

# Proof: Scalar Projection, Vector Projection and Finding the Angle Between Two Vectors

---

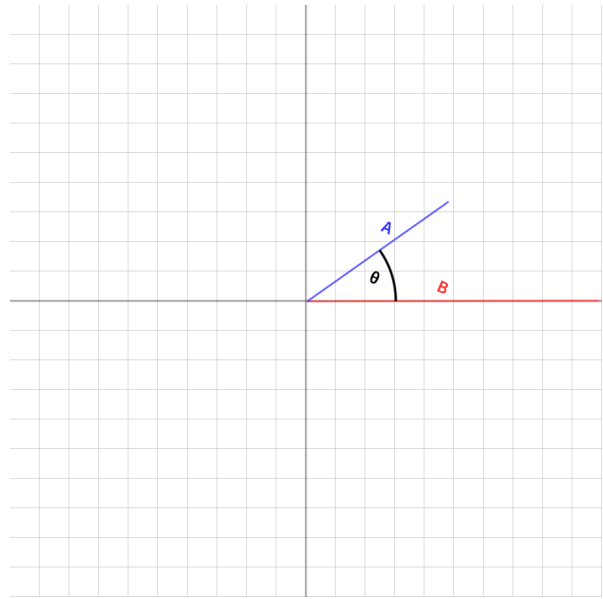
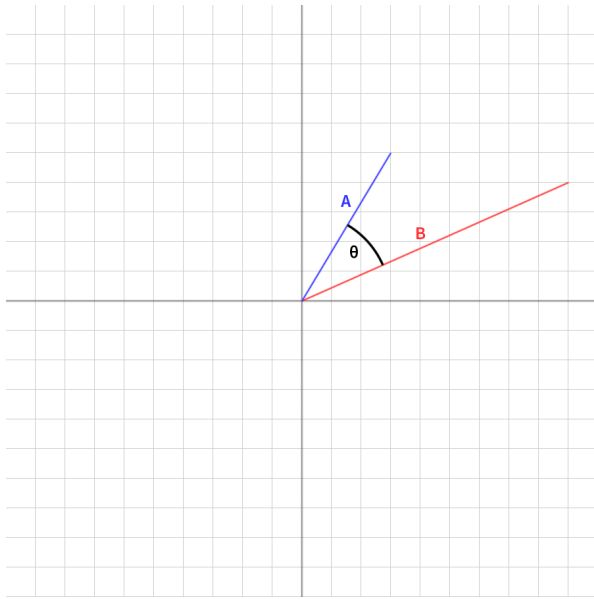
## Definitions

1. The Vector Projection can be thought of as “projecting” a given vector  $A$  onto another vector  $B$ . So if you have a given vector  $A = (7, 7)$  and  $B = (15, 0)$  the result,  $proj(A, B) = (7, 0)$ .
  2. The Scalar Projection can be thought of as the length or magnitude of the projection of a given vector  $A$  onto another vector  $B$ . So if you have a given vector  $A = (7, 7)$  and  $B = (15, 0)$  the result  $scalarProjection(A, B) = 7$ , which you can plainly see is also the magnitude or length of  $proj(A, B)$ .
  3. The angle between two vectors is pretty self-explanatory, but in this case we'll only talk about 2D until after the proof.
- 

## Relative Angle Orientation & Scalar Projection

The first thing to understand is that the relative angle between the two vectors ( $\theta$ ) is the only angle that matters for projections. This could be calculated if each vector had a known angle ( $\theta_A, \theta_B$ ) and taking the difference of them,  $\theta = \theta_A - \theta_B$  but this does not easily extrapolate into higher dimensions beyond 2D when you don't know the angles already. However a helpful understanding that can be grasped visually is the following,

---



This is to say that you can rotate both angles so that B is flat on the x-axis for easier comprehension of the system. In these circumstances we can also benefit from the polar-to-Cartesian transform of the x-axis component of a vector.

$$x = r * \cos(\theta)$$

$$r = \frac{x}{\cos(\theta)}$$

$$\theta = \cos^{-1}\left(\frac{x}{r}\right)$$

This tells us that we can get the x-axis length of A by multiplying the length of A ( $r$  here) by the cosine of the angle between A and B ( $\theta$ ). We can calculate the value of  $r$  with the following equation,

$$r = \sqrt{x^2 + y^2}$$

$$\|A\| = \sqrt{A_x^2 + A_y^2}$$

These two formula are the exact same thing, just written in two ways. From now on we'll refer to the length of  $A$  as  $\|A\|$  instead of  $r$ . So the formula becomes,

$$x = \|A\| * \cos(\theta)$$

Which is to say that  $x$  is the length of A whose component is projected onto B in the same way we treat normal vectors as breakdowns of components projected on the X and Y axis. This is the scalar product (*scalar Projection(A, B)*).

