

Mathematics for Analytics and Finance

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Module 3



Matrix Algebra



Vectors

Definition: A vector, in simple terms, is like a list of numbers arranged either in a single column or a single row.

Example: $\mathbf{a} = \begin{bmatrix} a_1 \\ a_2 \end{bmatrix}$, a_1 = number of products 1, a_2 = number of products 2

Vector addition and scalar multiplication are component-wise. For $\mathbf{a} = \begin{bmatrix} a_1 \\ a_2 \end{bmatrix} \in \mathbb{R}^2$ and $\mathbf{b} = \begin{bmatrix} b_1 \\ b_2 \end{bmatrix} \in \mathbb{R}^2$ we define:

- **Vector addition:** $\mathbf{a} + \mathbf{b} = \begin{bmatrix} a_1 + b_1 \\ a_2 + b_2 \end{bmatrix} \in \mathbb{R}^2$
- **Scalar Multiplication:** $\lambda \mathbf{a} = \begin{bmatrix} \lambda a_1 \\ \lambda a_2 \end{bmatrix} \in \mathbb{R}^2$

(this is extended to n dimensions)

Remark: $\mathbf{a} = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$ is also written as $\mathbf{a} = (1,2,3)$



Dot Products

Definition: The dot product or inner product of $\mathbf{a} = (a_1, a_2, \dots, a_n)$ and $\mathbf{b} = (b_1, b_2, \dots, b_n)$ is

$$\mathbf{a} \cdot \mathbf{b} = a_1a_2 + b_1b_2 + \dots + a_nb_n.$$

Example: The vectors $\mathbf{a} = (4,2)$ and $\mathbf{b} = (-1,2)$ have a **zero** dot product.

$$\mathbf{a} \cdot \mathbf{b} = (4)(-1) + (2)(2) = 0$$

Remark: Zero is a special case. It means that the two vectors are **perpendicular**.

Example: We have three products to buy. The prices are (p_1, p_2, p_3) and the quantities are (q_1, q_2, q_3) . Then, our total cost is:

$$\text{Total Cost} = (p_1, p_2, p_3) \cdot (q_1, q_2, q_3) = p_1q_1 + p_2q_2 + p_3q_3$$



Length and Unit Vectors

Definition: The length or norm of a vector $\mathbf{a} = (a_1, a_2, \dots, a_n)$ is the square root of $\mathbf{a} \cdot \mathbf{a}$:

$$\text{Length} = \|\mathbf{a}\| = \sqrt{\mathbf{a} \cdot \mathbf{a}} = \sqrt{a_1^2 + \dots + a_n^2}.$$

Example: The length of $\mathbf{a} = (3, 4)$ is $\sqrt{3^2 + 4^2} = 5$

Definition: A unit vector \mathbf{u} is a vector whose length is equal to one. That is, $\mathbf{u} \cdot \mathbf{u} = 1$

Example: $\mathbf{i} = (1, 0)$ and $\mathbf{j} = (0, 1)$ and $\mathbf{u} = \left(\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}}\right)$ are unit vectors.

Remark: Divide any nonzero vector \mathbf{a} by its length. Then the vector $\mathbf{u} = \frac{\mathbf{a}}{\|\mathbf{a}\|}$ is a unit vector in the same direction as \mathbf{a} .

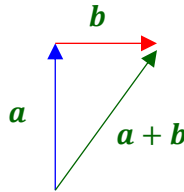


Angle between Two Vectors

If \mathbf{a} is perpendicular to \mathbf{b} , then $\mathbf{a} \cdot \mathbf{b} = 0$.

Proof: (For interested students)

By Pythagorean Theorem: $\|\mathbf{a}\|^2 + \|\mathbf{b}\|^2 = \|\mathbf{a} + \mathbf{b}\|^2 \Rightarrow (a_1^2 + b_2^2) + (b_1^2 + b_2^2) = (a_1 + b_1)^2 + (a_2 + b_2)^2$
 $\Rightarrow 2(a_1b_1 + a_2b_2) = 0 \Rightarrow a_1b_1 + a_2b_2 = 0$



