

# LECTURE 10

Q.

$$A = \begin{pmatrix} 1 & 1.05 \\ 1 & 1 \\ 1 & 0.95 \end{pmatrix} \quad B = \begin{pmatrix} 1 & 1/\sqrt{2} \\ 1 & 0 \\ 1 & -1/\sqrt{2} \end{pmatrix}$$

Show that the colspace of  $A = \text{colspace of } B$ .

(Chapter 4 — not on exam but some are review topics/tied to previous topics, useful as review)

Key concept We can solve linear systems of equations

- Gauss Elim for exact soln
- Normal Equations for best fit

But in the context of signal + noise, we care about the "data quality" — in our context, how robustly a set of vectors determines the span.

e.g. Nearly linearly independent columns:

$$A = \begin{pmatrix} 1 & 1.05 \\ 1 & 1 \\ 1 & 0.95 \end{pmatrix}$$

Nearly orthogonal columns:

$$B = \begin{pmatrix} 1 & 1/\sqrt{2} \\ 1 & 0 \\ 1 & -1/\sqrt{2} \end{pmatrix}$$

Same columnspace. What happens when I add some noise / small perturbation?

Motto "Good" basis vectors are "nearly orthogonal".

## § 4.1 Orthogonal Vectors & Matrices

- $x, y \in \mathbb{R}^n$  orthogonal if  $x^T y = 0$  i.e.  $\cos \theta(x, y) = 0$ ,  $y^T x = 0$ .

prop. Let  $q_j$  ( $j \in [n]$ ) be orthogonal (i.e. mutually orthogonal). Then they are linearly independent.

Q. Converse?

Pf. Easy -  $q_n = \sum_{i=1}^n \alpha_i q_i \Rightarrow q_n^T q_n = 0$ .  $\square$

- Set of orth. vefs  $\{q_j\}$  are normalized if they're  $\|q_j\|_2 = 1 \forall j$ .  
 $\Rightarrow$  they are orthonormal, and form an orthonormal basis for their span.
- A square matrix  $Q = (q_1, \dots, q_m) \in \mathbb{R}^{m \times m}$  with orthonormal cols is an orthogonal matrix

Q. Why do you think I chose square matrix?

Facts (Why are these true?)

- $Q^T Q = I$  Why?
- $Q^{-1} = Q^T$  Why?
- If  $Q, P$  orth  $\Rightarrow PQ$  also orth. Why?
- If  $Q_1 \in \mathbb{R}^{m \times k}$  ( $k < m$ ) with orthonormal columns, then  $\exists Q_2 \in \mathbb{R}^{m \times (m-k)}$  s.t.  $Q = [Q_1 \ Q_2]$  is orthogonal matrix.

Q. How would you prove this (these)?

\* What are the eigenvalues of an orthogonal matrix?