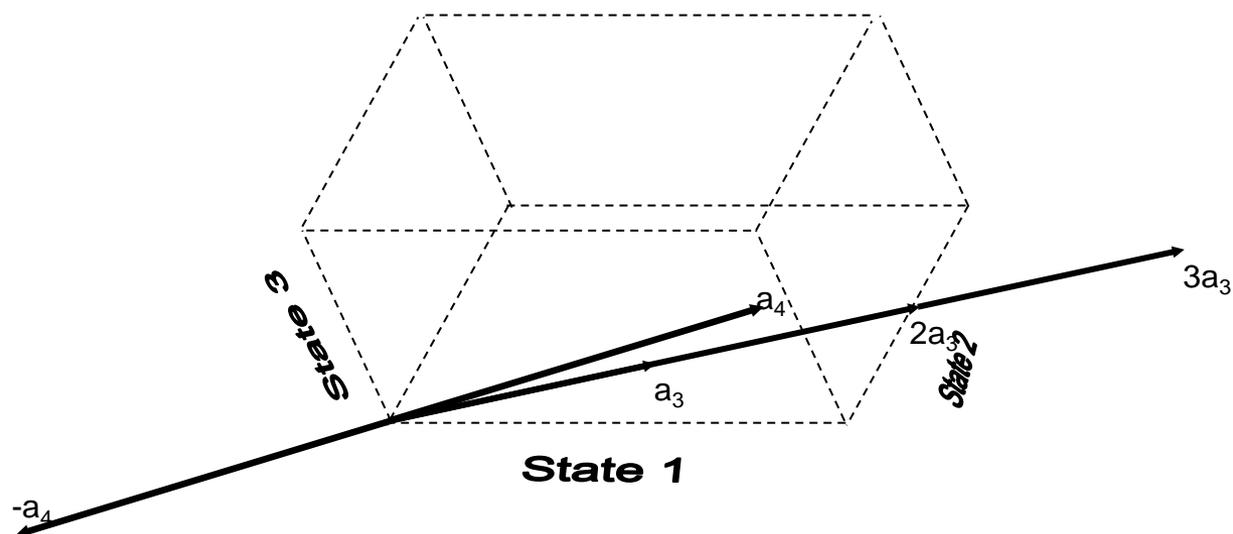


## Lecture 1

One-period model  
of financial markets  
& hedging  
problems



# One-period model of financial markets



# Aims of section

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- 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

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- 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

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- 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

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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

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# Payoffs as vectors

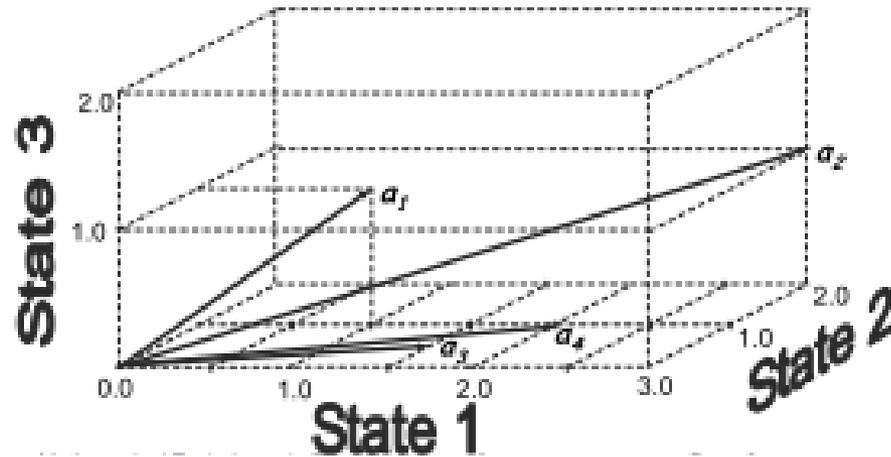
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- Risk free asset :  $a_1 = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$
- Stock :  $a_2 = \begin{bmatrix} 3 \\ 2 \\ 1 \end{bmatrix}$
- Options :  $a_3 = \begin{bmatrix} 1.5 \\ 0.5 \\ 0 \end{bmatrix}$ ,  $a_4 = \begin{bmatrix} 2 \\ 1 \\ 0 \end{bmatrix}$

# 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

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`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:  $2 * a_3$

