APMA 4007 | Sep 5, 2024

Lecture 2

Outline

- 1. Review
 - -Ax=b
 - dot product
- 2. Dot product apprication
- 3. Dot product is special case of matrix mult etc
- 4. Unique ness

1) A system of m equations in M unknowns can be written as a vector equation where the lhs is a L.C. of m vectors : the rhs is a a constant vector w/ m elements.

Example

$$2x_1 - 3x_2 - X_3 = 1$$

Take 1 minute & write in the form stated above, LC = const. What are min?

 $\chi_{1}\begin{bmatrix} 3 \\ -1 \\ 0 \end{bmatrix} + \chi_{2}\begin{bmatrix} -3 \\ 0 \\ 1 \end{bmatrix} + \chi_{3}\begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix} = \begin{bmatrix} 0 \\ 4 \end{bmatrix}$

In general $x_1 a_1 + \dots + x_n a_n = b$ m

OR $A \times = b$ where $x = \begin{bmatrix} x_1 \\ 1 \\ x_n \end{bmatrix}$

$$\Delta = \begin{bmatrix} 1 & 1 & 1 \\ a_1 & \cdots & a_n \\ 1 & 1 \end{bmatrix}$$

$$b = \begin{bmatrix} b_1 \\ b_m \end{bmatrix}$$

(2.) Ax is matrix-vector multiplication and Δx is/means L.C. of the columns of A.

Ax = b states that (for some x), b is a L.C. of the columns of A

The statement that Ax=b has no solly means that b is not a L.C. of the columns of A

Example $A = \begin{bmatrix} 1 & 3 \\ 2 & 4 \end{bmatrix}$ $A = \begin{bmatrix} 1 & 3 \\ 0 & 0 \end{bmatrix}$ $A = \begin{bmatrix} 1 & 3 \\ 0 & 0 \end{bmatrix}$ $A = \begin{bmatrix} 1 & 3 \\ 2 & 4 \end{bmatrix}$ $A = \begin{bmatrix} 1 & 3 \\ 0 & 0 \end{bmatrix}$ $A = \begin{bmatrix} 1 & 3 \\ 0 & 0 \end{bmatrix}$

Matrix-vector mult. has some properties of ordinary mult. but not all.

A(x+y) = Ax + Ay b/c $A(x+y) = B(x+y,)a_1 + ... + (x+y_m)a_n$ $Using properties = B(x_1a_1 + ... + x_na_n + y, a_1 + ... + y_ma_n)$ of scaling = Ax + Ay

AX \ XA g \ Ab \ does not compute!

NOTE Length of
$$\chi = ||x||$$
 len $(x) = m$

For
$$\vec{x} \neq \vec{0}$$
, $\frac{\vec{x}}{\|\vec{x}\|}$ is a unit vector b/c it has length = 1

length
$$\frac{\vec{\chi}}{|\vec{\chi}|} = \frac{\vec{\chi}}{|\vec{\chi}|} \cdot \frac{\vec{\chi}}{|\vec{\chi}|} = \frac{\vec{\chi} \cdot \vec{\chi}}{|\vec{\chi}|^2} = \frac{|\vec{\chi} \cdot \vec{\chi}|^2}{|\vec{\chi}|^2} = 1$$

not used so often in this class

notation is compact & concise compared

Example: Statistics (will also provide a geometrical picture)

$$\overrightarrow{X}_{1}$$
 $\overrightarrow{X}_{2} = \begin{bmatrix} X_{1} \\ Y_{N} \end{bmatrix}$ a vector containing n samples (measurements)

Mean
$$M = \frac{1}{n} \sum_{i=1}^{n} \chi_i = \frac{1}{n} \stackrel{?}{1} \circ \stackrel{?}{\chi}$$
 where $\stackrel{?}{1} = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$

Variance
$$T_{\chi^2} = \frac{1}{n-1} \sum_{i=1}^{n} (\chi_i - \mu_{\chi})^2 = \frac{1}{n-1} || \vec{\chi} - \mu_{\vec{\chi}} \vec{1} ||^2$$

If
$$M_{x}=0$$
, variance of $x=\frac{||x||^{2}}{|x-1|}$ proportional to length squared.