Vectors in R

Creating and Combining Vectors

```
c(first number or string,...,last number or character)
```

```
> c(1,1,2,3,5,8,13,21)
```

```
[1] 1 1 2 3 5 8 13 21
```

```
> c(1*pi, 2*pi, 3*pi, 4*pi)
```

```
[1] 3.141593 6.283185 9.424778 12.566371
```

```
> c("Everyone", "loves", "stats.")
```

```
[1] "Everyone" "loves" "stats."
```

Combining two vectors:

```
> v1 = c(1,2,3)
```

```
> v2 = c(4,5,6)
```

```
> c(v1,v2)
```

```
[1] 1 2 3 4 5 6
```



Vectors in R

Computing Basic Statistics

```
mean(x), median(x), sd(x), var(x), cor(x,y), cov(x,y), range(x), quantile(x)
```

```
summary(x) gives some of the summary statistics
```

```
> x = c(0,1,1,2,3,5,8,13,21,34)
```

```
> y = log(x+1)
```

```
> mean(x)
```

```
[1] 8.8
```

```
> median(x)
```

```
[1] 4
```

```
> sd(x)
```

```
[1] 11.03328
```

```
> summary(x)
```

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
0.00	1.25	4.00	8.80	11.75	34.00



Vectors in R

Creating Sequences

```
seq(from, to) or from:to  
seq(from, to, by= )  
seq(from, to, length= )  
rep(number, number of repetitions)
```

```
> 10:19
```

```
[1] 10 11 12 13 14 15 16 17 18 19
```

```
> 9:0
```

```
[1] 9 8 7 6 5 4 3 2 1 0
```

```
> seq(from=0, to=20)
```

```
[1] 0 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20
```

```
> seq(from=0, to=20, by=2)
```

```
[1] 0 2 4 6 8 10 12 14 16 18 20
```

```
> seq(from=1.0, to=2.0, length=5)
```

```
[1] 1.00 1.25 1.50 1.75 2.00
```



Vectors in R

Comparing Numbers and Vectors

The comparison operators compare two values and return TRUE or FALSE.

For two vectors, the comparison is component-wise. The result is a logical vector.

```
> a = 3
```

```
> a == pi
```

```
[1] FALSE
```

```
> a != pi
```

```
[1] TRUE
```

```
> a < pi
```

```
[1] TRUE
```

```
> v = c( 3, pi, 4)
```

```
> w = c(pi, pi, pi)
```

```
> v == w
```

```
[1] FALSE TRUE FALSE
```



Vectors in R

Comparing Numbers and Vectors

```
any(v == w) Return TRUE if any element of v equals the same element in w
all(v == w) Return TRUE if all element of v equals the same element in w
any(v == num) Return TRUE if any element of v equals the value num
all(v == num) Return TRUE if all element of v equals the value num
```

```
> v = c(3,pi,4)
> w = c(pi,pi,pi)
> v != w
[1] TRUE FALSE TRUE
> v <= w
[1] TRUE TRUE FALSE
> v != 3
[1] FALSE TRUE TRUE
> any(v != 3)
[1] TRUE
```



Vectors in R

Choosing Vector Elements

```
vec[a] Select element a  
vec[a:b] Select elements a through b  
vec(c(a,b,c)) Select elements a,b, and c.  
vec(-a:-b) or vec(-(a:b)) exclude elements a through b
```

```
> fib = c(0,1,1,2,3,5,8,13,21,34)
```

```
> fib[6]
```

```
[1] 05
```

```
> fib[4:9]
```

```
[1] 2 3 5 8 13 21
```

```
> fib[c(1,2,4,8)]
```

```
[1] 0 1 2 13
```

```
> fib[-1:-5]
```

```
[1] 5 8 13 21 34
```



Vectors in R

More Complex Element Selections

```
Vec[vec<=a] Select elements that are not greater than a  
Vec[vec>=a&vec<=b] Select elements that are between a and b inclusively  
More complex element selections can be coded similarly.
```

```
> fib = c(0,1,1,2,3,5,8,13,21,34)  
> fib[fib < 10] # Use that vector to select elements less than 10  
[1] 0 1 1 2 3 5 8  
> fib[fib > mean(fib)] # Use that vector to select elements less than 10  
[1] 13 21 34  
> fib[fib<=5|fib>=21] # | means "or"  
[1] 0 1 1 2 3 5 21 34  
> fib[fib>5&fib<21] # & means "and"  
[1] 8 13  
> w=c(1,2,4,3,5,6,8,2,7,0)  
> fib[w>5]  
[1] 5 8 21
```