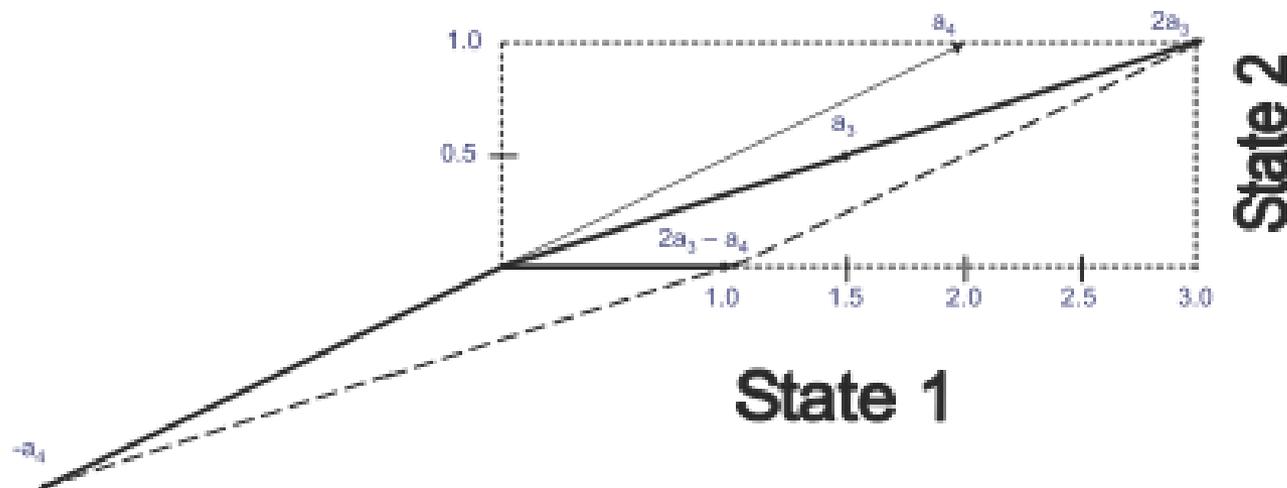
# Operation on securities/vectors (cont.)

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

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

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

# Matrix operation: Multiplication

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

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

- 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

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- 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

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- $Ax = b$ 
  - $A$  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.)

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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

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- 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

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- 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

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- 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

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- 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  $\psi_j$  is the price of an Arrow-Debreu security  $e_j$
- Vector  $\psi$  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 :**

- 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:**

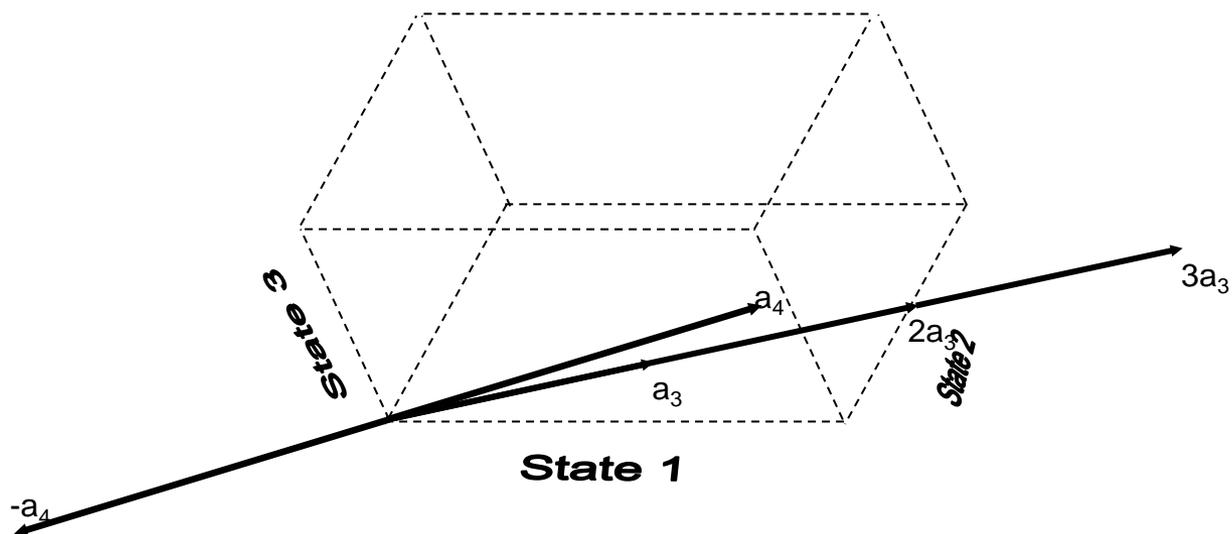
- 
- What is the price of  $b = \begin{bmatrix} 2 \\ 1 \\ 0 \end{bmatrix}$  given basis assets  $A = \begin{bmatrix} 1 & 3 & 1.5 \\ 1 & 2 & 0.5 \\ 1 & 1 & 0 \end{bmatrix}$  with

prices  $S = \begin{bmatrix} 1 \\ 2 \\ 0.6 \end{bmatrix}$

- By replication:

- By state price:

# Hedging in an incomplete market



# Aims of this section

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- 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  $A$  ?

