

# Machine Learning: Analytic Geometry

Hao Tang

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## Vectors in $\mathbb{R}^d$

- $x = (x_1, x_2, \dots, x_d)$
- $ax = (ax_1, ax_2, \dots, ax_d)$  for  $a \in \mathbb{R}$
- $u + v = (u_1 + v_1, \dots, u_d + v_d)$
- $u - v = u + (-1)v$

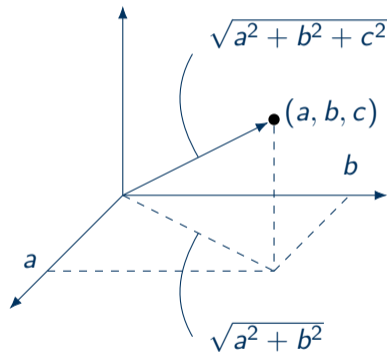
# Dot product

- $u^T v = u_1 v_1 + \cdots + u_d v_d = \sum_{i=1}^d u_i v_i$
- $(au)^T v = a(u^T v) = u^T(av)$  for  $a \in \mathbb{R}$
- $(u + v)^T w = u^T w + v^T w$
- $w^T(u + v) = w^T u + w^T v$
- $u^T v = v^T u$

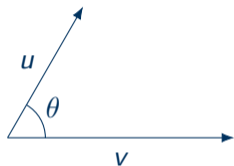
## The $\ell_2$ norm

- $\|v\|_2 = \sqrt{v^\top v} = \sqrt{v_1^2 + \dots + v_d^2}$
- $\|au\|_2 = |a|\|u\|_2$  for  $a \in \mathbb{R}$
- $\|u\|_2 \geq 0$
- If  $\|u\|_2 = 0$ , then  $u = 0$
- $\|u\| + \|v\| \geq \|u + v\|$

## The $\ell_2$ norm and length

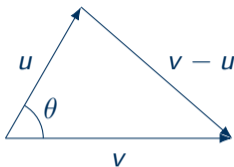


## Dot product and angle



$$\cos \theta = \frac{u^T v}{\|u\|_2 \|v\|_2} \quad (1)$$

## Dot product and angle



By the law of cosine,

$$\|v - u\|_2^2 = \|u\|_2^2 + \|v\|_2^2 - \|u\|_2 \|v\|_2 \cos \theta. \quad (2)$$

Comparing the above with

$$\|v - u\|_2^2 = \|v\|_2^2 - 2u^\top v + \|u\|_2^2, \quad (3)$$

we get

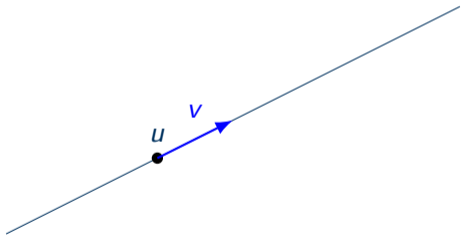
$$\cos \theta = \frac{u^\top v}{\|u\|_2 \|v\|_2}. \quad (4)$$

# A line



## A line

A line is a set of points  $\{x : x = u + tv \text{ for } t \in \mathbb{R}\}$  for any vector  $u$  and vector  $v \neq 0$ .



## A line in 2D

- Is  $y = ax + b$  a line?

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- Is  $y = ax + b$  a line?
- We can rewrite  $y = ax + b$  as

$$\begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 0 \\ b \end{pmatrix} + t \begin{pmatrix} 1 \\ a \end{pmatrix}. \quad (5)$$

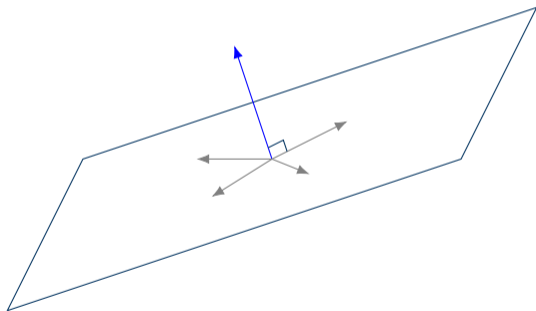
# A plane

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A plane (or hyperplane) is a set of points  $\{x : v^\top(x - u) = 0\}$  for any vector  $u$  and vector  $v \neq 0$ .

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## A plane in 3D

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- Is  $ax + by + cz + d = 0$  a plane?
- We can rewrite  $ax + by + cz + d = 0$  as

$$\begin{pmatrix} a \\ b \\ c \end{pmatrix}^T \left( \begin{pmatrix} x \\ y \\ z \end{pmatrix} - \begin{pmatrix} 0 \\ 0 \\ d/c \end{pmatrix} \right) = 0. \quad (6)$$



## A plane in $d + 1$ dimension

- Is  $y = w^T x + b$ , where  $w \in \mathbb{R}^d$  and  $b \in \mathbb{R}$ , a plane?

## A plane in $d + 1$ dimension

- Is  $y = w^\top x + b$ , where  $w \in \mathbb{R}^d$  and  $b \in \mathbb{R}$ , a plane?
- We can rewrite  $y = w^\top x + b$  as

$$\begin{pmatrix} w \\ -1 \end{pmatrix}^\top \left( \begin{pmatrix} x \\ y \end{pmatrix} - \begin{pmatrix} 0 \\ b \end{pmatrix} \right) = 0. \quad (7)$$

# Halfspaces

- Points  $\{x : v^\top(x - u) > 0\}$  are on one side of the plane  $\{x : v^\top(x - u) = 0\}$ .

