

MATRICES AND MATRIX OPERATIONS V

EIGENVALUES AND EIGENVECTORS

Linear Systems of the Form

$$Ax = \lambda x$$

Many applications of linear algebra are concerned with systems of n linear equations in n unknowns that are expressed in the form

$$Ax = \lambda x$$

where λ is a scalar. Such systems are really homogeneous linear systems in disguise, since the given system can be written as

$\lambda x - Ax = 0$ or, by inserting an identity matrix and factoring, as

$$(\lambda I - A)x = 0$$

Example: Finding $(\lambda I - A)$

The linear system

$$\begin{aligned}x_1 + 3x_2 &= \lambda x_1 \\4x_1 + 2x_2 &= \lambda x_2\end{aligned}$$

can be written in matrix form as

$$\begin{bmatrix} 1 & 3 \\ 4 & 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \lambda \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

$$A = \begin{bmatrix} 1 & 3 \\ 4 & 2 \end{bmatrix} \quad \text{and} \quad x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

This system can be written as

$$\lambda \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} - \begin{bmatrix} 1 & 3 \\ 4 & 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

or

$$\lambda \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} - \begin{bmatrix} 1 & 3 \\ 4 & 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

or

$$\begin{bmatrix} \lambda - 1 & -3 \\ -4 & \lambda - 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$\lambda I - A = \begin{bmatrix} \lambda - 1 & -3 \\ -4 & \lambda - 2 \end{bmatrix}$$

The primary problem of interest for linear systems of the form $(\lambda I - A)x = 0$ is to determine those values of λ for which the system has a nontrivial solution; such a value of λ is called **characteristic value** or an **eigenvalue** of A . If λ is eigenvalue of A , then the nontrivial solution of $(\lambda I - A)x = 0$ are called **eigenvectors** of A corresponding to λ .

Definition

If A is $n \times n$ matrix, then a nonzero vector x in R^n is called an **eigenvector** of A if Ax is a scalar multiple of x ; that is if

$$Ax = \lambda x$$

for some scalar λ . The scalar λ is called an **eigenvalue** of A , and x said to be an **eigenvector** of A corresponding to λ .

Example: Eigenvector of a 2x2 Matrix

The vector $x = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$ is an eigenvector of

$$A = \begin{bmatrix} 3 & 0 \\ 8 & -1 \end{bmatrix}$$

corresponding to the eigenvalue $\lambda = 3$, since

$$Ax = \begin{bmatrix} 3 & 0 \\ 8 & -1 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \end{bmatrix} = \begin{bmatrix} 3 \\ 6 \end{bmatrix} = 3 \begin{bmatrix} 1 \\ 2 \end{bmatrix} = 3x$$

To find the eigenvalues of $n \times n$ matrix A , we rewrite

$$Ax = \lambda x \quad \text{as} \quad Ax = \lambda Ix \quad \text{or equivalently,}$$

$$(\lambda I - A)x = 0$$

For λ to be an eigenvalue, there must be a nonzero solution of this equation.

Theorem Equation $(\lambda I - A)x = 0$ has a nonzero solution if and only if

$$\det(\lambda I - A) = 0$$

This is called the characteristic equation of A ; the scalars satisfying this equation are the eigenvalues of A .

Remark A $n \times n$ homogeneous system of linear equations has a unique solution (the trivial solution) if and only if its determinant is non-zero. If this determinant is zero, then the system has an infinite number of solutions.

When expanded the determinant, $\det(\lambda I - A) = 0$ is always a polynomial p in λ , called the **characteristic polynomial** of A :

If A is an $n \times n$ matrix, then the characteristic polynomial of A has degree n and the coefficient of λ^n is 1; that is, the characteristic polynomial $p(\lambda)$ of an $n \times n$ matrix has the form

$$p(\lambda) = \det(\lambda I - A) = \lambda^n + c_1 \lambda^{n-1} + \dots + c_n$$

It follows from the Fundamental Theorem of Algebra that the **characteristic equation**

$$\lambda^n + c_1 \lambda^{n-1} + \dots + c_n = 0$$

has at most n distinct solutions, so an $n \times n$ matrix has at most n distinct eigenvalues.

