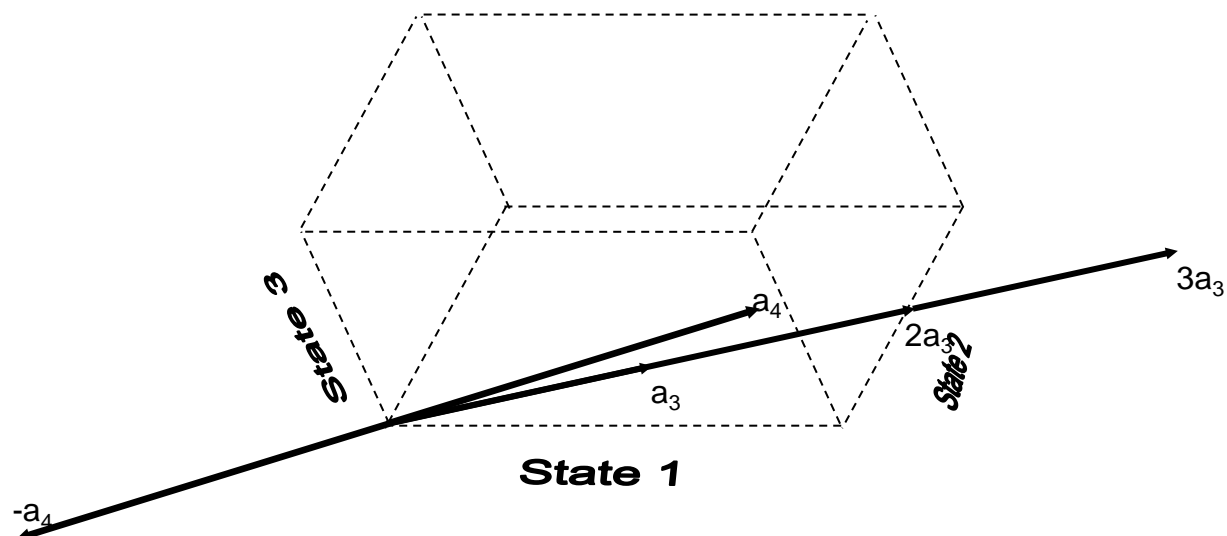


Lecture 1

One-period model
of financial markets
& hedging
problems



One-period model of financial markets



Aims of section

- Introduce one-period model with finite number of states and the basic asset-pricing terminology
- Formulate hedging problem and try to solve it
- Explain the concepts of complete and incomplete markets
- Explain the role of matrix inverse in hedging and pricing
- Introduce “state prices” and discuss two different methods for asset pricing:
 - 1) by replication
 - 2) using a pricing kernel

Introductory example

- There are four assets available
 - 1) Uncertain stock price – 3 scenarios
 - Stock value tomorrow = 3, 2, or 1 with probability $1/2$, $1/6$, $1/3$
 - 2) Risk-free asset with value 1 tomorrow
 - 3) Two derivative securities – options on the stock
 - 3.1) Call option #1 struck at $K = 1.5$
 - 3.2) Call option #2 struck at $K = 1$
- Task: sell option #2, and to reduce risk exposure construct a hedging portfolio consisting of stock and risk-free asset

Uncertainty in one-period model

- Two dates: today and tomorrow
- Value of all securities known today
- The tomorrow's payoffs are uncertain
- Organization of uncertainty:
 - Finite number of scenarios
 - Each scenario known in detail today
 - Probability of each scenario known today

Model asset payoffs

- Model stock and derivative payoffs

Probability	$\frac{1}{2}$	$\frac{1}{6}$	$\frac{1}{3}$
Stock	3	2	1
Call option #1 ($K = 1.5$)	1.5	0.5	0
Call option #2 ($K = 1$)	2	1	0

