



Chapter 5.

Mean-variance frontier and beta representations



Main contents

- An orthogonal characterization of mean-variance frontier
- Spanning the mean-variance frontier
- A compilation of properties of R^* , R^{e*} , x^*
- Mean-variance frontiers for m : H-J bounds



5.1 An orthogonal characterization of mean variance frontier



Introduction

- Method: geometric methods.
- Characterization: rather than write portfolios as combination of basis assets, and pose and solve the minimization problem, we describe the return by a three-way orthogonal decomposition, the mean variance frontier then pops out easily without any algebra.



Some definitions

- Definition of R^* : the return corresponding to the payoff x^* that can act as the discount factor.

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

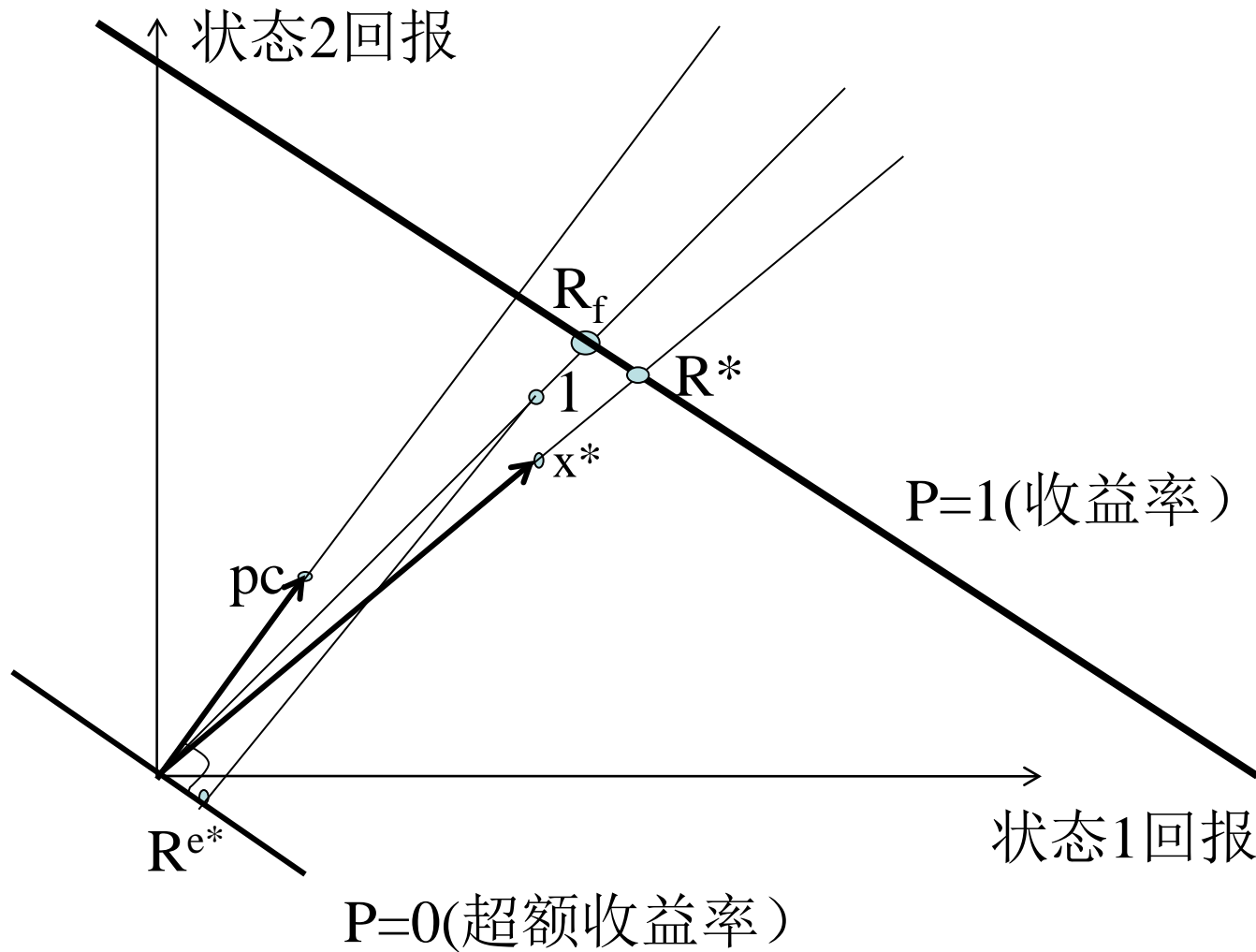
- Definition of R^{e*} :

$$R^{e*} \equiv \text{proj}(1 \mid \underline{R}^e),$$

假定 $\pi(1) = \pi(2)$

$$\underline{R}^e = \{x \in \underline{X}, \text{ s.t. } p(x) = 0\}$$

- R^{e*} 反映了对不同状态的偏好程度。在状态偏好中性世界中，它等于0。





Theorem:

- Every return R^i can be expressed as:

$$R^i = R^* + \omega^i R^{e*} + n^i$$

- Where ω^i is a number, and n^i is an excess return with the property $E(n^i)=0$.
- The three components are orthogonal,

$$E(R^* R^{e*}) = E(R^* n^i) = E(R^{e*} n^i) = 0$$



Theorem: two-fund theorem for MVF

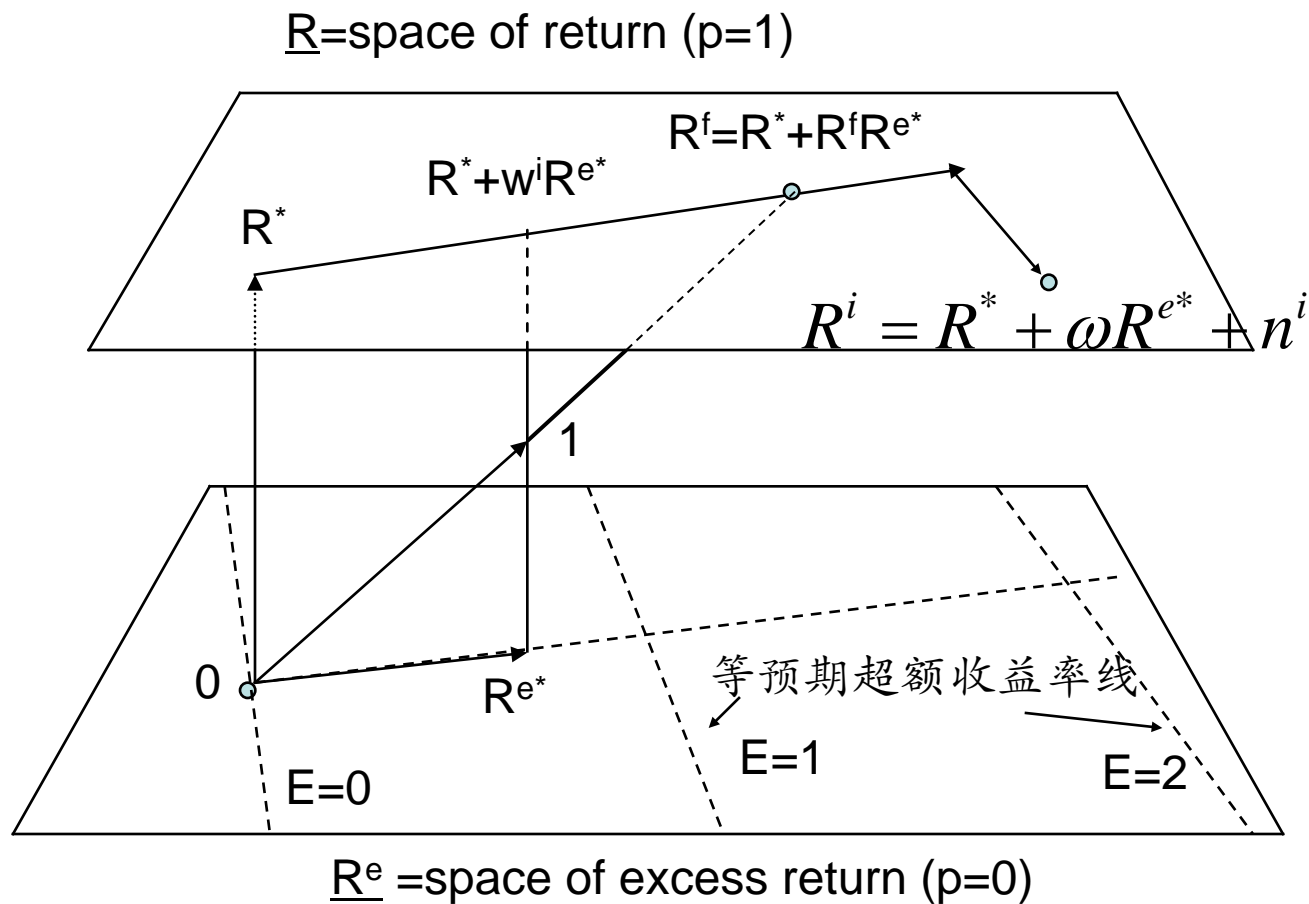
- R^{mv} is on the mean-variance frontier iff

