \vec{x}) = ((\pi_1, \cdots, \pi_S); (x_1, \cdots, x_S)))$ - People have preferences for these prospects
- Under the Axioms of Consumer Choice, exists continuous $U(\vec{\pi}; \vec{x})$ representing these pref.
- If we fix consequences; focus on probabilities

$$U(\vec{\pi}; \vec{x}) = U(\vec{\pi}) = U(\pi_1, \pi_2, \pi_3)$$

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States and Probabilities

• Assume $x_3 \succ x_2 \succ x_1$, can show all possible probabilities on 2D: $\vec{\pi} = (\pi_1, \pi_2, \pi_3)$



Compound Prospect (Compound Lottery)

• If I offer you $\vec{\pi}^1 = (\pi_1^1, \pi_2^1, \pi_3^1)$ with prob. p_1 , and $\vec{\pi}^2 = (\pi_1^2, \pi_2^2, \pi_3^2)$ with probability $p_2 = 1 - p_1$ π_3 Compound Prospect: $\vec{\pi}^c = (p_1, p_2 : \vec{\pi}^1, \vec{\pi}^2)$ $= p_1 \vec{\pi}^1 + (1 - p_1) \vec{\pi}^2$ π_1

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Are Indifference Curves Linear?

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- If you are indifferent between $\vec{\pi}^1 \text{and} \ \vec{\pi}^2$
- How would you feel about randomizing them? $\pi_3 \uparrow$ Indifferent !!

 $\vec{\pi}^{1} \sim \vec{\pi}^{2} \Rightarrow$ $\vec{\pi}^{1} \sim (p_{1}, 1 - p_{1} : \vec{\pi}^{1}, \vec{\pi}^{2})$ Indifference Curves $\vec{\pi}^{1} \quad \mathbf{Are \ Linear!}$

When Are Indifference Curves Parallel?

- Consider a third prospect \vec{r}

 $\pi_3^{\bullet} \operatorname{\mathsf{For}} \vec{q^1} = (1 - \lambda, \lambda : \vec{\pi}^1, \vec{r}), \ \vec{q}^2 = (1 - \lambda, \lambda : \vec{\pi}^2, \vec{r})$ Then, $\vec{\pi}^1 \sim \vec{\pi}^2 \Rightarrow \vec{q}^1 \sim \vec{q}^2$ $\vec{\pi}^1 \succeq \vec{\pi}^2 \Rightarrow \vec{q}^1 \succeq \vec{q}^2$ (if preferences are independent of irrelevant alternatives) **Parallel Indifference Curves!** π_1

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Independence Axiom(s)

• (IA) If $\vec{\pi}^1 \succeq \vec{\pi}^2$, then for any prospect \vec{r} and probabilities $p_1, p_2 > 0, p_1 + p_2 = 1$

$$\vec{q}^1 = (p_1, p_2 : \vec{\pi}^1, \vec{r}) \succeq (p_1, p_2 : \vec{\pi}^2, \vec{r}) = \vec{q}^2$$

• (IA') If $\vec{\pi}^m \succeq \vec{\mu}^m, m = 1, \cdots, M$, then for any probability vector $p = (p_1, \cdots, p_M)$ $(p_1, \cdots, p_M : \vec{\pi}^1, \cdots, \vec{\pi}^M)$ $\succeq (p_1, \cdots, p_M : \vec{\mu}^1, \cdots, \vec{\mu}^M)$

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Expected Utility

- For any prospect $\vec{\pi}$, consider (on $\pi_1 + \pi_3 = 1$):
- Extreme lottery $(1 v(\vec{\pi}), 0, v(\vec{\pi})) \sim \vec{\pi}$ π_3 $x_3 = v(x_3) = 1$ Can use $v(\vec{\pi})$ to represent pref.!! $(1 - v(\vec{\pi}), v(\vec{\pi}) : x_1, x_3) \Rightarrow v(\vec{\pi})$ $\Rightarrow v(x_2) \in (0,1)$ $v(x_1) = 0$ π_1 Joseph Tao-yi Wang Theory of Risky Choice 2019/10/9

Expected Utility

- In general, for any prospect $\vec{p} = (p_1, \cdots, p_S)$
- The consumer is indifferent between \vec{p} and playing the extreme lottery

$$\sum_{s=1}^{S} p_s v(x_s) | 0, \cdots, 0, 1 - \sum_{s=1}^{S} p_s v(x_s) \right)$$

- Hence, we can represent her preferences with the above expected win probabilities
 - Expected Utility!!