Vectors in R

Arithmetic Operations

`+, -, *, /, ^`: These arithmetic operations are component-wise

```
> v = c(11,12,13,14,15)
> w = c(1,2,3,4,5)
> v + w
[1] 12 14 16 18 20
> v - w
[1] 10 10 10 10 10
> v * w
[1] 11 24 39 56 75
> v / w
[1] 11.000000 6.000000 4.333333 3.500000 3.000000
> w ^ v
[1] 1 4096 1594323 268435456 30517578125
```



Vectors in R

Arithmetic Operations between a Vector and a Scalar

The operation is performed between every vector element and the scalar

```
> w + 2
```

```
[1] 3 4 5 6 7
```

```
> w - 2
```

```
[1] -1 0 1 2 3
```

```
> w * 2
```

```
[1] 2 4 6 8 10
```

```
> w / 2
```

```
[1] 0.5 1.0 1.5 2.0 2.5
```

```
> w ^ 2
```

```
[1] 1 4 9 16 25
```

```
> 2 ^ w
```

```
[1] 2 4 8 16 32
```



Vectors in R

Operation of Functions on Vectors

Functions operate on every element. The result is a vector.

```
> w
```

```
[1] 1 2 3 4 5
```

```
> sqrt(w)
```

```
[1] 1.000000 1.414214 1.732051 2.000000 2.236068
```

```
> log(w)
```

```
[1] 0.0000000 0.6931472 1.0986123 1.3862944 1.6094379
```

```
> sin(w)
```

```
[1] 0.8414710 0.9092974 0.1411200 -0.7568025 -0.9589243
```



Matrices

Definition: A matrix with m rows and n columns is called an $m \times n$ *matrix* or a *matrix of size* $m \times n$.

$$\mathbf{A} = [a_{ij}]_{m \times n} = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \cdots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{bmatrix}_{m \times n}$$

- **Row Vector:** A matrix that has exactly one row
- **Column Vector:** A matrix that has exactly one column
- **Remark:** Scalars can be treated as 1×1 matrices.

Equal Matrices: For $\mathbf{A} = [a_{ij}]_{m \times n}$ and $\mathbf{B} = [b_{ij}]_{m \times n}$

$$\mathbf{A} = \mathbf{B} \text{ if and only if } a_{ij} = b_{ij} \text{ for all } 1 \leq i \leq m, 1 \leq j \leq n$$



Special Matrices

Consider $\mathbf{A}_{m \times n} = [a_{ij}]$:

Zero matrix: If $a_{ij} = 0$ for all $1 \leq i \leq m, 1 \leq j \leq n$, $\mathbf{A}_{m \times n}$ is called a zero matrix and denoted with $\mathbf{O}_{m \times n}$.

Square Matrix: If $m = n$, then $\mathbf{A}_{m \times m}$ is called a square matrix and sometimes denoted with \mathbf{A}_m .

Diagonal Matrix: A square matrix $\mathbf{A}_{m \times m}$ is called a diagonal matrix if for all $i \neq j$, $a_{ij} = 0$.

Identity Matrix: A diagonal matrix with $a_{ii} = 1$. It is sometimes denoted with \mathbf{I}_m .



Matrices in R

Creating a Matrix

```
Mat=matrix(c(a1,...,an),nrow=,ncol=,byrow=)
```

`dim(Mat)` gives the matrix dimension

```
> mat=matrix(c(1,2,3,4),2,2,byrow=1)
```

```
> mat
```

```
      [,1] [,2]  
[1,]    1    2  
[2,]    3    4
```

Zero Matrix: `z= matrix(0,nrow=,ncol=)`

```
> matrix(0, ncol = 4, nrow = 6)
```

Diagonal Matrix: `d=diag(c(a1,...,an))`

```
> diag(c(1,4,6))
```

Identity Matrix: `I=diag(n)`

```
> diag(3)
```



Matrix Addition and Scalar Multiplication

For any two matrices $\mathbf{A} = [a_{ij}]_{m \times n}$ and $\mathbf{B} = [b_{ij}]_{m \times n}$:

- **Matrix Addition:** $\mathbf{A} + \mathbf{B} = [a_{ij} + b_{ij}]_{m \times n}$ (In R use: $\mathbf{A} + \mathbf{B}$)
- **Scalar Multiplication:** $k\mathbf{A} = [ka_{ij}]_{m \times n}$, $k \in R$ (In R use: $k * \mathbf{A}$)

Example:

$$\begin{bmatrix} -1 & 2 \\ 0 & 1 \end{bmatrix} + \begin{bmatrix} 1 & 3 \\ -1 & 2 \end{bmatrix} = \begin{bmatrix} -1+1 & 2+3 \\ 0-1 & 1+2 \end{bmatrix} = \begin{bmatrix} 0 & 5 \\ -1 & 3 \end{bmatrix}$$

$$3 \times \begin{bmatrix} -1 & 2 \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} -3 & 6 \\ 0 & 3 \end{bmatrix}$$



Matrix Addition and Scalar Multiplication

Properties

Matrix Addition:

- Commutativity: $A + B = B + A$
- Associativity: $(A + B) + C = A + (B + C)$
- Identity: $A + O = O + A = A$

Scalar Multiplication:

- Distributivity: $k(A + B) = kA + kB$
- Distributivity: $(k + l)A = kA + lA$
- Associativity: $k(lA) = (kl)A$

- Remark: $A - B = A + (-1)B = [a_{ij} - b_{ij}]_{m \times n}$



Transpose of A Matrix

For $\mathbf{A} \in \mathbb{R}^{m \times n}$, the matrix $\mathbf{B} \in \mathbb{R}^{n \times m}$ with $b_{ij} = a_{ji}$ is called the transpose of \mathbf{A} . We write $\mathbf{B} = \mathbf{A}^T$. (In R, use: `t(A)`)

Example:

$$\mathbf{A} = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{bmatrix} \Rightarrow \mathbf{A}^T = \begin{bmatrix} 1 & 4 \\ 2 & 5 \\ 3 & 6 \end{bmatrix}$$

Properties:

- $(\mathbf{A}^T)^T = \mathbf{A}$
- $(\mathbf{A} + \mathbf{B})^T = \mathbf{A}^T + \mathbf{B}^T$
- $\mathbf{A}^T = \mathbf{A}$, whenever \mathbf{A} symmetric

Example: If $\mathbf{A} = \begin{bmatrix} 6 & 0 \\ 2 & -1 \end{bmatrix}$ and $\mathbf{B} = \begin{bmatrix} 3 & -3 \\ 1 & 2 \end{bmatrix}$, find $\mathbf{A}^T - 2\mathbf{B}$

Solution:

$$\mathbf{A}^T - 2\mathbf{B} = \begin{bmatrix} 6 & 2 \\ 0 & -1 \end{bmatrix} - \begin{bmatrix} 6 & -6 \\ 2 & 4 \end{bmatrix} = \begin{bmatrix} 0 & 8 \\ -2 & -5 \end{bmatrix}$$



Matrix Multiplication

Matrix Multiplication: For any two matrices $\mathbf{A} = [a_{ij}]_{m \times n}$ and $\mathbf{B} = [b_{ij}]_{n \times p}$, the product \mathbf{AB} is the $m \times p$ matrix with the entry defined as

$$c_{ij} = \sum_{k=1}^n a_{ik}b_{kj} = a_{i1}b_{1j} + a_{i2}b_{2j} + \dots + a_{in}b_{nj} = (\text{row } i \text{ of } \mathbf{A}) \cdot (\text{column } j \text{ of } \mathbf{B})$$

In R use: `A %*% B`

Properties of Matrix Multiplication:

- $\mathbf{A(BC)} = (\mathbf{AB})\mathbf{C}$ (Associative property)
- $\mathbf{A(B + C)} = \mathbf{AB} + \mathbf{AC}$ (Distributive property)
- $\mathbf{(A + B)C} = \mathbf{AC} + \mathbf{BC}$ (Distributive property)
- $k(\mathbf{AB}) = \mathbf{A(kB)}$ (for any scalar k)
- $(\mathbf{AB})^T = \mathbf{B^T A^T}$
- Often $\mathbf{AB} \neq \mathbf{BA}$ (Not Commutative Law)



