Lecture # 03

Linear Algebra

Outline

- Vectors
 - Operations
- Matrix
 - Operations
- Transformations
 - Scaling
 - Rotation
 - Translation
- Singular value decomposition

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Vector

- Scalar: $x \in \mathbb{R}$
- Vector: $\mathbf{x} \in \mathbb{R}^N$
 - Row Vector $v \in \mathbb{R}^{1 \times n}$

$$\boldsymbol{x} = [\boldsymbol{x}_1 \ \boldsymbol{x}_2 \ \dots \ \boldsymbol{x}_n]$$

$$\mathbf{x} = [\mathbf{x}_1 \ \mathbf{x}_2 \ \dots \ \mathbf{x}_n]$$
• Column vector $\mathbf{v} \in \mathbb{R}^{n \times 1} : \mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = [x_1 \ x_2 \ \dots \ x_n]^T$

Transpose

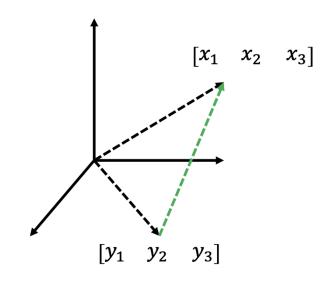
Vectors - use

- Store data in memory
 - Feature vectors
 - Pixel values
 - Any other data for processing

- Any point in coordinate system
 - Can be **n** dimensional

Difference between two points

$$[x1 - y1 \ x2 - y2 \ x3 - y3]$$



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Vector operations

- Norm size of the vector
- P-norm

• Euclidean norm

• L1-norm

$$\left\|x\right\|_{p} = \left(\sum_{i} \left|a_{i}\right|^{p}\right)^{\frac{1}{p}}$$

$$\left\|x\right\|_2 = \left(\sum_i \left|a_i\right|^2\right)^{1/2}$$

$$\|x\|_1 = \left(\sum_i |a_i|\right)$$

$$p \ge 1$$

- Inner product (dot product)
 - Scalar number
 - Multiply corresponding entries and odd

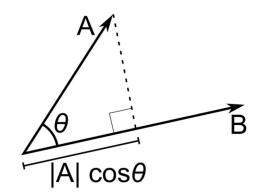
$$\mathbf{x}^T \mathbf{y} = \begin{bmatrix} x_1 & x_2 & \cdots & x_n \end{bmatrix} \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix} = \sum_{k=1}^n x_k y_k$$

Inner product (dot product)

$$\mathbf{x}_i^T \mathbf{x}_i = \sum_{k=1}^{n} (x_k^i)^2$$
 = squared norm of \mathbf{x}_i

A.B is also /A//B/cos (angle between A and B)

• If B is a unit vector, A.B gives projection of A on B



Outer product

$$\boldsymbol{x}_{i}\boldsymbol{x}_{j}^{T} = \begin{bmatrix} x_{1}^{i}x_{1}^{j} & x_{1}^{i}x_{2}^{j} & \cdots & x_{1}^{i}x_{n}^{j} \\ x_{2}^{i}x_{1}^{j} & x_{2}^{i}x_{2}^{j} & \cdots & x_{2}^{i}x_{2}^{j} \\ \vdots & \vdots & \ddots & \vdots \\ x_{n}^{i}x_{1}^{j} & x_{n}^{i}x_{2}^{j} & \cdots & x_{n}^{i}x_{m}^{j} \end{bmatrix} \Longrightarrow$$

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Matrix

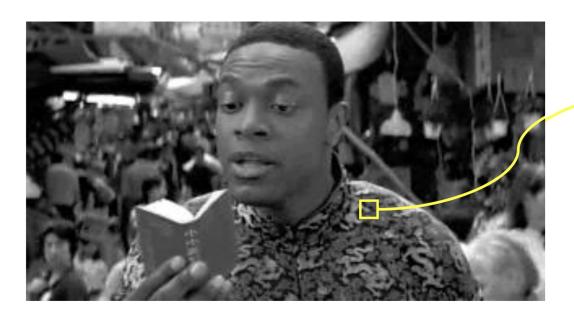
- Array $A \in \mathbb{R}^{m \times n}$ of numbers with shape m by n
 - o m rows and n columns

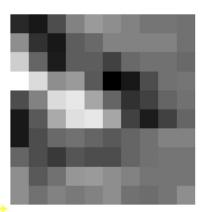
$$\mathbf{A} = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{bmatrix}$$

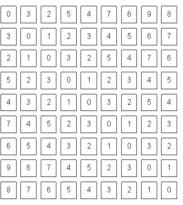
- A row vector is a matrix with single row
- A column vector is a matric with single column

Matrix — Use

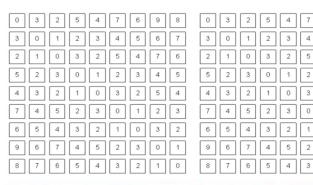
- Image representation grayscale
 - One number per pixel
 - Stored as **n x m** matrix

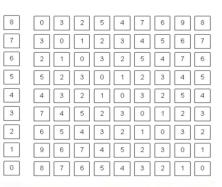






- