

10/29/2021

Last time:

→ finished up projections.

$$\underline{P} = \underline{A} (\underline{A}^T \underline{A})^{-1} \underline{A}^T \leftarrow \text{orthogonal projection onto } R(\underline{A}).$$

$$\underline{Q} = \underline{I} - \underline{P} \leftarrow \text{orthogonal projection onto } R(\underline{A})^\perp = N(\underline{A}^T).$$

Ex: Compute the orthogonal projection onto the subspace

$$V = \{ \underline{x} \in \mathbb{R}^3 : x_1 + x_2 + x_3 = 0 \}.$$

Note, if we define $\underline{B} = (1 \ 1 \ 1)$,

then $V = N(\underline{B})$.

To compute a basis for $N(\underline{B})$, look at:

$$\underline{B} \underline{x} = 0 \Leftrightarrow (1 \ 1 \ 1) \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = 0.$$

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$$\text{Then } \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} \in N(\underline{B}) \Leftrightarrow \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} -x_2 - x_3 \\ x_2 \\ x_3 \end{pmatrix}$$

$$= x_2 \begin{pmatrix} -1 \\ 1 \\ 0 \end{pmatrix} + x_3 \begin{pmatrix} -1 \\ 0 \\ 1 \end{pmatrix}.$$

So $\left\{ \begin{pmatrix} -1 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} -1 \\ 0 \\ 1 \end{pmatrix} \right\}$ is a basis

for $N(\underline{B})$. To compute the orthogonal projection onto this subspace, we can put the basis vectors in the columns of a matrix:

$$\underline{A} = \begin{pmatrix} -1 & -1 \\ 1 & 0 \\ 0 & 1 \end{pmatrix},$$

and compute $\underline{P} = \underline{A}(\underline{A}^T \underline{A})^{-1} \underline{A}^T$

$$= \frac{1}{3} \begin{pmatrix} 2 & -1 & -1 \\ -1 & 2 & -1 \\ -1 & -1 & 2 \end{pmatrix}.$$

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Alternative approach:

Note $N(\underline{B})^\perp = R(\underline{B}^T)$ by FTLA, so, first compute orthogonal proj onto $R(\underline{B}^T)$, call it $\underline{Q}$. Then $\underline{P} = \underline{I} - \underline{Q}$

will be the orthogonal proj. onto

$$R(\underline{B}^T)^\perp = (N(\underline{B})^\perp)^\perp = N(\underline{B}).$$

$$\leadsto \underline{Q} = \underline{B}^T (\underline{B} \underline{B}^T)^{-1} \underline{B} = \frac{1}{3} \begin{pmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{pmatrix}.$$

$$\leadsto \underline{P} = \underline{I} - \underline{Q} = \frac{1}{3} \begin{pmatrix} 2 & -1 & -1 \\ -1 & 2 & -1 \\ -1 & -1 & 2 \end{pmatrix}.$$

Application of Projections.

Gram-Schmidt orthogonalization.

main idea: we are given a set of vectors (assume linear independence)

$$\{\underline{a}_1, \underline{a}_2, \dots, \underline{a}_k\}.$$

$\leadsto$

we want to construct another set of vectors:

$$\{q_1, q_2, \dots, q_k\}$$

so that

$$\text{span}(a_1, \dots, a_k) = \text{span}(q_1, \dots, q_k)$$

$$\text{and } q_i^T q_j = \begin{cases} 1 & i=j \\ 0 & i \neq j \end{cases}$$

The latter condition is to say that the set $\{q_1, \dots, q_k\}$ is orthonormal, i.e. mutually orthogonal and 2-norm equal to one.

Why do we care? Suppose we want to compute the orthogonal projection of b onto

$$\text{span}(a_1, \dots, a_k)$$

We know we can build the matrix

$$\underline{A} = \begin{pmatrix} | & | & & | \\ a_1 & a_2 & \dots & a_k \\ | & | & & | \end{pmatrix} \quad \leadsto$$

and solve $\underline{A}^T \underline{A} \underline{x}^* = \underline{A}^T \underline{b}$.

We can equivalently do this with

$$\underline{Q} = \begin{pmatrix} | & & | \\ \underline{q}_1 & \dots & \underline{q}_k \\ | & & | \end{pmatrix},$$

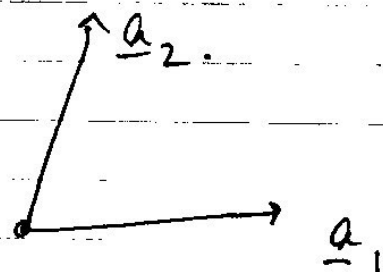
but notice $\underline{Q}^T \underline{Q} \underline{x}^* = \underline{Q}^T \underline{b} = \underline{x}^*$,

since $\underline{Q}^T \underline{Q} = \underline{I}$. Much cleaner!