Example: Eigen values and Eigenvectors

Find the eigenvalues and corresponding eigenvectors of the matrix

$$A = \begin{bmatrix} 1 & 3 \\ 4 & 2 \end{bmatrix}$$

The characteristic equation of A is

$$\det(\lambda I - A) = \begin{vmatrix} \lambda & 0 \\ 0 & \lambda \end{vmatrix} - \begin{vmatrix} 1 & 3 \\ 4 & 2 \end{vmatrix} = \begin{vmatrix} \lambda - 1 & -3 \\ -4 & \lambda - 2 \end{vmatrix} = 0$$

or

$$\lambda^2 - 3\lambda - 10 = 0$$

The factored form of this equation is

$$(\lambda + 2)(\lambda - 5) = 0, \text{ so the eigenvalues}$$

of A are $\lambda = -2, \lambda = 5$.

By definition,

$$x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

is an eigenvector of A if and only if x is a nontrivial solution of $(\lambda I - A)x = 0$; that is,

$$\begin{bmatrix} \lambda - 1 & -3 \\ -4 & \lambda - 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

If $\lambda = -2$, then

$$\begin{bmatrix} -3 & -3 \\ -4 & -4 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

Solving this system yields $x_1 = -x_2$ $x_2 = t$ so the eigenvectors corresponding to $\lambda = -2$ are the nonzero solutions of the form

$$x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} -t \\ t \end{bmatrix} = t \begin{bmatrix} -1 \\ 1 \end{bmatrix}$$

If $\lambda = 5$, then

$$\begin{bmatrix} 4 & -3 \\ -4 & 3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

Solving this system yields $x_1 = \frac{3}{4}t, x_2 = t$ so the eigenvectors corresponding to $\lambda = 5$ are the nonzero solutions of the form

$$x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} \frac{3}{4}t \\ t \end{bmatrix} = t \begin{bmatrix} \frac{3}{4} \\ 1 \end{bmatrix}$$

EIGEN VALUES AND EIGENVECTORS IN MATLAB

```
>> A=[1 3; 4 2]
```

```
A =
```

```
1 3
```

```
4 2
```

```
>> [v,d]=eig(A)
```

```
v =
```

```
-0.7071 -0.6000
```

```
0.7071 -0.8000
```

```
d =
```

```
-2 0
```

```
0 5
```

```
>> vv=[-1 1]
```

```
vv = -1 1
```

```
>> vv=vv'
```

```
vv =
```

```
-1
```

```
1
```

```
>> nor=norm(vv)
```

```
nor =
```

```
1.4142
```

```
>> vv=vv/nor
```

```
vv =
```

```
-0.7071
```

```
0.7071
```

```
>> vv'*vv
```

```
ans =
```

```
1.0000
```

Example: Eigen values and Eigenvectors

Find the eigenvalues and corresponding eigenvectors of the matrix

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

$$\det(\lambda I - A) = \begin{vmatrix} \lambda - 2 & 1 \\ 1 & \lambda - 2 \end{vmatrix} = 0$$

or

$$(\lambda - 2)^2 - 1 = 0$$

The roots of this equation $\lambda = 3, \lambda = 1$.

Consider the first Eigen value

$$\lambda = 3$$

$$\begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

Both rows of this matrix equation reduce to

$$x_1 = -x_2$$

$$x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} -t \\ t \end{bmatrix} = t \begin{bmatrix} -1 \\ 1 \end{bmatrix}$$

For $t=1$ the corresponding eigenvector is

$$\begin{bmatrix} -1 \\ 1 \end{bmatrix}.$$

Consider the second Eigen value

$$\lambda = 1$$

$$\begin{bmatrix} -1 & 1 \\ 1 & -1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

Similar process leads to $x_1 = x_2$

$$x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} t \\ t \end{bmatrix} = t \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