$$R^{mv} = R^* + \omega R^{e*}$$

for some real number ω .



Proof: Geometric method



NOTE: 1、回报空间为三维的。 2、横的平面必须与竖的平面垂直。 3、如果有无风险证券，则竖的平面过1点，否则不过，此时图上的1就是1在回报空间的投影。



- 若不存在无风险资产，则0与均值方差有效边界构成的平面则只能由两个不完全相关的证券构成，否则的话就可以用两个完全相关的证券复制出无风险证券。此时，均值方差有效边界上的点不完全相关。



Proof: Algebraic approach

- Directly from definition, we can get

$$E(n^i) = E(R^{e*} n^i) = 0$$

$$E(R^i) = E(R^*) + w^i E(R^{e*})$$

$$\sigma^2(R^i) = \sigma^2(R^* + w^i R^{e*}) + \sigma^2(n^i)$$

只有 $n^i = 0$ 时，方差在收益率给定情况下才最小。

注意：正交不等于不相关。但 $\text{cov}(R^*, n^i) = E(R^* n^i) - E(R^*) E(n^i) = 0$ ，而 $\text{cov}(R^*, R^{e*}) = E(R^* R^{e*}) - E(R^*) E(R^{e*}) = -E(R^*) E(R^{e*})$



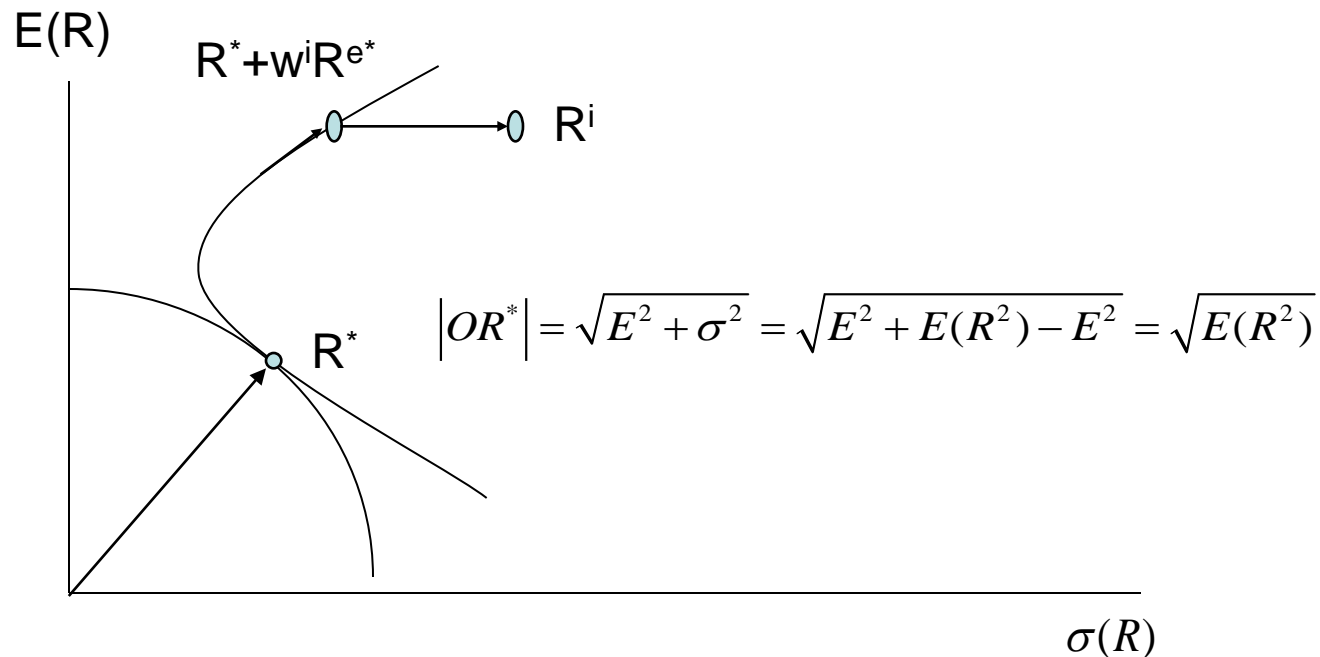
Decomposition in mean-variance space

- R^* is the minimum second moment return.
- Since $E(R^2) = E(R^{*2}) + \omega^2 E(R^{e*2}) + E(n^i2)$
- When $\omega=0$ and $n=0$, $E(R^2)$ is smallest.
- In mean-standard deviation space, the line is circles, thus the minimum second moment return is the smallest circle the intersect the set of all assets.
- It is generally on the lower, or inefficient segment of mean-variance frontier.



Remark

- The minimum second moment return is not the minimum variance return.(why?)





5.2 Spanning the mean variance frontier

Spanning the mean variance frontier



- With any two portfolios on the frontier, we can span the mean-variance frontier.
- Consider

$$R^\alpha = R^* + \gamma R^{e*}, \gamma \neq 0,$$

$$R^{e*} = \frac{R^\alpha - R^*}{\gamma},$$

$$R^* + wR^{e*} = R^* + \frac{w}{\gamma}(R^\alpha - R^*) = (1 - y)R^* + yR^\alpha$$

$$y = w / \gamma$$



5.3 A compilation of properties of R^* , R^{e*} , and x^*



Properties(1)

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

- Proof:

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

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

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



Properties(2)

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

- Proof:

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

