

# Projection

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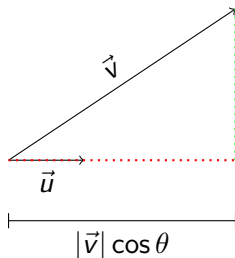
August 27, 2011

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Suppose  $|\vec{u}| = 1$ , then

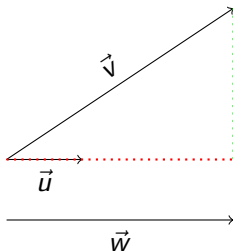
$$\vec{u} \cdot \vec{v} = |\vec{u}| |\vec{v}| \cos \theta = |\vec{v}| \cos \theta$$



How do we get a vector with the same direction as  $\vec{u}$  but with the magnitude of  $|\vec{v}| \cos \theta$ ?

Suppose  $|\vec{u}| = 1$ , and let

$$\vec{w} = (\vec{u} \cdot \vec{v})\vec{u} = (|\vec{v}| \cos \theta)\vec{u}$$



This is what we call “projection”.

What if  $|\vec{u}| \neq 1$ ? Consider the normalized vector  $\vec{u}_N$ .

$$\vec{u}_N = \frac{\vec{u}}{|\vec{u}|}$$

So we can get the same resultant vector by computing  $(\vec{u}_N \cdot \vec{v})\vec{u}_N$ .  
Expanding this:

$$\begin{aligned}(\vec{u}_N \cdot \vec{v})\vec{u}_N &= \left( \frac{\vec{u}}{|\vec{u}|} \cdot \vec{v} \right) \frac{\vec{u}}{|\vec{u}|} \\&= \frac{\vec{u} \cdot \vec{v}}{|\vec{u}|^2} \vec{u} \\&= \frac{\vec{u} \cdot \vec{v}}{\vec{u} \cdot \vec{u}} \vec{u}\end{aligned}$$

# The Projection Function

## Definition

Let  $\vec{u}, \vec{v} \in \mathbb{R}^n$  with  $\vec{u} \neq 0$ , the projection of vector  $\vec{v}$  into  $\vec{u}$  defined as:

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

## Remark

If  $|\vec{u}| = 1$  then the function simplifies to

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

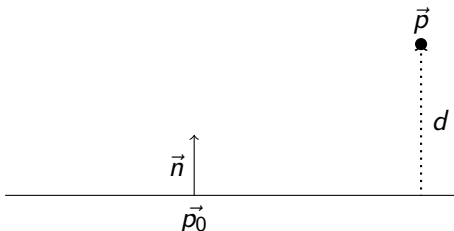
Suppose we have a vector  $u \in \mathbb{R}^m$  and a matrix  $A = [v_1 \ v_2 \ \cdots \ v_n]$  where  $v_1, v_2, \dots, v_n \in \mathbb{R}^m$  are linearly independent vectors. How do we come up with a matrix function  $\text{proj}_u(A)$  such that:

$$\text{proj}_u(A) = [\text{proj}_u(v_1) \ \text{proj}_u(v_2) \ \cdots \ \text{proj}_u(v_n)]?$$

$$\text{proj}_u(A) = u \frac{u^T A}{u^T u}$$

# Computing Distances from a Point to a Line/Hyperplane

Suppose we have a point  $\vec{p}$  and a line/hyperplane passing through point  $\vec{p}_0$  with normal  $\vec{n}$ .