Example

Matrix Multiplication

a. $\begin{bmatrix} 1 & 2 & 3 \end{bmatrix} \begin{bmatrix} 4 \\ 5 \\ 6 \end{bmatrix} =$

b. $\begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} \begin{bmatrix} 1 & 6 \end{bmatrix} =$

c. $\begin{bmatrix} 1 & 3 & 0 \\ -2 & 2 & 1 \\ 1 & 0 & -4 \end{bmatrix} \begin{bmatrix} 1 & 0 & 2 \\ 5 & -1 & 3 \\ 2 & 1 & -2 \end{bmatrix} =$

*



Solution

$$\text{a. } \begin{bmatrix} 1 & 2 & 3 \end{bmatrix} \begin{bmatrix} 4 \\ 5 \\ 6 \end{bmatrix} = [32]$$

$$\text{b. } \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} \begin{bmatrix} 1 & 6 \end{bmatrix} = \begin{bmatrix} 1 & 6 \\ 2 & 12 \\ 3 & 18 \end{bmatrix}$$

$$\text{c. } \begin{bmatrix} 1 & 3 & 0 \\ -2 & 2 & 1 \\ 1 & 0 & -4 \end{bmatrix} \begin{bmatrix} 1 & 0 & 2 \\ 5 & -1 & 3 \\ 2 & 1 & -2 \end{bmatrix} = \begin{bmatrix} 16 & -3 & 11 \\ 10 & -1 & 0 \\ -7 & 4 & 10 \end{bmatrix}$$



Example

The Cost Vector

Given the price and the quantities, calculate the total cost.

$$\mathbf{P} = [2 \ 3 \ 4], \quad \mathbf{Q} = \begin{bmatrix} 7 \\ 5 \\ 11 \end{bmatrix} \begin{array}{l} \text{units of A} \\ \text{units of B} \\ \text{units of C} \end{array}$$

Solution: The cost vector is

$$\mathbf{PQ} = [2 \ 3 \ 4] \begin{bmatrix} 7 \\ 5 \\ 11 \end{bmatrix} = [2 \times 7 + 3 \times 5 + 4 \times 11] = [73]$$



Example

Linear Equation

Write the system of equations in matrix form by using matrix multiplication.

$$\begin{cases} 2x_1 + 5x_2 = 12 \\ 8x_1 + 3x_2 = 14 \end{cases}$$



Solution

Let

$$\mathbf{A} = \begin{bmatrix} 2 & 5 \\ 8 & 3 \end{bmatrix} \quad \mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \quad \mathbf{b} = \begin{bmatrix} 12 \\ 14 \end{bmatrix}$$

then the single matrix equation is

$$\mathbf{Ax} = \mathbf{b}$$

$$\begin{bmatrix} 2 & 5 \\ 8 & 3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 12 \\ 14 \end{bmatrix}$$



Matrix Reduction



Solving Systems by Reducing Matrices

Elementary Row Operations

1. Interchanging two rows of a matrix
2. Multiplying a row of a matrix by a nonzero number
3. Adding a multiple of one row of a matrix to a different row of that matrix

Notation

Corresponding Row Operation

$$\mathbf{R}_i \leftrightarrow \mathbf{R}_j$$

Interchange rows \mathbf{R}_i and \mathbf{R}_j .

$$k\mathbf{R}_i$$

Multiply row \mathbf{R}_i by a nonzero constant k .

$$k\mathbf{R}_i + \mathbf{R}_j$$

Add k times row \mathbf{R}_i to row \mathbf{R}_j (but leave \mathbf{R}_j unchanged).



Properties of a Reduced Matrix

A matrix is said to be a reduced matrix if all the following are true:

- All zero-rows at the bottom.
- For each nonzero-row, leading entry is 1 and all other entries in the column of the leading entry are 0.
- Leading entry in each row is to the right of the leading entry in any row above it.

Remark:

- A **leading entry** is the first nonzero entry in a nonzero-row
- Each matrix is **equivalent** to a **unique reduced matrix**.