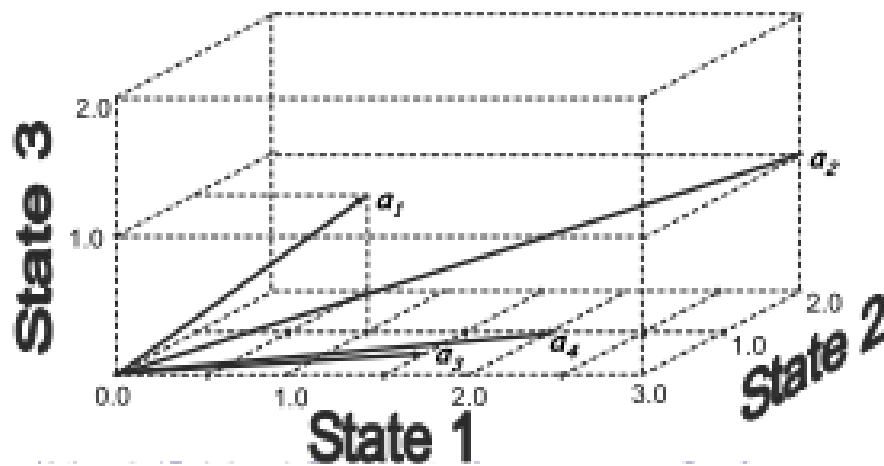
Payoffs as vectors

- Risk free asset : $a_1 = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$
- Stock : $a_2 = \begin{bmatrix} 3 \\ 2 \\ 1 \end{bmatrix}$
- Options : $a_3 = \begin{bmatrix} 1.5 \\ 0.5 \\ 0 \end{bmatrix}, a_4 = \begin{bmatrix} 2 \\ 1 \\ 0 \end{bmatrix}$

Graphical representation of the payoffs

- Payoffs in state-space form

$$a_1 = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} \quad a_2 = \begin{bmatrix} 3 \\ 2 \\ 1 \end{bmatrix} \quad a_3 = \begin{bmatrix} 1.5 \\ 0.5 \\ 0 \end{bmatrix} \quad a_4 = \begin{bmatrix} 2 \\ 1 \\ 0 \end{bmatrix}$$



Create vectors in Matlab

`a1 = ones(3,1) or a1=[1;1;1];`

`a2 = [3;2;1];`

`a3 = [1.5;0.5;0];`

`a4 = [2;1;0];`

Operation on securities/vectors

- Scalar multiplication: Leverage
 - Example: Buy two units of option #1

$$2a_3 = 2 \begin{bmatrix} 1.5 \\ 0.5 \\ 0 \end{bmatrix} = \begin{bmatrix} 3 \\ 1 \\ 0 \end{bmatrix}$$

- Matlab command: $2 * a_3$

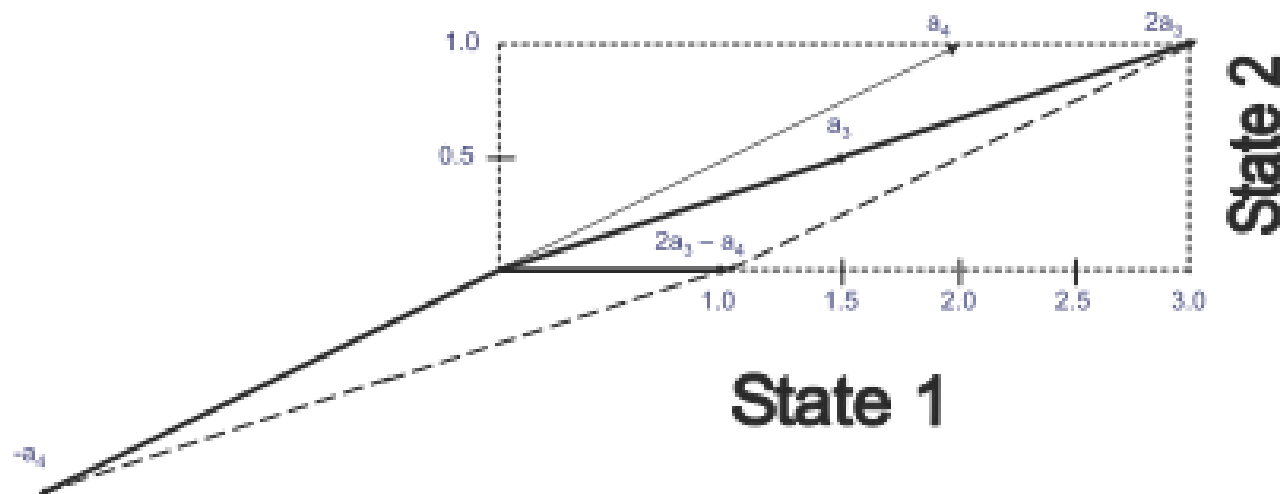
Operation on securities/vectors (cont.)

