# 1 Equivalent Representations

#### 1.1 Intro

- So far in this course, we have seen three basic representations of asset pricing models:
  - The  $\beta$  Representation: The expected return of any asset can be expressed as  $E(R^i) = \gamma + \beta_{i,m} \lambda_m$

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- Mean-Variance Frontier: If  $R^{mv}$  is on the mean-variance frontier, then  $m = a + R^{mv}$  can price all assets.
- Discount Factor Representation: Given m such that p = E(mx), then  $m, x^*, R^*$ , or  $R^* + wR^{e*}$  can be used to price all assets
- What we're going to show now is that these representations are "equivalent"
- Credit for the connections:
  - Roll (1976): Between the mean-variance frontier and beta pricing (the CAPM)
  - Ross (1978) and Dybvig and Ingersoll (1982): Between linear discount factors and beta pricing
  - Hansen and Richard (1987): Between a discount factor and the mean-variance frontier

#### 1.2 Discount Factors to Betas

• From Discount Factors to Betas:  $m, x^*$ , and  $R^*$  can all be the single factor in a single-beta representation

#### 1.2.1 Using m

- Given  $m, p = E\left(mx\right)$  implies  $E\left(R^{i}\right) = \gamma + \beta_{i,m}\lambda_{m}$ 
  - To see this...
  - Start with

$$1 = E\left(mR^{i}\right) = E\left(m\right)E\left(R^{i}\right) + cov\left(m, R^{i}\right)$$

- So

$$E\left(R^{i}\right) = \frac{1}{E\left(m\right)} - \frac{cov\left(m, R^{i}\right)}{E\left(m\right)}$$

- Now, define  $\gamma = 1/E(m)$ , so

$$\begin{split} E\left(R^{i}\right) &= \gamma + \left(\frac{cov\left(m, R^{i}\right)}{var\left(m\right)}\right) \left(-\frac{var\left(m\right)}{E\left(m\right)}\right) \\ &= \gamma + \beta_{i,m}\lambda_{m} \end{split}$$

- What this says is that expected returns should be linear in the regression betas of asset returns on m.
  - \* For example, m could be  $(c_{t+1}/c_t)^{-\gamma}$ , where  $\gamma$  is risk aversion.

#### 1.2.2 Using $x^*$

- Given  $x^*$ ,  $p = E(x^*x)$  implies  $E(R^i) = \gamma + \beta_{i,x^*}\lambda_{x^*}$ 
  - To see this...
  - Recall that we can write the price of an asset in terms of the mimicking portfolio,  $x^*$  as  $p = E(x^*x)$
  - Therefore,

$$1 = E(mR^{i}) = E(x^{*}R^{i}) = E(x^{*}) E(R^{i}) + cov(x^{*}, R^{i})$$

- So as before, we can use  $x^*$  instead of m and write

$$E(R^{i}) = \gamma + \left(\frac{cov(x^{*}, R^{i})}{var(x^{*})}\right) \left(-\frac{var(x^{*})}{E(x^{*})}\right)$$
$$= \gamma + \beta_{i,x^{*}}\lambda_{x^{*}}$$

- Note: the term  $1/E(x^*)$  is the zero-beta rate and applies when there is no riskfree asset.

### 1.2.3 Using $R^*$

- Given  $R^*$ ,  $p = E\left(mx\right)$  implies  $E\left(R^i\right) = \gamma + \beta_{i,R^*}\lambda_{R^*}$ 
  - To see this...
  - Recall the definition of  $R^*$  is

$$R^* = \frac{x^*}{E\left(x^{*2}\right)}$$

- Multiply both sides by  $R^*$  and take expectations

$$E\left(R^{*2}\right) = \frac{E\left(R^{*}x^{*}\right)}{E\left(x^{*2}\right)}$$

- And since the price of  $R^*$  is one,  $1 = E(R^*x^*)$  and

$$E\left(R^{*2}\right) = \frac{1}{E\left(x^{*2}\right)}$$

- Now we can re-arrange and combine this last expression with the first to get

$$x^* = R^* E(x^{*2}) = \frac{R^*}{E(R^{*2})}$$

- Now, we can substitute this expression for  $x^*$  into

$$E\left(R^{i}\right) = \gamma + \left(\frac{cov\left(x^{*}, R^{i}\right)}{var\left(x^{*}\right)}\right) \left(-\frac{var\left(x^{*}\right)}{E\left(x^{*}\right)}\right)$$

to get

$$E(R^{i}) = \gamma + \left(\frac{cov(R^{*}, R^{i})}{var(R^{*})}\right) \left(-\frac{var(R^{*})}{E(R^{*})}\right)$$
$$= \gamma + \beta_{i,R^{*}} \lambda_{R^{*}}$$

