

Introduction to computational programming

Appendix 3

Matrix review and use in *R*

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Overview

This guide is intended to serve as a review of matrices and linear algebra and an introduction to their use in *R*.

1 Matrix review

1.1 Linear systems of equations

Consider the following system of equations.

$$\begin{aligned}3a + 2b &= -1 \\ a - b &= 3\end{aligned}$$

A more compact form of this system can be written (by omitting everything but the coefficients) as

$$A = \begin{bmatrix} 3 & 2 & -1 \\ 1 & -1 & 3 \end{bmatrix}$$

This matrix, A , would be referred to as a 2×3 matrix, since it has 2 rows and three columns. It is easy to solve a system of linear equations in this form (see ??).

1.2 Special matrices

There are several *special* types of matrices with which you should be familiar.

- **Square Matrix** - A square matrix is a matrix with an equal number of rows and columns.
- **Augmented square matrix** - A matrix with n rows and $n + 1$ columns (such as the first example shown above). This form is often used to represent a simultaneous system of linear equations.
- **Identity matrix** - The identity matrix (abbreviated I_m) is an $m \times m$ matrix that contains only zeros and ones with ones on the top-left to bottom-right diagonal (this is called the *main diagonal*). The following are examples of identity matrices

$$I_2 = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$I_3 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Identity matrices are special in that $\forall k \times k$ square matrices A , $AI_k = A$ (more on matrix multiplication later).

- **Diagonal matrix** - A diagonal matrix is a square matrix for which all entries not on the main diagonal are zero.
- **Upper/lower triangular matrix** - An *upper triangular matrix* is a square $k \times k$ matrix where $\forall m, n \in \mathbb{Z} : m > n \quad a_{mn} = 0$. Similarly a *lower triangular matrix* is a square $k \times k$ matrix for which $\forall m, n \in \mathbb{Z} : m < n \quad a_{mn} = 0$.

$$\begin{bmatrix} 1 & 5 & 16 \\ 0 & 2 & -5 \\ 0 & 0 & 2 \end{bmatrix} \qquad \begin{bmatrix} 0 & 0 & -3 \\ 0 & 1 & -1 \\ 14 & 3 & 5 \end{bmatrix}$$

- **Row-echelon form** - A matrix in *row-echelon* form is an augmented square matrix where $\forall m, n \in \mathbb{Z} : m > n \quad a_{mn} = 0$ and $\forall m, n \in \mathbb{Z} : m = n \quad a_{mn} = 1$.

$$\begin{bmatrix} 1 & 3 & 2 & 1 \\ 0 & 1 & -1 & 9 \\ 0 & 0 & 1 & -14 \end{bmatrix}$$

- **Reduced row-echelon form** - Reduced row-echelon form is a subset of row-echelon form with the additional constraint that, (for a $k \times k + 1$ matrix) $\forall m, n \in \mathbb{Z} : m \neq n$ and $n \neq k + 1 \quad a_{mn} = 0$.

$$\begin{bmatrix} 1 & 0 & 0 & 16 \\ 0 & 1 & 0 & -5 \\ 0 & 0 & 1 & -14 \end{bmatrix}$$

1.3 Operations on matrices

There are five “basic” operations which can be performed on matrices.

1. **Scalar addition** A scalar can be added to any matrix. The result is the matrix of scalar by element sums.

$$3 + \begin{bmatrix} 1 & 3 \\ 5 & 7 \end{bmatrix} = \begin{bmatrix} 4 & 6 \\ 8 & 10 \end{bmatrix}$$

2. **Matrix addition** Two matrices can be added if and only if they are of equivalent dimensions. The sum of two matrices is the matrix of elementwise sums.

$$\begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} + \begin{bmatrix} 2 & 4 \\ 6 & 8 \end{bmatrix} = \begin{bmatrix} 1+2 & 2+4 \\ 3+6 & 4+8 \end{bmatrix} = \begin{bmatrix} 3 & 6 \\ 9 & 12 \end{bmatrix}$$

3. **Scalar multiplication** Any matrix can be multiplied by a scalar to yield a matrix of scalar by element products.

$$3 \cdot \begin{bmatrix} 1 & 3 \\ 5 & 7 \end{bmatrix} = \begin{bmatrix} 3 & 9 \\ 15 & 21 \end{bmatrix}$$

