

1.5 Dot Products and Projections in \mathbb{R}^n

Quote. "The important thing to remember about mathematics is not to be frightened" Richard Dawkins (1941-)

Vocabulary.

- Scalar Product: another name for the dot product (the result of the dot product is a scalar)
- Inner product: another name for the dot product.
- Projection: an everyday example of this is your shadow - a projection of you onto the ground.
- Unit Vector: a vector with a norm of 1
- Dimension of a space: the number of vectors in the basis that spans the space.
- Hyperplane: n -dimensional generalization of the plane. A subspace whose dimension is one less than the ambient space.

Here generalize the dot product from \mathbb{R}^2 and \mathbb{R}^3 to \mathbb{R}^n and we introduce the important idea of projection.

1. Dot Products, Orthogonality, and Norms in \mathbb{R}^n

By way of series of examples, we will see how these operations generalize to \mathbb{R}^n

$$\underset{\lambda}{(1, 2)} \cdot \underset{\lambda}{(-1, 5)} = (1)(-1) + (2)(5) = 9$$

$$(1, 2, 3, 4) \cdot (-1, 0, 0, 1) = (1)(-1) + (2)(0) + (3)(0) + (4)(1) = 3$$

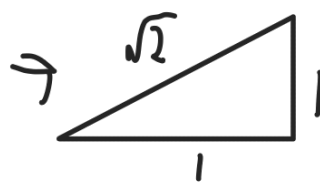
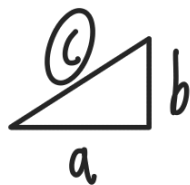
in any \mathbb{R}^r $\vec{u} \cdot \vec{v} = 0 \implies \vec{u} \perp \vec{v}$ (orthogonal)

2. The Triangle Inequality

$$\|\vec{x} + \vec{y}\| \leq \|\vec{x}\| + \|\vec{y}\|$$

c is less than
 $a+b$

in \mathbb{R}^2



$\sqrt{2}$ less than those 2

3. The Cauchy-Schwartz Inequality

(complicated to prove, not intuitive)

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

Fall 2020 video has a proof.

4. Hyperplane

Recall the method of finding the scalar equation of a plane in \mathbb{R}^3 :

• Test points \vec{x} and \vec{p} , \vec{p} fixed, \vec{x} any other point

$$\vec{n} \cdot (\vec{x} - \vec{p}) = 0 \text{ (normal line)}$$

Using the same method we can find scalar equation of the Hyperplane in \mathbb{R}^n

$$\vec{m} \text{ orthogonal to } \vec{x} - \vec{p} \quad \vec{m}, \vec{x}, \vec{p} \in \mathbb{R}^n$$

$$m_1 x_1 + m_2 x_2 + \dots + m_n x_n = \vec{m} \cdot \vec{p}$$

in the scalar eq'n of the hyperplane.

$$\vec{n} = \begin{pmatrix} n_1 \\ n_2 \\ n_3 \end{pmatrix}$$

$$n_1(x_1 - p_1) + n_2(x_2 - p_2) + n_3(x_3 - p_3) = 0, \quad n_1 x_1 + n_2 x_2 + n_3 x_3 = \vec{n} \cdot \vec{p}$$

5. Orthogonal projections onto lines in \mathbb{R}^n

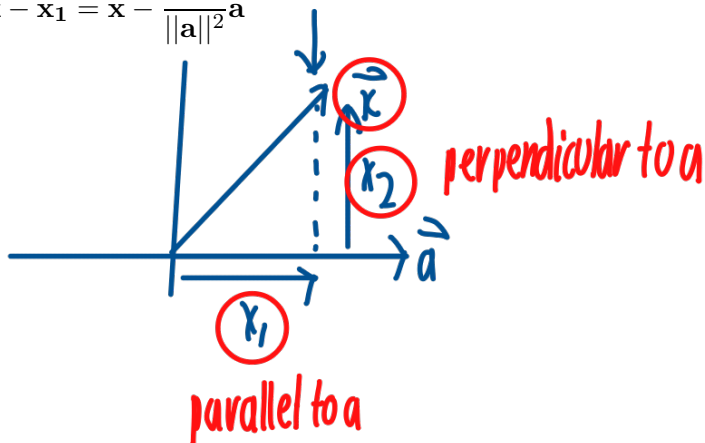
Theorem 0.1 If \mathbf{a} is a non-zero vector in \mathbb{R}^n , then every vector \mathbf{x} in \mathbb{R}^n can be expressed in exactly one way as

