

Lecture 4

Announcements

- HW2 + HW2 lab posted

Readings

- Strang I.5 and I.6

Today

- Vector norms
- Orthogonality (vectors, bases, subspaces, matrices, examples + properties)
- Eigenvalues and eigenvectors

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$$X = \begin{bmatrix} | & | & | & | \\ \vec{x}_1 & \vec{x}_2 & & \vec{x}_p \\ | & | & & | \end{bmatrix}$$

n observation \times p variables

What if we have rows that are linearly dependent?
ex one row is equal to another row
" " " "

Vector Norms

How do we quantify the "size" or "length" of a vector?

- a vector norm maps vectors in \mathbb{R}^n to a single number in \mathbb{R}

ex. $\|\vec{x}\|_1 = \sum_{i=1}^n |x_i|$ (1-norm)

$\|\vec{x}\|_2 = \sqrt{\sum_{i=1}^n x_i^2}$ (2-norm)

$\|\vec{x}\|_p = \left(\sum_{i=1}^n |x_i|^p \right)^{1/p}$ (p-norm)
 $p > 0$

$\|\vec{x}\|_{\infty} = \max_{1 \leq i \leq n} |x_i|$ (max-norm)

$$\begin{bmatrix} 1 \\ 1 \\ 10 \end{bmatrix} \quad \begin{bmatrix} 5 \\ 5 \\ 5 \end{bmatrix}$$

\vec{x} \vec{y}

$\|\vec{x}\|_1 = 12$

$\|\vec{y}\|_1 = 15$

$\|\vec{x}\|_2 = \sqrt{102}$

$\|\vec{y}\|_2 = \sqrt{75}$

$\|\vec{x}\|_{\infty} = 10$

$\|\vec{y}\|_{\infty} = 5$

note $\|\vec{x}\|_2^2 = \vec{x}^T \vec{x} = \vec{x} \cdot \vec{x} = \langle \vec{x}, \vec{x} \rangle$

Orthogonality

def Two vectors \vec{x} and \vec{y} in \mathbb{R}^n are orthogonal if $\vec{x}^T \vec{y} = \vec{y}^T \vec{x} = 0$

$$\vec{x}^T \vec{y} = x_1 y_1 + \dots + x_n y_n = 0$$

In \mathbb{R}^n this is equivalent to them being perpendicular.

note $\vec{x}^T \vec{y} = \|\vec{x}\| \|\vec{y}\| \cos \theta$ where θ is the angle between \vec{x} and \vec{y} .

What are three orthogonal vectors in \mathbb{R}^3 ?

$$\left\{ \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \right\}$$

\vec{e}_3 \vec{e}_2 \vec{e}_1

note These vectors form a basis for \mathbb{R}^3

theorem Let $\vec{q}_1, \dots, \vec{q}_n$ be orthogonal vectors (so $\vec{q}_i^T \vec{q}_j = 0$ for $i \neq j$)

Then they are linearly independent.

proof outline Assume linear dependence and prove by contradiction.

theorem Every subspace of \mathbb{R}^n has an orthogonal basis.

ex A plane in \mathbb{R}^3 spanned by independent vectors \vec{x} and \vec{y} .

- Set the first vector of our basis $\vec{q}_1 = \vec{x}$.

- define $\vec{q}_2 = \vec{y} - \frac{\vec{y}^T \vec{x}}{\vec{x}^T \vec{x}} \vec{x}$ (residual from projecting \vec{y} on \vec{x})

In general, this idea leads to Gram-Schmidt orthogonalization

Orthogonal Subspaces

ex The row space of A is orthogonal to the nullspace of A .

The column space of A is orthogonal to the nullspace of A^T

$$\begin{bmatrix} -\vec{r}_1- \\ \vdots \\ -\vec{r}_m- \end{bmatrix} \begin{bmatrix} 1 \\ \vec{x} \\ 1 \end{bmatrix} = \begin{bmatrix} \vec{0} \\ \vdots \\ \vec{0} \end{bmatrix}$$

$A \quad \vec{x} \quad = \vec{0}$

Orthonormal vectors

Let $\vec{q}_1, \dots, \vec{q}_n$ be orthogonal with all having length 1 so $\|\vec{q}_j\|_2 = 1$ for all j .

Then they are called orthonormal.

In particular if $\vec{q}_1, \dots, \vec{q}_n$ are n orthonormal vectors in \mathbb{R}^n , they form an orthonormal basis for \mathbb{R}^n .