For $t=1$ the corresponding eigenvector is

$$\begin{bmatrix} 1 \\ 1 \end{bmatrix}.$$

```
>> A=[2 1; 1 2]
```

```
A =
```

```
    2    1
```

```
    1    2
```

```
>> [v,d]=eig(A)
```

```
v =
```

```
 -0.7071    0.7071
```

```
  0.7071    0.7071
```

```
d =
```

```
    1    0
```

```
    0    3
```

Example: Eigen values of a 3x3 Matrix

Find the eigenvalues of

$$A = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 4 & -17 & 8 \end{bmatrix}$$

The characteristic polynomial of A is

$$\det(\lambda I - A) = \det \begin{bmatrix} \lambda & -1 & 0 \\ 0 & \lambda & -1 \\ -4 & 17 & \lambda - 8 \end{bmatrix} = \lambda^3 - 8\lambda^2 + 17\lambda - 4$$

The Eigenvalues of A must therefore satisfy the cubic equation

$$\lambda^3 - 8\lambda^2 + 17\lambda - 4 = 0$$

$$(\lambda - 4)(\lambda^2 - 4\lambda + 1) = 0$$

$$\lambda = 4, \lambda = 2 + \sqrt{3}, \lambda = 2 - \sqrt{3}$$

```
>> A=[0 1 0;0 0 1;4 -17,8]
A =
    0     1     0
    0     0     1
    4   -17     8
>> [v,d]=eig(A)
v =
   -0.9636   -0.0692   -0.0605
   -0.2582   -0.2582   -0.2421
   -0.0692   -0.9636   -0.9684
d =
    0.2679         0         0
         0    3.7321         0
         0         0    4.0000
>>
```

If $\lambda = 4$

$$\begin{bmatrix} 4 & -1 & 0 \\ 0 & 4 & -1 \\ -4 & 17 & -4 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} \quad \begin{bmatrix} 4 & -1 & 0 \\ 0 & 4 & -1 \\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

$$\begin{aligned} 4x_1 - x_2 &= 0 \\ 4x_2 - x_3 &= 0 \end{aligned}$$

Eigenvectors of A corresponding to $\lambda = 4$

$$x = t \begin{bmatrix} -1/16 \\ -1/4 \\ -1 \end{bmatrix}$$

Example:

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

The characteristic equation of matrix A is

$$\lambda^3 - 5\lambda^2 + 8\lambda - 4 = 0$$

Or in factored form,

$$(\lambda - 1)(\lambda - 2)^2 = 0$$

Thus the eigenvalues of A are $\lambda_1 = 1$ and $\lambda_{2,3} = 2$

By definition,

$$x = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}$$

is an eigenvector of A corresponding to λ if and only if

x is a nontrivial solution of $(\lambda I - A)x = 0$ that is, of

$$\begin{bmatrix} \lambda & 0 & 2 \\ -1 & \lambda - 2 & -1 \\ -1 & 0 & \lambda - 3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

If $\lambda = 1$

$$\begin{bmatrix} 1 & 0 & 2 \\ -1 & -1 & -1 \\ -1 & 0 & -2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

Solving this system yields (verify)

$$x_1 = -2s, x_2 = s, x_3 = s$$

Thus the eigenvectors of A corresponding to $\lambda = 1$ are the nonzero vectors of the form

$$\begin{bmatrix} -2s \\ s \\ s \end{bmatrix} = s \begin{bmatrix} -2 \\ 1 \\ 1 \end{bmatrix}$$

If $\lambda = 2$

$$\begin{bmatrix} 2 & 0 & 2 \\ -1 & 0 & -1 \\ -1 & 0 & -1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

Solving this system using Gaussian elimination yields (verify)

$$x_1 = -s, x_2 = t, x_3 = s$$

Thus the eigenvectors of A corresponding to $\lambda = 2$ are the nonzero vectors of the form

$$x = \begin{bmatrix} -s \\ t \\ s \end{bmatrix}$$

$$x = \begin{bmatrix} -s \\ t \\ s \end{bmatrix} = \begin{bmatrix} -s \\ 0 \\ s \end{bmatrix} + \begin{bmatrix} 0 \\ t \\ 0 \end{bmatrix} = s \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix} + t \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$$