Covariance
$$T_{xy} = \frac{1}{m-1} \sum_{N=1}^{m} (x_i - \mu_x) (y_i - \mu_y)$$

$$= \frac{1}{m-1} (\vec{x} - \mu_x \vec{1}) (\vec{y} - \mu_y \vec{1})$$

if
$$Mx = My = 0$$
, $Txy = \frac{\vec{x} \cdot \vec{y}}{m-1} \in Txy = 0$ iff $\vec{x} \perp \vec{y}$ correlation

$$\int xy = \frac{1}{\sqrt{x}} \int xy = \frac{1}{\sqrt{x}} \int xy = \frac{1}{\sqrt{x}} \int y = \frac{1}{\sqrt{x}} \int$$

(5)

Let us not pretend that matrix mult if the dot product are two different things.

$$\vec{\chi} \cdot \vec{y} = \sum_{i=1}^{m} \chi_{i} y_{i}$$

$$= \left[\chi_{1} - \chi_{n} \right] \begin{bmatrix} y_{1} \\ y_{n} \end{bmatrix} \text{ or } \left[y_{1} - y_{n} \right] \begin{bmatrix} \chi_{1} \\ \chi_{n} \end{bmatrix}$$
a matrix w/ one row $\begin{bmatrix} y_{n} \\ y_{n} \end{bmatrix}$
and n columns

The notation for [x1 — Xn] is x where "I" means transpose and that means was switch rows & columns,

- if x is $n \times 1$, x^T is $1 \times n$ $- (x^T)^T = x$

We will write xty (or yix) and think $\dot{x} \cdot \dot{y}$.

 $||x||^2 = x^T x$ $x^T y = 0. \quad ||H||^2 x \perp y$ etc

Let us not pretend that matrix mult. c, the dot product are two different things (II).

Example

$$\begin{bmatrix} 1 & 0 & 2 \\ 0 & 3 & 4 \end{bmatrix} \begin{bmatrix} 5 \\ -6 \end{bmatrix} = 5 \begin{bmatrix} 1 \\ 0 \end{bmatrix} 0 - 6 \begin{bmatrix} 0 \\ 3 \end{bmatrix} + \begin{bmatrix} 2 \\ 4 \end{bmatrix}$$

$$= \begin{bmatrix} 5 + 0 + 2 \\ 0 - 18 + 4 \end{bmatrix} = \begin{bmatrix} 7 \\ -14 \end{bmatrix}$$

$$= \begin{bmatrix} 1 \\ 0 \\ -14 \end{bmatrix}$$

look like

$$= \begin{bmatrix} 5+0+2 \\ 0-18+4 \end{bmatrix} = \begin{bmatrix} + \\ -14 \end{bmatrix}$$

$$\begin{bmatrix} 5.0 - 6.0 + 1.2 \\ - 5.0 - 6.3 + 1.4 \end{bmatrix}$$

$$\begin{bmatrix} 5 \cdot 1 - 6 \cdot 0 + 1 \cdot 2 \\ = \begin{bmatrix} 5 \cdot 0 - 6 \cdot 3 + 1 \cdot 4 \end{bmatrix}$$

$$= \begin{bmatrix} 0 \\ 2 \end{bmatrix} \cdot \begin{bmatrix} 5 \\ 6 \\ 1 \end{bmatrix}$$

$$= \begin{bmatrix} 0 \\ 3 \\ 4 \end{bmatrix} \cdot \begin{bmatrix} 5 \\ 6 \\ 1 \end{bmatrix}$$

$$= \begin{bmatrix} \begin{bmatrix} 1 & 0 & 2 \end{bmatrix} \begin{bmatrix} 5 \\ 6 \\ 1 \end{bmatrix} \\ \begin{bmatrix} 0 & 3 & 4 \end{bmatrix} \begin{bmatrix} 5 \\ 6 \\ 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 2 \\ 0 & 3 & 4 \end{bmatrix} \begin{bmatrix} 5 \\ 6 \\ 1 \end{bmatrix}$$

