Introduction to Scientific Computing I

Lecture 10

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Linear Algebra

Vector, Matrix

Linear Algebra was invented to solve equations like this:

$$5x_1 + 12x_2 = -2$$
$$-2x_1 - 3x_2 = 12$$

• By representing them as matrices like this.

$$\left(\begin{array}{ccc}
5 & 12 & -2 \\
-2 & -3 & 12
\end{array}\right)$$

• Or better yet:

$$A = \begin{pmatrix} 5 & 12 \\ -2 & -3 \end{pmatrix} \quad \vec{x} = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \quad \vec{y} = \begin{pmatrix} -2 \\ 12 \end{pmatrix}$$

• Inverse $A\vec{x} = \vec{y} \qquad \longrightarrow \qquad \vec{x} = A^{-1}\vec{y}$

Basics

$$x = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \quad A = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{bmatrix} \quad \begin{bmatrix} Columns & Rows \\ \begin{bmatrix} | & | & | & | \\ a_1 & a_2 & \cdots & a_n \\ | & | & | \end{bmatrix} \quad \begin{bmatrix} -a_1^T & - \\ -a_2^T & - \\ \vdots & \vdots & | \\ -a_m^T & - \end{bmatrix} \\ A_{:,j} = a_j \quad A_{i,:} = a_i^T$$

$$\left[\begin{array}{cccc} | & | & | \\ a_1 & a_2 & \cdots & a_n \\ | & | & | \end{array}\right]$$

$$A_{:,j} = a_j$$

$$egin{bmatrix} -&a_1^T&-\-&a_2^T&-\ &dots\-&a_m^T&- \end{bmatrix}$$

$$A_{i,:} = a_i^T$$

Transpose:
$$(A^T)_{ij} = A_{ji}$$

Sum: Elementwise
$$C_{ij} = A_{ij} + B_{ij}$$

$$C_{ij} = A_{ij} + B_{ij}$$

Identity:
$$I_{ij} = \left\{ \begin{array}{ll} 1 & i = j \\ 0 & i \neq j \end{array} \right.$$
 $\left(\begin{array}{ll} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{array} \right)$ $A^{-1}A = I = AA^{-1}$

$$\left(\begin{array}{cccc}
1 & 0 & 0 \\
0 & 1 & 0 \\
0 & 0 & 1
\end{array}\right)$$

Products

Vector-Vector (inner)

aka: dot product

$$x \cdot y = \begin{bmatrix} x_1 & x_2 & \cdots & x_n \end{bmatrix} \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix} = \sum_{i=1}^n x_i y_i. \qquad \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_m \end{bmatrix} \begin{bmatrix} y_1 & y_2 & \cdots & y_n \end{bmatrix} = \begin{bmatrix} x_1 y_1 & x_1 y_2 & \cdots & x_1 y_n \\ x_2 y_1 & x_2 y_2 & \cdots & x_2 y_n \\ \vdots & \vdots & \ddots & \vdots \\ x_m y_1 & x_m y_2 & \cdots & x_m y_n \end{bmatrix}$$

Vector-Vector (outer)

$$\begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_m \end{bmatrix} \begin{bmatrix} y_1 & y_2 & \cdots & y_n \end{bmatrix} = \begin{bmatrix} x_1y_1 & x_1y_2 & \cdots & x_1y_n \\ x_2y_1 & x_2y_2 & \cdots & x_2y_n \\ \vdots & \vdots & \ddots & \vdots \\ x_my_1 & x_my_2 & \cdots & x_my_n \end{bmatrix}$$

Matrix-Vector

$$y = Ax = \begin{bmatrix} \begin{vmatrix} & & & & \\ & & & \\ & & & \end{vmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} a_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} a_1 \\ a_2 \end{bmatrix} x_1 + \begin{bmatrix} a_2 \\ a_2 \end{bmatrix} x_2 + \dots + \begin{bmatrix} a_n \\ a_n \end{bmatrix} x_n \qquad y = Ax = \begin{bmatrix} - & a_1^T & - \\ - & a_2^T & - \\ \vdots & \vdots & - \\ - & a_m^T & - \end{bmatrix} x = \begin{bmatrix} a_1^T x \\ a_2^T x \\ \vdots \\ a_m^T x \end{bmatrix}$$

Matrix-Matrix

$$C_{ij} = \sum_{k=1}^{n} A_{ik} B_{kj} \qquad C = AB = \begin{bmatrix} - & a_1^T & - \\ - & a_2^T & - \\ \vdots & \vdots & - \\ - & a_m^T & - \end{bmatrix} \begin{bmatrix} | & | & | & | \\ b_1 & b_2 & \cdots & b_p \\ | & | & | & | \end{bmatrix} = \begin{bmatrix} a_1^T b_1 & a_1^T b_2 & \cdots & a_1^T b_p \\ a_2^T b_1 & a_2^T b_2 & \cdots & a_2^T b_p \\ \vdots & \vdots & \ddots & \vdots \\ a_m^T b_1 & a_m^T b_2 & \cdots & a_m^T b_p \end{bmatrix}$$

$$(AB)C = A(BC)$$

 $A(B+C) = AB = AC$
 $AB \neq BA$
 $AI = A$

Norms

$$||x||_1 = \sum_{i=1}^n |x_i|$$

I=1 Norm

$$||x||_2 = \sqrt{\sum_{i=1}^n x_i^2} \qquad ||x||_2^2 = x^T x$$

I=2 Norm

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

I=p Norm

$$||x||_{\infty} = \max_{i} |x_{i}|$$

I=∞ Norm

$$x \cdot y = ||x||_2 ||y||_2 \cos \theta$$

Law of cosines

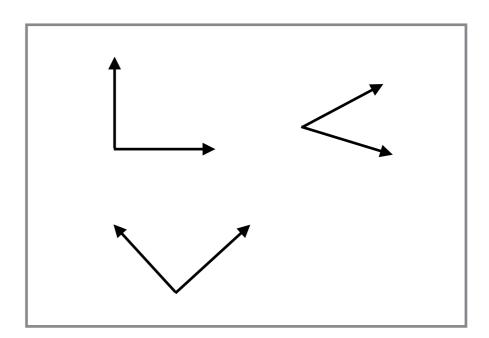
Linear (In)dependence

Consider a set of vectors: $\{x_1, x_2, \dots x_n\}$

Can your write one as sum of others? $x_n = \sum_{i=1}^{n} \alpha_i x_i$

Yes = Linear Dependent No = Linear Independent

Orthogonal: $x^Ty = 0$



Matrix Inverse

$$A\vec{x} = \vec{y} \qquad \longrightarrow \qquad \vec{x} = A^{-1}\vec{y}$$

Step 1: Let p = 0, d = 1;

Step 2: $p \Leftarrow p + 1$

Step 3: If $a_{p,p} = 0$ then cannot calculate inverse, go to step 10.

Step 4: $d' \Leftarrow d \times a_{p,p}$

Step 5: Calculate the new elements of the pivot row by:

$$a_{p,j}^{'} \leftarrow \frac{a_{p,j}}{a_{p,p}}, \text{ where } j = 1, \dots, n, j \neq p$$

Step 6: Calculate the new elements of the pivot column by:

$$a_{i,p}^{'} \leftarrow -\frac{a_{i,p}}{a_{p,p}}, \text{ where } i = 1, \dots, n, i \neq p$$

Step 7: Calculate the rest of the new elements by:

$$\mathbf{a_{i,j}^{'}} \Leftarrow a_{_{i,j}} + a_{_{p,j}} \times \ \mathbf{a_{i,p}^{'}} \ , \ \ where \quad \ i = 1, \cdots, n, j = 1, \cdots, n \ \& \ i, j \neq p$$

Step 8: Calculate the new value of the current pivot location:

$$a_{p,p}^{'} \Leftarrow \frac{1}{a_{p,p}}$$

Step 9: If p < n go to step 2 (n the dimension of the matrix A).

Step 10: Stop. If inverse exists- *A* contains the inverse and *d* is the determinant.