



# EECS 245 Fall 2025

## Math for ML

Lecture 7: Projections and Spans

→ Read: 2.3 (new), 2.4 (coming soon)

# Agenda

- ① Important (in)equalities from last class
  - ② The "Approximation Problem"
    - Motivation
    - Solution (Orthogonal Projections)
    - Many examples
  - ③ Span      } start Ch. 2.4
    - Equation of a line in  $\mathbb{R}^n$
- Ch. 2.3  
(new!)

## Some updates

- Thank you for giving feedback on Homework 3!
- Some changes:
  - Updated lecture recordings to not cut off the last few minutes
  - Future homeworks will be due on Thursdays (more time for OH)

# Inequalities (Ch. 22)

Triangle inequality

$$\|\vec{u}\| \quad \|\vec{v}\|$$

$$\|\vec{u} + \vec{v}\| \leq \|\vec{u}\| + \|\vec{v}\|$$



Cauchy-Schwarz

$$|\vec{u} \cdot \vec{v}| \leq \|\vec{u}\| \|\vec{v}\|$$

# Cauchy-Schwarz

$$|\vec{u} \cdot \vec{v}| \leq \|\vec{u}\| \|\vec{v}\|$$

Prove that the geometric mean of  $a, b$  is  $\leq$  arithmetic mean of  $a, b$

Hint:

$$\vec{u} = \begin{bmatrix} \sqrt{a} \\ \sqrt{b} \end{bmatrix}$$

$$\vec{v} = \begin{bmatrix} ? \\ ? \end{bmatrix}$$

$$\sqrt{ab} \leq \frac{a+b}{2}$$

Prove

$$\sqrt{ab} \leq \frac{a+b}{2}$$

AM-GM-HM  
inequality  
 $AM \geq GM \geq HM$   
Harmonic

$$\vec{u} = \begin{bmatrix} \sqrt{a} \\ \sqrt{b} \end{bmatrix} \quad \vec{v} = \begin{bmatrix} \sqrt{b} \\ \sqrt{a} \end{bmatrix}$$

$$C-Z: |\vec{u} \cdot \vec{v}| \leq \|\vec{u}\| \|\vec{v}\|$$

$$2\sqrt{ab} \leq \underbrace{\sqrt{(\sqrt{a})^2 + (\sqrt{b})^2}}_{\sqrt{a+b}} \underbrace{\sqrt{(\sqrt{b})^2 + (\sqrt{a})^2}}_{(\sqrt{a+b})^2 = a+b}$$

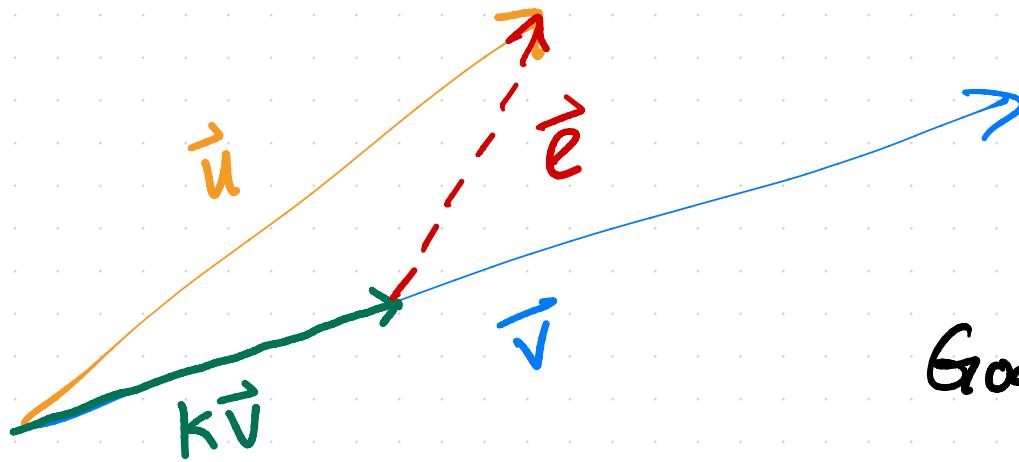
# The Approximation Problem (2.3)

$$\vec{u}, \vec{v} \in \mathbb{R}^n$$



Goal: of all vectors of the form  $K\vec{v}$ ,  
which is "closest" to  $\vec{u}$ ?