- Now this expression

$$E\left(R^{i}\right) = \gamma + \beta_{i,R^{*}} \lambda_{R^{*}}$$

has to hold for any  $R^i$ , including  $R^*$ , so we can write

$$E(R^*) = \gamma + \beta_{R^*,R^*} \lambda_{R^*} = \gamma + \lambda_{R^*}$$

- So the price of risk can be written as

$$\lambda_{R^*} = E\left(R^*\right) - \gamma$$

- And the return on any asset can be written as

$$E\left(R^{i}\right) = \gamma + \beta_{i,R^{*}} \left[E\left(R^{*}\right) - \gamma\right]$$

• Recall that the traditional CAPM states that

$$E\left(R^{i}\right) = R^{f} + \beta_{i,R^{m}} \left[E\left(R^{m}\right) - R^{f}\right]$$

and recall that  $\gamma = 1/E[m] = R^f$ , so we are close to the CAPM here.

- This is not quite the CAPM, since though  $R^*$  is on the mean-variance frontier, it is not the market portfolio.
  - We will see later that any return on the MV Frontier can be used in place of  $R^*$ . So if the market portfolio is on the frontier, then the CAPM will hold.
- Note that because of this, the CAPM will **only** work if the market portfolio is mean-variance efficient (this is Roll's critique).

## 1.3 From MV Frontier to m and beta

- Now, we'll go in the other direction...
- If  $R^{mv}$  is any return on the mean-variance frontier, then

$$m = a + bR^{mv}$$

will price assets via

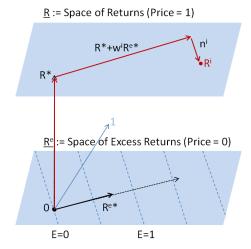
$$p = E(mx)$$

and so we can express the expected return of any asset as

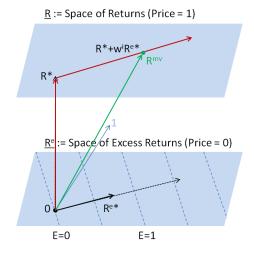
$$E(R^{i}) = \gamma + \beta_{i,R^{mv}} \left[ E(R^{mv} - \gamma) \right]$$

- Theorem: There is a discount factor of the form  $m = a + bR^{mv}$  iff  $R^{mv}$  is on the mean-variance frontier.
- Proof by pictures...

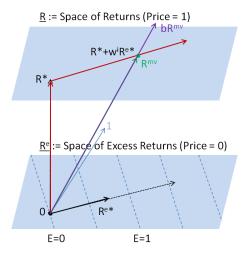
- Recall our characterization of the MV Frontier



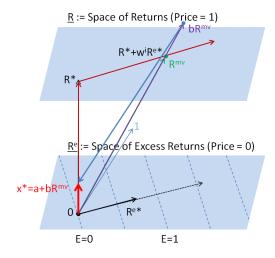
– The MV Frontier is swept out by  $R^{mv} = R^* + wR^{e*}$  as real number w varies.