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



Properties(3)

- R^* can be used in pricing.
- Proof:

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

- For returns,

$$1 = p(R) = E(x^* R) = \frac{E(R^* R)}{E(R^{*2})} \Rightarrow$$

$$E(R^* R) = E(R^{*2})$$



Properties(4)

- If a risk-free rate is traded,

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

- If not, this gives a “zero-beta rate” interpretation.



Properties(5)

- R^{e*} has the same first and second moment.
- Proof:

$$E(R^{e*}) = E(R^{e*} R^{e*}) = E(R^{e*2})$$

- Then

$$\text{var}(R^{e*}) = E(R^{e*2}) - E(R^{e*})^2 = E(R^{e*})(1 - E(R^{e*}))$$



Properties(6)

- If there is risk free rate,

$$R^f = R^* + R^f R^{e^*}$$

- Proof:

$$1 = \text{proj}(1 | \underline{R^e}) + \text{proj}(1 | R^*)$$

$$R^{e^*} = 1 - \text{proj}(1 | R^*) = 1 - \frac{1}{R_f} R^* \text{ (从上图的两个相似三角形可以看出)}$$

注意 R^* 前面的1和 R_f 都是标量，其余都是向量

$$R_f = R^* + R_f R^{e^*} \text{ (这里的第二个 } R_f \text{ 是标量，其余都是向量)}$$



If there is no risk free rate

- Then the 1 vector can not exist in payoff space since it is risk free.
Then we can only use

$$\begin{aligned} \text{proj}(1 | \underline{X}) &= \text{proj}(\text{proj}(1 | \underline{X}) | \underline{R}^e) + \text{proj}(\text{proj}(1 | \underline{X}) | R^*) \\ &= \text{proj}(1 | \underline{R}^e) + \text{proj}(1 | R^*) \\ R^{e*} &= \text{proj}(1 | \underline{X}) - \text{proj}(1 | R^*) \\ &= \text{proj}(1 | \underline{X}) - E(x^*)R^* \\ &= \text{proj}(1 | \underline{X}) - \frac{E(R^*)}{E(R^{*2})} R^* \end{aligned}$$



Properties(7)

- Since

$$x^* = p' E(xx')^{-1} x$$

$$p(x^*) = E(x^* x^*)$$

- We can get

$$R^* = \frac{x^*}{p(x^*)} = \frac{p' E(xx')^{-1} x}{p' E(xx')^{-1} p}$$



Properties(8)

- Following the definition of projection, we can get

$$R^{e*} = \text{proj}(1 | R^e) = E(R^e)' E(R^e R^{e'})^{-1} R^e$$

- If there is risk free rate, we can also get it by:

$$R^{e*} = 1 - \frac{1}{R_f} R^* = 1 - \frac{1}{R_f} \frac{p' E(xx')^{-1} x}{p' E(xx')^{-1} p}$$