Example

Reduced Matrices

For each of the following matrices, determine whether it is reduced or not reduced.

a. $\begin{bmatrix} 1 & 0 \\ 0 & 3 \end{bmatrix}$

d. $\begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$

b. $\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}$

e. $\begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}$

c. $\begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$

f. $\begin{bmatrix} 0 & 1 & 0 & 3 \\ 0 & 0 & 1 & 2 \\ 0 & 0 & 0 & 0 \end{bmatrix}$



Solution

- a. Not reduced because the leading entry in the second row is not 1
- b. Reduced
- c. Not reduced as the leading entry in the second row is not to the right of the leading entry in the first row
- d. Reduced
- e. Not reduced because the second row, which is a zero-row, is not at the bottom of the matrix
- f. Reduced



Reducing a Matrix

The Strategy for Reducing a Matrix

- Get the leading entry to be a 1 in the first row, the leading entry a 1 in the second row, and so on, until we arrive at a zero-row, if there are any.
- Work from left to right, because the leading entry in each row must be to the left of all other leading entries in the rows below it.

Example: Reduce the matrix

$$\begin{bmatrix} 0 & 0 & 1 & 2 \\ 3 & -6 & -3 & 0 \\ 6 & -12 & 2 & 11 \end{bmatrix}$$



Solution

$$\begin{bmatrix} 0 & 0 & 1 & 2 \\ 3 & -6 & -3 & 0 \\ 6 & -12 & 2 & 11 \end{bmatrix} \xrightarrow{R_1 \leftrightarrow R_2} \begin{bmatrix} 3 & -6 & -3 & 0 \\ 0 & 0 & 1 & 2 \\ 6 & -12 & 2 & 11 \end{bmatrix} \xrightarrow{\frac{1}{3}R_1} \begin{bmatrix} 1 & -2 & -1 & 0 \\ 0 & 0 & 1 & 2 \\ 6 & -12 & 2 & 11 \end{bmatrix} \xrightarrow{-6R_1+R_3} \begin{bmatrix} 1 & -2 & -1 & 0 \\ 0 & 0 & 1 & 2 \\ 0 & 0 & 8 & 11 \end{bmatrix}$$

$$\xrightarrow{R_2+R_1} \begin{bmatrix} 1 & -2 & 0 & 2 \\ 0 & 0 & 1 & 2 \\ 0 & 0 & -4 & 11 \end{bmatrix} \xrightarrow{-8R_2+R_3} \begin{bmatrix} 1 & -2 & 0 & 2 \\ 0 & 0 & 1 & 2 \\ 0 & 0 & 0 & -5 \end{bmatrix} \xrightarrow{-\frac{1}{5}R_3} \begin{bmatrix} 1 & -2 & 0 & 2 \\ 0 & 0 & 1 & 2 \\ 0 & 0 & 0 & 1 \end{bmatrix} \xrightarrow{-2R_3+R_1} \begin{bmatrix} 1 & -2 & 0 & 0 \\ 0 & 0 & 1 & 2 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

$$\xrightarrow{-2R_3+R_2} \begin{bmatrix} 1 & -2 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$



Solving a System by Reduction

Elimination does the same row operation to \mathbf{A} and to \mathbf{b} . We can include \mathbf{b} as an extra column and follow it through elimination. The matrix \mathbf{A} is augmented by an extra column \mathbf{b} :

$$[\mathbf{A}|\mathbf{b}] \rightarrow \cdots \rightarrow [\mathbf{R}|\mathbf{b}']$$

Example: By using matrix reduction, solve the system

$$\begin{cases} 2x + 3y = -1 \\ 2x + y = 5 \\ x + y = 1 \end{cases}$$



Solution

Reducing the augmented coefficient matrix of the system,

$$\left[\begin{array}{cc|c} 2 & 3 & -1 \\ 2 & 1 & 5 \\ 1 & 1 & 1 \end{array} \right] \xrightarrow{R_1 \leftrightarrow R_3} \left[\begin{array}{cc|c} 1 & 1 & 1 \\ 2 & 1 & 5 \\ 2 & 3 & -1 \end{array} \right] \xrightarrow{-2R_1 + R_2} \left[\begin{array}{cc|c} 1 & 1 & 1 \\ 0 & -1 & 3 \\ 2 & 3 & -1 \end{array} \right] \xrightarrow{-2R_1 + R_3} \left[\begin{array}{cc|c} 1 & 1 & 1 \\ 0 & -1 & 3 \\ 0 & 1 & -3 \end{array} \right] \xrightarrow{-R_2} \left[\begin{array}{cc|c} 1 & 1 & 1 \\ 0 & 1 & -3 \\ 0 & 1 & -3 \end{array} \right]$$

$$\xrightarrow{-R_2 + R_1} \left[\begin{array}{cc|c} 1 & 0 & 4 \\ 0 & 1 & -3 \\ 0 & 1 & -3 \end{array} \right] \xrightarrow{-R_2 + R_3} \left[\begin{array}{cc|c} 1 & 0 & 4 \\ 0 & 1 & -3 \\ 0 & 0 & 0 \end{array} \right]$$

