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 Col A.

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

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

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

Least Squares Solutions

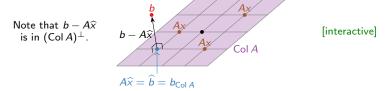
Let A be an $m \times n$ matrix.

Definition

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

$$||b - A\widehat{x}|| \le ||b - Ax||$$

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



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 \mathbb{R}^n such that

$$A\widehat{x} = \widehat{b} = b_{Col A}.$$

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 $A\hat{x} = \hat{b} = b_{Col A}$. Or, $A\hat{x} = b_W$ for W = 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

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

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

Least Squares Solutions Example

Find the least squares solutions of Ax = b where:

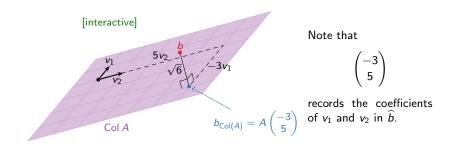
$$A = \begin{pmatrix} 0 & 1 \\ 1 & 1 \\ 2 & 1 \end{pmatrix} \qquad b = \begin{pmatrix} 6 \\ 0 \\ 0 \end{pmatrix}.$$

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

Least Squares Solutions

Example, continued

How close did we get?



Least Squares Solutions Second example

Find the least squares solutions of Ax = b where:

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

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} ?

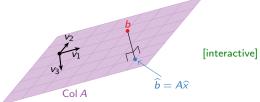
Theorem

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

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

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



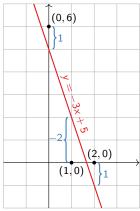
Note: $A^T A$ is always a square matrix, but it need not be invertible.

Application

Data modeling: best fit line

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

[interactive]



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

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

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

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

Row reduce:

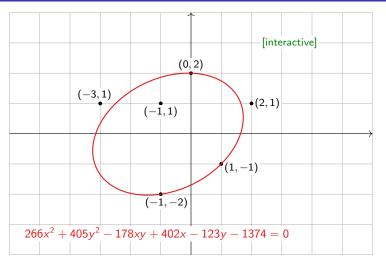
Best fit ellipse:

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

or

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

Application
Best fit ellipse, picture



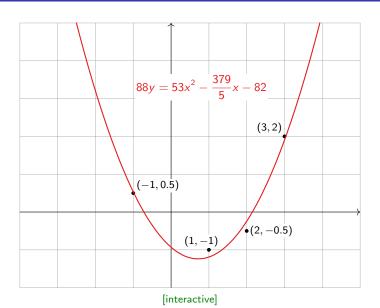
Remark: Gauss invented the method of least squares to do exactly this: he predicted the (elliptical) orbit of the asteroid Ceres as it passed behind the sun in 1801.

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



Application

Best fit linear function

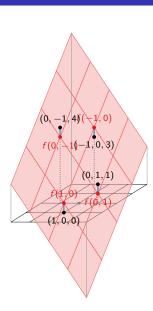
What least squares problem Ax = b finds the best linear function f(x, y) fitting the following data?

X	у	f(x,y)
1	0	0
0	1	1
-1	0	3
0	-1	4

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







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

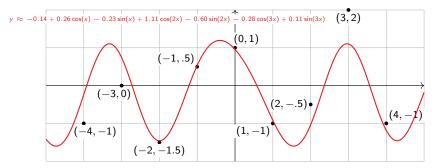
Application Best-fit Trigonometric Function

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

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

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.