Orthogonal Matrices

A square matrix $Q \in \mathbb{R}^{n \times n}$ with orthonormal columns is called an orthogonal matrix.

properties

• $Q^T Q = I$

• $Q Q^T = I$

• $Q^{-1} = Q^T$

$$\begin{bmatrix} -\vec{q}_1^T- \\ \vdots \\ -\vec{q}_n^T- \end{bmatrix} \begin{bmatrix} 1 & & \\ \vec{q}_1 & \dots & \vec{q}_n \\ 1 & & 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 & \dots & 0 \\ 0 & 1 & & \\ \vdots & & \ddots & 0 \\ 0 & & & 1 \end{bmatrix}$$

$Q^T \quad Q \quad = \quad I$

property

$$\|Q\vec{x}\|_2 = \|\vec{x}\|_2$$

$$\|Q\vec{x}\|_2^2 = (Q\vec{x}) \cdot (Q\vec{x}) = (Q\vec{x})^T (Q\vec{x}) = \vec{x}^T \overbrace{Q^T Q}^I \vec{x} = \vec{x}^T \vec{x} = \|\vec{x}\|_2^2$$

ex

$$Q_3 = \frac{1}{3} \begin{bmatrix} 2 & 2 & -1 \\ 2 & -1 & 2 \\ -1 & 2 & 2 \end{bmatrix}$$

Confirm this is orthogonal

$$Q_3^T Q_3 = I_3$$

$$Q_3 Q_3^T = I_3$$

$$Q_1 = \frac{1}{3} \begin{bmatrix} 2 \\ 2 \\ -1 \end{bmatrix}$$

Is this orthogonal? Not really

$$Q_1^T Q_1 = I_1$$

$$Q_1 Q_1^T = ?$$

It turns out that $Q_1 Q_1^T$ is the orthogonal projection matrix onto the column space of Q_1 .

$$\begin{matrix} \begin{bmatrix} 2/3 \\ 2/3 \\ -1/3 \end{bmatrix} & \begin{bmatrix} 2/3 & 2/3 & -1/3 \end{bmatrix} & \begin{bmatrix} x \\ y \\ z \end{bmatrix} \\ Q_1 & Q_1^T & \vec{b} \end{matrix}$$

compare with $\frac{\vec{q}_1^T \vec{b}}{\vec{q}_1^T \vec{q}_1} \vec{q}_1$

In general, given an orthonormal basis $\vec{q}_1, \dots, \vec{q}_n$ for \mathbb{R}^n

any vector $\vec{x} \in \mathbb{R}^n$ can be written

$$\vec{x} = c_1 \vec{q}_1 + \dots + c_n \vec{q}_n$$

$$c_1 = \vec{q}_1^T \vec{x} \quad c_2 = \vec{q}_2^T \vec{x} \quad \dots \quad c_n = \vec{q}_n^T \vec{x}$$

Proof exercise

Eigenvalues and Eigenvectors

def A vector \vec{x} is an eigenvector of A if $A\vec{x} = \lambda\vec{x}$ for some number λ .
 λ is called an eigenvalue of \vec{x} and of A .

note if \vec{x} is an eigenvector of A , it is also an eigenvector of A^2

$$A^2 \vec{x} = A A \vec{x} = A(\lambda \vec{x}) = \lambda(A\vec{x}) = \lambda^2 \vec{x}$$

ex $S = \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix}$ has eigenvectors $\begin{bmatrix} 1 \\ 1 \end{bmatrix}$ and $\begin{bmatrix} 1 \\ -1 \end{bmatrix}$ $\lambda_1 = 3$
 $S \begin{bmatrix} 1 \\ 1 \end{bmatrix} = 3 \begin{bmatrix} 1 \\ 1 \end{bmatrix}$

What are the eigenvalues?

$\lambda = 1$
 $S \begin{bmatrix} 1 \\ -1 \end{bmatrix} = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$

$i = \sqrt{-1}$

ex $Q = \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}$ has imaginary eigenvalues i and $-i$
 $Q \begin{bmatrix} 1 \\ -i \end{bmatrix} = i \begin{bmatrix} 1 \\ -i \end{bmatrix}$ $Q \begin{bmatrix} 1 \\ i \end{bmatrix} = -i \begin{bmatrix} 1 \\ i \end{bmatrix}$

Computing eigenvalues

Note that $A\vec{x} = \lambda\vec{x}$ so $A\vec{x} - \lambda\vec{x} = \vec{0} \Rightarrow (A - \lambda I)\vec{x} = \vec{0}$

Assuming \vec{x} is non-zero $(A - \lambda I)$ is not invertible and must be singular.

$\det(A - \lambda I) = 0$

This is an n^{th} -degree polynomial in λ .

$\rightarrow \det(A - \lambda I) = 0$ has n solutions.

ex $A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$ $\det(A) = ad - bc$

$A - \lambda I = \begin{bmatrix} a-\lambda & b \\ c & d-\lambda \end{bmatrix}$ $\det(A - \lambda I) = (a-\lambda)(d-\lambda) - bc = 0$
 $\lambda^2 - (a+d)\lambda + (ad - bc) = 0$

Solve this eqn to get two roots λ_1 and λ_2 .

ex $A = \begin{bmatrix} 8 & 3 \\ 2 & 7 \end{bmatrix}$

$A - \lambda I = \begin{bmatrix} 8-\lambda & 3 \\ 2 & 7-\lambda \end{bmatrix}$

$\det(A - \lambda I) = (8-\lambda)(7-\lambda) - 6$
 $= \lambda^2 - 15\lambda + 50$

$\boxed{\begin{matrix} \lambda_1 = 10 \\ \lambda_2 = 5 \end{matrix}}$

$A\vec{x} = 10\vec{x}$

$\begin{bmatrix} 8 & 3 \\ 2 & 7 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 10x_1 \\ 10x_2 \end{bmatrix}$