4. **Matrix multiplication** A $m \times n$ matrix A and a $n \times l$ matrix B can be multiplied together to yield a $m \times l$ matrix as follows:

$$\begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \cdot \begin{bmatrix} 2 & 4 \\ 6 & 8 \end{bmatrix} = \begin{bmatrix} 1 \cdot 2 + 2 \cdot 6 & 1 \cdot 4 + 2 \cdot 8 \\ 3 \cdot 2 + 4 \cdot 6 & 3 \cdot 4 + 4 \cdot 8 \end{bmatrix} = \begin{bmatrix} 14 & 18 \\ 24 & 48 \end{bmatrix}$$

5. **Determinant** The *determinant* of a matrix A is the sum the products of the “top-left” to “bottom-right” diagonals minus the products of the the “top-right” to “bottom-left” diagonals. For a 2×2 matrix

$$\begin{vmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{vmatrix} = a_{11} \cdot a_{22} - a_{12} \cdot a_{21}$$

$$\begin{vmatrix} a & b & c \\ d & e & f \\ g & h & i \end{vmatrix} = aei + bfg + cdh - ceg - bdi - afh$$

1.4 Matrix transformations

There are three basic “elementary row operations” for modifying a (square) matrix.

- **Row-switching operations** Swap two rows of a matrix. This reverses the sign of the matrix determinant.

$$\begin{bmatrix} a & b & c \\ d & e & f \\ g & h & i \end{bmatrix} \Rightarrow \begin{bmatrix} a & b & c \\ g & h & i \\ d & e & f \end{bmatrix}$$

- **Row-multiplication operations** Multiply a row of a matrix by a constant. The effect of multiplying a single row of A by k causes the determinant of A to increase by a factor of k as well. $k|A| = |A'|$.

$$\begin{bmatrix} a & b & c \\ d & e & f \\ g & h & i \end{bmatrix} \Rightarrow \begin{bmatrix} a & b & c \\ kg & kh & ki \\ d & e & f \end{bmatrix}$$

- **Linear combinations of rows** - Given a matrix A with row vectors $a_m = (a_{m1}, a_{m2}, \dots)$ and $a_n = (a_{n1}, a_{n2}, \dots)$, one of rows m and n can be replaced with a linear combination of a_m and a_n . This does not change the determinant of the matrix.

$$\begin{bmatrix} a & b & c \\ d & e & f \\ g & h & i \end{bmatrix} \Rightarrow \begin{bmatrix} a & b & c \\ 3a - g & 3b - h & 3c - i \\ d & e & f \end{bmatrix}$$

Basic row operations can be used to transform a matrix into row-(echelon) form. For an augmented square matrix representing a set of simultaneous linear equations, this corresponds to solving the linear system by *Gauss-Jordan* elimination.

1.5 Matrix differential equations

Consider the system of differential equations

$$\begin{aligned} \frac{dy}{dx} &= 3y - 4z \\ \frac{dz}{dx} &= 4y - 7z \end{aligned}$$

with initial conditions $y(0) = 1$ and $z(0) = 3$

$$\begin{aligned} \begin{bmatrix} y'(x) \\ z'(x) \end{bmatrix} &= \begin{bmatrix} 3 & -4 \\ 4 & -7 \end{bmatrix} \begin{bmatrix} y \\ z \end{bmatrix} \\ \det \left(\begin{bmatrix} 3 & -4 \\ 4 & -7 \end{bmatrix} - \lambda \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \right) &= 0 \\ (3 - \lambda)(-7 - \lambda) + 16 &= 0 \\ \lambda^2 + 4\lambda - 5 &= 0 \\ \lambda &\in \{1, -5\} \end{aligned}$$

$$\begin{aligned} \begin{bmatrix} 3 & -4 \\ 4 & -7 \end{bmatrix} \begin{bmatrix} \alpha \\ \beta \end{bmatrix} &= \lambda_1 \begin{bmatrix} \alpha \\ \beta \end{bmatrix} \\ \begin{bmatrix} 3 & -4 \\ 4 & -7 \end{bmatrix} \begin{bmatrix} \alpha \\ \beta \end{bmatrix} &= \begin{bmatrix} \alpha \\ \beta \end{bmatrix} \\ \alpha &= 2\beta \\ \begin{bmatrix} 3 & -4 \\ 4 & -7 \end{bmatrix} \begin{bmatrix} \alpha \\ \beta \end{bmatrix} &= \lambda_2 \begin{bmatrix} \alpha \\ \beta \end{bmatrix} \\ \begin{bmatrix} 3 & -4 \\ 4 & -7 \end{bmatrix} \begin{bmatrix} \alpha \\ \beta \end{bmatrix} &= -5 \begin{bmatrix} \alpha \\ \beta \end{bmatrix} \\ 2\alpha &= \beta \end{aligned}$$