- Addition: Portfolios

- Example: Buy two units of option #1, sell one unit of option #2

$$2a_3 - a_4 = \begin{bmatrix} 2 * 1.5 - 2 \\ 2 * 0.5 - 1 \\ 2 * 0 - 0 \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$$

- Matlab command : $2 * a3 - a4$



Matrix as a collection of securities/vectors

- It is common to work with several vectors at once and it is natural to form a matrix

$$\text{Vectors: } a_1 = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} \quad a_2 = \begin{bmatrix} 3 \\ 2 \\ 1 \end{bmatrix} \quad a_3 = \begin{bmatrix} 1.5 \\ 0.5 \\ 0 \end{bmatrix} \quad a_4 = \begin{bmatrix} 2 \\ 1 \\ 0 \end{bmatrix}$$

$$\text{A matrix: } \begin{bmatrix} 1 & 3 & 1.5 & 2 \\ 1 & 2 & 0.5 & 1 \\ 1 & 1 & 0 & 0 \end{bmatrix}$$

- Matlab command: $A = [a_1 \ a_2 \ a_3 \ a_4]$

Matrix operation: Transposition

- Sometimes we need a *row vector* rather than a *column vector*.
- Achieved by *transposition* of a column vector

$$x = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}, \quad x^* = [x_1 \quad x_2 \quad \cdots \quad x_n]$$

- Conversely, transposition of a *row vector* gives a *column vector*

Matrix notation

- Matrix notation: $M \in \mathbb{R}^{m \times n}$
 - Where m is the number of rows and n is the number of columns
- The element in the i -th row and j -th column is denoted M_{ij}
- The entire j -th column is denoted $M_{.j}$
- The entire i -th row is denoted $M_i.$

Matrix notation (cont.)

- $M \in \mathbb{R}^{m \times n}$

$$M = \begin{bmatrix} M_{11} & M_{12} & \dots & M_{1n} \\ M_{21} & M_{22} & \dots & M_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ M_{m1} & M_{m2} & \dots & M_{mn} \end{bmatrix} = \begin{bmatrix} M_{1\cdot} \\ M_{2\cdot} \\ \vdots \\ M_{m\cdot} \end{bmatrix} = [M_{\cdot 1} \quad M_{\cdot 2} \quad \dots \quad M_{\cdot n}]$$

Matrix operation: Transposition

- $M \equiv M_{ij} \rightarrow M^* \equiv M_{ji}$
- $M \in \mathbb{R}^{m \times n} \rightarrow M^* \in \mathbb{R}^{n \times m}$

Example

$$1) \quad x = \begin{bmatrix} 1.5 \\ 0.5 \\ 0 \end{bmatrix}, \quad x^* =$$

$$2) \quad A = \begin{bmatrix} 1 & 3 & 1.5 \\ 1 & 2 & 0.5 \\ 1 & 1 & 0 \end{bmatrix}, \quad A^* =$$