Since $\begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix}$ and $\begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$ are **orthogonal**, these vectors

form a basis for the **eigenspace** corresponding to $\lambda = 2$.

```
A=[0 0 -2;1 2 1;1 0 3]
```

```
A =
```

```
0 0 -2
```

```
1 2 1
```

```
1 0 3
```

```
>> [v,d]=eig(A)
```

```
v =
```

```
0 -0.8165 0.7071
```

UNIT VECTORS

```
1.0000 0.4082 0
```

```
0 0.4082 -0.7071
```

```
d =
```

```
2 0 0
```

```
0 1 0
```

```
0 0 2
```

```
>>
```

Example: Eigenvalues of an Upper Triangular Matrix

Find the eigenvalues of the upper triangular matrix

$$A = \begin{bmatrix} a_{11} & a_{12} & a_{13} & a_{14} \\ 0 & a_{22} & a_{23} & a_{24} \\ 0 & 0 & a_{33} & a_{34} \\ 0 & 0 & 0 & a_{44} \end{bmatrix}$$

$$\begin{aligned} \det(\lambda I - A) &= \det \begin{bmatrix} \lambda - a_{11} & -a_{12} & -a_{13} & -a_{14} \\ 0 & \lambda - a_{22} & -a_{23} & -a_{24} \\ 0 & 0 & \lambda - a_{33} & -a_{34} \\ 0 & 0 & 0 & \lambda - a_{44} \end{bmatrix} \\ &= (\lambda - a_{11})(\lambda - a_{22})(\lambda - a_{33})(\lambda - a_{44}) \end{aligned}$$

Thus, the characteristic equation is

$$(\lambda - a_{11})(\lambda - a_{22})(\lambda - a_{33})(\lambda - a_{44}) = 0$$

and the eigenvalues are

$$\lambda_1 = a_{11}, \lambda_2 = a_{22}, \lambda_3 = a_{33}, \lambda_4 = a_{44}$$

which are precisely the diagonal entries of A.

Theorem: *If A is an $n \times n$ triangular matrix (upper triangular, lower triangular, or diagonal), then the eigenvalues of A are the entries on the main diagonal of A .*

Example: Eigenvalues of a Lower Triangular Matrix

By inspection, the eigenvalues of the lower triangular matrix

$$A = \begin{bmatrix} \frac{1}{2} & 0 & 0 \\ -1 & \frac{2}{3} & 0 \\ 5 & -8 & -\frac{1}{4} \end{bmatrix}$$

are $\lambda = \frac{1}{2}, \lambda = \frac{2}{3}, \lambda = -\frac{1}{4}.$

Complex Eigenvalues

It is possible for the characteristic equation of a matrix with real entries to have complex solutions. In fact, because the eigenvalues of an $n \times n$ matrix are the roots of a polynomial of precise degree n , every $n \times n$ matrix has exactly n eigenvalues if we count them as we count the roots of a polynomial (meaning that they may be repeated, and may occur in complex conjugate pairs).

Example: Complex eigenvalues

$$A = \begin{bmatrix} -2 & -1 \\ 5 & 2 \end{bmatrix}$$

The characteristic polynomial of the given matrix is

$$\det(\lambda I - A) = \det \begin{bmatrix} \lambda + 2 & 1 \\ -5 & \lambda - 2 \end{bmatrix} = \lambda^2 + 1$$

So the characteristic equation is $\lambda^2 + 1 = 0$, the solutions of which are the imaginary numbers

$$\lambda = i, \lambda = -i.$$

Thus we are forced to consider complex eigenvalues, even for real matrices.

Eigenvalues and Eigenvectors of the power of A

Once the eigenvalues and eigenvectors of a matrix A are found, it is simple matter to find the eigenvalues and eigenvectors of any positive integer power of A ; for example, if λ is an eigenvalue of A and x is a corresponding eigenvector, then

$$A^2 x = A(Ax) = A(\lambda x) = \lambda(\lambda x) = \lambda^2 x$$

which shows that λ^2 is an eigenvalue of A^2 and that x is a corresponding eigenvector.