Interestingly enough to understand this another way we can use a different equation. If you recall from trigonometry that  $\cos(\theta)$  is the x-axis part of a vector whose length is 1, and that you can reduce a vector to a “unit vector” (a vector with a length of 1) via the formula,

$$\hat{B} = \left( \frac{B_x}{B_x + B_y}, \frac{B_y}{B_x + B_y} \right)$$

Then you can figure out that

$$\frac{B_x}{B_x + B_y} = \cos(\theta)$$

Given this you can alternatively calculate the scalar projection of (A,B) via the formula,

$$A \cdot \hat{B} = A_x \hat{B}_x + A_y \hat{B}_y$$

Where the  $\cdot$  symbol means “dot product”, which is defined as,

$$A \cdot B = \sum_{i=1}^n A_i B_i$$

for n-dimensions. This means the above formula ( $A \cdot \hat{B}$ ) turns out to the following,

$$A \cdot \hat{B} = A_x \hat{B}_x + A_y 0 = A_x \hat{B}_x$$

Due to the relative rotation of B acting as the x-axis and therefore not having a relative Y component. Just imagine it as the version of the two angles in the second figure above - with B laying flat on the X axis.

This means that you can equate the scalarProjection(A,B) (which we’ll just write as S(A,B)) to the two formula:

$$S(A, B) = A \cdot \hat{B} = \|A\| \cos(\theta)$$

---

## Calculating Cosine

Yet, how do we calculate cosine if we only have  $A$  and  $B$  and are unable to rotate them like in the above graphical representation? Well with some algebraic rearrangement we can deduce a formula from the definition of the scalar projection.

We know that,

$$A \cdot \hat{B} = \|A\| \cos(\theta)$$

Further we can define the relationship between  $B$  and  $\hat{B}$  via the formula,

$$B = \hat{B} * \|B\|$$
$$\hat{B} = \frac{B}{\|B\|}$$

And because dot-products have the commutative multiplication property,

$$A \cdot Bc = Ac \cdot B = c(A \cdot B)$$

We can deduce,

$$A \cdot B * \|B\|^{-1} = \frac{A \cdot B}{\|B\|}$$

Therefore,

$$\frac{A \cdot B}{\|B\|} = \|A\| \cos(\theta)$$

Which we can simplify into,

$$\frac{A \cdot B}{\|A\| * \|B\|} = \cos(\theta)$$

and derive the value of  $\theta$  via,

$$\theta = \cos^{-1}\left(\frac{A \cdot B}{\|A\| * \|B\|}\right)$$

Which is the angle between the two vectors.

---

## Vector Projection

Now, given that we know the angle between two vectors ( $\theta$ ) and the scalar project ( $s = S(A, B)$ ) we can now calculate the vector projection ( $v = V(A, B)$ ) very easily.

Recall the vector projection is a projection of the components of A onto B. All we have to do to do this is get the normalized vector of B ( $\hat{B}$ ) and multiply it by the length of A projected onto B - aka the Scalar Product ( $s$ ),

$$v = s\hat{B} = \|A\|\cos\theta * \frac{B}{\|B\|}$$

Now that might seem like a rather poor example, but what we can consider it as is that we are taking the vector length of A projected on B ( $s$ ) and then we're multiplying it by a unit-vector representing the direction of B ( $\hat{B}$ ) which gives us a vector in the direction of B with the length of the A vector components when projected onto B.

If that is confusing, an example will likely clear things up far better than words!

---

### Example 1

Given two vectors,

$$\begin{aligned}A &= (7, 7) \\B &= (15, 0) \\ \hat{B} &= (1, 0)\end{aligned}$$

The scalar projection is,

$$s = A \cdot \hat{B} = (7 * 1) + (7 * 0) = 7$$

Which - keep in mind is not the distance of A but the length of A projected onto B. Now given that, we can find the angle between the two,

$$\theta = \cos^{-1}\left(\frac{A \cdot B}{\|A\| * \|B\|}\right)$$

$$A \cdot B = (7 * 15) + (7 * 0) = 105$$

$$\|A\| = \sqrt{7^2 + 7^2} = \sqrt{98} = 9.8995$$

$$\|B\| = \sqrt{15^2 + 0^2} = \sqrt{15^2} = 15$$

$$\|A\| * \|B\| = 148.4925$$

$$\frac{A \cdot B}{\|A\| * \|B\|} = \frac{105}{148.4925} = 0.707$$

$$\cos^{-1}(0.707) = 0.7856 \text{ or } 45.01^\circ$$

Which makes sense - A is rising and extending at the same amount and B is an x-axis only vector. So there's a 45\* angle between them, with the 0.01 being the margin of error from rounding the various roots in the formula.

Given this, we can then perform the vector projection,

$$v = s\hat{B} = 7 * (1, 0) = (7, 0)$$

So we now can see - we projected the components of A onto B, yielding a vector in the direction of B ((1, 0)) but with length of the A component (7), yielding the vector ((7, 0)).

Hopefully that makes some sense.

## Conclusion

This has been a review of the manner I worked through in order to understand how Scalar Projection and Vector Projection work, as well as how to calculate the angle between two vectors. A useful bonus - and why this method is better than simple trigonometry is that it *scales to any number of dimensions*. Given we can define the magnitude operations (ex:  $\|A\|$ ) and the dot product on two vectors, we can calculate the angle between them, and perform all manner of projection on them.

Thus I felt - should I ever forget this, or need to explain to another person this would be a worthwhile explanation and/or proof.