Eigenvalues

Eigenvalue Decomposition

An eigenvalue and eigenvector of a square matrix A are, respectively, a scalar λ and a nonzero vector v that satisfy

$$Av = \lambda v$$

With the eigenvalues on the diagonal matrix Λ and the corresponding eigenvectors forming the columns of matrix V, you have

$$AV = V \wedge$$

If *V* is nonsingular, this becomes the eigenvalue decomposition

$$A = V \wedge V^{-1}$$

A good example is the coefficient matrix of the differential equation dx/dt = Ax.

$$A = \begin{bmatrix} 0 & -6 & -1 \\ 6 & 2 & -16 \\ -5 & 20 & -10 \end{bmatrix}$$

$$\begin{array}{ccccccc}
A & = & 3 \times 3 & & & & \\
0 & & -6 & & -1 & \\
6 & & 2 & & -16 & \\
-5 & & 20 & & -10 & & \\
\end{array}$$

The solution to this equation is expressed in terms of the matrix exponential $x(t) = e^{tA}x(0)$. The statement

```
lambda = eig(A)
```

```
lambda = 3x1 complex
-3.0710 + 0.0000i
-2.4645 +17.6008i
-2.4645 -17.6008i
```

Produces a column vector containing the eigenvalues of A. For this matrix, the eigenvalues are complex.

The real part of each of the eigenvalues is negative, so $e^{\lambda t}$ approaches zero as t increases. The nonzero imaginary part of rtwo of the eigenvalues, $\pm w$, contributes to oscillary component, sin(wt), to the slution of the differential equation.

With two output arguments, eig computes the eigenvectors and stores the eigenvalue in a diagonal matrix:

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

```
0.0000 + 0.0000i -2.4645 +17.6008i 0.0000 + 0.0000i
0.0000 + 0.0000i 0.0000 + 0.0000i -2.4645 -17.6008i
```

There s a double eigenvalue at $\lambda = 1$. The first and second columns of V are the same. For this matrix, a full set of linearly independent eigenvectors does not exist.

Schur Decomposition

Many advanced matrix computations do not require eigenvalue decomposition. They are based, instead, on the Schur decomposition

```
A = USU^{\prime}
```

where U is an orthogonal matrix and S is a block upper-triangular matrix with 1-by-1 and 2-by-2 blocks on the diagonal. The eigenvalues are revealed by the diagonal elements and blocks of S, while the columns U provide an orthogonal basis, which has much better numerical properties than a set of eigenvectors.

For example, compare the eigenvalue and Schur decomposition of this defective matrix:

```
A = [6]
          12
                 19
      -9 - 20 - 33
      4
            9
                 15];
[V,D]=eig(A)
V = 3 \times 3 complex
 -0.4741 + 0.0000i -0.4082 - 0.0000i -0.4082 + 0.0000i
  0.8127 + 0.0000i 0.8165 + 0.0000i
                                        0.8165 + 0.0000i
 -0.3386 + 0.0000i -0.4082 + 0.0000i -0.4082 - 0.0000i
D = 3 \times 3 complex
  -1.0000 + 0.0000i 0.0000 + 0.0000i
                                        0.0000 + 0.0000i
  0.0000 + 0.0000i 1.0000 + 0.0000i
                                        0.0000 + 0.0000i
   0.0000 + 0.0000i
                     0.0000 + 0.0000i
                                        1.0000 - 0.0000i
```

```
[U,S]=schur(A)
U = 3 \times 3
   -0.4741
               0.6648
                          0.5774
               0.0782
                          0.5774
    0.8127
              -0.7430
   -0.3386
                          0.5774
S = 3 \times 3
   -1.0000
              20.7846 -44.6948
               1.0000
                         -0.6096
          0
          0
               0.0000
                           1.0000
```

The matris A is defective since it does not have a full set of linearly independent eigenvectors (the second and third columns of V are the same). Since not all columns of V are linearly independent, it has a large condition number about ~1e8. However, schur is able to calcuate three different basis vectors in U. Since U is orthogonal, cond(U)=1.

The matrix S has the real eigenvalue as the first entry on the diagonal and the repeated eigenvalue represented by the lower right 2-by-2 block. The eigenvalues of the 2-by-2 block are also eigenvalues of A:

```
eig(S(2:3,2:3))
```

ans = 2×1 complex

1.0000 + 0.0000i 1.0000 - 0.0000i