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Expected Utility Rule

- Assume (IA'), then
- Preferences over prospects

$$(\vec{p};\vec{x}) = (p_1,\cdots,p_S;x_1,\cdots,x_S)$$

 Can be represented by the Von Neumann-Morgenstern utility function

$$u(\vec{p}, \vec{x}) = \sum_{s=1}^{S} p_s v(x_s)$$

• Proof:

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Expected Utility Rule

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- Proof: S consequences, best is x^* , worse is x_*
- Can assign probability for extreme lotteries: $\vec{e}^s \equiv (v(x_s), 1 - v(x_s) : x^*, x_*) \sim x_s$
- (IA') implies $(\vec{p}; \vec{x}) \sim (p_1, \cdots, p_S : \vec{e}^1, \cdots, \vec{e}^S)$ $\sim (u(\vec{p}, \vec{x}), 1 - u(\vec{p}, \vec{x}) : x^*, x_*)$ where $u(\vec{p}, \vec{x}) = \sum_{s=1}^{S} p_s v(x_s)$ - (by reducing compound prospects)

Experimental Anomalies

- Allais Paradox
- Ellsberg Paradox
- Bayes' Rule Paradoxes
 - Soft vs. Hard Probabilities
 - Game Show Paradox
- Rabin Paradox

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Allais Paradox

- Consider four prospects:
- A. \$1 million for sure
- B. 90% chance \$5 million (& 10% chance zero)
 Among A and B, you choose...
- C. 10% chance \$1 million (& 90% chance zero)
- D. 9% chance \$5 million (& 91% chance zero)
 - Among C and D, you choose...
- Is this consistent with Expected Utility???

Allais Paradox * 1,000

A. \$1 billion for sure

- B. 90% chance \$5 billion (& 10% chance zero)
 - Among A and B, you choose...
- C. 10% chance \$1 billion (& 90% chance zero)
- D. 9% chance \$5 billion (& 91% chance zero)
 - Among C and D, you choose...
- Are your answers (still) consistent with Expected Utility? Why or why not?

Ellsberg Paradox

- One urn: 30 Black balls, and 60 "other balls"
 Other balls could be either Red or Green
- 1. One ball is drawn. You win \$100 if the ball is (a) Black or (b) Green. You pick...?
- Now you win \$50 if the ball is "either Red or another color you choose." Would you choose (a) Black or (b) Green?
- What did you choose? Did it violate EU?

Ellsberg Paradox

- One ball is drawn. You win \$100 if the ball is
 (a) Black or (b) Green.
- Picking Black = Believe <30 Green balls
- 2. Now you win if "either Red or another color." You choose (a) Black or (b) Green?
- Picking Green = Believe >30 Green balls
- Since it is the same urn, this is inconsistent!
 - Can this be due to hedging (risk aversion)?
 - Maybe, but can fix this by paying only 1 round...

Bayes' Rule Paradoxes: Soft vs. Hard Prob.

- Two urns, each contain 100 balls.
- 1. Urn 1 has 60 Yellow balls.
- 2. Urn 2 has 75 or 25 Yellow balls with equal chance.
- You win a prize if you draw a Yellow ball.
- A ball is drawn from Urn 2 and it is Yellow.
- Which Urn should you choose?
- Did you do Bayesian updating correctly?

Bayes' Rule Paradoxes: Soft vs. Hard Prob.

- Prior to draw, Pr(draw a Y) = 0.5. After:
- $Pr(Y \mid 75 Y) = 0.75, Pr(Y \mid 25 Y) = 0.25$
- Pr(75 Y | Y) = $0.5 \times 0.75 / 0.5 = 0.75$
- Pr(25 Y | Y) = 1 0.75 = 0.25
- Pr(draw another Y |Y) = Pr(75 - Y |Y) × Pr(Y | 75 - Y) + Pr(25 - Y |Y) × Pr(Y | 25 - Y) = $0.75 \times 0.75 + 0.25 \times 0.25 = 0.625 > 0.6$
- So you should pick Urn 2!! (Did you do that?)

