

Quiz

You are given a procedure `transformation(A)` with the following spec:

- ▶ *input*: Mat A
- ▶ *output*: Mat M such that $M * A$ is a Mat in echelon form

Your job is to write each of the following procedures:

```
A = Mat(({ 'a', 'b' }, { 'A', 'B' }), { ('a', 'A'):one, ('b', 'B'):one, ('a', 'B'):one, ('b', 'A'):one })
>>> M=transformation(A)
>>> print(M*A)
```

```
          A  B
      -----
0  |  one one
1  |   0  0
```

The input to these is a list of Vecs:

- ▶ `rank(L)`
- ▶ `is_independent(L)`
- ▶ `basis(L)` returns a list of Vecs forming a basis for Span L

- ▶ `null_space_basis(A)` where A is a Mat
- ▶ `solve(A, b)` where A is a Mat and b is a Vec

For `solve`, you can also assume a procedure `echelon_solve(A,b)` that requires A to be in echelon form and solves $A\mathbf{x} = \mathbf{b}$.

Properties of orthogonality

To solve the Fire Engine Problem, we will use the Pythagorean Theorem in conjunction with the following simple observations:

Orthogonality Properties:

Property O1: If \mathbf{u} is orthogonal to \mathbf{v} then \mathbf{u} is orthogonal to $\alpha \mathbf{v}$ for every scalar α .

Property O2: If \mathbf{u} and \mathbf{v} are both orthogonal to \mathbf{w} then $\mathbf{u} + \mathbf{v}$ is orthogonal to \mathbf{w} .

Proof:

1. $\langle \mathbf{u}, \alpha \mathbf{v} \rangle = \alpha \langle \mathbf{u}, \mathbf{v} \rangle = \alpha 0 = 0$
2. $\langle \mathbf{u} + \mathbf{v}, \mathbf{w} \rangle = \langle \mathbf{u}, \mathbf{w} \rangle + \langle \mathbf{v}, \mathbf{w} \rangle = 0 + 0$

Example: $[1, 2] \cdot [2, -1] = 0$ so $[1, 2] \cdot [20, -10] = 0$

Example:

$$\begin{array}{rcl} [1, 2, 1] \cdot [1, -1, 1] & = & 0 \\ [0, 1, 1] \cdot [1, -1, 1] & = & 0 \\ \hline ([1, 2, 1] + [0, 1, 1]) \cdot [1, -1, 1] & = & 0 \end{array}$$

Decomposition of \mathbf{b} into parallel and perpendicular components

Definition: For any vector \mathbf{b} and any vector \mathbf{a} , define vectors $\mathbf{b}^{\parallel\mathbf{a}}$ and $\mathbf{b}^{\perp\mathbf{a}}$ to be the *projection of \mathbf{b} onto $\text{Span}\{\mathbf{a}\}$* and the *projection of \mathbf{b} orthogonal to \mathbf{a}* if

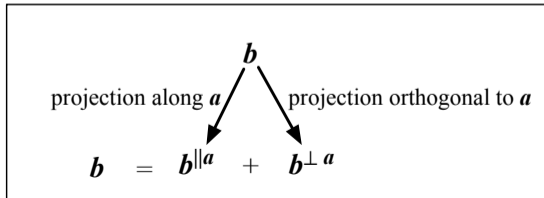
$$\mathbf{b} = \mathbf{b}^{\parallel\mathbf{a}} + \mathbf{b}^{\perp\mathbf{a}}$$

and there is a scalar $\sigma \in \mathbb{R}$ such that

$$\mathbf{b}^{\parallel\mathbf{a}} = \sigma \mathbf{a}$$

and

$\mathbf{b}^{\perp\mathbf{a}}$ is orthogonal to \mathbf{a}



Decomposition of \mathbf{b} into $\mathbf{b}^{\parallel\mathbf{a}}$ and $\mathbf{b}^{\perp\mathbf{a}}$

$\mathbf{b} = \mathbf{b}^{\parallel\mathbf{a}} + \mathbf{b}^{\perp\mathbf{a}}$ and there is a scalar $\sigma \in R$ such that $\mathbf{b}^{\parallel\mathbf{a}} = \sigma \mathbf{a}$ and $\mathbf{b}^{\perp\mathbf{a}}$ is orthogonal to \mathbf{a}

Example: $\mathbf{b} = [b_1, b_2]$, $\mathbf{a} = [1, 0]$.

Then $\mathbf{b}^{\parallel\mathbf{a}} = [b_1, 0]$ Note that $[b_1, 0] = \mathbf{b}_1 [1, 0]$.

$$\mathbf{b}^{\perp\mathbf{a}} = \mathbf{b} - \mathbf{b}^{\parallel\mathbf{a}} = [b_1, b_2] - [b_1, 0] = [0, b_2]$$

Example: $\mathbf{b} = [10, 20, 30]$ and $\mathbf{a} = [-1, 2, 1]$.