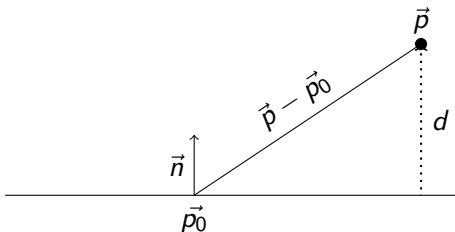




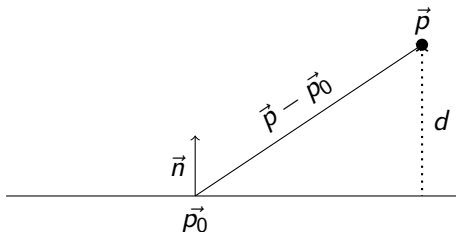
To find the distance between point  $\vec{p}$  and line  $L$ . Simply compute

### Point-Hyperplane Distance Formula

$$d = \frac{\vec{n} \cdot (\vec{p} - \vec{p}_0)}{|\vec{n}|} = \pm |\text{proj}_{\vec{n}} (\vec{p} - \vec{p}_0)|$$



$$d = \frac{\vec{n} \cdot (\vec{p} - \vec{p}_0)}{|\vec{n}|}$$



### Exercise

- What does the sign of the result mean?
- How would you do this for the implicit form?

## Exercise

Suppose we have a line  $L : \vec{l} = \vec{p}_0 + t\vec{u}$  and a point  $\vec{p}$ . What is the distance from  $\vec{p}$  to line  $L$ ?

## Solution

Simply compute

$$\frac{(\vec{p} - \vec{p}_0) \cdot \vec{u}^\perp}{|\vec{u}|}$$

Or alternatively, if we're in  $\mathbb{R}^2$ :

$$\frac{\vec{u} \times (\vec{p} - \vec{p}_0)}{|\vec{u}|}$$

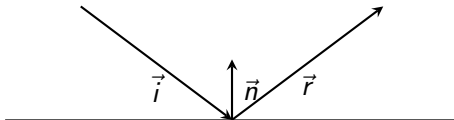
To find the intersection between a line  $L$  and a hyperplane  $P$ , simply find the solution for:

$$\frac{(\vec{p} - \vec{p}_L) \cdot \vec{u}}{|\vec{u}|} = 0$$

Where  $\vec{u}$  is the normal of  $P$ ,  $\vec{p}$  is a point in  $P$  and  $\vec{p}_L$  is the point in  $L$  to be solved.

### Exercise

How would you find the intersection of two lines in parametric form?



## Definition

The reflection of an incident vector  $\vec{i}$ , hitting a plane with normal  $\vec{n}$  is:

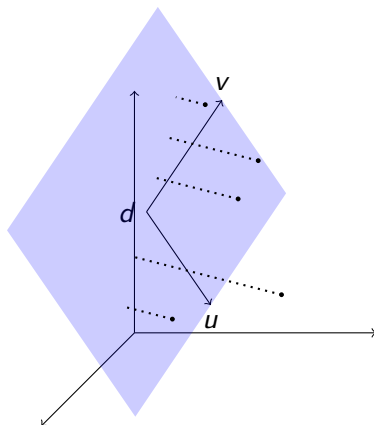
$$\vec{r}(\vec{i}, \vec{n}) = \vec{i} - 2 \operatorname{proj}_{\vec{n}}(\vec{i})$$

# Projection of Points into a Plane

Suppose we have a set of points  $S = \{p_1, p_2, \dots, p_n\} \subset \mathbb{R}^n$  and a plane  $P(s, t) = su + tv + d$  (where  $u, v, d \in \mathbb{R}^n$ ). How do we project  $S$  orthogonally to  $P$ ? That is, find the set  $S' = \{p'_1, p'_2, \dots, p'_n\}$  such that:

- ① All points in  $S'$  lie in  $P$ .
- ②  $p_i - p'_i$  is orthogonal to both  $u$  and  $v$  for all  $i$  (i.e.  $|p_i - p'_i|$  is minimized).

# Point Projection Visualization



This is how many algorithms compute shadows.