5.4 Mean-Variance Frontiers for Discount Factors: The Hansen- Jagannathan Bounds

Mean-variance frontier for m : H-J bounds



- The relationship between the Sharpe ratio of an excess return and volatility of discount factor.

$$E(mR^e) = E(m)E(R^e) + \rho_{m,R^e} \sigma(m)\sigma(R^e) = 0,$$

$$|\rho_{m,R^e}| = \left| \frac{E(m)E(R^e)}{\sigma(m)\sigma(R^e)} \right| \leq 1,$$

$$\frac{\sigma(m)}{E(m)} \geq \frac{|E(R^e)|}{\sigma(R^e)}$$

- 从经济意义上讲， m 的波动率不会太大，所有夏普比率也不应太大。
- If there is risk free rate, $E(m) = 1 / R_f$



Remark

- We need very volatile discount factors with a mean near one to price the stock returns.



The behavior of Hansen and Jagannathan bounds

- For any hypothetical risk free rate, the highest Sharpe ratio is the tangency portfolio.
- Note: there are two tangency portfolios, the higher absolute Sharpe ratio portfolio is selected.
- If risk free rate is less than the minimum variance mean return, the upper tangency line is selected, and the slope increases with the declination of risk free rate, which is equivalent to the increase of $E(m)$.

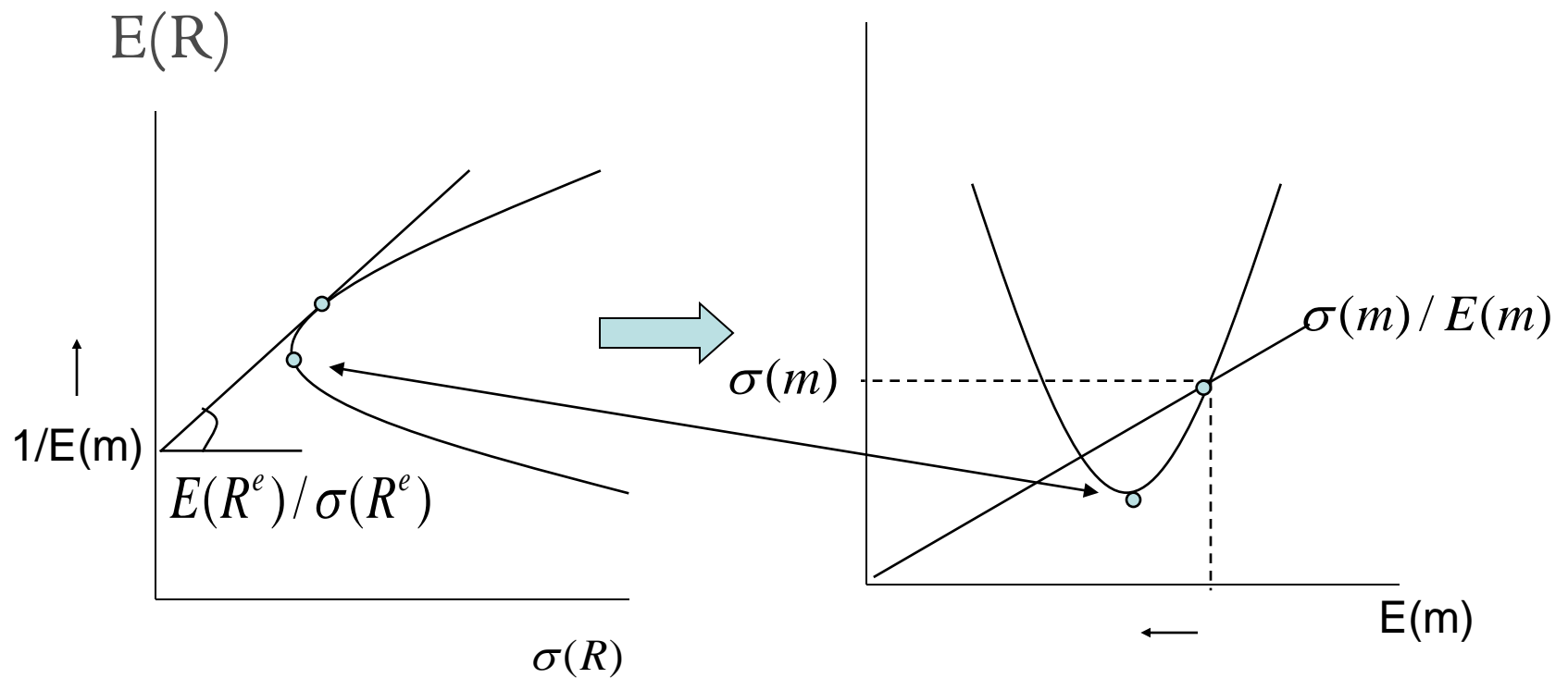
The behavior of Hansen and Jagannathan bounds