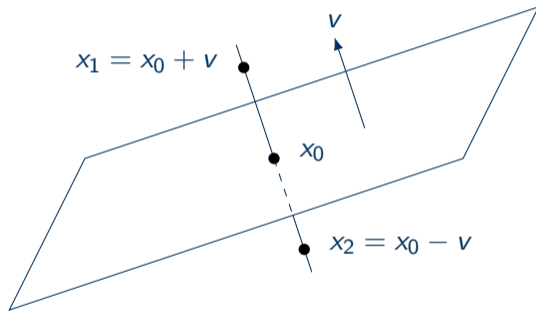
# Halfspaces

- Points  $\{x : v^\top(x - u) > 0\}$  are on one side of the plane  $\{x : v^\top(x - u) = 0\}$ .
- Consider the line  $\{x : x = x_0 + tv \text{ for } t \in \mathbb{R}\}$  for some  $x_0$  on the plane.

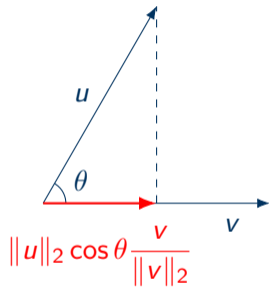
$$v^\top(x_0 + tv - u) = v^\top(x_0 - u) + t\|v\|_2^2 = t\|v\|_2^2 \quad (8)$$

When  $t > 0$ ,  $v^\top(x_0 + tv - u) > 0$ .

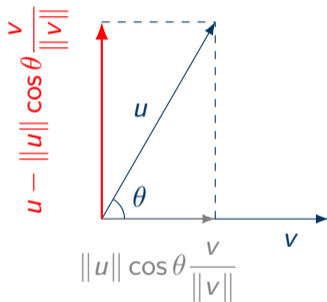
# Halfspaces



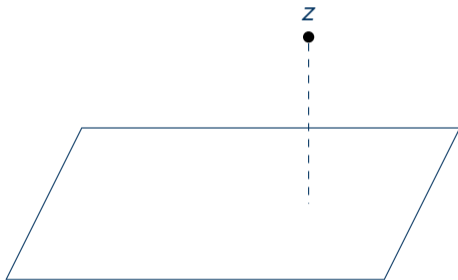
# Projection



## The perpendicular part after projection

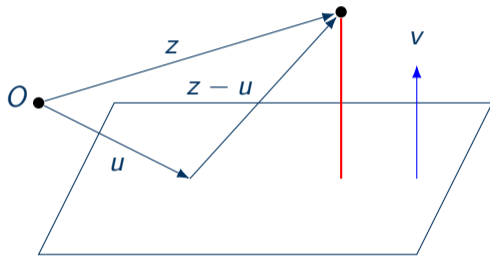


## The distance between a plane and a point





## The distance between a plane and a point



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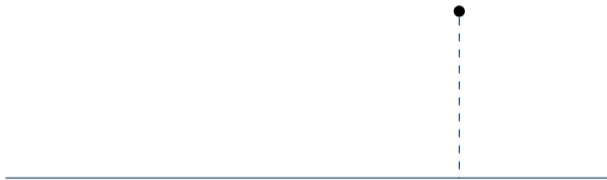
The projection of  $z - u$  on  $v$  is

$$\frac{v^\top(z - u)}{\|v\|_2} \frac{v}{\|v\|_2}, \quad (9)$$

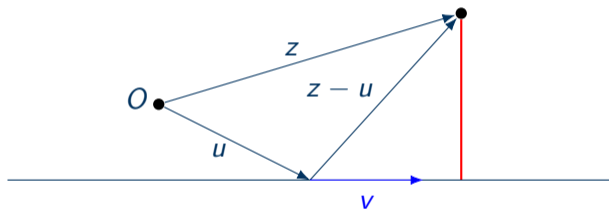
which has an  $\ell_2$  norm of

$$\frac{|v^\top(z - u)|}{\|v\|_2}. \quad (10)$$

## The distance between a line and a point



## The distance between a line and a point



## The distance between a line and a point

The perpendicular part after the projection of  $z - u$  on  $v$  is

$$(z - u) - \frac{v^T(z - u)}{\|v\|_2} \frac{v}{\|v\|_2}, \quad (11)$$

with a norm of

$$\left\| (z - u) - \frac{v^T(z - u)}{\|v\|_2} \frac{v}{\|v\|_2} \right\|_2 \quad (12)$$

## Cauchy–Schwarz inequality

For any vector  $u$  and vector  $v$ ,

$$(u^T v)^2 \leq \|u\|_2^2 \|v\|_2^2. \quad (13)$$

## Cauchy–Schwarz inequality

$$\|u\|_2^2 \|v\|_2^2 - (u^\top v)^2 = \|u\|_2^2 \left( \|v\|_2^2 - \frac{(u^\top v)(u^\top v)}{\|u\|_2^2} \right) \quad (14)$$

$$= \|u\|_2^2 \left[ \|v\|_2^2 - 2 \left( \frac{u^\top v}{\|u\|_2^2} u \right)^\top v + \left( \frac{u^\top v}{\|u\|_2^2} u \right)^\top \left( \frac{u^\top v}{\|u\|_2^2} u \right) \right] \quad (15)$$

$$= \|u\|_2^2 \left\| v - \frac{u^\top v}{\|u\|_2^2} u \right\|^2 \geq 0 \quad (16)$$