

EECS 16B

Module 2, Lecture 5

March 11, 2021.

- Midterm

Last time: Controllability

Today: Machine Learning / Robotics.

- Handling model changes / unknown models.
↳ Repeated Least Squares.
- Gram-Schmidt Orthonormalization.
↳ Projections.

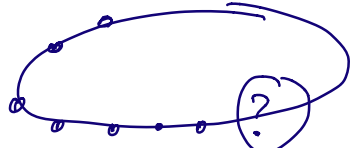
Recap: Least Squares. $A\vec{x} \approx \vec{b}$

$$\begin{bmatrix} \vec{a}_1 & \vec{a}_2 & \dots & \vec{a}_m \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = \vec{b}$$

A

Unknowns: x_1, x_2, \dots, x_n .
Weights m columns of
 $A: \vec{a}_1, \vec{a}_2, \dots, \vec{a}_m$

16A: ① Imaging Lab : Unknowns: pixels.

② Polynomial fitting: 



$$\alpha x^2 + \beta x + \gamma = 0$$

Unknowns: α, β, γ

③ System Identification:

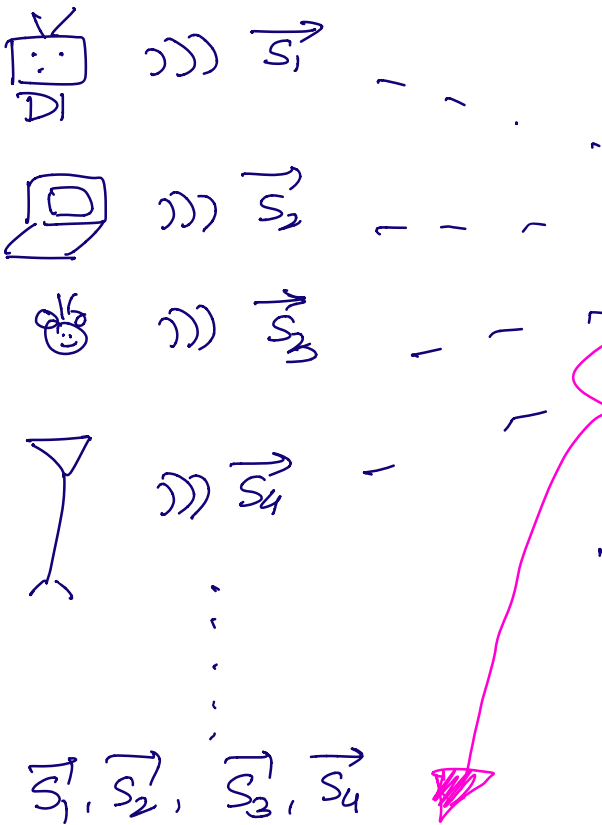
$$x(t+1) = ax(t) + bu(t)$$

Polynomial fitting:

- Choice 1: $\alpha x^2 + \beta x + \gamma = 0$
- Choice 2: $\alpha x^3 + \beta x^2 + \gamma x + \delta = 0$

Example for today:

"Internet of Things"



$$\vec{r} = \alpha_1 \vec{s}_1 + \alpha_2 \vec{s}_2 + \dots$$

I want to know:
Who is transmitting?

$$\begin{bmatrix} \vec{s}_1 & \vec{s}_2 & \vec{s}_3 & \vec{s}_4 \end{bmatrix} \begin{bmatrix} \alpha_1 \\ \alpha_2 \\ \alpha_3 \\ \alpha_4 \end{bmatrix} = \begin{bmatrix} \vec{r} \end{bmatrix}$$

What if I only find out about each of the devices one by one?

Initially: \vec{s}_1, \vec{r}

• Does \vec{s}_1 explain \vec{r} ?

$$\text{LS: } \begin{bmatrix} \vec{s}_1 \end{bmatrix} [\alpha_1] = \begin{bmatrix} \vec{r} \end{bmatrix}$$

Find α_1 using least squares.

↳ get some estimate.

