ECON 4360: Empirical Finance

Risk-Neutral Probabilities, Risk-Sharing, State Diagrams

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Theory Lecture #08

What are we doing today?

- Risk-Neutral Probabilities
- Consumers and Risk-Sharing
- State Diagrams

• Now, we're going to define a new type of probability - a *risk-neutral* probability, $\pi^*\left(s\right)$

$$\pi^{*}\left(s\right)=R^{f}m\left(s\right)\pi\left(s\right)=R^{f}\phi\left(s\right)$$

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 - (Recall, $R^f = 1/E(m) = 1/\sum \phi(s)$)

- These probabilities $\pi^*(s)$ are called risk-neutral because we can use them to take the expected value of a payoff and discount at the risk-free rate to get prices.
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- Why would we want to do this?
 - It allows us to think of asset pricing as if people are risk-neutral, but with probabilities $\pi^*(s)$ in place of $\pi(s)$...

Theory Lecture #08

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 - (What does this mean in terms of consumption?)
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- So with $\pi^*(s)$, we pay attention to states that are highly likely to occur (large $\pi(s)$), or may not be likely to occur, but may have disastrous consequences if they do occur (large m(s)).

Exercise 3: Risk-Neutral Probabilities

• Given the state-prices and corresponding probabilities, what are the risk-neutral probabilities $\pi^*(s)$?

State (s)	$\phi(s)$	$\pi(s)$
1	0.2	0.25
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So

$$\pi^* (1) = (1.11) (0.2) = 0.222$$

$$\pi^*(2) = (1.11)(0.4) = 0.444$$

$$\pi^*(3) = (1.11)(0.3) = 0.333$$

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Bringing the Consumer Back In

- Even though we don't need utility functions in the contingent claims context, we would like to see how this matches up with our consumer-investor's FOCs.
- Let's look again at the definition of the SDF based on marginal utility

$$m(s) = \beta \frac{u'(c(s))}{u'(c)}$$

- Here, s represents some future state, and m(s) is the SDF a random variable whose value depends on the realization of state s
 - Note that this equation has to hold for any future state.
- Note that we can write

$$m(s) = \beta \frac{u'(c(s))}{u'(c)} = \frac{\phi(s)}{\pi(s)}$$



Bringing the Consumer Back In

Consider two different possible future states

$$m\left(\mathbf{s}_{1}
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$$\frac{m(s_1)}{m(s_2)} = \frac{u'(c(s_1))}{u'(c(s_2))}$$

• Recall that $m\left(s\right)=\phi\left(s\right)/\pi\left(s\right)$, so we can write the previous equation as

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• Thinking in terms of contingent claims, what this says is that the price of giving up a unit of consumption in state 1 for an additional unit in state 2, $\left(\frac{\phi(s_1)}{\phi(s_2)}\right)$, has to equal the ratio of expected happiness lost in state 1 to expected happiness gained in state 2.

• Suppose $\pi\left(s_1\right)=0.2$, $\pi\left(s_2\right)=0.4$, $u'\left(c\left(s_1\right)\right)=0.1$, $u'\left(c\left(s_2\right)\right)=0.2$, $\phi_1=0.1$, and $\phi_2=0.8$. Does the first-order condition hold?

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• We get that

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- So we see that *m* can give us the MRS between both *date* and *state* contingent claims

Risk-Sharing

- If markets are complete, it is possible to insure against any state by using contingent claims. Why? And what does that get us?
 - In complete contingent claims markets, all investors share all risks...
 - We just found that the MRS (for anyone) equals the contingent claims price ratio.
 - Since prices are the same for everyone, we find that marginal utility growth should be the same for everyone.

$$m\left(s_{t+1}\right) = \frac{\phi\left(s_{t+1}\right)}{\pi\left(s_{t+1}\right)} = \beta \frac{u'\left(c_{t+1}^{i}\right)}{u'\left(c_{t}^{i}\right)} = \beta \frac{u'\left(c_{t+1}^{j}\right)}{u'\left(c_{t}^{j}\right)}$$

- What does this say? What doesn't this say?
- In reality, markets are not complete. Keep in mind, though, that a big part of financial innovation is to bring products to market that better enable people to share risks.

- Now, we're going to talk a bit about state-space geometry.
- Random variables can be represented by vectors, with each element representing a different possible outcome.
 - We can think of contingent claims prices and asset payoffs as vectors in \mathbb{R}^S , where S is the total number of states.
 - Let's work in R² for now...
 - Ex: A payoff of \$7 in state 1 and \$3 in state 2 can be represented by the following vector in \mathbb{R}^2

$$x = \left[\begin{array}{c} 7 \\ 3 \end{array} \right]$$

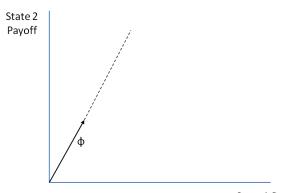
• Ex: Contingent claims prices can be represented by the following

$$\phi = \left[\begin{array}{c} 0.2 \\ 0.4 \end{array} \right]$$



• Note that the contingent claims vector always points in the positive orthant, since marginal utility is always non-negative:

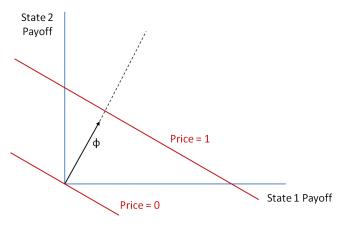
$$\phi(s) = m(s)\pi(s)$$



State 1 Payoff

- Think about the equation $p(x) = \Sigma \phi(s) x(s)$. If we interpret ϕ and x as vectors, we can interpret price as the inner product of the contingent claims prices and the payoffs.
- Recall, that two orthogonal vectors (vectors points out from the origin at right angles) have an inner product of zero.
- Where does the set of all zero-price payoffs lie?

- The plane of price = 0 payoffs is the plane of excess returns.
- Similarly, the plane of price = 1 payoffs is the plane of returns.



Since we can write the price of any risky payoff as an inner product,

$$p(x) = \sum_{s} \phi(s) x(s)$$

$$= \phi \cdot x$$

$$= |\phi| \times |proj(x|\phi)|$$

$$|\phi| \times |x| \times \cos(\theta)$$

- This gives us the result that the set of all payoffs with the same price lie in a plane that is perpendicular to the contingent claims vector
- We also get the result that p(x) is a linear pricing function:

$$p(ax + by) = ap(x) + bp(y)$$

since, e.g.,

$$p(2x) = \Sigma_s \phi(s) 2x(s) = 2p(x)$$

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- What about a state-contingent claim to a 1 unit payoff in state 1?
 - ullet At the intersection of the horizontal axis and the price $=\phi_1$ plane.

End of Today's Lecture.

• That's all for today. Today's material corresponds roughly to Chapter 3 in Cochrane (2005).