$$\mathbf{x} = \mathbf{x}_1 + \mathbf{x}_2$$

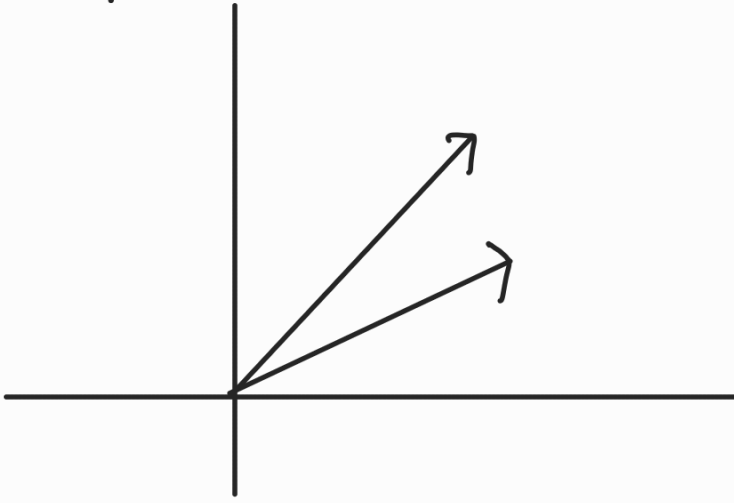
where \mathbf{x}_1 is in $\text{span}\{\mathbf{a}\}$ and \mathbf{x}_2 is perpendicular to \mathbf{a} . The vectors \mathbf{x}_1 and \mathbf{x}_2 are given by

$$\mathbf{x}_1 = \frac{\mathbf{x} \cdot \mathbf{a}}{\|\mathbf{a}\|^2} \mathbf{a} \quad \text{and} \quad \mathbf{x}_2 = \mathbf{x} - \mathbf{x}_1 = \mathbf{x} - \frac{\mathbf{x} \cdot \mathbf{a}}{\|\mathbf{a}\|^2} \mathbf{a}$$

projection of \vec{x}
onto \vec{a}



Disregard...



Definition: **orthogonal projection onto line in \mathbb{R}^n**

If \mathbf{a} is a non-zero vector in \mathbb{R}^n and \mathbf{x} is any vector in \mathbb{R}^n , then the **orthogonal projection of \mathbf{x} onto the line $\text{span}\{\mathbf{a}\}$** is denoted by $\text{proj}_{\mathbf{a}}\mathbf{x}$ and defined by

$$\text{proj}_{\mathbf{a}}\mathbf{x} = \frac{\mathbf{x} \cdot \mathbf{a}}{\|\mathbf{a}\|^2} \mathbf{a}$$

The vector $\text{proj}_{\mathbf{a}}\mathbf{x}$ is also called the **vector component of \mathbf{x} along \mathbf{a}** and $\mathbf{x} - \text{proj}_{\mathbf{a}}\mathbf{x}$ is called the **vector component of \mathbf{x} orthogonal to \mathbf{a}** .

Note: the portion of the vector that is perpendicular to the projection is denoted $\text{perp}_{\mathbf{a}}\mathbf{x}$.

Example.

$$\vec{v} = (1/3, -2/3, 2/3) \quad \vec{u} = (4, 1, -3)$$

Projection of \vec{u} into \vec{v}
Don't get direction wrong!
Know what you're trying to calculate

$$\text{proj}_{\vec{v}}(\vec{u}) \text{ and } \text{perp}_{\vec{v}}(\vec{u})$$

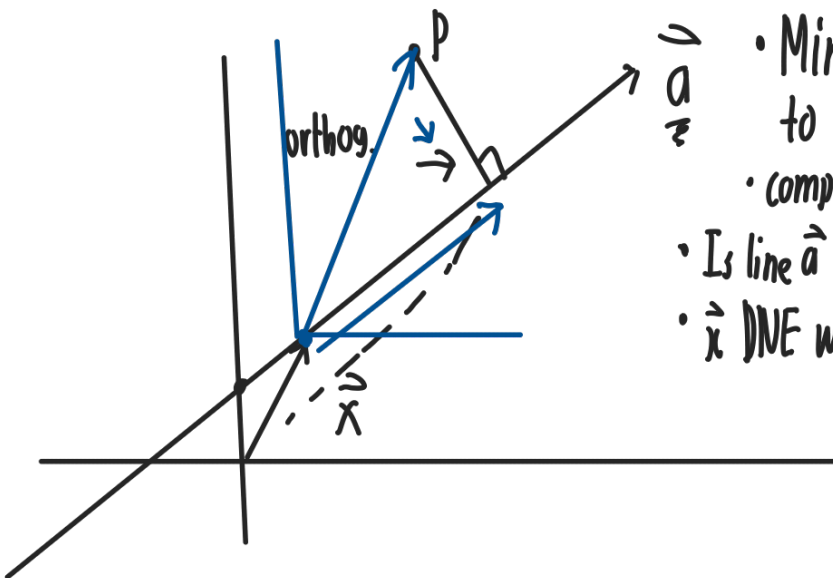
$$\text{proj}_{\vec{v}}(\vec{u}) = \frac{\vec{u} \cdot \vec{v}}{\|\vec{v}\|^2} \vec{v}$$

$$= \frac{(4, 1, -3) \cdot (1/3, -2/3, 2/3)}{(1/3)^2 + (-2/3)^2 + (2/3)^2} \cdot (1/3, -2/3, 2/3) = \frac{(-4/9, 8/9, -8/9)}{1/3} = (-4/3, 8/3, -8/3)$$

$$\vec{u} - \text{proj}_{\vec{v}}(\vec{u}) = \text{perp}_{\vec{v}}(\vec{u})$$

• Should length affect?
NO!
• remove dependency of it!