$$\begin{aligned} \begin{bmatrix} y \\ z \end{bmatrix} &= Ae^{\lambda_1 x} v_1 + Be^{\lambda_2 x} v_2 \\ y &= 2Ae^x + Be^{-5x} \\ z &= Ae^x + 2Be^{-5x} \\ 1 &= 2A + B \\ 3 &= A + 2B \\ B &= \frac{5}{3} \\ A &= -\frac{1}{3} \\ y &= -\frac{2}{3}e^x + \frac{5}{3}e^{-5x} \\ z &= -\frac{2}{3}e^x + \frac{10}{3}e^{-5x} \end{aligned}$$

Solving this system using sage:

```
x = var('x')
y = function('y', x)
z = function('z', x)
DE1 = diff(y, x) == 3 * y - 4 * z
DE2 = diff(z, x) == 4 * y - 7 * z
iVals = [0, 1, 3]
desolve_system([DE1, DE2], [y, z], iVals)
```

$$\left[y(x) = \frac{5}{3}e^{-5x} - \frac{2}{3}e^x, z(x) = \frac{10}{3}e^{-5x} - \frac{1}{3}e^x \right]$$

2 Creating matrices in *R*

1. `matrix()` - matrices are created in *R* using the `matrix()` function. The number of rows (or columns) can be specified using the `nrow=` and `ncol=` arguments. By default, values created as such are placed in the matrix in *column-major* format; you can change this with the `byrow=TRUE` parameter.

```
> M <- matrix(1:9, nrow=3);
> N <- matrix(1:9, nrow=3, byrow=TRUE);
> M
```

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

```
> N
```

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

```
> vec <- sample(1:100, 20);
> vec
```

```
[1] 59 19 7 53 17 50 41 80 66 55 60 62 92 88 77 15 58 29 6 72
```

```
> O <- matrix(vec, nrow=4, byrow=TRUE);
> O
```

```
      [,1] [,2] [,3] [,4] [,5]
[1,]    59    19     7    53    17
[2,]    50    41    80    66    55
[3,]    60    62    92    88    77
[4,]    15    58    29     6    72
```

```
> O[1,];
```

```
[1] 59 19 7 53 17
```

```
> O[,2];
```

```
[1] 19 41 62 58
```

```
> O[3,4];
```

```
[1] 88
```

Using sage:

```
A = matrix(QQ, [[1, 2, -3], [4, -5, -6], [7, 8, 19]])
A
```

$$\begin{pmatrix} 1 & 2 & -3 \\ 4 & -5 & -6 \\ 7 & 8 & 19 \end{pmatrix}$$

2. `diag()` - Diagonal matrices can be created using the `diag()` function. Identity matrices can be created using the function as well.

```
> I2 <- diag(1, nrow=2);
> I2;
```

```
      [,1] [,2]
[1,]     1     0
[2,]     0     1
```

```
> I5 <- diag(1, nrow=5);
> I5;
```

```
      [,1] [,2] [,3] [,4] [,5]
[1,]     1     0     0     0     0
[2,]     0     1     0     0     0
[3,]     0     0     1     0     0
[4,]     0     0     0     1     0
[5,]     0     0     0     0     1
```

```
> diag(c(1, 2, 4), nrow=3);
```

```
      [,1] [,2] [,3]
[1,]     1     0     0
[2,]     0     2     0
[3,]     0     0     4
```

3. `cbind()` - matrices can be “joined” columnwise using the `cbind()` command.
4. `rbind()` - matrices can be “joined” rowwise using the `rbind()` command.

```
> cbind(M, N);
```

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

```
> rbind(M, N);
```

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

2.1 Basic manipulation

1. `det()` Calculate the determinant of a matrix.

```
> set.seed(12345);
> M <- matrix(sample((1:10) - 5, 9), nrow=3);
> M
```

```

      [,1] [,2] [,3]
[1,]    3    4   -3
[2,]    5   -2   -1
[3,]    2   -4    1

```

```
> det(M);
```

```
[1] 2
```

```
A.determinant()
```

```
-484
```

2. + - Add either a scalar to a matrix *or* add two matrices together. *WARNING:*

```
> N <- matrix(sample((1:10) - 5, 9), nrow=3);
> N;
```

```

      [,1] [,2] [,3]
[1,]    5    1    0
[2,]   -4    4   -1
[3,]   -3    3    2