~~⊗~~ Expanded set: $\{\vec{s}_1, \vec{s}_2\}$.

$$\text{LS: } \begin{bmatrix} \vec{s}_1 & \vec{s}_2 \end{bmatrix} \begin{bmatrix} \alpha_1 \\ \alpha_2 \end{bmatrix} = \begin{bmatrix} \vec{r} \end{bmatrix}$$

New set: $\{\vec{s}_1, \vec{s}_2, \vec{s}_3\}$

$$\text{LS: } \underbrace{\begin{bmatrix} \vec{s}_1 & \vec{s}_2 & \vec{s}_3 \end{bmatrix}}_A \begin{bmatrix} \alpha_1 \\ \alpha_2 \\ \alpha_3 \end{bmatrix} = \begin{bmatrix} \vec{r} \end{bmatrix}$$

Observe: A keeps getting bigger

$ATA \rightarrow$ more work.

$(ATA)^{-1} \rightarrow$ more work.

Properties of orthonormality.

Least-Squares is easy if the columns of matrix A are orthonormal.

$$\begin{bmatrix} \vec{q}_1 & \vec{q}_2 & \dots & \vec{q}_n \end{bmatrix}$$

$$\|\vec{q}_i\| = 1 = \langle \vec{q}_i, \vec{q}_i \rangle \quad \text{Normalized}$$

$$\langle \vec{q}_i, \vec{q}_j \rangle = 0 \quad \forall i, j, i \neq j$$

Orthogonal.

$$A \vec{x} = \vec{b}$$

$$A = \begin{bmatrix} \vec{q}_1 & \vec{q}_2 \end{bmatrix} \quad \text{Orthonormal.}$$

$$\hat{x} = (A^T A)^{-1} A^T \cdot \vec{b}$$

$$= \left(\begin{bmatrix} -\vec{q}_1^T \\ -\vec{q}_2^T \end{bmatrix} \begin{bmatrix} \vec{q}_1 & \vec{q}_2 \end{bmatrix} \right)^{-1} A^T \cdot \vec{b}$$

$$= \left(\begin{bmatrix} \vec{q}_1^T \vec{q}_1 & \vec{q}_1^T \vec{q}_2 \\ \vec{q}_2^T \vec{q}_1 & \vec{q}_2^T \vec{q}_2 \end{bmatrix} \right)^{-1} A^T \vec{b}$$

$$= \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} -\vec{q}_1^T \\ -\vec{q}_2^T \end{bmatrix} \vec{b}$$

$$= \begin{bmatrix} \langle \vec{q}_1, \vec{b} \rangle \\ \langle \vec{q}_2, \vec{b} \rangle \end{bmatrix}$$

$$A = \begin{bmatrix} \vec{q}_1 & \vec{q}_2 & \vec{q}_3 \end{bmatrix} \quad A \vec{x} = \vec{b}$$

Orthogonal.

$$A^T A = I = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

$$\hat{\vec{x}} = (A^T A)^{-1} A^T \vec{b}$$

$$= \begin{bmatrix} -\vec{q}_1^T \\ -\vec{q}_2^T \\ -\vec{q}_3^T \end{bmatrix} \begin{bmatrix} \vec{b} \end{bmatrix}$$

$$\langle \vec{x}, \vec{y} \rangle = \vec{x}^T \vec{y} = \vec{y}^T \vec{x}$$

$$= \begin{bmatrix} \langle \vec{q}_1, \vec{b} \rangle \\ \langle \vec{q}_2, \vec{b} \rangle \\ \langle \vec{q}_3, \vec{b} \rangle \end{bmatrix}$$



We can reuse previous work! Yay!

Goal: To find \vec{x} as a linear combination of $\vec{s}_1, \vec{s}_2, \vec{s}_3 \dots$, where we only find out the \vec{s}_i vectors one at a time.