- We can pick any vector  $R^{mv}$  on the frontier and then stretch it to length  $bR^{mv}$ 



- Then we can add (subtract) an amount a of the 1 vector, which is simply the constant. This gets us back to the tip of  $x^*$ 



- Therefore, we can find some a and b, such that  $x^* = a + bR^{mv}$
- Now, we can get to m by adding an orthogonal piece  $\varepsilon$  to  $x^*$ :

$$p = E(mx) = E((x^* + \varepsilon)x) = E(x^*x)$$

- Recall that there are an infinite number of m's that can price assets:  $m = x^* + \varepsilon$
- Perhaps some a + bR with R not on the MV Frontier could be a discount factor?
- Turns out this can't be the case...
- For any R, we can write  $R = R^* + wR^{e^*} + n$ 
  - \* Recall that frontier returns have n = 0...
- If n is not zero, there is no way to express  $x^*$  as a linear form a + bR.
- To see this, pick some R not on the frontier where  $n \neq 0$ .
  - \* Now, if we stretch this R to any bR, there is no way to subtract the constant (the 1 vector) to get back to  $x^*$ .
- Hence, only returns that are on the frontier can be transformed by  $a + bR^{mv}$  into a valid SDF.

#### 1.4 Factor Models and Discount Factors

- We've shown that p = E(mx) implies a single beta representation.
- Now let's go the other way. Given an expected-return beta model (e.g., CAPM), what discount factor does that imply?
- Theorem: Given the model

$$m = a + b'f, 1 = E\left(mR^i\right)$$

where b is a vector of constants and f is a vector of factors, we can find  $\gamma$  and  $\lambda$  such that

$$E\left(R^{i}\right) = \gamma + \lambda' \beta_{i}$$

where the  $\beta_i$  are multiple regression coefficients of  $R^i$  on f with a constant. Conversely, given  $\gamma$  and  $\lambda$  in a factor model, one can find a and b such that m = a + b'f.

• The most common example of this type of model is the CAPM:

$$E\left(R^{i}\right) = R^{f} + \beta_{i,R^{m}} \left[E\left(R^{m}\right) - R^{f}\right]$$

- We would first regress a time-series of stock returns on our factor (the market portfolio) to get a  $\beta_i$  for each stock. Then we would see if  $\beta$  could explain stock returns in the cross-section:  $E\left(R^i\right) = \gamma + \lambda \beta_i$
- To prove this theorem, we just have to find a relation between  $(\lambda, \gamma)$  and (a, b) and show that it works.
  - Start with m = a + b'f and  $1 = E(mR^i)$  and fold the means of f into a so that E(f) = 0.
  - We have

$$1 = E(mR^{i}) = E[(a + b'f)R^{i}] = aE(R^{i}) + b'E(fR^{i})$$

- Solve for  $E(R^i)$ :

$$E(R^{i}) = \frac{1}{a} - \frac{E(R^{i}f')b}{a}$$

– If we regress  $R^i$  on f, the fitted coefficient vector is  $\beta_i = E(ff')^{-1} E(fR^i)$ , so we can incorporate  $\beta$  now and continue with...

$$E(R^{i}) = \frac{1}{a} - \frac{E(R^{i}f')b}{a}$$

$$= \frac{1}{a} - \frac{E(R^{i}f')E(ff')^{-1}E(ff')b}{a}$$

$$= \frac{1}{a} - \left[E(R^{i}f')E(ff')^{-1}\right]\frac{E(ff')b}{a}$$

$$= \frac{1}{a} - \beta_{i}\frac{E(ff')b}{a}$$

- And define  $\gamma$  and  $\lambda$  to make it work...
- Define

$$\gamma := \frac{1}{E(m)} = \frac{1}{a}$$

$$\lambda := -\frac{1}{a}E(ff')b = -\gamma E(ff')b = -\gamma E(mf) = -\gamma p(f)$$

- So

$$E\left(R^{i}\right) = \gamma + \lambda' \beta_{i}$$

- So we can:
  - Start with the discount factor (a, b) and get the factor model  $(\lambda, \gamma)$ . For this, we have to rule out an infinite riskfree rate, E(m) = 0

- Or we can start with a factor model  $(\lambda, \gamma)$  and get the discount factor (a, b). Note that since  $a = 1/\gamma$ , we have to rule out the riskfree rate being zero  $(\gamma = 0)$ . And since  $b = -\lambda a E(ff')^{-1}$ , we have to rule out E(ff') being singular.

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– We can go from a multiple beta model to a single beta model, just use m = a + b'f as the single factor.

# 2 Implications

- So what are the implications of these equivalences?
  - We're now going to discuss several of these in turn...