```

```
> 3 + N;
```

```

      [,1] [,2] [,3]
[1,]    8    4    3
[2,]   -1    7    2
[3,]    0    6    5

```

```
> M + N;    ## Probably not what you want
```

```

      [,1] [,2] [,3]
[1,]    8    5   -3
[2,]    1    2   -2
[3,]   -1   -1    3

```

```
> matrix(M + N, nrow=3, byrow=TRUE);
```

```

      [,1] [,2] [,3]
[1,]    8    1   -1
[2,]    5    2   -1
[3,]   -3   -2    3

```

```
3 * A
```

```

      (  3    6   -9 )
      ( 12  -15  -18 )
      ( 21   24   57 )

```

3. %*% - Calculate the matrix product of two matrices.

```
> M %*% N;
```



```

      [,1] [,2] [,3]
[1,]    8   10 -10
[2,]   36   -6    0
[3,]   23  -11    6

```

```
> N %*% M;
```

```

      [,1] [,2] [,3]
[1,]   20   18 -16
[2,]    6  -20    7
[3,]   10  -26    8

```

```

B = matrix(QQ, [[1,3, -5],[2, -4, -16], [1, -1, 11]])
A * B;

```

$$\begin{pmatrix} 2 & -2 & -70 \\ -12 & 38 & -6 \\ 42 & -30 & 46 \end{pmatrix}$$

4. %/% - Inverse matrix product.

```

> O <- N %/% M;
> O;

```

```

      [,1] [,2] [,3]
[1,]    1    0    0
[2,]   -1   -2    1
[3,]   -2   -1    2

```

```
> O %*% M;
```

```

      [,1] [,2] [,3]
[1,]    3    4   -3
[2,]  -11   -4    6
[3,]   -7  -14    9

```

5. t() - Transpose a matrix.

```
> M;
```

```

      [,1] [,2] [,3]
[1,]    3    4   -3
[2,]    5   -2   -1
[3,]    2   -4    1

```

```
> t(M) ;
```

```

      [,1] [,2] [,3]
[1,]    3    5    2
[2,]    4   -2   -4
[3,]   -3   -1    1

```

```
A.transpose()
```

$$\begin{pmatrix} 1 & 4 & 7 \\ 2 & -5 & 8 \\ -3 & -6 & 19 \end{pmatrix}$$

6. `library(Matrix)` - The `library()` command loads an *R* package - a collection of functions and other *R* objects. The *Matrix* package contains a large number of functions for the efficient manipulation of matrices.

2.2 Letting *R* do the heavy lifting

Using *R* invert a matrix

Not all matrices have an inverse. A matrix is *invertible* if and only if the determinant of the matrix is nonzero. The inverse of a square matrix A is the matrix A^{-1} such that $AA^{-1} = I$. *R* can calculate the inverse of a matrix as follows.

```
> A <- matrix(c(1, 5, 2, 8), nrow=2, byrow=TRUE);
> Ainv <- solve(A);
> A
```

```
      [,1] [,2]
[1,]    1    5
[2,]    2    8
```

```
> Ainv
```

```
      [,1] [,2]
[1,]   -4  2.5
[2,]    1 -0.5
```

```
> A %*% Ainv
```

```
      [,1] [,2]
[1,]    1    0
[2,]    0    1
```

Using sage instead

```
A = matrix(QQ, [[1, 5], [2, 8]]);
A.inverse();
```

$$\begin{pmatrix} -4 & \frac{5}{2} \\ 1 & -\frac{1}{2} \end{pmatrix}$$

Similarly, the matrix equation $\mathbf{A} \cdot \mathbf{X} = \mathbf{B}$ for a given matrix A and column vector B can be solved for vector X using *R* as shown below. This method fails when the matrix is not invertible.

```
> A
```

```

      [,1] [,2]
[1,]    1    5
[2,]    2    8

```

```

> B <- matrix(c(3,4), nrow=2);
> x <- solve(A, B);
> x;

```

```

      [,1]
[1,]   -2
[2,]    1

```

```

> A %*% x;

```

```

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

```

Using sage ...

```

B = matrix(QQ, [[3], [4]])
A.solve_right(B)

```

$$\begin{pmatrix} -2 \\ 1 \end{pmatrix}$$

This corresponds to solution to the simultaneous system of linear equations described in the beginning. Moreover, the “failure” which occurs when the determinant is 0 (i.e. A is not invertible) implies that the equation lacks a unique solution (i.e. the system of equations is underdetermined).