I claim $\mathbf{b}^{\parallel\mathbf{a}} = [-10, 20, 10]$ and therefore $\mathbf{b}^{\perp\mathbf{a}} = [10, 20, 30] - [-10, 20, 10] = [20, 0, 20]$.

Are these correct?

- ▶ Check if $\mathbf{b}^{\parallel\mathbf{a}} = \sigma \mathbf{a}$ for some σ ... Yes, $\sigma = 10$
- ▶ Check if $\mathbf{b}^{\perp\mathbf{a}}$ is orthogonal to \mathbf{a} ... $[20, 0, 20] \cdot [-1, 2, 1] = 0$, so yes

Orthogonality helps solve the *fire engine* problem

Fire Engine Lemma:

- ▶ Let \mathbf{b} be a vector.
- ▶ Let \mathbf{a} be a nonzero vector

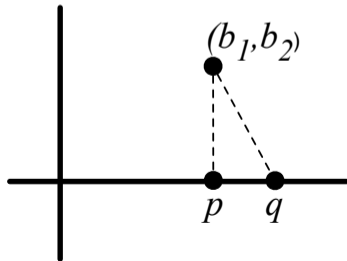


Then $\mathbf{b}^{\parallel\mathbf{a}}$ is the point on the line $\{\alpha \mathbf{a} : \alpha \in \mathbb{R}\}$ that is closest to \mathbf{b} , and the distance is $\|\mathbf{b}^{\perp\mathbf{a}}\|$.

Example: Line is the x-axis, i.e. the set $\{(x, y) : y = 0\}$, and point is (b_1, b_2) .

Lemma states: closest point on the line is $\mathbf{p} = (b_1, 0)$.

- ▶ For any other point \mathbf{q} , the points $\mathbf{b} = (b_1, b_2)$, $\mathbf{b}^{\parallel\mathbf{a}}$, and \mathbf{q} form a right triangle.
- ▶ Since \mathbf{q} is different from $\mathbf{b}^{\parallel\mathbf{a}}$, the base is nonzero.
- ▶ By the Pythagorean Theorem, the hypotenuse's length is greater than the height.



Orthogonality helps solve the *fire engine* problem

Fire Engine Lemma:

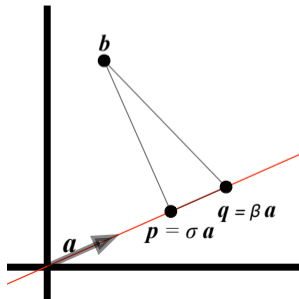
- ▶ Let \mathbf{b} be a vector.
- ▶ Let \mathbf{a} be a nonzero vector
- ▶

Then $\mathbf{b}^{\parallel\mathbf{a}}$ is the point on the line $\{\alpha \mathbf{a} : \alpha \in \mathbb{R}\}$ that is closest to \mathbf{b} , and the distance is $\|\mathbf{b}^{\perp\mathbf{a}}\|$.

Proof: Let L be the line. Let $\mathbf{p} = \mathbf{b}^{\parallel\mathbf{a}}$. Let \mathbf{q} be any point on L . The three points \mathbf{q} , \mathbf{p} , and \mathbf{b} form a triangle.

- ▶ Since \mathbf{p} and \mathbf{q} are both on L , they are both multiples of \mathbf{a} , so their difference $\mathbf{p} - \mathbf{q}$ is also a multiple of \mathbf{a} .
- ▶ Hence, since $\mathbf{b}^{\perp\mathbf{a}}$ is orthogonal to \mathbf{a} , it is also orthogonal to $\mathbf{p} - \mathbf{q}$. Note $\mathbf{b}^{\perp\mathbf{a}} = \mathbf{b} - \mathbf{p}$.
- ▶ Hence by the Pythagorean Theorem,

$$\|\mathbf{b} - \mathbf{q}\|^2 = \|\mathbf{p} - \mathbf{q}\|^2 + \|\mathbf{b} - \mathbf{p}\|^2.$$



Orthogonality helps solve the *fire engine* problem

Fire Engine Lemma:

- ▶ Let \mathbf{b} be a vector.
- ▶ Let \mathbf{a} be a nonzero vector

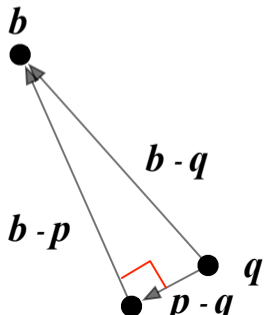
▶

Then $\mathbf{b}^{\parallel\mathbf{a}}$ is the point on the line $\{\alpha \mathbf{a} : \alpha \in \mathbb{R}\}$ that is closest to \mathbf{b} , and the distance is $\|\mathbf{b}^{\perp\mathbf{a}}\|$.

Proof: Let L be the line. Let $\mathbf{p} = \mathbf{b}^{\parallel\mathbf{a}}$. Let \mathbf{q} be any point on L . The three points \mathbf{q} , \mathbf{p} , and \mathbf{b} form a triangle.

- ▶ Since \mathbf{p} and \mathbf{q} are both on L , they are both multiples of \mathbf{a} , so their difference $\mathbf{p} - \mathbf{q}$ is also a multiple of \mathbf{a} .
- ▶ Hence, since $\mathbf{b}^{\perp\mathbf{a}}$ is orthogonal to \mathbf{a} , it is also orthogonal to $\mathbf{p} - \mathbf{q}$. Note $\mathbf{b}^{\perp\mathbf{a}} = \mathbf{b} - \mathbf{p}$.
- ▶ Hence by the Pythagorean Theorem,

$$\|\mathbf{b} - \mathbf{q}\|^2 = \|\mathbf{p} - \mathbf{q}\|^2 + \|\mathbf{b} - \mathbf{p}\|^2.$$



Decomposition of \mathbf{b} into parallel and perpendicular components: example

For any vector \mathbf{b} and any vector \mathbf{a} , define vectors $\mathbf{b}^{\parallel\mathbf{a}}$ and $\mathbf{b}^{\perp\mathbf{a}}$

- ▶ $\mathbf{b} = \mathbf{b}^{\parallel\mathbf{a}} + \mathbf{b}^{\perp\mathbf{a}}$, and
- ▶ there is a scalar $\sigma \in R$ such that $\mathbf{b}^{\parallel\mathbf{a}} = \sigma \mathbf{a}$, and
- ▶ $\mathbf{b}^{\perp\mathbf{a}}$ is orthogonal to \mathbf{a}

Example: What if \mathbf{a} is the zero vector?

In this case, the only vector $\mathbf{b}^{\parallel\mathbf{a}}$ satisfying the second equation is the zero vector.

According to first equation, $\mathbf{b}^{\perp\mathbf{a}}$ must equal \mathbf{b} .

Fortunately, this choice of $\mathbf{b}^{\perp\mathbf{a}}$ does satisfy third equation: $\mathbf{b}^{\perp\mathbf{a}}$ is orthogonal to \mathbf{a} .

Indeed, every vector is orthogonal to \mathbf{a} when \mathbf{a} is the zero vector.

What is the point in $\text{Span}\{\mathbf{0}\}$ closest to \mathbf{b} ?

The *only* point in $\text{Span}\{\mathbf{0}\}$ is the zero vector...

so that must be the closest point to \mathbf{b} , and the distance to \mathbf{b} is $\|\mathbf{b}\|$.

Computing the projections

$$\mathbf{b} = \mathbf{b}^{\parallel \mathbf{a}} + \mathbf{b}^{\perp \mathbf{a}}$$

$$\mathbf{b}^{\parallel \mathbf{a}} = \sigma \mathbf{a}$$

$\mathbf{b}^{\perp \mathbf{a}}$ is orthogonal to \mathbf{a}

If $\mathbf{a} = \mathbf{0}$ then $\mathbf{b}^{\parallel \mathbf{a}} = \mathbf{0}$

What if $\mathbf{a} \neq \mathbf{0}$? Need to compute σ ...

▶ $\langle \mathbf{b}^{\perp \mathbf{a}}, \mathbf{a} \rangle = 0$. Substitute for $\mathbf{b}^{\perp \mathbf{a}}$: $\langle \mathbf{b} - \mathbf{b}^{\parallel \mathbf{a}}, \mathbf{a} \rangle = 0$.

▶ Substitute for $\mathbf{b}^{\parallel \mathbf{a}}$: $\langle \mathbf{b} - \sigma \mathbf{a}, \mathbf{a} \rangle = 0$.

▶ Using linearity and homogeneity of inner product,

$$\langle \mathbf{b}, \mathbf{a} \rangle - \sigma \langle \mathbf{a}, \mathbf{a} \rangle = 0$$

▶ Solving for σ , we obtain

$$\sigma = \frac{\langle \mathbf{b}, \mathbf{a} \rangle}{\langle \mathbf{a}, \mathbf{a} \rangle}$$

In the special case in which $\|\mathbf{a}\| = 1$, the denominator $\langle \mathbf{a}, \mathbf{a} \rangle = 1$ so

$$\sigma = \langle \mathbf{b}, \mathbf{a} \rangle$$

Quiz: Write `project_along(b, a)` to return the vector $\mathbf{b}^{\parallel \mathbf{a}}$

Answer: `def project_along(b, a): return ((b*a)/(a*a))*a` **Almost.**

Best: `def project_along(b, a): return ((b*a)/(a*a) if a*a != 0 else 0)*a`

Computing the projections

$$\mathbf{b} = \mathbf{b}^{\parallel\mathbf{a}} + \mathbf{b}^{\perp\mathbf{a}}$$

$$\mathbf{b}^{\parallel\mathbf{a}} = \sigma \mathbf{a}$$

$\mathbf{b}^{\perp\mathbf{a}}$ is orthogonal to \mathbf{a}

▶ $\sigma = \frac{\langle \mathbf{b}, \mathbf{a} \rangle}{\langle \mathbf{a}, \mathbf{a} \rangle}$

▶ However, if $\mathbf{a} = \mathbf{0}$ then $\sigma = 0$.

▶ `def project_along(b, a):`
 `sigma = (b*a)/(a*a) if a*a != 0 else 0`
 `return sigma * a`

Quiz: Use `project_along(b, a)` to write the procedure
 `project_orthogonal_1(b, a)`

that returns $\mathbf{b}^{\perp\mathbf{a}}$

```
def project_orthogonal_1(b, a): return b - project_along(b, a)
```

Projecting along “nearly zero” vectors

Mathematically, this procedure is correct:

```
def project_along(b, a):  
    sigma = (b*a)/(a*a) if a*a != 0 else 0  
    return sigma * a
```

However, because of floating-point roundoff error, we need to make a slight change.

Often the vector **a** will be not a truly zero vector but practically it will be zero.

If the entries of **a** are tiny, the procedure should treat **a** as a zero vector: `sigma` should be assigned zero.

We will consider **a** to be a zero vector if its squared norm is no more than, say, 10^{-20} .

Revised version:

```
def project_along(b, a):  
    sigma = (b*a)/(a*a) if a*a > 1e-20 else 0  
    return sigma * a
```

Solution to the *fire engine* problem

Example:

$\mathbf{a} = [6, 2]$ and $\mathbf{b} = [2, 4]$.

The closest point on the line $\{\alpha \mathbf{a} : \alpha \in \mathbb{R}\}$ is the point $\mathbf{b}^{\parallel \mathbf{a}} = \sigma \mathbf{a}$ where

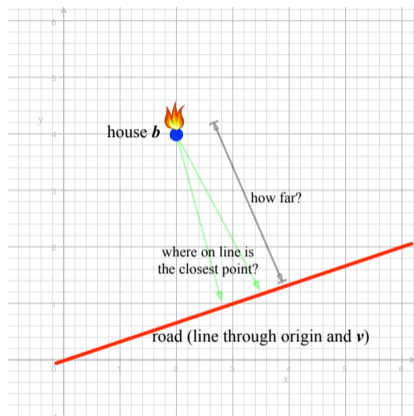
$$\begin{aligned}\sigma &= \frac{\mathbf{a} \cdot \mathbf{b}}{\mathbf{a} \cdot \mathbf{a}} \\ &= \frac{6 \cdot 2 + 2 \cdot 4}{6 \cdot 6 + 2 \cdot 2} \\ &= \frac{20}{40} \\ &= \frac{1}{2}\end{aligned}$$

Thus the point closest to \mathbf{b} is $\frac{1}{2} [6, 2] = [3, 1]$.

The distance to \mathbf{b} is

$$\|\mathbf{b}^{\perp \mathbf{a}}\| = \|[2, 4] - [3, 1]\| = \|[-1, 3]\| = \sqrt{10}$$

which is just under 3.5, the length of the firehose.



Best approximation

The *fire engine* problem can be restated as finding the vector on the line that “best approximates” the given vector **b**.

By “best approximation”, we just mean closest.

This notion of “best approximates” comes up again and again:

- ▶ in least-squares, a fundamental data analysis technique,
- ▶ image compression,
- ▶ in principal component analysis, another data analysis technique, and
- ▶ in latent semantic analysis, an information retrieval technique.

Towards solving the higher-dimensional version of *best approximation*

The fire engine problem can be stated thus:

Computational Problem: *Closest point in the span of a single vector*

Given a vector \mathbf{b} and a vector \mathbf{a} over the reals,
find the vector in $\text{Span}\{\mathbf{a}\}$ closest to \mathbf{b} .

A natural generalization of the *fire engine* problem is this:

Computational Problem: *Closest point in the span of several vectors*

Given a vector \mathbf{b} and vectors $\mathbf{a}_1, \dots, \mathbf{a}_n$ over the reals,
find the vector in $\text{Span}\{\mathbf{a}_1, \dots, \mathbf{a}_n\}$ closest to \mathbf{b} .

We will study this problem next.