We have $\begin{cases} x = 4 \\ y = -3 \end{cases}$



Example

Parametric Form of a Solution

Using matrix reduction, solve

$$\begin{cases} 2x_1 + 3x_2 + 2x_3 + 6x_4 = 10 \\ x_2 + 2x_3 + x_4 = 2 \\ 3x_1 - 3x_3 + 6x_4 = 9 \end{cases}$$



Solution

Reducing the augmented coefficient matrix of the system,

$$\left[\begin{array}{cccc|c} 2 & 3 & 2 & 6 & 10 \\ 0 & 1 & 2 & 1 & 2 \\ 3 & 0 & -3 & 6 & 9 \end{array} \right] \xrightarrow{\frac{1}{2}R_1} \left[\begin{array}{cccc|c} 1 & 3/2 & 1 & 3 & 5 \\ 0 & 1 & 2 & 1 & 2 \\ 3 & 0 & -3 & 6 & 9 \end{array} \right] \xrightarrow{-3R_1+R_3} \left[\begin{array}{cccc|c} 1 & 3/2 & 1 & 3 & 5 \\ 0 & 1 & 2 & 1 & 2 \\ 0 & -9/2 & -6 & -3 & -6 \end{array} \right]$$

$$\xrightarrow{-\frac{3}{2}R_2+R_1} \left[\begin{array}{cccc|c} 1 & 0 & -2 & 3/2 & 2 \\ 0 & 1 & 2 & 1 & 2 \\ 0 & -9/2 & -6 & -3 & -6 \end{array} \right] \xrightarrow{\frac{9}{2}R_2+R_3} \left[\begin{array}{cccc|c} 1 & 0 & -2 & 3/2 & 2 \\ 0 & 1 & 2 & 1 & 2 \\ 0 & 0 & 3 & 3/2 & 3 \end{array} \right] \xrightarrow{\frac{1}{3}R_3} \left[\begin{array}{cccc|c} 1 & 0 & -2 & 3/2 & 2 \\ 0 & 1 & 2 & 1 & 2 \\ 0 & 0 & 1 & 1/2 & 1 \end{array} \right]$$

$$\xrightarrow{2R_3+R_1} \left[\begin{array}{cccc|c} 1 & 0 & 0 & 5/2 & 4 \\ 0 & 1 & 2 & 1 & 2 \\ 0 & 0 & 1 & 1/2 & 1 \end{array} \right] \xrightarrow{-2R_3+R_2} \left[\begin{array}{cccc|c} 1 & 0 & 0 & 5/2 & 4 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 1/2 & 1 \end{array} \right]$$



Solution (cont'd)

This matrix is reduced and corresponds to the system with x_4 being a **free variable**. The answer can be written as:

$$\begin{cases} x_1 + \frac{5}{2}x_4 = 4 \\ x_2 = 0 \\ x_3 + \frac{1}{2}x_4 = 1 \end{cases} \Rightarrow \mathbf{x} = \begin{bmatrix} 4 \\ 0 \\ 1 \\ 0 \end{bmatrix} + x_4 \begin{bmatrix} -\frac{5}{2} \\ 0 \\ -\frac{1}{2} \\ 1 \end{bmatrix}$$

Definition: The variables whose columns do not have a leading entry are called **free variables**. The variables whose columns have a leading entry are called **pivot variables**.



Matrix Inverse



Inverses

The rank of a matrix \mathbf{A} is the number of its leading entries, k . It is equal to the number of nonzero rows of \mathbf{A} . It is also equal to the number of linearly independent columns as well as the number of the linearly independent rows of \mathbf{A} .

$\mathbf{A} \in \mathbb{R}^{n \times n}$ is called **full rank** if $k = n$. If \mathbf{A} is full-rank, there exists another matrix \mathbf{A}^{-1} such that

$$\mathbf{A}\mathbf{A}^{-1} = \mathbf{A}^{-1}\mathbf{A} = \mathbf{I}$$

\mathbf{A}^{-1} is called the **inverse of \mathbf{A}** and \mathbf{A} is called an **invertible matrix**.

Example: Let $\mathbf{A} = \begin{bmatrix} 1 & 2 \\ 3 & 7 \end{bmatrix}$ and $\mathbf{C} = \begin{bmatrix} 7 & -2 \\ -3 & 1 \end{bmatrix}$. Determine whether \mathbf{C} is an inverse of \mathbf{A} .

Solution:

$$\mathbf{CA} = \begin{bmatrix} 7 & -2 \\ -3 & 1 \end{bmatrix} \begin{bmatrix} 1 & 2 \\ 3 & 7 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = \mathbf{I}$$

Thus, matrix \mathbf{C} is an inverse of \mathbf{A} .



Method to Find the Inverse of a Matrix

If \mathbf{A} is an $n \times n$ matrix, form the $n \times (2n)$ matrix $[\mathbf{A}|\mathbf{I}]$ and perform elementary row operations until the first n columns form a reduced matrix. Assume that the result is $[\mathbf{R}|\mathbf{B}]$ so that we have

$$[\mathbf{A}|\mathbf{I}] \rightarrow \cdots \rightarrow [\mathbf{R}|\mathbf{B}]$$

When matrix is reduced,

- If $\mathbf{R} = \mathbf{I}$, \mathbf{A} is invertible and $\mathbf{A}^{-1} = \mathbf{B}$.
- If $\mathbf{R} \neq \mathbf{I}$, \mathbf{A} is not invertible, meaning that \mathbf{A}^{-1} does not exist.

Remark: To determine the inverse matrix in **R** you can use `solve(A)`



Example

Finding The Inverse by Reducing

Find the inverse of $A = \begin{bmatrix} 1 & 0 \\ 2 & 2 \end{bmatrix}$ if it exists.



Solution

$$[A|I] = \left[\begin{array}{cc|cc} 1 & 0 & 1 & 0 \\ 2 & 2 & 0 & 1 \end{array} \right] \xrightarrow{-2R_1+R_2} \left[\begin{array}{cc|cc} 1 & 0 & 1 & 0 \\ 0 & 2 & -2 & 1 \end{array} \right] \xrightarrow{\frac{1}{2}R_2} \left[\begin{array}{cc|cc} 1 & 0 & 1 & 0 \\ 0 & 1 & -1 & 1/2 \end{array} \right]$$

Matrix A is invertible where

$$A^{-1} = \begin{bmatrix} 1 & 0 \\ -1 & \frac{1}{2} \end{bmatrix}$$

Remark:

To create the segmented matrix $[A|I]$ in R type this code:

```
A=matrix(c(1,0,2,2),2,2,byrow=1)
```

```
I=diag(2)
```

```
AA=cbind(A,I)
```

To find A^{-1} directly type: `solve(A)`



Example 5

Using the Inverse to Solve a System

Solve the system by finding the inverse of the coefficient matrix.

$$\begin{cases} x_1 - 2x_3 = 1 \\ 4x_1 - 2x_2 + x_3 = 2 \\ x_1 + 2x_2 - 10x_3 = -1 \end{cases}$$



Solution

We have

$$\mathbf{A} = \begin{bmatrix} 1 & 0 & -2 \\ 4 & -2 & 1 \\ 1 & 2 & -10 \end{bmatrix}$$

For inverse,

$$\mathbf{A}^{-1} = \begin{bmatrix} -9 & 2 & 2 \\ 41 & 4 & 9 \\ -\frac{2}{2} & 4 & \frac{2}{2} \\ -5 & 1 & 1 \end{bmatrix}$$

The solution is given by $\mathbf{x} = \mathbf{A}^{-1}\mathbf{B}$:

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} -9 & 2 & 2 \\ 41 & 4 & 9 \\ -\frac{2}{2} & 4 & \frac{2}{2} \\ -5 & 1 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ -1 \end{bmatrix} = \begin{bmatrix} -7 \\ -17 \\ -4 \end{bmatrix}$$

Remark: To find the solution $\mathbf{x} = \mathbf{A}^{-1}\mathbf{B}$ in R use: `solve(A,b)`

