

Announcements

Wednesday, November 28

- ▶ Please fill out your CIOS survey!

If 85% of the class completes the survey by 11:59pm on December 7, then we will drop two quizzes instead of one.

- ▶ Final exam time: **Tuesday, December 11, 6–8:50pm.**
- ▶ WeBWork 6.6, 7.1, 7.2 are due today.
- ▶ No quiz on Friday! But this is the only recitation on chapter 7.
- ▶ My office is Skiles 244 and Rabinoffice hours are: Mondays, 12–1pm; Wednesdays, 1–3pm.

Section 7.5

The Method of Least Squares

Motivation

We now are in a position to solve the motivating problem of this third part of the course:

Problem

Suppose that $Ax = b$ does not have a solution. What is the best possible approximate solution?

To say $Ax = b$ does not have a solution means that b is not in $\text{Col } A$.

The closest possible \hat{b} for which $Ax = \hat{b}$ does have a solution is $\hat{b} = b_{\text{Col } A}$.

Then $A\hat{x} = \hat{b}$ is a consistent equation.

A solution \hat{x} to $A\hat{x} = \hat{b}$ is a **least squares solution**.

Least Squares Solutions

Let A be an $m \times n$ matrix.

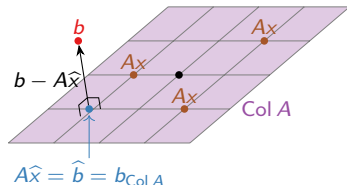
Definition

A **least squares solution** of $Ax = b$ is a vector \hat{x} in \mathbf{R}^n such that

$$\|b - A\hat{x}\| \leq \|b - Ax\|$$

for all x in \mathbf{R}^n .

Note that $b - A\hat{x}$
is in $(\text{Col } A)^\perp$.



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In other words, a least squares solution \hat{x} solves $Ax = b$ as closely as possible.

Equivalently, a least squares solution to $Ax = b$ is a vector \hat{x} in \mathbf{R}^n such that

$$A\hat{x} = \hat{b} = b_{\text{Col } A}.$$

This is because \hat{b} is the closest vector to b such that $A\hat{x} = \hat{b}$ is consistent.

Least Squares Solutions

Computation

We want to solve $A\hat{x} = \hat{b} = b_{\text{Col } A}$. Or, $A\hat{x} = b_W$ for $W = \text{Col } A$.

To compute b_W we need to solve $A^T A v = A^T b$; then $b_W = A v$.

Conclusion: \hat{x} is just a solution of $A^T A v = A^T b$!

Theorem

The least squares solutions of $Ax = b$ are the solutions of

$$(A^T A)\hat{x} = A^T b.$$

Note we compute \hat{x} directly, without computing \hat{b} first.

Least Squares Solutions

Example

Find the least squares solutions of $Ax = b$ where:

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

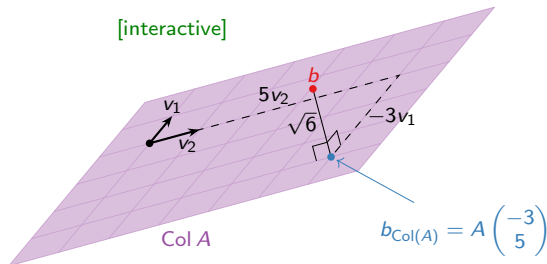
So the only least squares solution is $\hat{x} = \begin{pmatrix} -3 \\ 5 \end{pmatrix}$.

Least Squares Solutions

Example, continued

How close did we get?

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Note that

$$\begin{pmatrix} -3 \\ 5 \end{pmatrix}$$

records the coefficients of v_1 and v_2 in \hat{b} .

Least Squares Solutions

Second example

Find the least squares solutions of $Ax = b$ where:

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

So the only least squares solution is $\hat{x} = \begin{pmatrix} 1/3 \\ -1/3 \end{pmatrix}$.

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Least Squares Solutions

Uniqueness

When does $Ax = b$ have a *unique* least squares solution \hat{x} ?

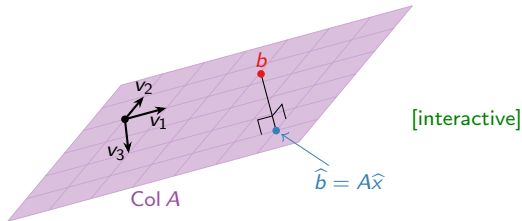
Theorem

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 \mathbf{R}^n .
2. The columns of A are linearly independent.
3. $A^T A$ is invertible.

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

Why? If the columns of A are linearly *dependent*, then $A\hat{x} = \hat{b}$ has many solutions:



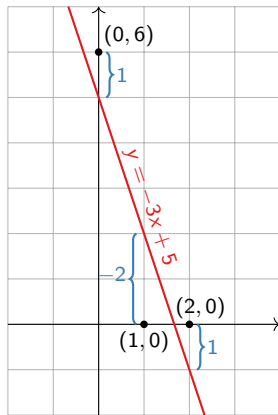
Note: $A^T A$ is always a square matrix, but it need not be invertible.

Application

Data modeling: best fit line

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

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$$A \begin{pmatrix} 5 \\ -3 \end{pmatrix} - \begin{pmatrix} 6 \\ 0 \\ 0 \end{pmatrix} = \begin{pmatrix} 1 \\ -2 \\ 1 \end{pmatrix}$$

Application

Best fit ellipse

Find the best fit ellipse for the points $(0, 2)$, $(2, 1)$, $(1, -1)$, $(-1, -2)$, $(-3, 1)$, $(-1, -1)$.

The general equation for an ellipse is

$$x^2 + Ay^2 + Bxy + Cx + Dy + E = 0$$

So we want to solve:

$$(0)^2 + A(2)^2 + B(0)(2) + C(0) + D(2) + E = 0$$

$$(2)^2 + A(1)^2 + B(2)(1) + C(2) + D(1) + E = 0$$

$$(1)^2 + A(-1)^2 + B(1)(-1) + C(1) + D(-1) + E = 0$$

$$(-1)^2 + A(-2)^2 + B(-1)(-2) + C(-1) + D(-2) + E = 0$$

$$(-3)^2 + A(1)^2 + B(-3)(1) + C(-3) + D(1) + E = 0$$

$$(-1)^2 + A(-1)^2 + B(-1)(-1) + C(-1) + D(-1) + E = 0$$

In matrix form:

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

Application

Best fit ellipse, continued

$$A = \begin{pmatrix} 4 & 0 & 0 & 2 & 1 \\ 1 & 2 & 2 & 1 & 1 \\ 1 & -1 & 1 & -1 & 1 \\ 4 & 2 & -1 & -2 & 1 \\ 1 & -3 & -3 & 1 & 1 \\ 1 & 1 & -1 & -1 & 1 \end{pmatrix} \quad b = \begin{pmatrix} 0 \\ -4 \\ -1 \\ -1 \\ -9 \\ -1 \end{pmatrix}.$$

$$A^T A = \begin{pmatrix} 36 & 7 & -5 & 0 & 12 \\ 7 & 19 & 9 & -5 & 1 \\ -5 & 9 & 16 & 1 & -2 \\ 0 & -5 & 1 & 12 & 0 \\ 12 & 1 & -2 & 0 & 6 \end{pmatrix} \quad A^T b = \begin{pmatrix} -19 \\ 17 \\ 20 \\ -9 \\ -16 \end{pmatrix}$$

Row reduce:

$$\left(\begin{array}{ccccc|c} 36 & 7 & -5 & 0 & 12 & -19 \\ 7 & 19 & 9 & -5 & 1 & 17 \\ -5 & 9 & 16 & 1 & -2 & 20 \\ 0 & -5 & 1 & 12 & 0 & -9 \\ 12 & 1 & -2 & 0 & 6 & -16 \end{array} \right) \rightsquigarrow \left(\begin{array}{ccccc|c} 1 & 0 & 0 & 0 & 0 & 405/266 \\ 0 & 1 & 0 & 0 & 0 & -89/133 \\ 0 & 0 & 1 & 0 & 0 & 201/133 \\ 0 & 0 & 0 & 1 & 0 & -123/266 \\ 0 & 0 & 0 & 0 & 1 & -687/133 \end{array} \right)$$

Best fit ellipse:

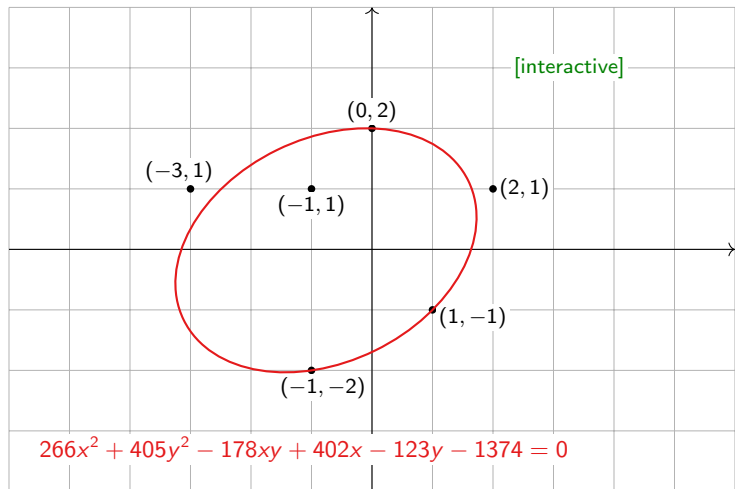
$$x^2 + \frac{405}{266}y^2 - \frac{89}{133}xy + \frac{201}{133}x - \frac{123}{266}y - \frac{687}{133} = 0$$

or

$$266x^2 + 405y^2 - 178xy + 402x - 123y - 1374 = 0.$$

Application

Best fit ellipse, picture



Remark: Gauss invented the method of least squares to do exactly this: he predicted the (elliptical) orbit of the asteroid Ceres as it passed behind the sun in 1801.

Application

Best fit parabola

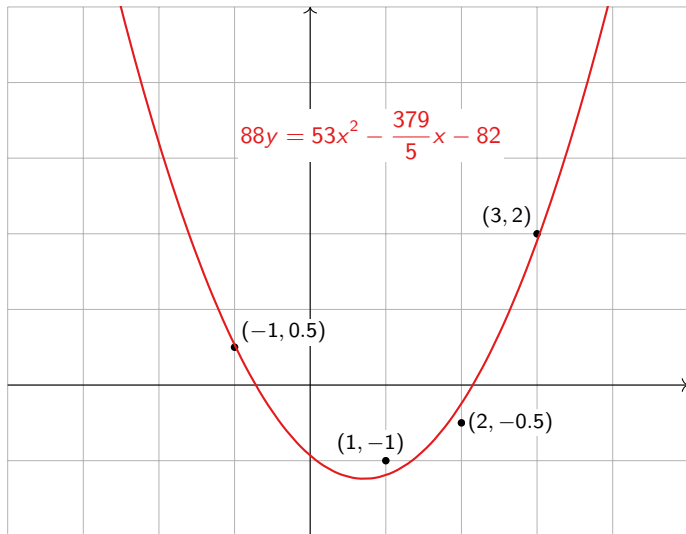
What least squares problem $Ax = b$ finds the best parabola through the points $(-1, 0.5)$, $(1, -1)$, $(2, -0.5)$, $(3, 2)$?

Answer:

$$88y = 53x^2 - \frac{379}{5}x - 82$$

Application

Best fit parabola, picture



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Application

Best fit linear function

What least squares problem $Ax = b$ finds the best linear function $f(x, y)$ fitting the following data?

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

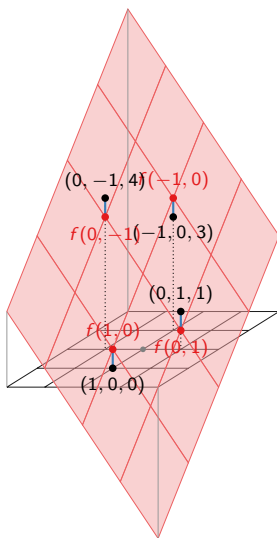
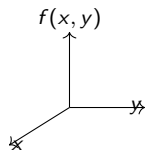
Answer:

$$f(x, y) = -\frac{3}{2}x - \frac{3}{2}y + 2$$

Application

Best fit linear function, picture

[interactive]



Graph of

$$f(x, y) = -\frac{3}{2}x - \frac{3}{2}y + 2$$

Application

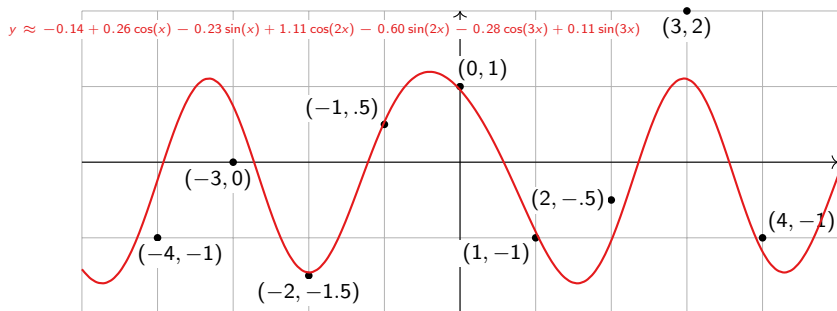
Best-fit Trigonometric Function

For fun: what is the best-fit function of the form

$$y = A + B \cos(x) + C \sin(x) + D \cos(2x) + E \sin(2x) + F \cos(3x) + G \sin(3x)$$

passing through the points

$$\begin{pmatrix} -4 \\ -1 \end{pmatrix}, \begin{pmatrix} -3 \\ 0 \end{pmatrix}, \begin{pmatrix} -2 \\ -1.5 \end{pmatrix}, \begin{pmatrix} -1 \\ .5 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \end{pmatrix}, \begin{pmatrix} 1 \\ -1 \end{pmatrix}, \begin{pmatrix} 2 \\ -.5 \end{pmatrix}, \begin{pmatrix} 3 \\ 2 \end{pmatrix}, \begin{pmatrix} 4 \\ -1 \end{pmatrix}?$$



[interactive]

Summary

- ▶ A **least squares solution** of $Ax = b$ is a vector \hat{x} such that $\hat{b} = A\hat{x}$ is as close to b as possible.
- ▶ This means that $\hat{b} = b_{\text{Col } A}$.
- ▶ One way to compute a least squares solution is by solving the system of equations

$$(A^T A)\hat{x} = A^T b.$$

Note that $A^T A$ is a (symmetric) square matrix.

- ▶ Least-squares solutions are unique when the columns of A are linearly independent.
- ▶ You can use least-squares to find best-fit lines, parabolas, ellipses, planes, etc.