

Glossary of Linear Algebra Terms

basis for a subspace:

A basis for a [subspace](#) W is a set of vectors $\mathbf{v}_1, \dots, \mathbf{v}_k$ in W such that:

1. $\mathbf{v}_1, \dots, \mathbf{v}_k$ are [linearly independent](#); and
2. $\mathbf{v}_1, \dots, \mathbf{v}_k$ [span](#) W .

characteristic polynomial of a matrix:

The characteristic polynomial of a n by n matrix A is the polynomial in t given by the formula $\det(A - tI)$.

column space of a matrix:

The column space of a matrix is the [subspace spanned](#) by the columns of the matrix considered as vectors. See also: [row space](#).

consistent linear system:

A system of linear equations is consistent if it has at least one solution. See also: [inconsistent](#).

defective matrix:

A matrix A is defective if A has an [eigenvalue](#) whose [geometric multiplicity](#) is less than its [algebraic multiplicity](#).

diagonalizable matrix:

A matrix is diagonalizable if it is [dimension](#) of a subspace:

The dimension of a [subspace](#) W is the number of vectors in any [basis](#) of W . (If W is the subspace $\{\mathbf{0}\}$, we say that its dimension is 0.)

row **echelon form** of a matrix:

A matrix is in row echelon form if:

1. all rows that consist entirely of zeros are grouped together at the bottom of the matrix; and
2. the first (counting left to right) nonzero entry in each nonzero row appears in a column to the right of the first nonzero entry in the preceding row (if there is a preceding row).

reduced row **echelon form** of a matrix:

A matrix is in reduced row echelon form if:

1. the matrix is in [row echelon form](#);
2. the first nonzero entry in each nonzero row is the number 1; and
3. the first nonzero entry in each nonzero row is the only nonzero entry in its column.

eigenspace of a matrix:

The eigenspace associated with the [eigenvalue](#) c of a matrix A is the [null space](#) of $A - cI$.

eigenvalue of a matrix:

An eigenvalue of a n by n matrix A is a scalar c such that $A*\mathbf{x} = c*\mathbf{x}$ holds for some nonzero vector \mathbf{x} (where \mathbf{x} is an n -tuple). See also: [eigenvector](#).

eigenvector of a matrix:

An eigenvector of a n by n matrix A is a nonzero vector \mathbf{x} such that $A*\mathbf{x} = c*\mathbf{x}$ holds for some scalar c . See also: [eigenvalue](#).

equivalent linear systems:

Two systems of linear equations in n unknowns are equivalent if they have the same set of solutions.

homogeneous linear system:

A system of linear equations $A*\mathbf{x} = \mathbf{b}$ is homogeneous if $\mathbf{b} = \mathbf{0}$.

inconsistent linear system:

A system of linear equations is inconsistent if it has no solutions. See also: [consistent](#).

inverse of a matrix:

The matrix B is an inverse for the matrix A if $A*B = B*A = I$.

invertible matrix:

A matrix is invertible if it has an [inverse](#).

least-squares solution of a linear system:

A least-squares solution to a system of linear equations $A*\mathbf{x} = \mathbf{b}$ is a vector \mathbf{x} that minimizes the length of the vector $A*\mathbf{x} - \mathbf{b}$.

linear combination of vectors:

A vector \mathbf{v} is a linear combination of the vectors $\mathbf{v}_1, \dots, \mathbf{v}_k$ if there exist scalars a_1, \dots, a_k such that $\mathbf{v} = a_1*\mathbf{v}_1 + \dots + a_k*\mathbf{v}_k$.

linearly dependent vectors:

The vectors $\mathbf{v}_1, \dots, \mathbf{v}_k$ are linearly dependent if the equation $a_1*\mathbf{v}_1 + \dots + a_k*\mathbf{v}_k = \mathbf{0}$ has a solution where not all the scalars a_1, \dots, a_k are zero.

linearly independent vectors:

The vectors $\mathbf{v}_1, \dots, \mathbf{v}_k$ are linearly independent if the only solution to the equation $a_1*\mathbf{v}_1 + \dots + a_k*\mathbf{v}_k = \mathbf{0}$ is the solution where all the scalars a_1, \dots, a_k are zero.

linear transformation :

A linear transformation from V to W is a function T from V to W such that:

1. $T(\mathbf{u}+\mathbf{v}) = T(\mathbf{u}) + T(\mathbf{v})$ for all vectors \mathbf{u} and \mathbf{v} in V ; and
2. $T(a*\mathbf{v}) = a*T(\mathbf{v})$ for all vectors \mathbf{v} in V and all scalars a .

algebraic **multiplicity** of an eigenvalue:

The algebraic multiplicity of an [eigenvalue](#) c of a matrix A is the number of times the factor $(t-c)$ occurs in the [characteristic polynomial](#) of A .

geometric **multiplicity** of an eigenvalue:

The geometric multiplicity of an [eigenvalue](#) c of a matrix A is the [dimension](#) of the [eigenspace](#) of c .

nonsingular matrix:

An n by n matrix A is nonsingular if the only solution to the equation $A^* \mathbf{x} = \mathbf{0}$ (where \mathbf{x} is an n -tuple) is $\mathbf{x} = \mathbf{0}$. See also: [singular](#).

null space of a matrix:

The null space of a m by n matrix A is the set of all n -tuples \mathbf{x} such that $A^* \mathbf{x} = \mathbf{0}$.

null space of a linear transformation:

The null space of a [linear transformation](#) T is the set of vectors \mathbf{v} in its domain such that $T(\mathbf{v}) = \mathbf{0}$.

nullity of a matrix:

The nullity of a matrix is the [dimension](#) of its [null space](#).

nullity of a linear transformation:

The nullity of a [dimension](#) of its [null space](#).

orthogonal set of vectors:

A set of n -tuples is orthogonal if the dot product of any two of them is 0.

orthogonal matrix:

A matrix A is orthogonal if A is [invertible](#) and its [orthogonal linear transformation](#): A [linear transformation](#) T from V to W is orthogonal if $T(\mathbf{v})$ has the same length as \mathbf{v} for all vectors \mathbf{v} in V .

orthonormal set of vectors:

A set of n -tuples is orthonormal if it is [orthogonal](#) and each vector has length 1.

range of a matrix:

The range of a m by n matrix A is the set of all m -tuples $A^* \mathbf{x}$, where \mathbf{x} is any n -tuple.

range of a linear transformation:

The range of a [linear transformation](#) T is the set of all vectors $T(\mathbf{v})$, where \mathbf{v} is any vector in its domain.

rank of a matrix:

The rank of a matrix is the number of nonzero rows in any [row equivalent](#) matrix that is in [row echelon form](#).

rank of a linear transformation:

The rank of a linear transformation (and hence of any matrix regarded as a [linear transformation](#)) is the [dimension](#) of its [range](#). Note: A theorem tells us that the two definitions of rank of a matrix are equivalent.

row equivalent matrices:

Two matrices are row equivalent if one can be obtained from the other by a sequence of elementary [row operations](#):

The elementary row operations performed on a matrix are:

- interchange two rows;
- multiply a row by a nonzero scalar;
- add a constant multiple of one row to another.

row space of a matrix:

The row space of a matrix is the [subspace spanned](#) by the rows of the matrix considered as vectors. See also: [similar matrices](#):

Matrices A and B are similar if there is a square [nonsingular](#) matrix S such that $S^{-1}AS = B$.

singular matrix:

An n by n matrix A is singular if the equation $A\mathbf{x} = \mathbf{0}$ (where \mathbf{x} is an n -tuple) has a nonzero solution for \mathbf{x} . See also: [nonsingular](#).

span of a set of vectors:

The span of the vectors $\mathbf{v}_1, \dots, \mathbf{v}_k$ is the [subspace](#) V consisting of all [linear combinations](#) of $\mathbf{v}_1, \dots, \mathbf{v}_k$. One also says that the subspace V is spanned by the vectors $\mathbf{v}_1, \dots, \mathbf{v}_k$ and that these vectors span V .

subspace:

A subset W of n -space is a subspace if:

1. the zero vector is in W ;
2. $\mathbf{x} + \mathbf{y}$ is in W whenever \mathbf{x} and \mathbf{y} are in W ; and
3. $a\mathbf{x}$ is in W whenever \mathbf{x} is in W and a is any scalar.

symmetric matrix:

A matrix is symmetric if it equals its transpose.