Matrix operation: Multiplication

- $U \in \mathbb{R}^{m \times k}$ and $V \in \mathbb{R}^{k \times n} \rightarrow UV \in \mathbb{R}^{m \times n}$

$$UV = \begin{bmatrix} U_{1\cdot} \\ U_{2\cdot} \\ \vdots \\ U_{m\cdot} \end{bmatrix} \begin{bmatrix} V_{\cdot 1} & V_{\cdot 2} & \dots & V_{\cdot n} \end{bmatrix} = \begin{bmatrix} U_{1\cdot}V_{\cdot 1} & U_{1\cdot}V_{\cdot 2} & \dots & U_{1\cdot}V_{\cdot n} \\ U_{2\cdot}V_{\cdot 1} & U_{2\cdot}V_{\cdot 2} & \dots & U_{2\cdot}V_{\cdot n} \\ \vdots & \vdots & \ddots & \vdots \\ U_{m\cdot}V_{\cdot 1} & U_{m\cdot}V_{\cdot 2} & \dots & U_{m\cdot}V_{\cdot n} \end{bmatrix}$$

- Example

$$U = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}, V = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \quad UV = ?$$

Properties of matrix multiplication

- Not commutative

$$UV \neq VU$$

- Associative

$$(UV)W = U(VW)$$

- Transposition reverses order of multiplication !

$$(UV)^* = V^*U^*$$

Matrix and its application to portfolios

- Example: Suppose we sell 2 units of option #1, one unit of option #2, buy two units of stocks and borrow one unit of bond. What are the payoffs of this portfolio ?

Given:

$$bond = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}, stock = \begin{bmatrix} 3 \\ 2 \\ 1 \end{bmatrix}, option\#1 = \begin{bmatrix} 1.5 \\ 0.5 \\ 0 \end{bmatrix} \text{ and } option\#2 = \begin{bmatrix} 2 \\ 1 \\ 0 \end{bmatrix}$$

- Assuming prices of securities are $S_{bond} = 1, S_{stock} = 3, S_{option\#1} = 1, S_{option\#2} = 2$, what is the price of this portfolio?

Hedging Problem

- $Ax = b$
 - A denotes payoff matrix of basis assets
 - x denotes portfolio of basis assets
 - b denotes payoff of a *focus asset*
- Bank wants to issue a security b
- To offset risk, it separately buys hedging portfolio x
- The risk is reduced to $Ax - b$
 - If $Ax - b = 0$, then the bank's position is perfectly hedged
- The bank will price the security at $S^*x + \text{overheads and risk premium}$.

Hedging Problem (cont.)

- Example:

Given:

$$b = \begin{bmatrix} 2 \\ 1 \\ 0 \end{bmatrix}, a1 = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}, a2 = \begin{bmatrix} 3 \\ 2 \\ 1 \end{bmatrix}, a3 = \begin{bmatrix} 1.5 \\ 0.5 \\ 0 \end{bmatrix}$$

Your client wants to buy a security with payoff b , knowing there are assets $a1$, $a2$, and $a3$ traded in the markets. What is the strategy that gives you a perfect hedge? And how much you would charge the client if $S_{a1} = \frac{1}{1.05}$ and $S_{a2} = 2$?

Hedging Problem (cont.)

Solution:

Hedging Complications

- If inverse of A exists: $x = A^{-1}b$ is a perfect hedge and the solution is also unique

- Now consider $\begin{bmatrix} 1 & 3 & 2 \\ 1 & 2 & 1 \\ 1 & 1 & 0 \end{bmatrix} x = \begin{bmatrix} 1.5 \\ 0.5 \\ 0 \end{bmatrix}$
 - No solution, Inverse does not exist

- Consider $\begin{bmatrix} 1 & 3 \\ 1 & 2 \\ 1 & 1 \end{bmatrix} x = \begin{bmatrix} 2 \\ 1 \\ 0 \end{bmatrix}$
 - Solution exists, inverse does not