To make the derivation easier, suppose that the plane passes through the origin (so  $d = 0$ ). Let  $p_i \in S$ , we need to find an  $a, b \in \mathbb{R}$  such that:

$$p_i = au + bv = \begin{bmatrix} u & v \end{bmatrix} \begin{bmatrix} a \\ b \end{bmatrix}$$

However, since  $p_i$  is not in the  $\text{span}(u, v)$ , we have an inconsistent solution. However, we know that  $p_i - p'_i$  is orthogonal to both  $u$  and  $v$  so:

$$\left. \begin{aligned} (p_i - (au + bv)) \cdot u &= 0 \\ (p_i - (au + bv)) \cdot v &= 0 \end{aligned} \right\} \Rightarrow \left. \begin{aligned} p_i \cdot u &= (au + bv) \cdot u \\ p_i \cdot v &= (au + bv) \cdot v \end{aligned} \right\}$$



The following system

$$\left. \begin{aligned} p_i \cdot u &= (au + bv) \cdot u \\ p_i \cdot v &= (au + bv) \cdot v \end{aligned} \right\} \Rightarrow \left. \begin{aligned} u^T p_i &= u^T (au + bv) \\ v^T p_i &= v^T (au + bv) \end{aligned} \right\}$$

can be expressed in terms of a matrix equation:

$$\begin{bmatrix} u & v \end{bmatrix}^T p_i = \begin{bmatrix} u & v \end{bmatrix}^T \begin{bmatrix} u & v \end{bmatrix} \begin{bmatrix} a \\ b \end{bmatrix}$$

Note how this differs from the initial system that we formed that was inconsistent. Let  $B = \begin{bmatrix} u & v \end{bmatrix}$ , the system looks like:

$$B^T p_i = B^T B \begin{bmatrix} a \\ b \end{bmatrix}$$

Since  $u$  and  $v$  are linearly independent,  $B^T B$  is non-singular (verify). So,  $B^T B$  has an inverse, therefore:

$$\begin{bmatrix} a \\ b \end{bmatrix} = (B^T B)^{-1} B^T p_i$$

Here,  $a$  and  $b$  are the coordinates of  $p'_i$  on the basis  $\{u, v\}$ . To get the actual value of  $p'_i$ , simply do a linear combination:

$$p'_i = au + bv = B \begin{bmatrix} a \\ b \end{bmatrix} = B(B^T B)^{-1} B^T p_i$$

So the projection of point  $p_i$  to a plane defined by  $u, v$  passing through the origin is:

$$\text{proj}_B(p_i) = B(B^T B)^{-1} B^T p_i$$

where  $B = \begin{bmatrix} u & v \end{bmatrix}$ .

### Remark

This formula is applicable for any matrix  $B$  formed by linearly independent vectors. (i.e. if  $B = \begin{bmatrix} u_1 & u_2 & \cdots & u_n \end{bmatrix}$  such that  $u_i \in \mathbb{R}^m$  for all  $i$  and  $p_i \in \mathbb{R}^m$ ).

If the plane passes through a point  $p_0$ , then the function can be modified to:

$$\text{proj}_B(p_i - p_0) + p_0 = B(B^T B)^{-1} B^T (p_i - p_0) + p_0.$$

We've managed to define

$$\text{proj}_u(v)$$

when  $u$  and  $v$  are both vectors. We also defined

$$\text{proj}_B(v) \quad \text{and} \quad \text{proj}_u(A)$$

when  $B$  and  $A$  is a matrix.

Can we combine both to form

$$\text{proj}_B(A)?$$

# General Orthogonal Projection Function

Let  $S = \{u_1, u_2, \dots, u_n\} \subset \mathbb{R}^m$  be a linearly independent set and let  $T = \{v_1, v_2, \dots, v_n\} \subset \mathbb{R}^m$ . Construct  $B = [u_1 \ u_2 \ \cdots \ u_n]$  and  $A = [v_1 \ v_2 \ \cdots \ v_n]$ . The projection of  $T$  into  $\text{span}(S)$  is then:

$$\text{proj}_B(A) = B(B^T B)^{-1} B^T A$$

## Remarks

Note that the projection function simple degrades to the vector-vector and vector-matrix and matrix-vector projections when  $A$  or  $B$  is a  $1 \times m$  matrix (i.e. a vector).