a scalar  
multiple of  
 $\vec{v}$



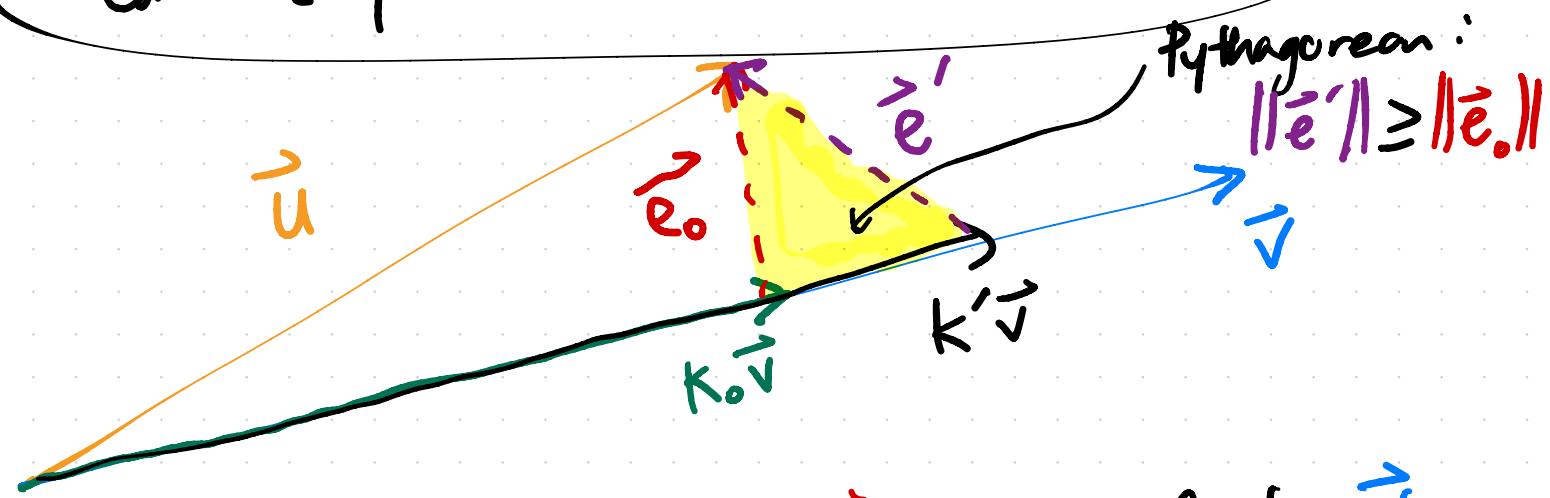
$$\underbrace{\vec{e}}_{\text{error vector}} = \vec{u} - k\vec{v}$$

Goal : Minimize  
 $\|\vec{e}\|$   
 $=$   
 $\|\vec{u} - k\vec{v}\|$ ,  
 function of  $k$ !

Guess: The best  $k$  is the one where

$\vec{e}$  is orthogonal to  $\vec{v}$

Can we prove that we're right?



Let  $k_0$  be the  $k$  such that  $\vec{e}_0$  orthogonal to  $\vec{v}$   
Let  $k'$  be some other  $k \neq k_0$ , error  $\vec{e}'$

Shown that the best  $k$ ,  $k^*$ , makes

$\vec{e}$  orthogonal to  $\vec{v}$

How do we find  $k^*$ ?