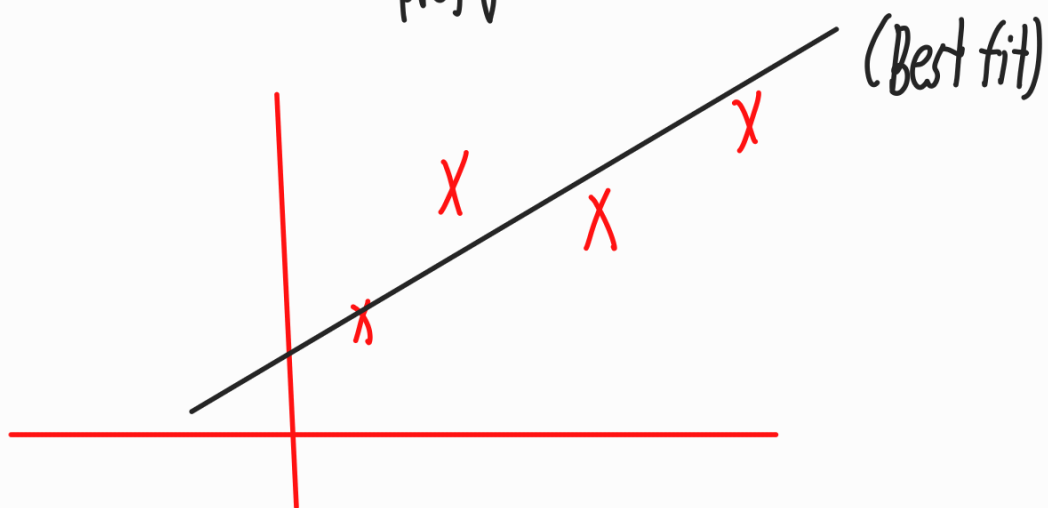
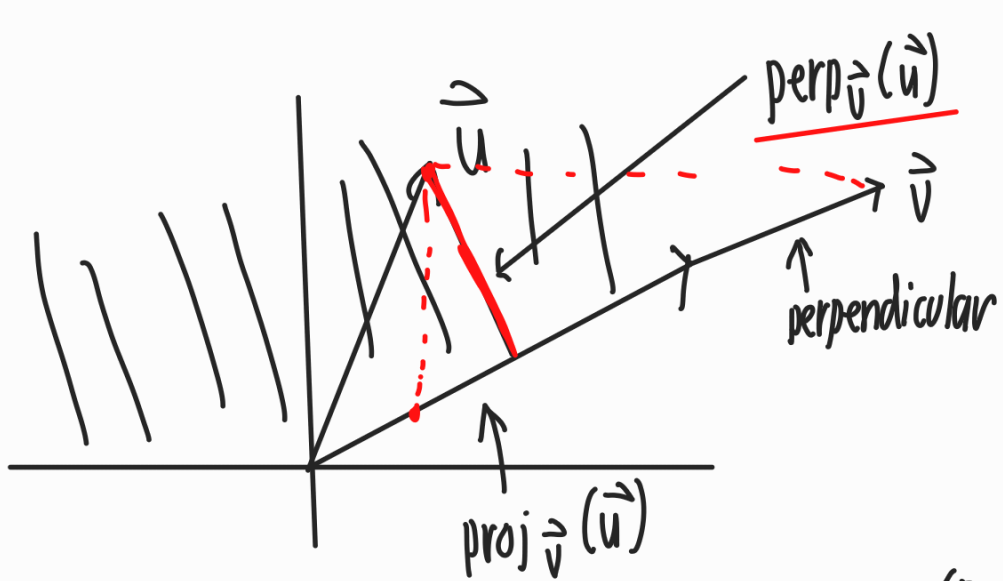
6. The ideas of minimum distance, nearest point, and best approximation



- Min dist from P to \vec{a}
- complication
- Is line \vec{a} a subspace? No!
- \vec{x} DNE when put in origin!

$(4, 1, -3) - (-4/3, 8/3, -8/3) = (40/3, 1/3, -19/3)$
Be prepared of fractions on midterm or final (More below)

least squares.



7. A bunch of exercises to do!

- (a) Find the orthogonal projection of \mathbf{x} onto the $\text{span}\{(2, 3)\}$. Draw the graph.
- (b) Find the orthogonal projection of \mathbf{x} onto the line $x + 3y = 0$. Draw the graph.
- (c) Given $\mathbf{x} = (7, -5, 9, -1)$ and $\mathbf{a} = (3, 0, 1, 2)$, find the vector components of \mathbf{x} along \mathbf{a} and orthogonal to \mathbf{a} .

Quiz 2 -1.3
- 1.5 from assignment 2
Not 2.1