$\vec{s}_1, \vec{s}_2, \vec{s}_3 \rightarrow$ Convert them to $\vec{q}_1, \vec{q}_2, \vec{q}_3$

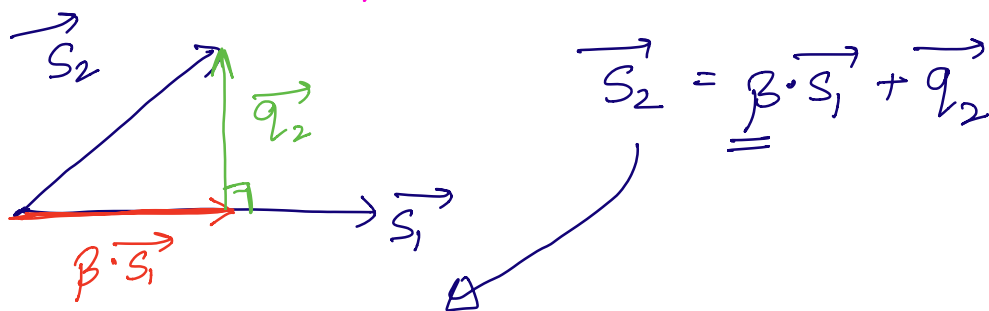
such that $\vec{q}_1, \vec{q}_2, \vec{q}_3$ etc. are all orthonormal.

Focus now: How to orthogonalize signatures
as they come in.

$\{\vec{s}_1, \vec{s}_2\}$.

\vec{s}_2 in terms of \vec{s}_1

$$\vec{s}_2 = \underbrace{\beta \cdot \vec{s}_1}_{\text{aligned to } \vec{s}_1} + \underbrace{\vec{q}_2}_{\text{brand "new"}}$$



$$\vec{q}_2 = \vec{s}_2 - \beta \vec{s}_1$$

$$\langle \vec{s}_1, \vec{q}_2 \rangle = 0$$

Review of
projections

$$\Rightarrow \langle \vec{s}_1, \vec{s}_2 - \beta \vec{s}_1 \rangle = 0$$

$$\Rightarrow \langle \vec{s}_1, \vec{s}_2 \rangle - \beta \cdot \|\vec{s}_1\|^2 = 0$$

$$\Rightarrow \beta = \frac{\langle \vec{s}_1, \vec{s}_2 \rangle}{\|\vec{s}_1\|^2}$$

Gram-Schmidt Algorithm / Procedure

Given $\{\vec{s}_1, \vec{s}_2, \dots, \vec{s}_n\}$

Convert this to a set

$$\{\vec{q}_1, \vec{q}_2, \dots, \vec{q}_n\}$$

such that: $\langle \vec{q}_i, \vec{q}_i \rangle = \|\vec{q}_i\|^2 = 1$

$$\langle \vec{q}_i, \vec{q}_j \rangle = 0$$

and:

$$\text{span}\{\vec{s}_1\} = \text{span}\{\vec{q}_1\}$$

$$\text{span}\{\vec{s}_1, \vec{s}_2\} = \text{span}\{\vec{q}_1, \vec{q}_2\}$$

\vdots

$$\text{span}\{\vec{s}_1, \vec{s}_2, \dots, \vec{s}_n\} = \text{span}\{\vec{q}_1, \vec{q}_2, \dots, \vec{q}_n\}$$

Consider: linearly independent set.

$$\{\vec{s}_1, \vec{s}_2, \dots, \vec{s}_n\} \text{ lin indep.}$$

Gram-Schmidt - Alg.

$$\{\vec{s}_1, \vec{s}_2, \vec{s}_3\}$$

① $\{\vec{s}_1\} \rightarrow \vec{q}_1$ find.

$$\frac{\vec{s}_1}{\|\vec{s}_1\|} = \vec{q}_1 \rightarrow \text{unit norm.}$$

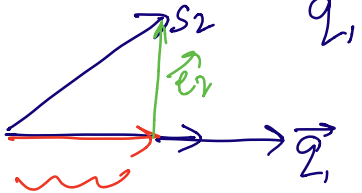
$$\text{span}\{\vec{q}_1\} = \text{span}\{\vec{s}_1\} \checkmark$$

② $\{\vec{s}_1, \vec{s}_2\}$

$$\vec{q}_1$$

What is new in \vec{s}_2 , that is not captured by \vec{q}_1 .