Theorem: *If k is a positive integer, λ is an eigenvalue of a matrix A , and x is corresponding eigenvector, then λ^k is an eigenvalue of A^k and x is a corresponding eigenvector.*

Example:

We showed that the eigenvalues of

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

are $\lambda = 2$ and $\lambda = 1$.

$\lambda = 2^5 = 32$ and $\lambda = 1^5 = 1$ are eigenvalues of A^5 .

$$A^5 = \begin{bmatrix} -30 & 0 & -62 \\ 31 & 32 & 31 \\ 31 & 0 & 63 \end{bmatrix}$$

```
>> A=[0 0 -2; 1 2 1; 1 0 3]
```

```
A =
```

```
0    0   -2
```

```
1    2    1
```

```
1    0    3
```

```
>> A5=A^5
```

```
A5 =
```

```
-30    0  -62
```

```
31   32   31
```

```
31    0   63
```



```
>> [v,d]=eig(A5)
```

```
v =
```

```
    0  -0.8165  0.7071
```

```
  1.0000  0.4082    0
```

```
    0  0.4082 -0.7071
```

```
d =
```

```
   32    0    0
```

```
    0    1    0
```

```
    0    0   32
```

```
>>
```

We also showed that

$\begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix}$ and $\begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$ are eigenvectors of A

corresponding to eigenvalue $\lambda = 2$. They are also eigenvectors of A^5 . Similarly, the eigenvector

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

of A corresponding to the eigenvalue $\lambda = 1$ is also an eigenvector of A^5 .

Theorem: A square matrix A is invertible if and only if $\lambda = 0$ is not an eigenvalue of A .

Assume that A is an $n \times n$ matrix and observe first that $\lambda = 0$ is a solution of the characteristic equation

$$\lambda^n + c_1 \lambda^{n-1} + \dots + c_n = 0$$

if and only if the constant term c_n is zero. Thus it suffices to prove that A is invertible if and only if $c_n \neq 0$. But

$$\det(\lambda I - A) = \lambda^n + c_1 \lambda^{n-1} + \dots + c_n$$

Or, on setting $\lambda = 0$,

$$\det(-A) = c_n \quad \text{or} \quad (-1)^n \det(A) = c_n$$

It follows from the last equation that $\det(A) = 0$ if and only if $c_n = 0$, and this in turn implies that A is invertible if and only if $c_n \neq 0$.

Example:

We showed that the eigenvalues of

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

are $\lambda_1 = 1$ and $\lambda_{2,3} = 2$

The matrix A is invertible since it has eigenvalues $\lambda = 2$ and $\lambda = 1$ neither of which is zero.

```
>> A=[0 0 -2; 1 2 1; 1 0 3]
```

```
A =
```

```
    0    0   -2
```

```
    1    2    1
```

```
    1    0    3
```

```
>> inv(A)
```

```
ans =
```

```
    1.5000    0    1.0000
```

```
   -0.5000    0.5000   -0.5000
```

```
   -0.5000    0    0
```

Eigenvalues Relationships

If A is a square n -by- n matrix with real or complex entries and if $\lambda_i, i=1..n$ are the (complex and distinct) eigenvalues of A , then

$$\text{trace}(A) = \sum_{i=1}^n \lambda_i.$$

The product of the eigenvalues of a square matrix is equal to the determinant of that matrix.

$$\det(A) = \prod_{i=1}^n \lambda_i$$

```
>> A=[0 0 -2;1 2 1;1 0 3]
A =
    0    0   -2
    1    2    1
    1    0    3
>> det(A)
ans = 4
>> eig(A)
ans =
    2
    1
    2
>> trace(A)
ans = 5
```