Linear independence and redundant assets

- Some complications in the hedging problem are caused by redundant assets
- An asset is redundant if it can be replicated by other assets
- Securities A_1, \dots, A_n are linearly independent if none of them is a portfolio payoff of the remaining $n-1$ securities

Dimension of marketed subspace

- Marketed subspace = all portfolio payoffs Ax generated by all possible basis assets x
 - Mathematically $Span(A_{.1}, \dots, A_{.n})$
- Each linearly independent asset adds a new dimension to the marketed subspace
- Maximum number of lin. ind. assets = dimension of marketed subspace
- Dimensionality Theorem: The lin. ind. assets can be chosen in many ways, but their number is always the same = dimension

Hedging in a complete market without redundant basis assets

- Complete market \Leftrightarrow
of linearly independent basis assets = # of states

- **Theorem:**

Suppose we have m states and a complete market with m basis assets. Then the payoff matrix is invertible and the hedging portfolio for any focus asset b is given by

$$x = A^{-1}b$$

Suppose prices of basis assets are stored in vector S . Under frictionless trading the only possible price of the focus asset equals

$$S^*x = S^*A^{-1}b$$

Find dimension of marketed subspace

- Concepts
 - Suppose A_1, \dots, A_k are linearly independent. Then either A_{k+1} is redundant or A_1, \dots, A_{k+1} are linearly independent
 - With m states, there cannot be more than m linearly independent assets.
- How we do it
 - Sort securities into two baskets
 - Linearly independent
 - Redundant

- Which of the following assets are linearly independent and redundant assets? Is this market complete?

$$A_1 = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}, A_2 = \begin{bmatrix} 3 \\ 2 \\ 1 \end{bmatrix}, A_3 = \begin{bmatrix} 1.5 \\ 0.5 \\ 0 \end{bmatrix}, A_4 = \begin{bmatrix} 2 \\ 1 \\ 0 \end{bmatrix}$$

Dimension and matrix rank

- Maximum number of linearly independent columns in a matrix A is called rank, $r(A)$
- If A is a payoff matrix of basis assets then $r(A)$ = dimension of marketed subspace
- Facts about rank
 - $r(A^*A) = r(A)$
 - $r(AB) \leq \min(r(A), r(B))$
 - $r(A) = r(A^*)$
 - If A is $m \times n$ then $r(A) \leq \min(m, n)$
- When $r(A) = \min(m, n)$, we say A has full rank
- Square matrices with full rank are invertible (non-singular matrices)

Identity matrix and Arrow-Debreu securities

- Arrow-Debreu (elementary) security for state j , denoted e_j , has payoff 1 in state j and payoff zero in all other states
- With m scenarios stacking all elementary securities into a matrix gives an $m \times m$ identity matrix

- For every square matrix A with full rank there is a matrix B such that $AB = BA = I$
- Matrix B is unique, it is called the inverse to matrix A and it is commonly denoted by A^{-1}
- Facts:
 - If C, D are invertible then $(CD)^{-1} = D^{-1}C^{-1}$
 - $(A^{-1})^{-1} = A$

Inverse matrix and hedging portfolios

- Interpretation of inverse matrix
- Let split $AA^{-1} = I$ by columns: $AA_{\bullet j}^{-1} = e_j$
- This looks like $Ax = e_j$
- Therefore j – th column of A^{-1} gives replicating portfolio to Arrow-Debreu security e_j
- Example: if $A = \begin{bmatrix} 1 & 3 & 1.5 \\ 1 & 2 & 0.5 \\ 1 & 1 & 0 \end{bmatrix}$ and knowing that $A^{-1} = \begin{bmatrix} 1 & -3 & 3 \\ -1 & 3 & -2 \\ 2 & -4 & 2 \end{bmatrix}$, show that Arrow-Debreu security e_j can be replicated by portfolio $A_{\bullet j}^{-1}$

- State price ψ_j is the price of an Arrow-Debreu security e_j
- Vector ψ is called the “state price vector”
- Assuming complete market and no redundant assets
 - State price vector

$$\psi^* = S^* A^{-1},$$

where S is the vector of prices of the basis assets and A is the matrix of the payoffs of the basis assets