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Bayes' Rule Paradoxes: Game Show Paradox



One door hides the prize (a car). Remaining two doors hides a goat (non-prize). Suppose you choose door number 1...

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Game Show Paradox (Monty Hall Problem)



Door 3 is opened for you... Obviously the car is not behind door 3... Would you want to switch to door 2?

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Depends on how door is opened...

• Rule to open one door:

The Host must open one "other" door without the prize. If he has a choice between more than one door, he will randomly open one of the possible (goat) doors.

• The Game Show Paradox is also known as the Monty Hall Problem, named after the name of the TV show host "Monty Hall"

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If You Ricked the Right Door (33.3%)



Host randomizes between door 2 and 3 If host opens door 3... (Prob=33.3%*50%) You should not switch (but you don't know)

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If You Ricked the Wrong Door (66.7%)



Host cannot open door 2 (contains car) See host opening door 3... (Prob.=66.70%*100%) You should switch (but you don't know)

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Bayesian Updating in the Monty Hall Problem

- Pr(host opens Door 3 & car in Door 1) = 33.3%*50% = 16.7%
- Pr(host opens Door 3 & car in Door 2) = 33.3%*100% = 33.3%
- Pr(host opens Door 3 & car in Door 3) = 0
 Host never opens Door 3!
- Pr (car in Door 1 | Host opens Door 3)
 = [16.7%] / [16.7% + 33.3% + 0%] = 1/3

Game Show Paradox Plus: Modified Monty Hall



One door hides the prize (a car). Remaining two doors hides a goat (non-prize). Door 3 is transparent (and you see the goat) Suppose you choose door number 1...

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Game Show Paradox Plus: Modified Monty Hall



Door 3 is opened for you... Obviously the car is not behind door 3 (and you knew that already)... Would you want to switch to door 2?

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If You Ricked the Right Door (50%)



Host randomizes between door 2 and 3 (50-50) If host opens door 2... (Prob.=50%*50%) You should definitely not switch!

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If You Ricked the Right Door (50%)



Host randomizes between door 2 and 3 If host opens door 3... (Prob=50%*50%) You should still not switch (but you don't know)

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If You Ricked the Wrong Door (50%)

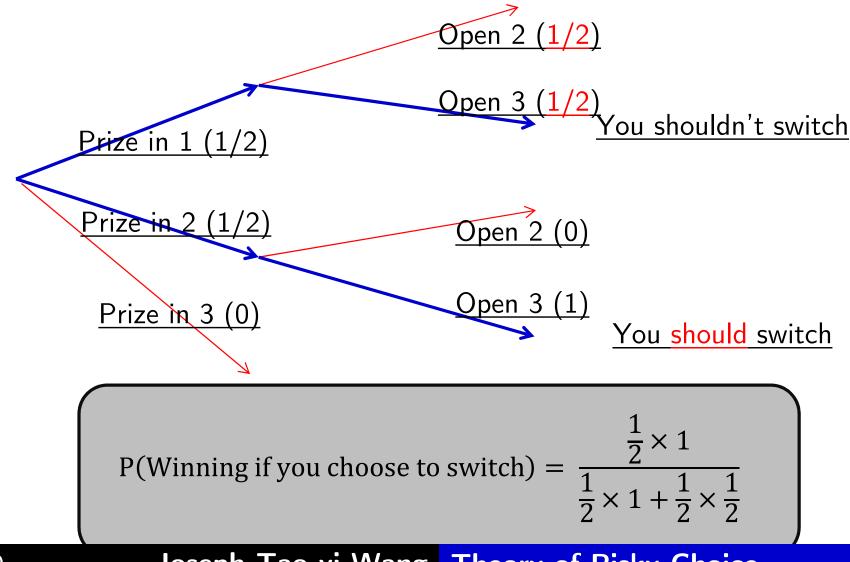


Host cannot open door 2 (contains car) See host opening door 3... (Prob.=50%*100%) You should switch (but you don't know)

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Bayesian Solution (Monty Hall Plus)

Door #3 is transparent...