$$\vec{e} \cdot \vec{v} = 0$$

$$(\vec{u} - k\vec{v}) \cdot \vec{v} = 0$$

$$\vec{u} \cdot \vec{v} - (k^* \vec{v}) \cdot \vec{v} = 0$$

$$\vec{u} \cdot \vec{v} - k^* (\vec{v} \cdot \vec{v}) = 0$$

$$\Rightarrow k^* = \frac{\vec{u} \cdot \vec{v}}{\vec{v} \cdot \vec{v}}$$

$$\|\vec{v}\|^2$$

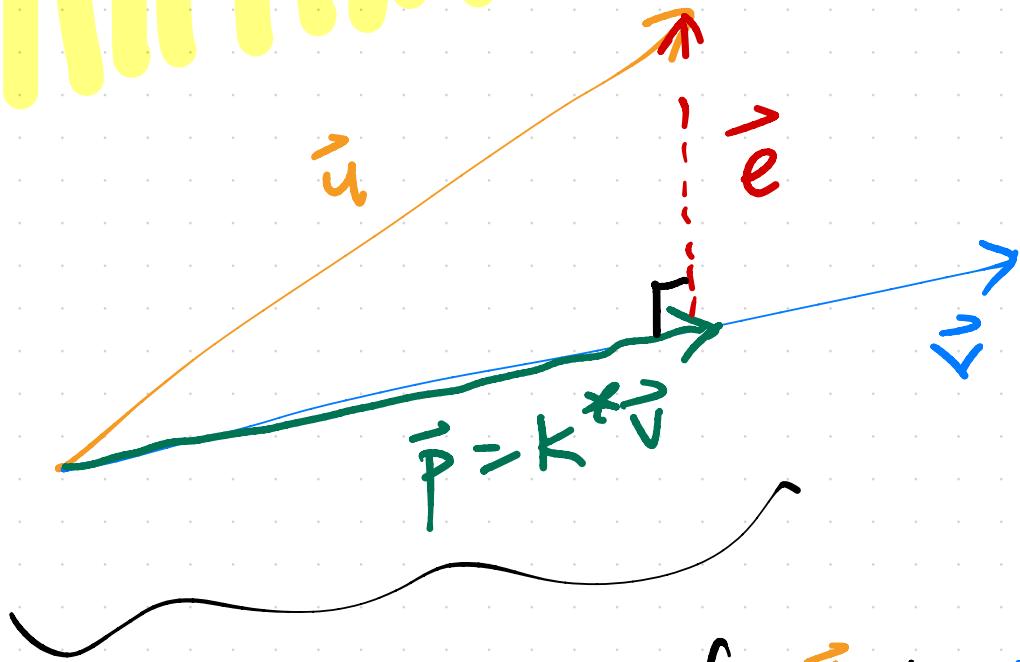
The orthogonal projection of  $\vec{u}$  onto  $\vec{v}$

is the **vector**

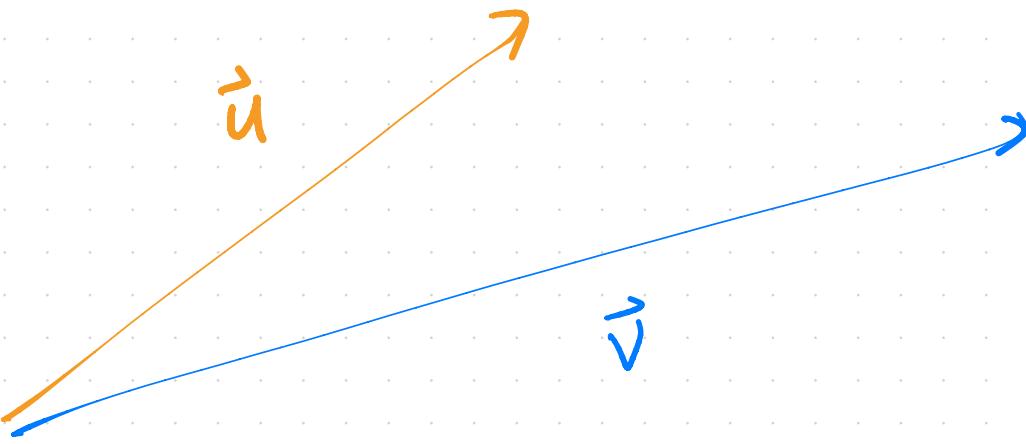
$$\vec{p} = \left( \frac{\vec{u} \cdot \vec{v}}{\vec{v} \cdot \vec{v}} \right) \vec{v}$$

of all vectors of the form  $k\vec{v}$ ,  
 $\vec{p}$  has the shortest error vector,

$$\vec{e} = \vec{u} - \vec{p}.$$



$\vec{p}$  is shadow of  $\vec{u}$  on  $\vec{v}$



$$\vec{u} = \begin{bmatrix} 1 \\ 2 \\ 2 \end{bmatrix} \quad \vec{v} = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$$

