Linear Algebra

Lecture 14

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Today

- Matrices, contd.
- Examples and exercises

Matrix vector multiplication A mxn matrix x n-vector def: Matrix-vector product $y = A \hat{x}$ $\mathcal{Y}_{i} = \sum_{k=1}^{n} A_{ik} X_{k} = A_{ii} X_{1} + \dots + A_{in} X_{n}, n = 1, \dots, m$ = (b_{i}, X) $e_{g}: \begin{bmatrix} 0 & 2 & -1 \\ -2 & 1 & 1 \end{bmatrix} \begin{bmatrix} 2 \\ 1 \\ -1 \end{bmatrix} := \begin{bmatrix} 0 \cdot 2 + 2 \cdot 1 + f_{1} \cdot f_{1} \\ +21 \cdot 2 + (1 \cdot 1(1) + (\Lambda(-1)) \end{bmatrix} = \begin{bmatrix} 3 \\ -4 \end{bmatrix}$

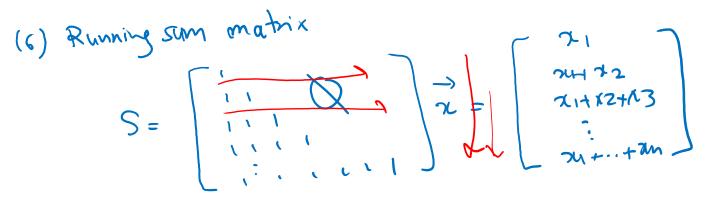
Matrix vector multiplication

$$y_{i} = \langle b_{i}, x \rangle$$
, b_{i} is low $i \circ f A$
 (b_{i}, x)
 $k-th$ column vector of A a_{k} :
 $y_{i}^{2} - A \widehat{x} = a_{i} \widehat{a}_{i} + s_{2} \widehat{a}_{3} + \dots + s_{m} \widehat{a}_{m}$
verify:
 $y_{i}^{2} = x_{i} a_{i}(\widehat{x}) + \dots + x_{m} a_{m} (\widehat{x})$
 $= x_{i} A_{i1} + \dots + x_{m} A_{im}$

Matrix vector multiplication
examples A max matrix,
$$\vec{z}$$
 multiplication
(n) zero matrix. $k = 0$: $A\vec{x} = 0 \neq \vec{x}$
(2) Identify matrix. $I\vec{x} = \vec{x}$
(3) Picking att columns and rows : $A\vec{e}_{3} = \vec{a}_{3}$
 $\begin{pmatrix} t_{11} \dots t_{2n} \\ A_{2n} \end{pmatrix} \begin{bmatrix} 0 \\ 0 \end{bmatrix} = \begin{pmatrix} 1 \\ A_{2n} \end{bmatrix} \begin{bmatrix} 1 \\ -A_{2n} \end{bmatrix} \begin{bmatrix} 1 \\ -A_$

Matrix vector multiplication

(5) Difference matrix



Application examples. Polynomial evaluation et multiple points P(t) = c1 + c2t+... + Chyth + Cuth - 1 value of p(+) at ti,..., tm $y = \begin{bmatrix} P(H_1) \\ \vdots \\ P(H_n) \end{bmatrix}, y = P(H_1)$ $A = \begin{bmatrix} 1 & t_1 & \dots & t_n^{m-2} & t_n^{m-1} \\ 1 & t_2 & t_2^{m-2} & t_2^{m-1} \\ \vdots & \vdots & \vdots & \vdots \\ 1 & t_n & t_n & \vdots \end{bmatrix}, \text{ then } AC = cy$ $\vec{z} = \begin{bmatrix} z \\ \vdots \\ z \\ x \end{bmatrix}$ Inner product: a,b, ab = y y'is 1-dim
Inner product: a,b, ab , -velder

· Linear dependence of columns

$$A = \begin{bmatrix} \vec{\alpha}_1 & \dots & \vec{\alpha}_n \end{bmatrix}$$
, $\vec{\alpha}_1$'s ar L. D if
 $A\vec{x} = 0$ for some $\vec{z} \neq 0$
and L. I is A $\vec{x} = 0$ for some $\vec{z} \neq 0 = 0$
 $F(A\vec{x} = a_1\vec{\alpha}_1 + \dots + a_n\vec{\alpha}_n = 0)$ for some $\vec{z} \neq 0 = 0$.
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 $F(A\vec{x} = a_1\vec{\alpha}_1 + \dots + a_n\vec{\alpha}_n = 0)$ for any n-vector \vec{b} , there is a
unique m-vector $\vec{a} : A\vec{x} = \vec{b}$
(since $f\vec{\alpha}_1$'s basis : $\vec{b} = \vec{x} \cdot \vec{\alpha}_1 + \dots + a_n\vec{\alpha}_n$)
Properties of matrix-vector $A\vec{z}$
 $= A(\vec{u} + \vec{v}) = A\vec{u} - A\vec{v}$ (distributes) $(\vec{a} + \vec{b})\vec{u} = a \cdot (A\vec{u})$
 $= a + \vec{u}$



Suppose A is the adjacency matrix of a directed graph. The reversed graph is obtained by reversing the directions of all the edges of the original graph. What is the adjacency matrix of the reversed graph? (Express your answer in terms of A.)

$$f_{ij} = \begin{cases} 2 \text{ if } (i,j) \text{ edge } (Gr \ (i,j) \in R \\ G \ ON \end{cases}$$

$$(sqhure, mxn)$$

$$onswer: \quad (i,j) \in R \\ (j,i) \in R^{rev}$$

$$Aij = 1 \implies Bji = 1 \implies B = A^{T}$$

$$Aij = 0 \implies Bji = 0$$

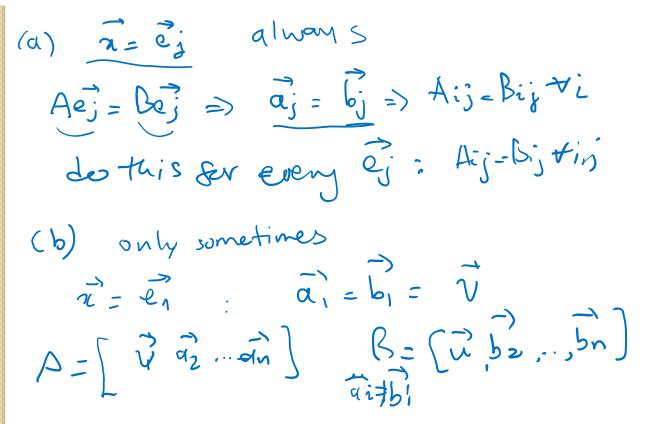
Matrix-vector multiplication. For each of the following matrices, describe in words how x and y = Ax are related. In each case x and y are n-vectors, with n = 3k.

(a)
$$A = \begin{bmatrix} 0 & 0 & I_k \\ 0 & I_k & 0 \\ I_k & 0 & 0 \end{bmatrix}$$
. If is a 3k-dimytector column
(b) $A = \begin{bmatrix} E & 0 & 0 \\ 0 & E & 0 \\ 0 & 0 & E \end{bmatrix}$, where E is the $k \times k$ matrix with all entries $1/k$.
 $\mathcal{H} = \begin{bmatrix} \chi_1 \\ \vdots \\ \chi_{3k} \end{bmatrix}$, $\begin{bmatrix} \chi_{1} \\ \vdots \\ \chi_{2k+1} \end{bmatrix}$, $\begin{bmatrix} \chi_{2k+1} \\ \vdots \\ \chi_{3k} \end{bmatrix}$,

 $\vec{x}_{(2)} = \begin{bmatrix} 0 \cdot \vec{x}_{(1)} + 0 \cdot \vec{x}_{(2)} + J_{k} \cdot \vec{x}_{(3)} \\ \delta \cdot \vec{x}_{(1)} + J_{k} \cdot \vec{x}_{(2)} + 0 \cdot \vec{x}_{(3)} \\ \overline{x}_{(2)} = \int \mathbf{x}_{(1)} + 0 \cdot \vec{x}_{(2)} + 0 \cdot \vec{x}_{(3)} \\ \overline{y}_{(2)} = \int \mathbf{x}_{(2)} \cdot \mathbf{x}_{(2)} + 0 \cdot \vec{x}_{(2)} + 0 \cdot \vec{x}_{(3)} \end{bmatrix}$ Exercises OJR O $= \begin{bmatrix} x_{(2)} \\ \overline{x_{(2)}} \\ \overline{x_{(2)}} \\ \overline{x_{(2)}} \end{bmatrix} = \begin{bmatrix} x_{2k+1} \\ \overline{x_{2k}} \\ \overline{x_{k+1}} \\ \overline{x_{2k}} \\ \overline{x_{2k}} \\ \overline{x_{2k}} \\ \overline{x_{2k}} \\ \overline{x_{k}} \end{bmatrix}$ 2 (b) $E_{\overline{x(i)}} = E_{\overline{x(i)}}$

Let A and B be two $m \times n$ matrices. Under each of the assumptions below, determine whether A = B must always hold, or whether A = B holds only sometimes.

- (a) Suppose Ax = Bx holds for all *n*-vectors *x*.
- (b) Suppose Ax = Bx for some nonzero *n*-vector *x*.



Norm of matrix-vector product. Suppose A is an $m \times n$ matrix and x is an n-vector. A famous inequality relates ||x||, ||A||, and ||Ax||:

 $||Ax|| \le ||A|| ||x||.$

The left-hand side is the (vector) norm of the matrix-vector product; the right-hand side is the (scalar) product of the matrix and vector norms. Show this inequality. *Hints*. Let a_i^T be the *i*th row of A. Use the Cauchy–Schwarz inequality to get $(a_i^T x)^2 \leq ||a_i||^2 ||x||^2$. Then add the resulting m inequalities.