- On the other hand, if the risk free rate is larger than the minimum variance mean return, the lower tangency line is selected, and the slope decreases with the declination of risk free rate, which is equivalent to the increase of $E(m)$.
- In all, when $1/E(m)$ is less than the minimum variance mean return, the H-J bound is the decreasing function of $E(m)$. When $1/E(m)$ is larger than the minimum variance mean return, the H-J bound is an increasing function.



Graphic construction





Duality

- A duality between discount factor volatility and Sharpe ratios.

$$\min_{\{ \text{all } m \text{ that price } x \in X \}} \frac{\sigma(m)}{E(m)} = \max_{\{ \text{all excess returns } R^e \text{ in } \underline{X} \}} \frac{E(R^e)}{\sigma(R^e)}$$



Explicit calculation

- A representation of the set of discount factors is

$$m = E(m) + [p - E(m)E(x)]\Sigma^{-1}[x - E(x)] + \varepsilon,$$
$$\Sigma = \text{cov}(x, x'), E(\varepsilon) = 0, E(\varepsilon x) = 0$$

- Proof:

$$\begin{aligned} E(mx) &= E((E(m) + [p - E(m)E(x)]\Sigma^{-1}[x - E(x)] + \varepsilon)x) \\ &= E(m)E(x) + [p - E(m)E(x)]\Sigma^{-1}E[(x - E(x))x] \\ &= E(m)E(x) + p - E(m)E(x) = p \end{aligned}$$