①  $\vec{p} = \left( \frac{\vec{u} \cdot \vec{v}}{\vec{v} \cdot \vec{v}} \right) \vec{v}$

$$= \frac{5}{3} \vec{v} = \begin{bmatrix} 5/3 \\ 5/3 \\ 5/3 \end{bmatrix}$$

① find the projection  
of  $\vec{u}$  onto  $\vec{v}$

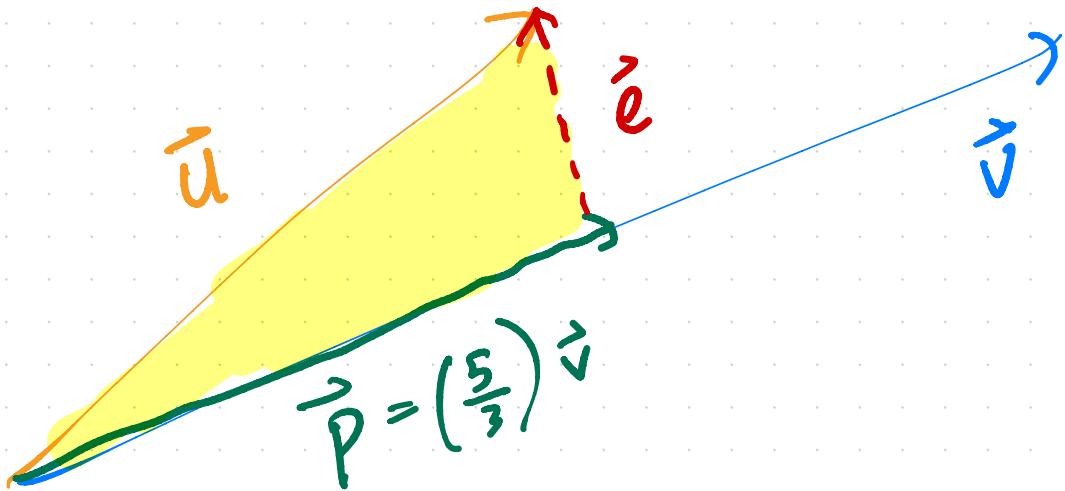
② find the error  
vector

③ show that the  
error vector  
is orthogonal  
to  $\vec{v}$

②  $\vec{e} = \vec{u} - \vec{p} = \begin{bmatrix} 1 \\ 2 \\ 2 \end{bmatrix} - \begin{bmatrix} 5/3 \\ 5/3 \\ 5/3 \end{bmatrix}$

$$= \begin{bmatrix} -2/3 \\ 1/3 \\ 1/3 \end{bmatrix}$$

③  $\vec{e} \cdot \vec{v} = 0 \rightarrow \begin{bmatrix} -2/3 \\ 1/3 \\ 1/3 \end{bmatrix} \cdot \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} = 0 \quad \checkmark$



$$\vec{u} = \underbrace{\vec{p}}_{\text{parallel to } \vec{v}} + \underbrace{\vec{e}}_{\text{orthogonal to } \vec{v}}$$

"orthogonal  
decomposition"  
of  $\vec{u}$   
with respect  
to  $\vec{v}$

Write  $\vec{u} = \begin{bmatrix} 3 \\ 5 \end{bmatrix}$  as a linear combination of

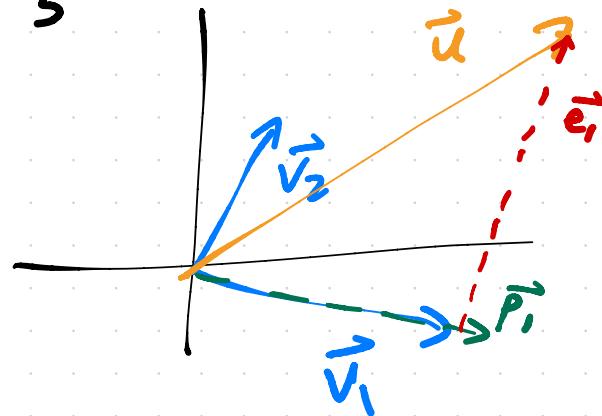
$$\vec{v}_1 = \begin{bmatrix} 6 \\ -2 \end{bmatrix}$$

$$\vec{v}_2 = \begin{bmatrix} 1 \\ 3 \end{bmatrix}$$

$$a_1 \begin{bmatrix} 6 \\ -2 \end{bmatrix} + a_2 \begin{bmatrix} 1 \\ 3 \end{bmatrix} = \begin{bmatrix} 3 \\ 5 \end{bmatrix}$$

$$6a_1 + a_2 = 3 \rightarrow a_1 = \frac{1}{5} \quad a_2 = \frac{9}{5}$$

$$-2a_1 + 3a_2 = 5$$



$$\vec{u} = \begin{bmatrix} 3 \\ 5 \end{bmatrix}$$

$$\vec{v}_1 = \begin{bmatrix} 6 \\ -2 \end{bmatrix}$$

$$\vec{v}_2 = \begin{bmatrix} 1 \\ 3 \end{bmatrix}$$

$$\frac{\vec{u} \cdot \vec{v}_1}{\vec{v}_1 \cdot \vec{v}_1} = \frac{8}{40} = \frac{1}{5}$$

