

# Announcements April 20

- Final exam Mon Apr 27 6 am – Tue Apr 28 11:59 pm.
- Final exam open for 3 hrs, planned to be a 1 hour exam.
- CIOS-like survey on Canvas Quizzes
- Chapter 6 WeBWorkS are not graded.
- My office hours Monday 3-4, Wed 2-3, and by appointment
- Stay tuned for review sessions...
- TA office hours on Blue Jeans (you can go to any of these!)
  - ▶ Isabella Wed 11-12
  - ▶ Kyle Wed 3-5, Thu 1-3
  - ▶ Kalen Mon/Wed 1-2
  - ▶ Sidhanth Tue 10-12
- Supplemental problems & practice exams on master web site
- Counseling Center: <http://counseling.gatech.edu> ▶ Click

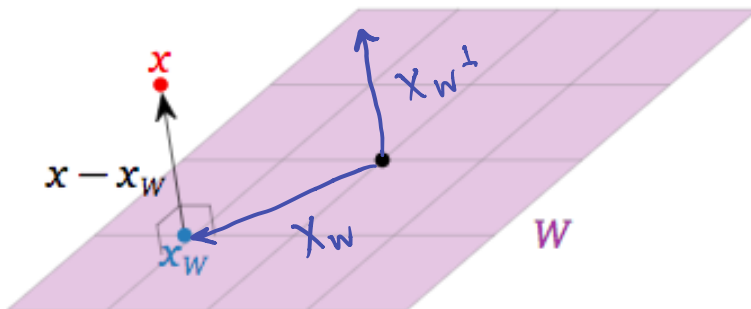
# Section 6.3

## Orthogonal projection

# Orthogonal Projections

Let  $v$  be a vector in  $\mathbb{R}^n$  and  $W$  a subspace of  $\mathbb{R}^n$ .

The **orthogonal projection** of  $v$  onto  $W$  is the vector obtained by drawing a line segment from  $v$  to  $W$  that is perpendicular to  $W$ .



**Fact.** The following three things are all the same:

- The orthogonal projection of  $v$  onto  $W$
- The vector  $v_W$  (the  $W$ -part of  $v$ )
- The closest vector in  $W$  to  $v$

# Orthogonal Projections

**Theorem.** Let  $W = \text{Col}(A)$ . For any vector  $v$  in  $\mathbb{R}^n$ , the equation

$$A^T A x = A^T v$$

is consistent and the orthogonal projection  $v_W$  is equal to  $Ax$  where  $x$  is any solution.

**Why?** Choose  $\hat{x}$  so that  $A\hat{x} = v_W$ . We know  $v - v_W = v - A\hat{x}$  is in  $W^\perp = \text{Nul}(A^T)$  and so

$$0 = A^T(v - A\hat{x}) = A^T v - A^T A \hat{x}$$

$$\rightsquigarrow A^T A \hat{x} = A^T v$$

$$\left( \begin{array}{c} \equiv \\ \equiv \\ \equiv \end{array} \right) \left( \begin{array}{c} \\ \\ \end{array} \right) = 0$$

A

Point: Being  $\perp$  to  $\text{Col}(A)$

$\iff$  being in  $\text{Nul } A^T$

geometry

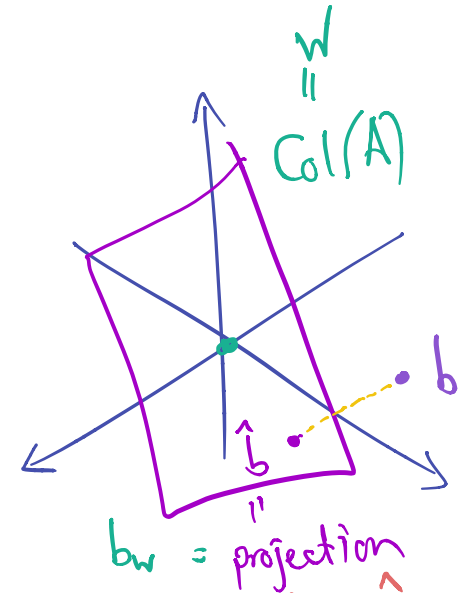
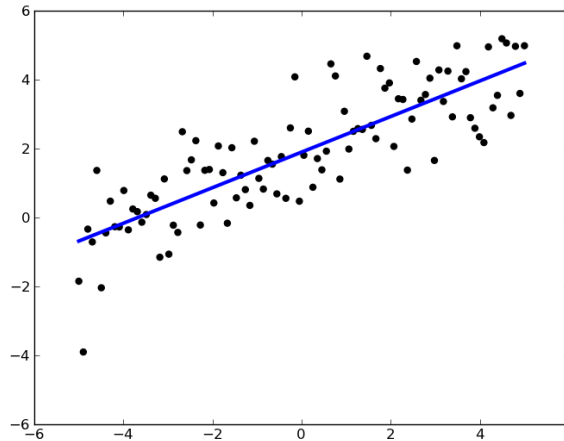
algebra

# Section 6.5

## Least Squares Problems

# Least Squares problems

What if we can't solve  $Ax = b$ ? How can we solve it as closely as possible?



$\leadsto$  Solve  $Ax = \hat{b}$   
 $\leadsto Ax = \hat{b}$   
looking for

To solve  $Ax = b$  as closely as possible, we orthogonally project  $b$  onto  $\text{Col}(A)$ ; call the result  $\hat{b}$ . Then solve  $Ax = \hat{b}$ . This is the *least squares solution* to  $Ax = b$ .

# Outline of Section 6.5

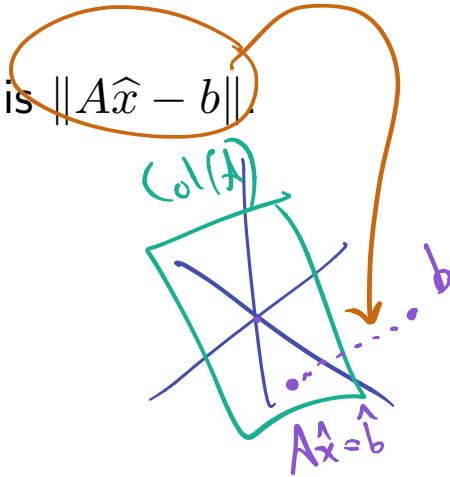
- The method of least squares
- Application to best fit lines/planes
- Application to best fit curves

# Least squares solutions

$A = m \times n$  matrix.

A **least squares solution** to  $Ax = b$  is an  $\hat{x}$  in  $\mathbb{R}^n$  so that  $A\hat{x}$  is as close as possible to  $b$ .

The error is  $\|A\hat{x} - b\|$ .



The  $A\hat{x}$ 's are all vectors in...  $\text{Col}(A)$

► Demo



# Least squares solutions

A **least squares solution** to  $Ax = b$  is an  $\hat{x}$  in  $\mathbb{R}^n$  so that  $A\hat{x}$  is as close as possible to  $b$ .

The error is  $\|A\hat{x} - b\|$ .

**Theorem.** The least squares solutions to  $Ax = b$  are the solutions to

$$(A^T A)x = (A^T b)$$

So this is just like what we did before when we were finding the projection of  $b$  onto  $\text{Col}(A)$ . But now we just solve and don't (necessarily) multiply the solution by  $A$ .

# Least squares solutions

## Example

**Theorem.** The least squares solutions to  $Ax = b$  are the solutions to

$$(A^T A)x = (A^T b)$$

Find the least squares solutions to  $Ax = b$  for this  $A$  and  $b$ :

$$A = \begin{pmatrix} 0 & 1 \\ 1 & 1 \\ 2 & 1 \end{pmatrix}$$

$$b = \begin{pmatrix} 6 \\ 0 \\ 0 \end{pmatrix}$$

What is the error?

Is there an actual solution?  
(consistent?)

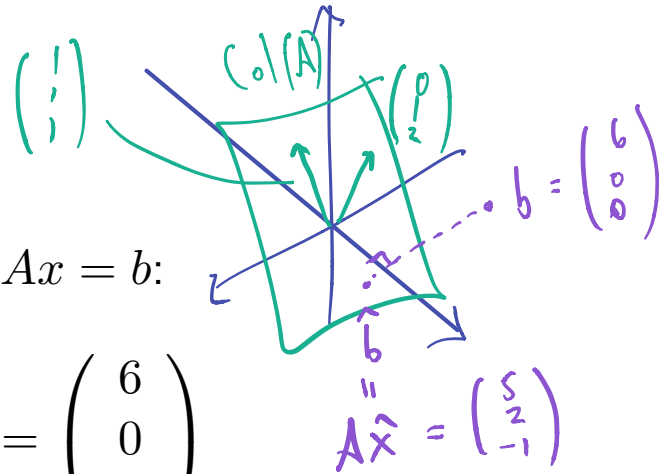
No.

# Least squares solutions

Example

$$\text{Formula: } (A^T A)x = (A^T b)$$

Find the least squares solution/error to  $Ax = b$ :



$$A = \begin{pmatrix} 0 & 1 \\ 1 & 1 \\ 2 & 1 \end{pmatrix} \quad b = \begin{pmatrix} 6 \\ 0 \\ 0 \end{pmatrix}$$

$$A^T A = \begin{pmatrix} 0 & 1 & 2 \\ 1 & 1 & 1 \end{pmatrix} \begin{pmatrix} 0 & 1 \\ 1 & 1 \\ 2 & 1 \end{pmatrix} = \begin{pmatrix} 5 & 3 \\ 3 & 3 \end{pmatrix} \quad \begin{pmatrix} 0 & 1 & 2 \\ 1 & 1 & 1 \end{pmatrix} \begin{pmatrix} 6 \\ 0 \\ 0 \end{pmatrix} = \begin{pmatrix} 0 \\ 6 \end{pmatrix} \quad \text{setup}$$

$$\begin{pmatrix} 5 & 3 \\ 3 & 3 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 6 \end{pmatrix} \rightsquigarrow \left( \begin{array}{cc|c} 5 & 3 & 0 \\ 3 & 3 & 6 \end{array} \right) \rightsquigarrow \left( \begin{array}{cc|c} 1 & 1 & 2 \\ 5 & 3 & 0 \end{array} \right) \quad \text{row red.}$$

$$\rightsquigarrow \left( \begin{array}{cc|c} 1 & 1 & 2 \\ 0 & -2 & -10 \end{array} \right) \rightsquigarrow \left( \begin{array}{cc|c} 1 & 1 & 2 \\ 0 & 1 & 5 \end{array} \right) \rightsquigarrow \left( \begin{array}{cc|c} 1 & 0 & -3 \\ 0 & 1 & 5 \end{array} \right)$$

$$\rightsquigarrow \begin{matrix} x_1 = -3 \\ x_2 = 5 \end{matrix} \quad \hat{x} = \begin{pmatrix} -3 \\ 5 \end{pmatrix}$$

$$\hat{b} = \begin{pmatrix} 0 & 1 \\ 1 & 1 \\ 2 & 1 \end{pmatrix} \begin{pmatrix} -3 \\ 5 \end{pmatrix} = \begin{pmatrix} 5 \\ 2 \\ -1 \end{pmatrix}$$

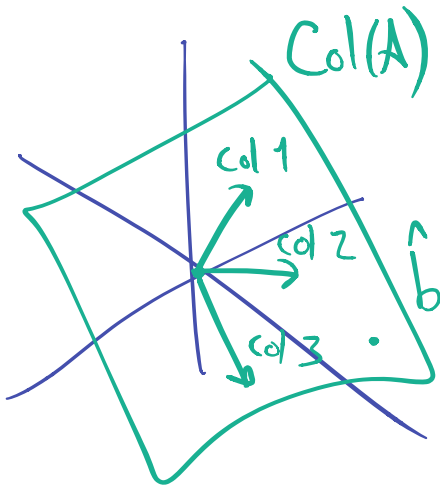
$$\text{Error: } \|b - \hat{b}\| = \left\| \begin{pmatrix} 6 \\ 0 \\ 0 \end{pmatrix} - \begin{pmatrix} 5 \\ 2 \\ -1 \end{pmatrix} \right\| = \left\| \begin{pmatrix} 1 \\ -2 \\ 1 \end{pmatrix} \right\| = \sqrt{6}$$

# Least squares solutions

**Theorem.** Let  $A$  be an  $m \times n$  matrix. The following are equivalent:

1.  $Ax = b$  has a unique least squares solution for all  $b$  in  $\mathbb{R}^m$
2. The columns of  $A$  are linearly independent
3.  $A^T A$  is invertible

In this case the least squares solution is  $(A^T A)^{-1}(A^T b)$ .



$$\boxed{A^T A} x = A^T b$$

unique soln  $\iff A^T A$  invertible.

# Application

## Best fit lines

**Problem.** Find the best-fit line through  $(0, 6)$ ,  $(1, 0)$ , and  $(2, 0)$ .

Want  $y = Mx + B$  (as close as possible)

Given points  $\rightsquigarrow$   $6 = M \cdot 0 + 1 \cdot B$

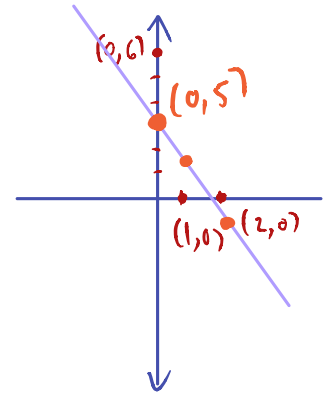
$$0 = M \cdot 1 + 1 \cdot B$$

$$0 = M \cdot 2 + 1 \cdot B$$

$$\rightsquigarrow \begin{matrix} & M & B \\ \begin{pmatrix} 0 & 1 \\ 1 & 1 \\ 2 & 1 \end{pmatrix} & X & = & \begin{pmatrix} 6 \\ 0 \\ 0 \end{pmatrix} \end{matrix}$$

as above  $\rightsquigarrow \hat{X} = \begin{pmatrix} -3 \\ 5 \end{pmatrix}$

$$\rightsquigarrow M = -3, B = 5 \rightsquigarrow y = -3x + 5$$



y-int: 6, 2.4, 6.5  
slope: -2, -4  
-3, -1

▶ Demo

# Best fit lines

## Poll

What does the best fit line minimize?

1. the sum of the squares of the distances from the data points to the line
2. the sum of the squares of the vertical distances from the data points to the line
3. the sum of the squares of the horizontal distances from the data points to the line
4. the maximal distance from the data points to the line

*y*-vals of orig pts  
↓

$$\left\| \begin{pmatrix} 6 \\ 0 \\ 0 \end{pmatrix} - \begin{pmatrix} 5 \\ 2 \\ -1 \end{pmatrix} \right\| = \sqrt{(6-5)^2 + (0-2)^2 + (0-(-1))^2}$$

# Least Squares Problems

## More applications

Determine the least squares problem  $Ax = b$  to find the best parabola  $y = Cx^2 + Dx + E$  for the points:

$$(0, 0), (2, 0), (3, 0), (0, 1)$$

$$(0, 0): 0 = 0 \cdot C + 0 \cdot D + 1 \cdot E$$

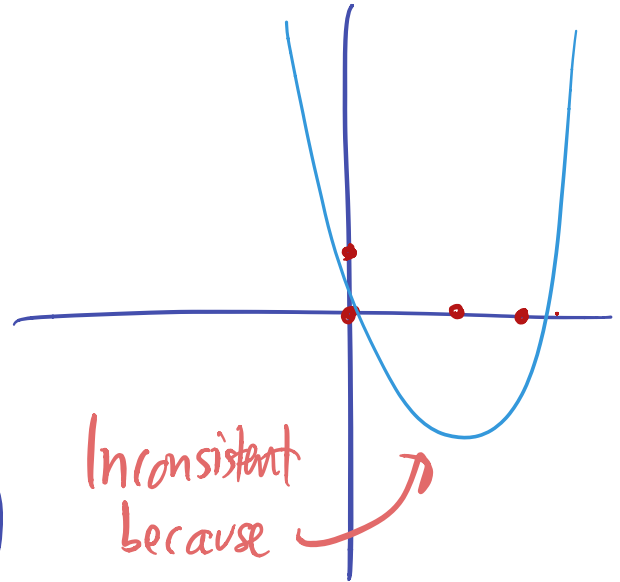
$$(2, 0): 0 = 4 \cdot C + 2 \cdot D + 1 \cdot E$$

$$(3, 0): 0 = 9 \cdot C + 3 \cdot D + 1 \cdot E$$

$$(0, 1): 1 = 0 \cdot C + 0 \cdot D + 1 \cdot E$$

► Demo

$$\begin{pmatrix} 0 & 0 & 1 \\ 4 & 2 & 1 \\ 9 & 3 & 1 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} C \\ D \\ E \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \\ 1 \end{pmatrix}$$

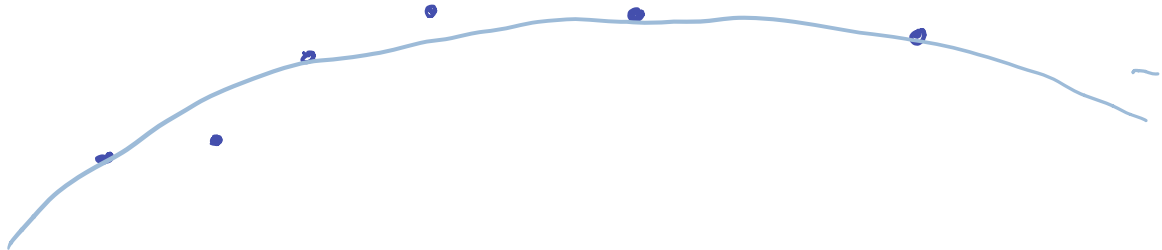


# Least Squares Problems

## More applications

Determine the least squares problem  $Ax = b$  to find the best fit ellipse  $Cx^2 + Dxy + Ey^2 + Fx + Gy + H = 0$  for the points:

$$(0, 0), (2, 0), (3, 0), (0, 1)$$



Gauss invented the method of least squares to predict the orbit of the asteroid Ceres as it passed behind the sun in 1801.

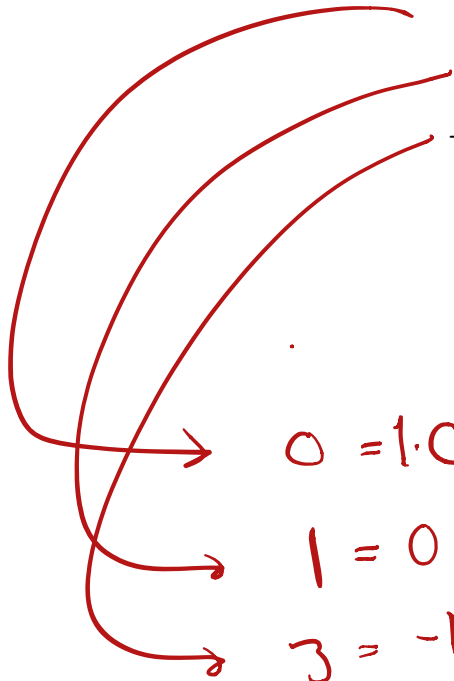


# Least Squares Problems

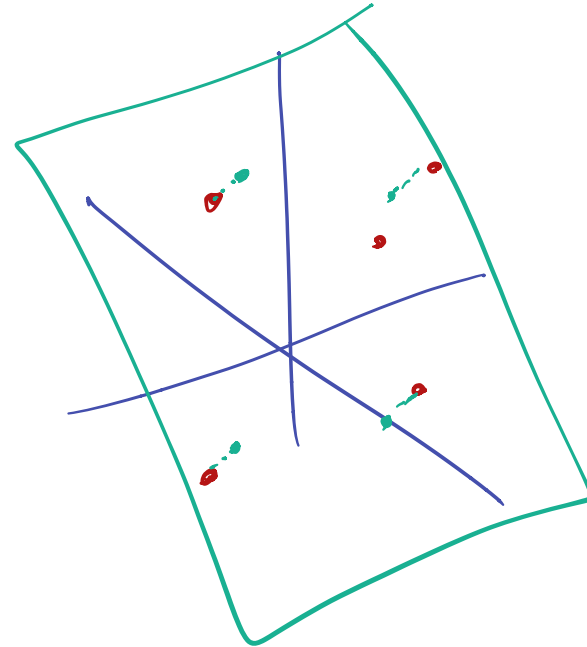
Best fit plane

Determine the least squares problem  $Ax = b$  to find the best fit linear function  $f(x, y) = Cx + Dy + E$

$x$	$y$	$f(x, y)$
1	0	0
0	1	1
-1	0	3
0	-1	4



$$\begin{aligned}0 &= 1 \cdot C + 0 \cdot D + 1 \cdot E \\1 &= 0 \cdot C + 1 \cdot D + 1 \cdot E \\3 &= -1 \cdot C + 0 \cdot D + 1 \cdot E \\&\text{etc...}\end{aligned}$$



## Summary of Section 6.5

- A **least squares solution** to  $Ax = b$  is an  $\hat{x}$  in  $\mathbb{R}^n$  so that  $A\hat{x}$  is as close as possible to  $b$ .
- The error is  $\|A\hat{x} - b\|$ .
- The least squares solutions to  $Ax = b$  are the solutions to  $(A^T A)x = (A^T b)$ .
- To find a best fit line/parabola/etc. write the general form of the line/parabola/etc. with unknown coefficients and plug in the given points to get a system of linear equations in the unknown coefficients.