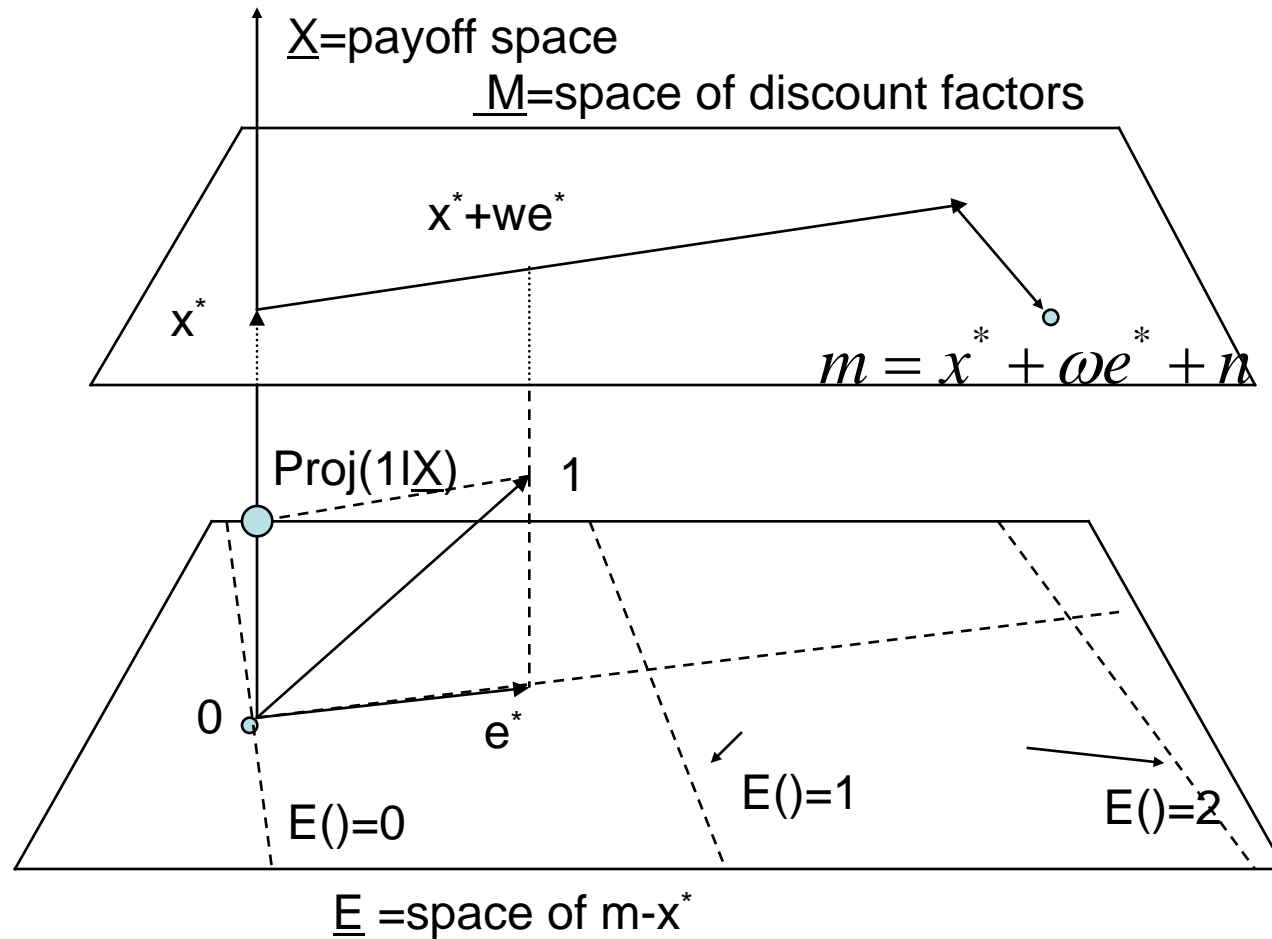
An explicit expression for H-J bounds

$$\sigma^2(m) \geq (p - E(m)E(x))' \Sigma^{-1} (p - E(m)E(x))$$

- Proof:

$$\begin{aligned} \sigma^2(m) &= \left([p - E(m)E(x)]' \Sigma^{-1} \right)^2 \Sigma + \sigma^2(\varepsilon) \\ &= [p - E(m)E(x)]' \Sigma^{-1} [p - E(m)E(x)] + \sigma^2(\varepsilon) \\ &\geq [p - E(m)E(x)]' \Sigma^{-1} [p - E(m)E(x)] \end{aligned}$$

Graphic Decomposition of discount factor



NOTE:横的平面必须与竖的平面垂直。



Decomposition of discount factor

- Any discount factor must lie in the plane perpendicular to payoff space through x^* .

$$m = x^* + we^* + n$$

- Where

$$e^* \equiv 1 - \text{proj}(1 | \underline{X}) = \text{proj}(1 | \underline{E})$$

e^* 反映了无风险资产不存在而产生的定价偏差。

- The mean-variance frontier of m is given by

$$m = x^* + we^*$$



Special case

- If unit payoff is in payoff space,

$$e^* = 1 - \text{proj}(1 | \underline{X}) = 0$$

- The frontier and bound are just $m = x^*$
- And

$$\sigma^2(m) \geq \sigma^2(x^*)$$

- This is exactly like the case of state preference neutrality for return mean-variance frontiers, in which the frontier reduces to the single point R^* .



Mathematical construction

- We have got

$$\begin{aligned}m^* &= x^* + we^* \\&= p' E(xx')^{-1} x + w \left(1 - \text{proj}(1 | \underline{X})\right) \\&= p' E(xx')^{-1} x + w \left(1 - E(x)' E(xx')^{-1} x\right) \\&= w + \left[p - wE(x)\right]' E(xx')^{-1} x \\E(m^*) &= w + \left[p - wE(x)\right]' E(xx')^{-1} E(x) \\ \sigma^2(m^*) &= \left[p - wE(x)\right]' \text{cov}(xx')^{-1} \left[p - wE(x)\right]\end{aligned}$$



Some development

- H-J bounds with positivity. It solves
$$\min \sigma^2(m), s.t. p = E(mx), m > 0, E(m) \text{ 固定}$$
- This imposes the no arbitrage condition.
- Short sales constraint and bid-ask spread is developed by Luttmer(1996).
- A variety of bounds is studied by Cochrane and Hansen(1992).

The End