#### 2.1 Ex-Ante and Ex-Post

- Roll showed that mean-variance efficiency implies a single-beta representation.
  - Problem: You can always find some mean-variance efficient return ex-post
    - \* So you can always find a single-beta representation that works.
- The proper way to test is not ex-post, but ex-ante.
- Theory should tell us what the reference portfolio should be
  - And then we should test to see if the reference portfolio is efficient
  - Example: The CAPM tells us that the reference portfolio should be the market portfolio. Therefore, the only test of the CAPM is if the market portfolio is mean-variance efficient.
- The expression p = E(mx) is just an updated re-statement of Roll's theorem.
  - -p = E(mx) always works for some m, like  $x^*$
- Key: Use theory to guide us in writing m = f(data)
- Equivalent to this, if the **sample** covariance matrix for a set of returns is non-singular, then there exists an **ex-post** mean-variance efficient portfolio where sample average returns line up exactly with sample betas.
  - Ex-post, you can always find a portfolio that makes the asset pricing model work.
  - You need theory to put restrictions on the reference portfolio (or the m). Then you can test the
    asset pricing model.
- Danger: A lot of asset pricing research proposes a set of ad hoc factors, tries them all, gets a few that work "pretty well" and then claims success, in that the model is not statistically rejected.

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- What can be done? There are two possible solutions:
  - Use economic theory to guide us on what m = f (data) should look like
  - Use out-of-sample (new time periods) and cross-sample (e.g., different countries) tests to confirm
    the results
- Problems:
  - The factors that current theory states should work well don't. We need new theories
  - What about with the second?

### 2.2 Market Efficiency

- If investors are irrational and markets are inefficient (as some people claim), is finding an asset pricing model hopeless?
- Actually... All we need is the absence of arbitrage to give us the existence of m.
  - Since consistent arbitrage opportunities do not exist, we know that there is an SDF that can be used to price payoffs.
- Markets can appear to be irrational without generating arbitrage opportunities if and only if the discount factors that generate asset prices are disconnected from MRS's.

### 2.3 Number of Factors

- Some researchers have focused on the number of factors needed to explain the cross-section of returns.
  - For example, we could have m = b'f where f is a  $5 \times 1$  vector.
- We have already shown that this is really not that meaningful...
  - We can always reduce a model with 5 factors to a model with 1 factor, m, and find betas not with respect to the factors, but with respect to m.

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#### 2.4 Historical Context

- Historically, asset pricing began by putting means on the y-axis and standard deviations on the x-axis.
  - We then described consumer preferences over means and variances by way of a utility function  $u(\mu, \sigma)$ .
  - Given this framework, the expected return  $E(R^i)$  measured the security's contribution to an overall portfolio expected return and its beta  $\beta$  measured the contribution to portfolio variance...
- From the CAPM, we know the relation between expected return and risk:

$$E(R^{i}) = R^{f} + \beta_{im} \left( E(R^{m}) - R^{f} \right)$$

But while the form above deals with expectations, the following deals with actual outcomes:

$$R^{i} = R^{f} + \beta_{im} \left( R^{m} - R^{f} \right) + \varepsilon^{i}$$

- Note that  $R^i$ ,  $R^m$ , and  $\varepsilon^i$  are the only random variables... everything else is a constant...
- Looking at the latter, we can figure out how the variance of a stock's return relates to the CAPM...
- Take the variance of both sides of the previous equation to get

$$var\left[R^{i}\right] = \beta_{im}^{2} var\left[R^{m}\right] + var\left[\varepsilon^{i}\right]$$

where

- $-var\left[R^{i}\right]$  is the variance of returns (total risk) for stock i
- $-\beta_{im}^2 var[R^m]$  is the market (systematic) risk for stock i
- and  $var\left[\varepsilon^{i}\right]$  is the non-market (unique, systematic) risk for stock i
- In words,

- In the p = E(mx) framework, we have a much more direct mapping of finance into microeconomics.
  - Instead of dealing with heuristics like mean and variance, we can specify preference and budget constraints over state-contingent consumption.
- Why would we prefer this representation?
  - It is much simpler... We get all the classical results, plus the p = E(mx) framework can be used to price any asset.