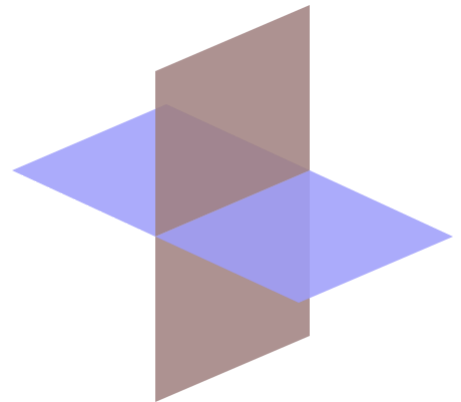


# Hyperplane

In geometry, a **hyperplane** is a subspace whose dimension is one less than that of its ambient space. If a space is 3-dimensional then its hyperplanes are the 2-dimensional planes, while if the space is 2-dimensional, its hyperplanes are the 1-dimensional lines. This notion can be used in any general space in which the concept of the dimension of a subspace is defined. In machine learning, hyperplanes are a key tool to create support vector machines for such tasks as computer vision and natural language processing.<sup>[1]</sup>

In different settings, the objects which are hyperplanes may have different properties. For instance, a hyperplane of an  $n$ -dimensional affine space is a flat subset with dimension  $n - 1$ .<sup>[2]</sup> By its nature, it separates the space into two half spaces. A hyperplane of an  $n$ -dimensional projective space does not have this property.

The difference in dimension between a subspace  $\mathbf{S}$  and its ambient space  $\mathbf{X}$  is known as the codimension of  $\mathbf{S}$  with respect to  $\mathbf{X}$ . Therefore, a necessary condition for  $\mathbf{S}$  to be a hyperplane in  $\mathbf{X}$  is for  $\mathbf{S}$  to have codimension one in  $\mathbf{X}$ .



Two intersecting planes in three-dimensional space. A plane is a hyperplane of dimension 2, when embedded in a space of dimension 3.

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## Technical description

In geometry, a **hyperplane** of an  $n$ -dimensional space  $V$  is a subspace of dimension  $n - 1$ , or equivalently, of codimension 1 in  $V$ . The space  $V$  may be a Euclidean space or more generally an affine space, or a vector space or a projective space, and the notion of hyperplane varies correspondingly since the definition of subspace differs in these settings; in all cases however, any hyperplane can be given in coordinates as the solution of a single (due to the "codimension 1" constraint) algebraic equation of degree 1.

If  $V$  is a vector space, one distinguishes "vector hyperplanes" (which are linear subspaces, and therefore must pass through the origin) and "affine hyperplanes" (which need not pass through the origin; they can be obtained by translation of a vector hyperplane). A hyperplane in a Euclidean space separates that space into two half spaces, and defines a reflection that fixes the hyperplane and interchanges those two half spaces.

## Special types of hyperplanes

Several specific types of hyperplanes are defined with properties that are well suited for particular purposes. Some of these specializations are described here.

### Affine hyperplanes

An **affine hyperplane** is an affine subspace of codimension 1 in an affine space. In Cartesian coordinates, such a hyperplane can be described with a single linear equation of the following form (where at least one of the  $a_i$ 's is non-zero and  $b$  is an arbitrary constant):

$$a_1x_1 + a_2x_2 + \cdots + a_nx_n = b.$$

In the case of a real affine space, in other words when the coordinates are real numbers, this affine space separates the space into two half-spaces, which are the connected components of the complement of the hyperplane, and are given by the inequalities

$$a_1x_1 + a_2x_2 + \cdots + a_nx_n < b$$

and

$$a_1x_1 + a_2x_2 + \cdots + a_nx_n > b.$$

As an example, a point is a hyperplane in 1-dimensional space, a line is a hyperplane in 2-dimensional space, and a plane is a hyperplane in 3-dimensional space. A line in 3-dimensional space is not a hyperplane, and does not separate the space into two parts (the complement of such a line is connected).

Any hyperplane of a Euclidean space has exactly two unit normal vectors.

Affine hyperplanes are used to define decision boundaries in many machine learning algorithms such as linear-combination (oblique) decision trees, and perceptrons.

### Vector hyperplanes

In a vector space, a vector hyperplane is a subspace of codimension 1, only possibly shifted from the origin by a vector, in which case it is referred to as a flat. Such a hyperplane is the solution of a single linear equation.

### Projective hyperplanes

**Projective hyperplanes**, are used in projective geometry. A projective subspace is a set of points with the property that for any two points of the set, all the points on the line determined by the two points are contained in the set.<sup>[3]</sup> Projective geometry can be viewed as affine geometry with vanishing points (points at infinity) added. An affine hyperplane together with the associated points at infinity forms a projective hyperplane. One special case of a projective hyperplane is the **infinite** or **ideal hyperplane**, which is defined with the set of all points at infinity.

In projective space, a hyperplane does not divide the space into two parts; rather, it takes two hyperplanes to separate points and divide up the space. The reason for this is that the space essentially "wraps around" so that both sides of a lone hyperplane are connected to each other.

## Dihedral angles

The dihedral angle between two non-parallel hyperplanes of a Euclidean space is the angle between the corresponding normal vectors. The product of the transformations in the two hyperplanes is a rotation whose axis is the subspace of codimension 2 obtained by intersecting the hyperplanes, and whose angle is twice the angle between the hyperplanes.

## Support hyperplanes

A hyperplane  $H$  is called a "support" hyperplane of the polyhedron  $P$  if  $P$  is contained in one of the two closed half-spaces bounded by  $H$  and  $H \cap P \neq \emptyset$ .<sup>[4]</sup> The intersection of between  $P$  and  $H$  is defined to be a "face" of the polyhedron. The theory of polyhedrons and the dimension of the faces are analyzed by the looking at these intersections involving hyperplanes.

## See also

- Hypersurface
- Decision boundary
- Ham sandwich theorem
- Arrangement of hyperplanes
- Separating hyperplane theorem
- Supporting hyperplane theorem

## References

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## External links

- Weisstein, Eric W. "Hyperplane" (<http://mathworld.wolfram.com/Hyperplane.html>). *MathWorld*.
- Weisstein, Eric W. "Flat" (<http://mathworld.wolfram.com/Flat.html>). *MathWorld*.

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