

Positive-definite matrices

1 Tests

Eigenvalues > 0

Leading submatrices test

Pivots positive

$$\begin{bmatrix} a & b \\ b & c \end{bmatrix} \quad (1)$$

$\lambda_1 > 0, \lambda_2 > 0$ work out what this means

$$a > 0, ac - b^2 > 0$$

$$a > 0, \frac{ac-b^2}{a} > 0$$

But probably the most important test / definition is

$$\langle x | Ax \rangle > 0$$

$$\begin{bmatrix} 2 & 6 \\ 6 & 18 \end{bmatrix} \quad (2)$$

positive semi-definite

$$\lambda = 0, 20$$

$$\text{pivots} = 2$$

$$\text{Now let's look at } x^T Ax = 2x_1^2 + 12x_1x_2 + 18x_2^2 = 2(x + 3y)^2$$

the coefficients are $a, 2b, c$

quadratic form

The form will be positive-definite if we can factor it into a sum of squares. This shows the connection between positive-definiteness and pivots. The things that go outside the squares are the pivots. Then we have a "bowl". Otherwise we could have a saddle point, minimum in one direction and maximum in another. The important directions to look turns out to be the eigenvector directions.

For a function f , the matrix of second derivatives at a point is positive-definite implies it is a true minimum. The matrix of second derivatives is obviously symmetric.

If I cut the bowl at a constant, it's obviously an ellipse.

$$\begin{bmatrix} 2 & -1 & 0 \\ -1 & 2 & -1 \\ 0 & -1 & 2 \end{bmatrix} \quad (3)$$

dets 2,3,4 pivots 2,3/2,4/3 eigenvalues $2 - \sqrt{2}, 2, 2 + \sqrt{2}$

ellipsoid, axes in the directions of the eigenvectors