Question: How do we build

$\{\underline{q}_1, \dots, \underline{q}_k\}$ from $\{\underline{a}_1, \dots, \underline{a}_k\}$?

To start. let $k=2$.



First step. set $\underline{q}_1 = \frac{\underline{a}_1}{\|\underline{a}_1\|_2}$.

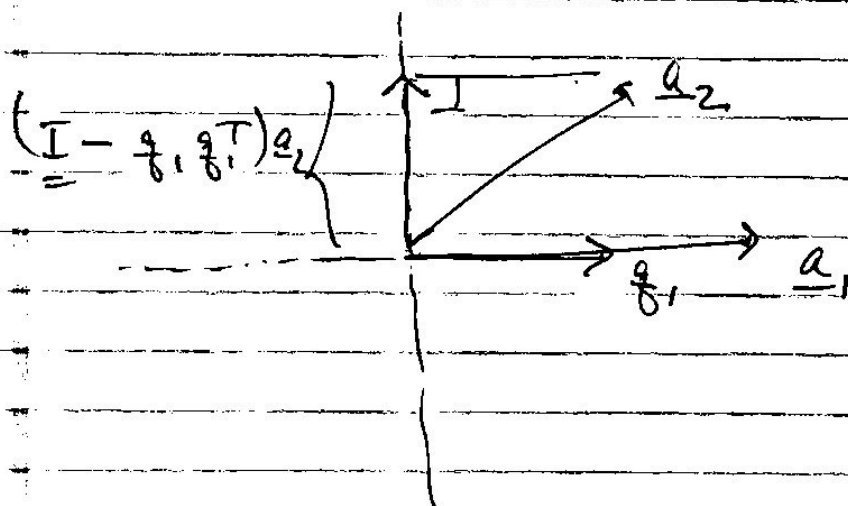
Second step: write $\underline{a}_2 = \underline{q}_1 \underline{q}_1^T \underline{a}_2 + (\underline{I} - \underline{q}_1 \underline{q}_1^T) \underline{a}_2$.



$$\underline{a}_2 = \underbrace{q_1 q_1^T \underline{a}_2}_{\in \text{span}(q_1)} + \underbrace{(\underline{I} - q_1 q_1^T) \underline{a}_2}_{\in \text{span}(q_1)^\perp}$$

So, we throw away $q_1 q_1^T \underline{a}_2$ and define:

$$\underline{q}_2 = \frac{(\underline{I} - q_1 q_1^T) \underline{a}_2}{\|(\underline{I} - q_1 q_1^T) \underline{a}_2\|_2}$$



Remark: By construction,

$$\text{span}(\underline{a}_1, \underline{a}_2) = \text{span}(q_1, \underline{a}_2)$$

$$= \text{span}(q_1, q_1 q_1^T \underline{a}_2, (\underline{I} - q_1 q_1^T) \underline{a}_2) \rightsquigarrow \Delta$$

$$\begin{aligned}
 &= \text{span} \left(\underline{q}_1, (\underline{I} - \underline{q}_1 \underline{q}_1^T) \underline{a}_2 \right) \\
 &= \text{span} \left(\underline{q}_1, \underline{q}_2 \right).
 \end{aligned}$$

Ex: Create an orthonormal basis

for the subspace $\text{span} \left(\begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} \right)$.

$$\text{let } \underline{a}_1 = \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}, \quad \underline{a}_2 = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}.$$

$$\text{step 1: set } \underline{q}_1 = \frac{\underline{a}_1}{\|\underline{a}_1\|} = \begin{pmatrix} \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \\ 0 \end{pmatrix}.$$

$$\text{step 2: calculate } \underline{b}_2 = (\underline{I} - \underline{q}_1 \underline{q}_1^T) \underline{a}_2.$$

$$\begin{aligned}
 &\rightarrow \left(\begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} - \begin{pmatrix} \frac{1}{2} & \frac{1}{2} & 0 \\ \frac{1}{2} & \frac{1}{2} & 0 \\ 0 & 0 & 0 \end{pmatrix} \right) \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} \\
 &= \begin{pmatrix} \frac{1}{2} & -\frac{1}{2} & 0 \\ -\frac{1}{2} & \frac{1}{2} & 0 \\ 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} = \begin{pmatrix} \frac{1}{2} \\ -\frac{1}{2} \\ 0 \end{pmatrix}.
 \end{aligned}$$

$$\text{set } \underline{q}_2 = \frac{\underline{b}_2}{\|\underline{b}_2\|} = \begin{pmatrix} \frac{1}{\sqrt{2}} \\ -\frac{1}{\sqrt{2}} \\ 0 \end{pmatrix}.$$