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Rabin Paradox: Which Cells Will You Accept?

Payoff if Green Ball	Nur	nber (ou	Payoff if Red Ball			
100	52	55	60	66	70	-100
1000	13	20	33	46	57	-100
5000	7	18	33	46	57	-100
25000	7	18	33	46	57	-100

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- Suppose your risk preference follows EU.
- For initial Wealth is ω
- Consider the prospect $(p, 1 p : \omega + g, \omega g)$
- If you reject this lottery, this implies: $v(\omega) \ge (1-p) \cdot v(\omega-g) + p \cdot v(\omega+g)$

• Or, $\begin{bmatrix} v(\omega+g) - v(\omega) \end{bmatrix} \leq \frac{1-p}{p} \cdot \begin{bmatrix} v(\omega) - v(\omega-g) \end{bmatrix}$(1)

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- Now consider initial wealth $\omega' = \omega + g$
- If you reject the prospect $(p, 1 p : \omega' + g, \omega' g)$
- Then: $v(\omega') \ge (1-p) \cdot v(\omega'-g) + p \cdot v(\omega'+g)$

• Or,

$$\begin{bmatrix} v(\omega + 2g) - v(\omega + g) \end{bmatrix} = \begin{bmatrix} v(\omega' + g) - v(\omega') \end{bmatrix}$$

$$\leq \frac{1 - p}{p} \cdot \begin{bmatrix} v(\omega') - v(\omega' - g) \end{bmatrix}$$

$$= \frac{1 - p}{p} \cdot \begin{bmatrix} v(\omega + g) - v(\omega) \end{bmatrix}$$

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• Combining the two inequalities:

$$\begin{bmatrix} v(\omega+2g) - v(\omega+g) \end{bmatrix} \\ \leq \frac{1-p}{p} \cdot \left[v(\omega+g) - v(\omega) \right] \\ \leq \left(\frac{1-p}{p}\right)^2 \cdot \left[v(\omega) - v(\omega-g) \right] \dots (2)$$

- Only required one to reject the fair gamble at both wealth levels ω and $\omega'=\omega+g$

- Suppose you reject the fair gamble at all wealth levels between ω and $\omega^{(n)}=\omega+ng$

• Then,

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$$\begin{bmatrix} v(\omega + ng) - v(\omega + (n-1)g) \end{bmatrix}$$

$$\leq \frac{1-p}{p} \cdot \left[v(\omega + (n-1)g) - v(\omega + (n-2)g) \right]$$

$$\leq \dots \leq \left(\frac{1-p}{p}\right)^n \cdot \left[v(\omega) - v(\omega - g) \right] \dots \dots (n)$$

• Summing (1) through (n):

 $\left[v(\omega+q) - v(\omega)\right] + \left[v(\omega+2g) - v(\omega+g)\right]$ $+ \cdot + \left[v(\omega + ng) - v(\omega + (n-1)g) \right]$ $= \left[v(\omega + ng) - v(\omega) \right]$ $\leq \left[\frac{1-p}{p} + \dots + \left(\frac{1-p}{p}\right)^n\right] \cdot \left[v(\omega) - v(\omega - g)\right] =$ $\frac{1-p}{p} \cdot \left[v(\omega) - v(\omega - g) \right] + \left(\frac{1-p}{p} \right)^2 \cdot \left[v(\omega) - v(\omega - g) \right]$ $+\cdots + \left(\frac{1-p}{p}\right)^n \cdot \left[v(\omega) - v(\omega - g)\right]$

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Let
$$s(n,p) = 1 + \frac{1-p}{p} + \dots + \left(\frac{1-p}{p}\right)^n$$

 $\begin{bmatrix} v(\omega + ng) - v(\omega) \end{bmatrix}$
 $\leq \begin{bmatrix} v(\omega) - v(\omega - g) \end{bmatrix} \cdot \begin{bmatrix} \frac{1-p}{p} + \dots + \left(\frac{1-p}{p}\right)^n \end{bmatrix}$
 $\Rightarrow \begin{bmatrix} v(\omega + ng) + (s(n,p) - 1)v(\omega - g) \end{bmatrix} \leq s(n,p) \cdot \underline{v(\omega)}$