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

# 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

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- 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

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

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

- Optimal hedge = least-squares hedging portfolio

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

# Least Squares Hedge – Example

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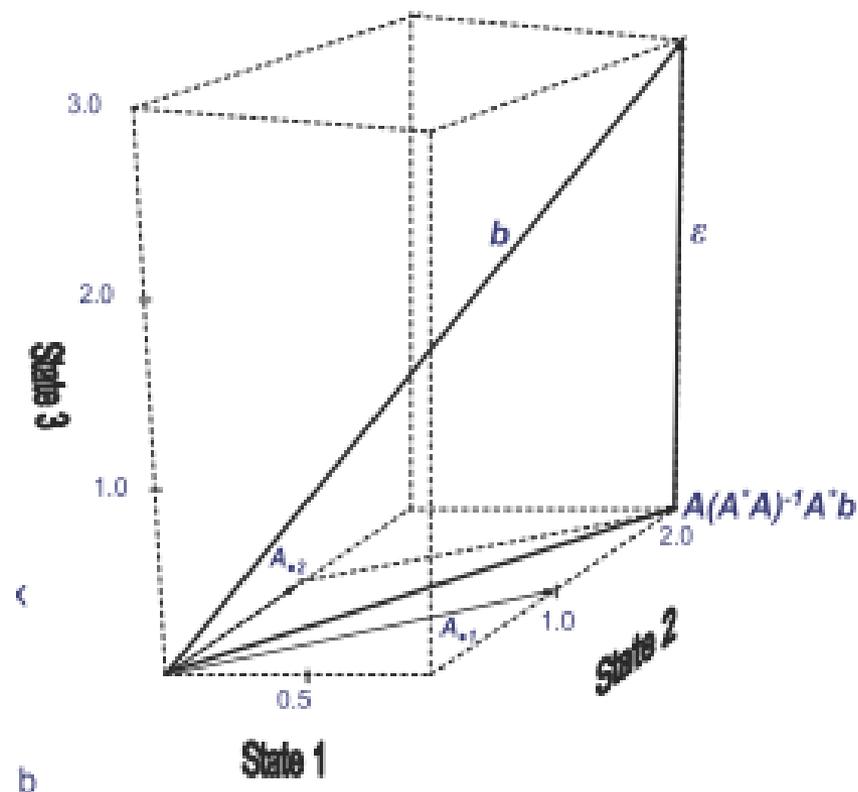
- Focus asset  $b^* = [1 \ 2 \ 3]$
- Basis assets:  $A_{.1}^* = [1 \ 1 \ 0]$  and  $A_{.2}^* = [0 \ 1 \ 0]$

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

Solution:

# Least squares hedge - Geometry

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



# 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 :

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

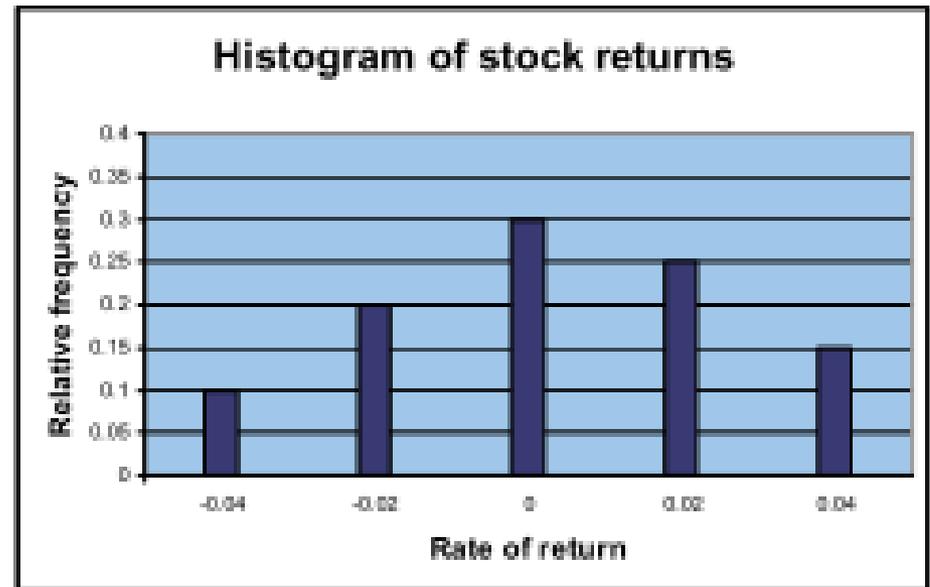
where  $\tilde{A}_i \equiv \sqrt{p_i} A_i$ . and  $\tilde{b}_i \equiv \sqrt{p_i} b_i$

# Working out the minimal ESRE hedge

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- Proof
  - Transform ESRE into SSRE

- 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