Example:

$$A = \begin{bmatrix} 1 & 3 \\ 4 & 2 \end{bmatrix}$$

The characteristic equation of A is

$$\det(\lambda I - A) = \begin{vmatrix} \lambda - 1 & -3 \\ -4 & \lambda - 2 \end{vmatrix} = 0$$

or

$$\lambda^2 - 3\lambda - 10 = 0$$

The factored form of this equation is

$$(\lambda + 2)(\lambda - 5) = 0, \text{ so the eigenvalues}$$

of A are $\lambda = -2, \lambda = 5$.

```
>> A=[1 3;4 2]
A =
     1     3
     4     2
>> eig(A)
ans =
    -2
     5
>> det(A)
ans = -10
>> trace(A)
ans = 3
```

Example:

$$A = \begin{bmatrix} 3 & 6 \\ 1 & 4 \end{bmatrix}.$$

Let's find both of the eigenvalues of the matrix

$$\begin{aligned} A - \lambda I &= \begin{bmatrix} 3 - \lambda & 6 \\ 1 & 4 - \lambda \end{bmatrix} \\ \det(A - \lambda I) &= (3 - \lambda)(4 - \lambda) - 6 \\ &= \lambda^2 - 7\lambda + 6 \\ &= (\lambda - 6)(\lambda - 1) \end{aligned}$$

Therefore, $\lambda = 6$ or $\lambda = 1$. We now know our eigenvalues.

$$\begin{aligned} \text{Product of eigenvalues} &= \det(A) \\ 6 * 1 &= 12 - 6 \\ 6 &= 6 \end{aligned}$$

Example

$$A = \begin{bmatrix} 2 & 6 \\ 2 & -2 \end{bmatrix},$$

For

$$\begin{aligned} \text{Product of eigenvalues} &= \det(A) \\ 4 * (-4) &= -4 - 12 \\ -16 &= -16 \end{aligned}$$

$$\begin{vmatrix} 2-\lambda & 6 \\ 2 & -2-\lambda \end{vmatrix} = (2-\lambda)(-2-\lambda) - 12 = 0$$

$$(\lambda^2 - 16) = 0 \quad \lambda_{1,2} = \pm 4 \quad \lambda_1 \lambda_2 = -16$$

$$\text{Trace}(A) = 0 = \sum_{i=1}^2 \lambda_i = 4 - 4 = 0$$

Example

$$A = \begin{bmatrix} 1 & 2 & 3 & 4 \\ 5 & 6 & 7 & 9 \\ 6 & 8 & 3 & 4 \\ 9 & 8 & 2 & 4 \end{bmatrix}$$

```
A=[1 2 3 4; 5 6 7 9; 6 8 3 4; 9 8 2 4]
```

```
A =
```

```
     1     2     3     4
     5     6     7     9
     6     8     3     4
     9     8     2     4
```

```
>> [V,D]=eig(A)
```

```
V =
```

```
    0.2499    0.3566    0.4375    0.0725
    0.6496    0.5444   -0.3859   -0.1563
    0.5016   -0.4121    0.6595   -0.7518
    0.5137   -0.6377   -0.4741    0.6365
```

```
D =
```

```
   20.4460         0         0         0
         0   -6.5683         0         0
         0         0   -0.5759         0
         0         0         0    0.6982
```

```
>> trace(D)
```

```
ans = 14.0000
```

```
>> TRACE(A)
```

```
ans = 14
```

```
>> det(A)
```

```
ans = 54
```


Square root of a matrix

A matrix B is said to be a square root of A if the [matrix product](#) $B \cdot B$ is equal to A ^[1].

[\[edit\]](#) Computation by diagonalization

The square root of a [diagonal matrix](#) D is formed by taking the square root of all the entries on the diagonal. This suggests the following methods for general matrices:

An $n \times n$ matrix A is [diagonalizable](#) if there is a matrix V such that $D = V^{-1}AV$ is a [diagonal matrix](#). This happens if and only if A has n [eigenvectors](#) which constitute a basis for \mathbf{C}^n ; in this case, V can be chosen to be the matrix with the n eigenvectors as columns.

Now, $A = VDV^{-1}$, and hence the square root of A is

$$A^{1/2} = VD^{1/2}V^{-1}.$$

This approach works only for [diagonalizable matrices](#). For non-diagonalizable matrices one can calculate the [Jordan normal form](#) followed by a series expansion, similar to the approach described in [logarithm of a matrix](#).