Proof :

Pricing formulae

- Suppose A is invertible (square, full rank)
- Complete market, no redundant basis assets
- Perfect hedge $Ax = b$
- Two ways to find the price of focus asset b
 1. By replication:

$$\text{Focus asset price} = S^*x$$

2. Using state price (pricing kernel):

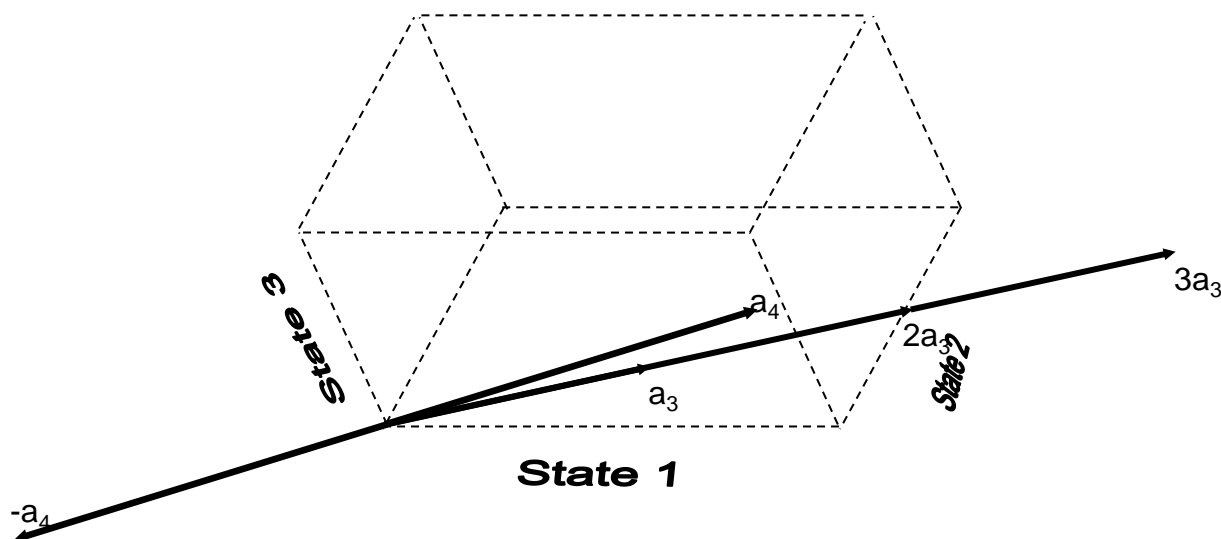
$$\text{Focus asset price} = \psi^*b$$

Proof of the second formula:

Example

-
- What is the price of $b = \begin{bmatrix} 2 \\ 1 \\ 0 \end{bmatrix}$ given basis assets $A = \begin{bmatrix} 1 & 3 & 1.5 \\ 1 & 2 & 0.5 \\ 1 & 1 & 0 \end{bmatrix}$ with prices $S = \begin{bmatrix} 1 \\ 2 \\ 0.6 \end{bmatrix}$
 - By replication:
 - By state price:

Hedging in an incomplete market



Aims of this section

- Explain how to compute replicating portfolios in an incomplete market

Hedging formula for incomplete markets, no redundant basis assets

- $r(A) = n < m$, fewer assets than states

$$Ax = b \quad (1)$$

- multiply by A^* from the left

$$A^*Ax = A^*b \quad (2)$$

- A^*A is square with full rank, it has an inverse

$$x = (A^*A)^{-1} A^*b \quad (3)$$

- x solves (2) but does it solve (1)?
- Hedging error = $Ax - b = A(A^*A)^{-1} A^*b - b$
- If hedging error = 0 then solution exists and is given by (3), otherwise there is no solution