$a_1$

$$\frac{\vec{u} \cdot \vec{v}_2}{\vec{v}_2 \cdot \vec{v}_2} = \frac{18}{10} = \frac{9}{5}$$

$a_2$

# Span

$n$ : # of dimensions  
 $d$ : # of vectors

The span of  $\vec{v}_1, \vec{v}_2, \dots, \vec{v}_d \in \mathbb{R}^n$  is  
the set of all possible linear combinations  
of those vectors.

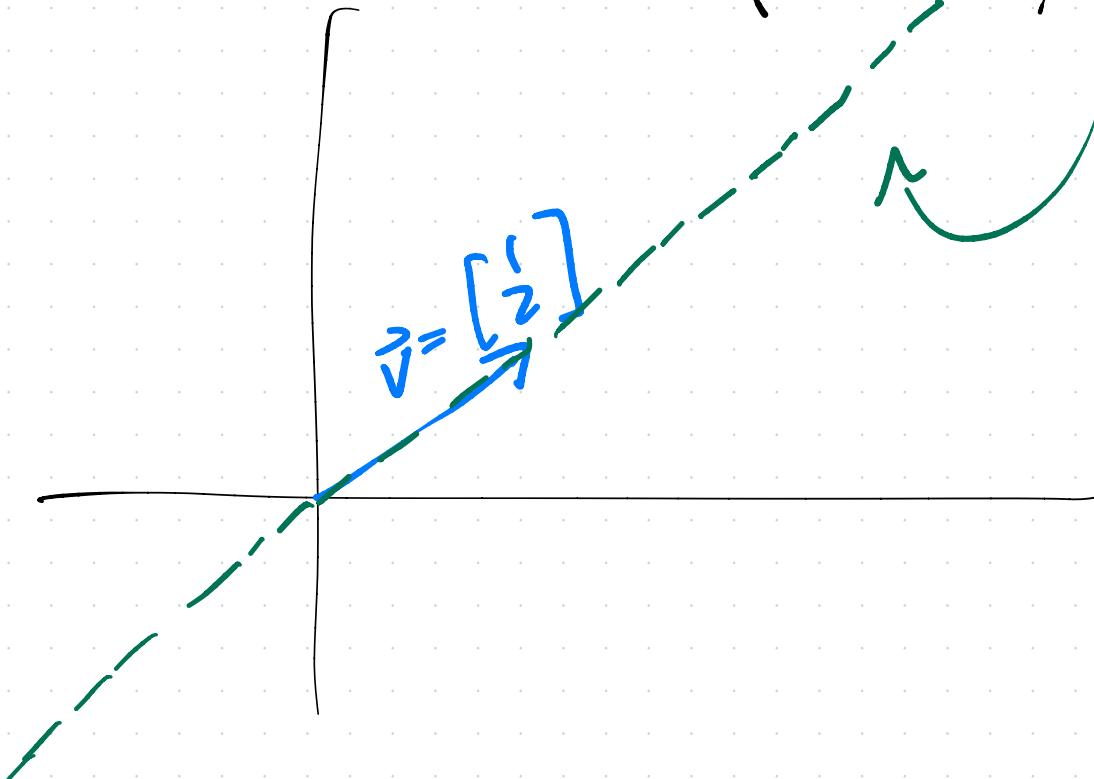
span  $(\{\vec{v}_1, \vec{v}_2, \dots, \vec{v}_d\})$

$$= \left\{ a_1 \vec{v}_1 + a_2 \vec{v}_2 + \dots + a_d \vec{v}_d \mid \begin{array}{l} a_1, a_2, \dots, a_d \\ \in \mathbb{R} \end{array} \right\}$$

Span of a single vector

$$\text{span} \left( \left\{ \begin{bmatrix} 1 \\ 2 \end{bmatrix} \right\} \right)$$

is a line



span includes

$$(1, 2)$$

AND

$$(0, 0)$$

# Parametric equation of line

$$L = \begin{bmatrix} 3 \\ 1 \\ 4 \end{bmatrix} + t \begin{bmatrix} 2 \\ -5 \\ 2 \end{bmatrix}, \quad t \in \mathbb{R}$$