- Or, $v(\omega) \ge \frac{1}{s(n,p)}v(\omega+ng) + (1-\frac{1}{s(n,p)})v(\omega-g)$
- This means rejecting

$$(\frac{1}{s(n,p)}, 1 - \frac{1}{s(n,p)} : \omega + ng, \omega - g)$$

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- We have shown that:
- If you reject prospect $(p, 1 p : \omega + g, \omega g)$
- For all wealth levels $[\omega, \omega + ng]$
- →You would also reject the more favorable prospect (¹/_{s(n,p)}, 1 ¹/_{s(n,p)} : ω + ng, ω g)
 s(n,p) = 1 + ^{1-p}/_p + ··· + (^{1-p}/_p)ⁿ → ¹/<sub>1 ^{1-p}/_p
 This is true for any large n! ¹/_{s(n,p)} → ^{2p-1}/_∞
 </sub>

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Rabin Paradox: Which Cells Will You Accept?

Payoff if Green Ball	Nur	nber (ou	Payoff if Red Ball			
100	5 2	55	X	66	70	-100
1000	X	20	33	46	57	-100
5000	X	×	33	46	57	-100
25000	X)	33	46	57	-100

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Continuous Probability Distribution

• Let state $s \in \mathcal{S} = [\alpha, \beta]$

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- CDF is $F(t) = \Pr\{s \le t\}$ $F(\alpha) = 0, F(\beta) = 1$
- Probability of being in C = [s, s'] is:
- Probability Measure $\pi(C) = F(s') F(s)$
- Can generalize and assign probability measures over closed convex hypercube $C \in \mathbb{R}^n$

Support of the Continuous Distribution

- x is in the support of the distribution if for every neighborhood $N(x, \delta)$ of x, $\pi(N(x, \delta)) > 0$
- Example: $\mathcal{S} = [0, 3]$

$$F(\theta) = \begin{cases} \frac{1}{2}\theta, & 0 \le \theta \le 1\\ \frac{1}{2}, & 1 < \theta < 2\\ \frac{1}{2}(\theta - 1), & 2 \le \theta \le 3 \end{cases}$$

• What is the support?

 $[0,1] \cup [2,3]$

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Summary of 7.1

- Preferences over prospects
- Indifference Curves
 - Linear: "Reduction of Compound Lotteries"
 - Parallel: "Independent of Irrelevant Alternatives"
- Expected Utility
- Anomalies: Allais paradox, Ellsberg paradox, Bayes' Rule paradoxes (Soft vs. Hard prob. and Monty Hall Problem) and Rabin paradox
- Continuous State Space
- Homework: Exercise 7.1-4 (Optional: 7.1-3)

In-Class Homework: Exercise 7.1-1 IA ⇔ IA'

a) For M = 2, show that IA implies IA' -(IA) If $\pi^1 \succeq \pi^2$, then for any prospect r and probabilities $p_1, p_2 > 0, p_1 + p_2 = 1$ $q^1 = (p_1, p_2 : \pi^1, r) \succeq (p_1, p_2 : \pi^2, r) = q^2$ -(IA') If $\pi^m \succeq \hat{\pi}^m, m = 1, \cdots, M$, then for any probability vector $p = (p_1, \cdots, p_M)$ $(p:\pi^1,\cdots,\pi^M) \succeq (p:\hat{\pi}^1,\cdots,\hat{\pi}^M)$ b) Show that if the proposition holds for M =

k-1, then it must also hold for M = k.

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In-Class Homework: Exercise 7.1-2 Allais

- A. \$1 million for sure -(0, 1, 0)
- B. 90% chance \$5 million (0.90, 0, 0.10)
- C. 10% chance \$1 million -(0, 0.10, 0.90)
- D. 9% chance \$5 million -(0.09, 0, 0.91)
- 1. Draw tree diagrams showing that C and D can be represented as compound gambles between A and B, respectively, and (0,0,1), where the probability of (0,0,1) is the same.
- 2. Show that the ranking of A and B should be the same as the ranking of C and D.