Example: Incomplete markets without redundant basis assets

- Is there a perfect hedge of a focus asset b considering that we can trade portfolio A ?

$$A = [A_{\bullet 1} \ A_{\bullet 2}] = \begin{bmatrix} 1 & 3 \\ 1 & 2 \\ 1 & 1 \end{bmatrix} \quad b = A_{\bullet 3} = \begin{bmatrix} 1.5 \\ 0.5 \\ 0 \end{bmatrix}$$

Hedging problem with redundant basis assets

- Redundant basis assets do not affect existence of a solution, they merely add free parameters to the solution
- $A = A_1 \mid A_2$, $A_2 = A_1 C$, $r(A)=r(A_1)$, $x = x_1 \mid x_2$
 - If A_1 is square, market is complete
$$x_1 = (A_1)^{-1}b - Cx_2$$
 - If A_1 is not square, market is not complete
$$x_1 = (A_1^* A_1)^{-1} A_1^* b - Cx_2$$
 - Hedging error = $A_1 (A_1^* A_1)^{-1} A_1^* b - b$
 - x_2 represents the free parameters in the solution
- Example 2.3 pp. 28-29

The Least-Squares Hedge

- In practice most markets are incomplete $r(A) = n < m$, fewer assets than number of states
- This means we cannot always find a perfect hedge (replication) for a focus asset $b \rightarrow$ mathematically, $Ax = b$ cannot always be solved.
- Instead, we would like to find the best *approximate hedge* according to some criterion

The Least-Squares Hedge

- Replication error : $\varepsilon = Ax - b$
- Criterion: Minimize the Sum of Squared Replication Errors

$$SSRE = \varepsilon_1^2 + \varepsilon_2^2 + \cdots + \varepsilon_m^2$$

- Optimal hedge = least-squares hedging portfolio

$$\mathbf{x} = (\mathbf{A}^* \mathbf{A})^{-1} \mathbf{A}^* \mathbf{b}$$

Least Squares Hedge – Example

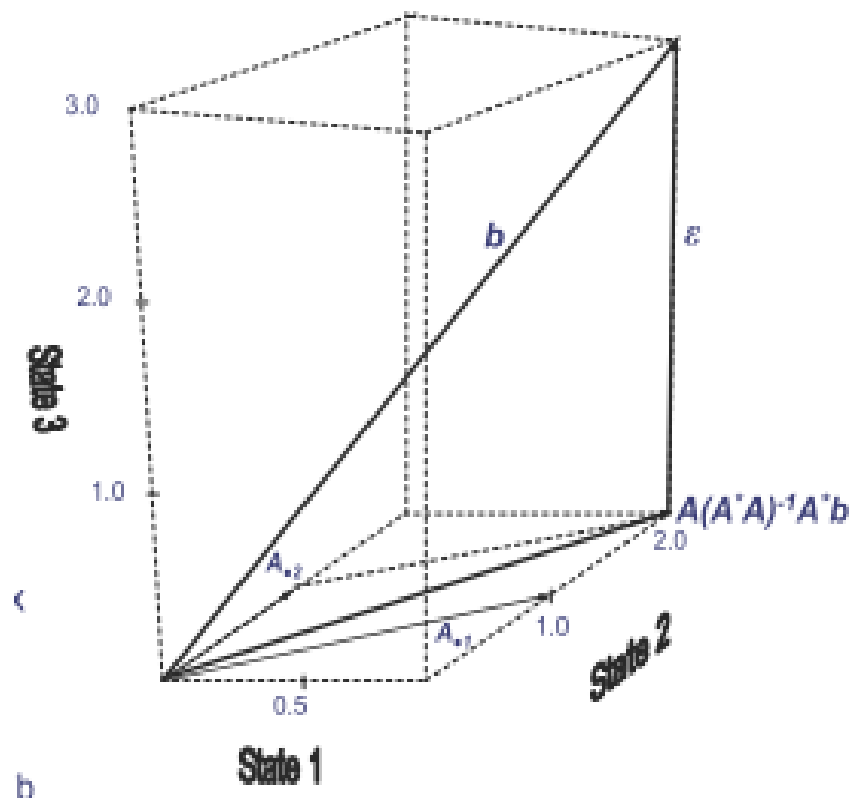
- Focus asset $b^* = [1 \ 2 \ 3]$
- Basis assets: $A_{.1}^* = [1 \ 1 \ 0]$ and $A_{.2}^* = [0 \ 1 \ 0]$

What is the least-squares hedge of the focus asset?