Remove from \vec{s}_2 , the projection of \vec{s}_2 onto \vec{q}_1



$$\vec{e}_2 = \vec{s}_2 - \frac{\langle \vec{s}_2, \vec{q}_1 \rangle}{\|\vec{q}_1\|^2} \cdot \vec{q}_1$$

$$\vec{e}_2 = \vec{s}_2 - \langle \vec{s}_2, \vec{q}_1 \rangle \cdot \vec{q}_1$$

$$\vec{q}_2 = \frac{\vec{e}_2}{\|\vec{e}_2\|} \rightarrow \text{unit norm}$$

Check: $\langle \vec{q}_2, \vec{q}_1 \rangle = \langle \frac{\vec{e}_2}{\|\vec{e}_2\|}, \vec{q}_1 \rangle$

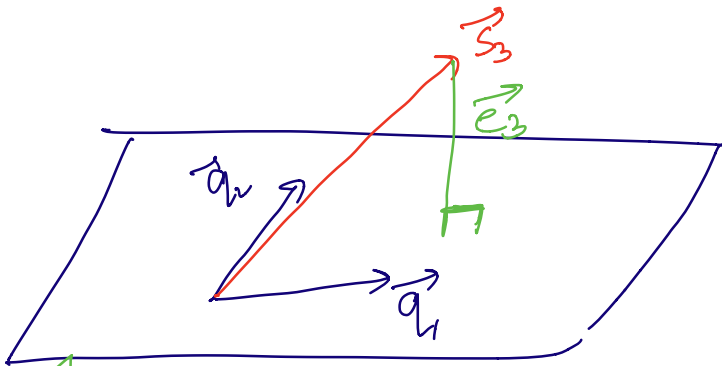
$$= \left\langle \frac{\vec{s}_2 - \langle \vec{s}_2, \vec{q}_1 \rangle \cdot \vec{q}_1}{\|\vec{e}_2\|}, \vec{q}_1 \right\rangle$$

$$= \frac{1}{\|\vec{e}_2\|} \left[\langle \vec{s}_2, \vec{q}_1 \rangle - \langle \vec{s}_2, \vec{q}_1 \rangle \underbrace{\langle \vec{q}_1, \vec{q}_1 \rangle}_1 \right]$$

$$= 0 \quad \text{Furthermore: } \text{span}\{\vec{s}_1, \vec{s}_2\} = \text{span}\{\vec{q}_1, \vec{q}_2\}$$

$$\textcircled{3} \quad \{\vec{s}_1, \vec{s}_2, \vec{s}_3\}$$

$$\boxed{\vec{q}_1, \vec{q}_2}$$



Project \vec{s}_3 onto $\text{span}\{\vec{q}_1, \vec{q}_2\}$

$$\text{proj} = \begin{bmatrix} \langle \vec{s}_3, \vec{q}_1 \rangle \\ \langle \vec{s}_3, \vec{q}_2 \rangle \end{bmatrix} \begin{bmatrix} \vec{q}_1 & \vec{q}_2 \end{bmatrix}$$

$$= \underbrace{\langle \vec{s}_3, \vec{q}_1 \rangle \vec{q}_1 + \langle \vec{s}_3, \vec{q}_2 \rangle \vec{q}_2}$$

true by projection formula
and because \vec{q}_1 and \vec{q}_2 are orthogonal.

$$\vec{e}_3 = \vec{s}_3 - \left(\langle \vec{s}_3, \vec{q}_1 \rangle \vec{q}_1 + \langle \vec{s}_3, \vec{q}_2 \rangle \vec{q}_2 \right)$$

$$\vec{q}_3 = \frac{\vec{e}_3}{\|\vec{e}_3\|} \quad \checkmark$$

Check: $\text{span} \{ \vec{q}_1, \vec{q}_2, \vec{q}_3 \} = \text{span} \{ \vec{s}_1, \vec{s}_2, \vec{s}_3 \}$

Check: $\langle \vec{q}_3, \vec{q}_1 \rangle = \langle \vec{q}_3, \vec{q}_2 \rangle = 0.$

$$\underbrace{\begin{bmatrix} \vec{s}_1 & \vec{s}_2 & \vec{s}_3 \\ | & | & | \\ 1 & 1 & 1 \end{bmatrix}}_A \longleftrightarrow \underbrace{\begin{bmatrix} \vec{q}_1 & \vec{q}_2 & \vec{q}_3 \\ | & | & | \\ 1 & 1 & 1 \end{bmatrix}}$$

Columnspace

Basis. for columnspace
of A.