By the same logic"

i-th element of Ax

[AX]i is the i-th row of A dothod with x,

 $[Ax]_{i} = \sum_{j=1}^{n} A_{ij} x_{j} = A_{ij} \circ x_{j}$ i-th elementof A

row i

column j

wx vow x column

Using the same formula $\begin{bmatrix}
A \times \end{bmatrix}_{i}^{i} = \begin{bmatrix}
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\end{bmatrix} = \begin{bmatrix}
X_{i}^{$

2 views of Ax:

1. L.C. of the columns of A

a. Dotting rows of A in x.

Both are useful

Practice/computation

- if \$ x has many zeros #1

1s good A[0] = ? by column of A

of A, #2 is nice.

Theory

- Ax=b has solly of by

L.C. of columns of A comes

from #1. Existence

- #2 is handy for statements about uniqueness

Suppose that there are 2 solutions is in (and \$\frac{1}{2} \dig \frac{1}{y}) so that Ax=b & Ay=b.

By math

$$A(x-y) = Ax - Ay = b - b = \vec{0}$$

$$\vec{2} \neq \vec{0}$$

III the soll is not unique, there is a vector $2 \neq 0$ such that Az=0.

[this is unlike scalar mult, where are = 0 means that either a=0 or z=0

$$\begin{bmatrix} 1 & -2 & 1 \\ 2 & -2 & 0 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \leftarrow \text{ each }$$

$$\begin{bmatrix} 1 & -2 & 1 \\ 2 & -2 & 0 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \begin{bmatrix} 2 \\ 3 \end{bmatrix}$$

$$\text{not unique } 1.$$

$$\text{not zevo}$$

$$\text{1 } -2 \quad 1 \end{bmatrix} \begin{bmatrix} 0 \\ -1 \end{bmatrix} = \begin{bmatrix} 2 \\ 3 \end{bmatrix}$$

$$\begin{bmatrix} 1 & -2 & 1 \\ 2 & -2 & 0 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 2 \\ 3 \end{bmatrix}$$

$$= \begin{bmatrix} 3 \\ 3 \end{bmatrix}$$

such that

The logic if soll not unique, then Iz > Az = 0, goes both ways,

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Suppose $A \not\equiv 0$ and $\not\equiv \not\equiv \vec{0}$. Then a soly of A x = b is not unique. Why?

A(x+z) = Az + Ax = b $\pm wo$ soll's $x \in x+z$ and they differ b/c $\pm z+\overline{o}$,

Actually more than 2

 $A(\vec{\chi} + c\vec{z}) = Ax + cAz = b$ Seglar

for any scalar C, so an ∞ # of sol_s .

As we saw in the 2D example the possibilities are $0,1,\infty$ # of sol_s .

If a soll exists, it is either unique or there are an infinite # of soll's

A statement about uniqueness.

A solh of Ax=b is unique iff the only solution of Az=0 is z=0.

Note the uniqueness statement does not involve b.

(A does not express a view on existence)

Recall the 2nd view of AX — dotting vows of A

AZ=3 means that Z is I to all the rows of A.

Uniqueness of soll of Ax=b depends on whether there is a non-zero vector Z I to all the rows of A.

$$A = \begin{bmatrix} 2 & 0 & 6 & 1 \\ -8 & 4 & 2 & 6 \\ 3 & -3 & 9 & 1 \\ 4 & 6 & 4 & 0 \end{bmatrix}$$

$$A = \begin{bmatrix} 0 \\ 4 \\ -3 \\ 6 \end{bmatrix}$$
what is the solf-

Since b is the and column of A $\chi = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$. Next time a mean general method.