Solution:

Least squares hedge - Geometry

- Optimal criteria
 - Minimize $\varepsilon^* \varepsilon$
(square length of vector ε)
- Point of minimal distance of Ax from b :
 - ε must be orthogonal to all vectors in marketed subspace A
 - Mathematically $A^* \varepsilon = 0$
 - $A^*(b - Ax) = 0 \Rightarrow$
 $x = (A^*A)^{-1}A^*b$



Minimizing the Expected Squared Replication Error (ESRE)

- Not all scenarios are equally likely: we need criterion putting more weight on likely scenarios:

$$ESRE = p_1 \varepsilon_1^2 + p_2 \varepsilon_2^2 + \dots + p_m \varepsilon_m^2$$

- The optimal hedge is :

$$\hat{x} = (\tilde{A}^* \tilde{A})^{-1} \tilde{A}^* \tilde{b}$$

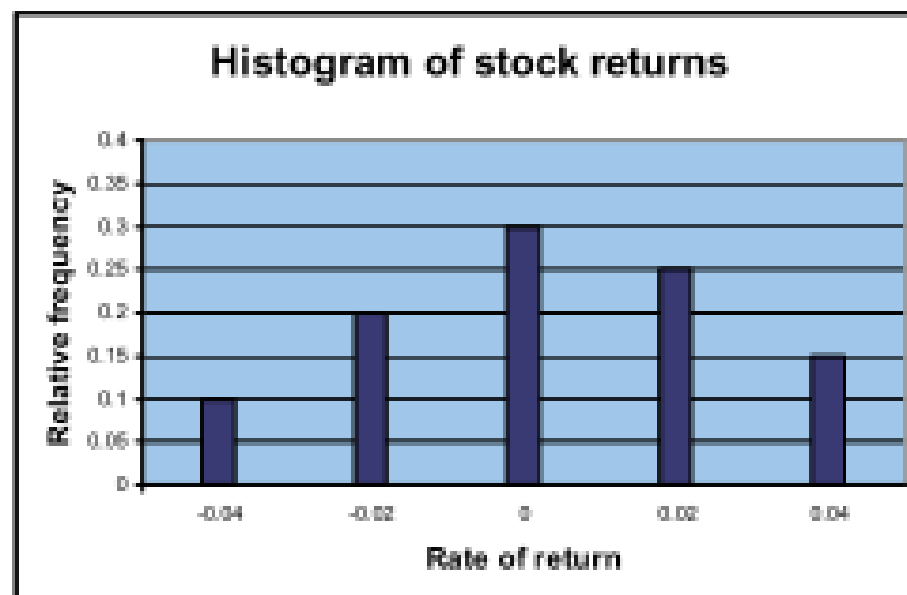
where $\tilde{A}_i. \equiv \sqrt{p_i} A_i.$ and $\tilde{b}_i \equiv \sqrt{p_i} b_i$

Working out the minimal ESRE hedge

- Proof
 - Transform ESRE into SSRE

Example

- Exercise 2.7 p.54
 - Risk-free rate 0%
 - Find optimal hedge of exposure X
- Share price £2
 - Share pay-off = $(1 + \text{rate of return}) \times \text{initial investment}$
 - Evaluate the cost of the best hedge



PFCo rate of return	Amount X to be hedged
0.04	£2000
0.02	£1000
0	£100
-0.02	£20
-0.04	£0