

Residue of a function  $f$  at a pole of order  $n$   $z_0$ ,

$$\text{Res}(f, z_0) = \frac{1}{(n-1)!} \lim_{z \rightarrow z_0} \frac{d^{n-1}}{dz^{n-1}} (z - z_0)^n f(z).$$

Residue theorem:

$$\int_C f(z) dz = 2\pi i \left( \sum_{k=1}^n \text{Res}(f, z_k) \right),$$

where  $z_k$  are singularities of  $f$  which lie *inside* the contour  $C$ .

Inner product for  $\mathbf{u}, \mathbf{v} \in \mathbb{R}^n$

$$\mathbf{u}^T \mathbf{v} = \sum_{k=1}^n u_k v_k = \|\mathbf{u}\| \|\mathbf{v}\| \cos \theta.$$

Note that the inner product of two vectors is a scalar, i.e.,  $\mathbf{u}^T \mathbf{v} \in \mathbb{R}$ . This is different than outer product  $\mathbf{u} \mathbf{v}^T$ , which is a matrix of size  $n \times n$ .

Multiplying two outer products is equal to

$$\begin{aligned} (\mathbf{u} \mathbf{v}^T)(\mathbf{u} \mathbf{v}^T) &= \mathbf{u}(\mathbf{v}^T \mathbf{u})\mathbf{v}^T \\ &= (\mathbf{v}^T \mathbf{u})(\mathbf{u} \mathbf{v}^T) \end{aligned}$$

because  $\mathbf{v}^T \mathbf{u}$  is a *scalar*.

Norm of a vector:

$$\|\mathbf{u}\| = \sqrt{\mathbf{u}^T \mathbf{u}}.$$

Two vectors are *orthogonal* if  $\mathbf{u}^T \mathbf{v} = 0$ .

$\mathbf{u}$  is a *unit vector* if  $\|\mathbf{u}\| = 1$ .

A set of vectors  $B = \{\mathbf{u}_1, \dots, \mathbf{u}_k\}$  are *linearly independent* if and only if

$$c_1 \mathbf{u}_1 + \dots + c_k \mathbf{u}_k = \mathbf{0}$$

only when  $c_1 = \dots = c_k = 0$ .

The  $n \times n$  *identity matrix*  $\mathbf{I}_n$  has all zeros, except for ones on the diagonal. For a matrix  $\mathbf{A} \in \mathbb{R}^{n \times n}$  and vector  $\mathbf{x} \in \mathbb{R}^n$ , the following are true:

$$\mathbf{I}_n \mathbf{A} = \mathbf{A}, \quad \mathbf{I}_n \mathbf{x} = \mathbf{x}.$$

A matrix  $\mathbf{A}$  is *symmetric* if  $\mathbf{A} = \mathbf{A}^T$ .

The *LU decomposition*  $\mathbf{A} = \mathbf{L}\mathbf{U}$  finds a lower triangular matrix  $\mathbf{L}$  and an upper triangular matrix  $\mathbf{U}$ . The diagonal values on  $\mathbf{L}$  are all 1.

If you know  $\mathbf{A} = \mathbf{L}\mathbf{U}$ , you can easily solve a system  $\mathbf{A}\mathbf{x} = \mathbf{b}$  by solving two triangular systems:

$$\begin{aligned} \mathbf{L}\mathbf{y} &= \mathbf{b}, \\ \mathbf{U}\mathbf{x} &= \mathbf{y}. \end{aligned}$$

The *inverse* of a matrix  $\mathbf{A} \in \mathbb{R}^{n \times n}$  is defined as

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

where  $\mathbf{I}_n$  is the  $n \times n$  identity matrix.

Basic matrix properties

$$\begin{aligned} (\mathbf{A}\mathbf{B})^T &= \mathbf{B}^T \mathbf{A}^T \\ (\mathbf{A} + \mathbf{B})^T &= \mathbf{A}^T + \mathbf{B}^T \\ (\mathbf{A}\mathbf{B})^{-1} &= \mathbf{B}^{-1} \mathbf{A}^{-1} \end{aligned}$$

$$(\mathbf{A} + \mathbf{B})(\mathbf{C} + \mathbf{D}) = \mathbf{A}\mathbf{C} + \mathbf{A}\mathbf{D} + \mathbf{B}\mathbf{C} + \mathbf{B}\mathbf{D}$$

A matrix  $\mathbf{V}$  is *orthogonal* if  $\mathbf{V}^T \mathbf{V} = \mathbf{I}_n$ .

The eigenvalue equation is

$$\mathbf{A}\mathbf{x} = \lambda \mathbf{x}.$$

$\mathbf{x}$  is an *eigenvector* and  $\lambda$  is an *eigenvalue*.

If  $\mathbf{A}$  is symmetric, the eigenvectors of  $\mathbf{A}$  are orthogonal.

The spectral decomposition of a matrix  $\mathbf{A}$  is

$$\mathbf{A} = \mathbf{X}\mathbf{\Lambda}\mathbf{X}^{-1}.$$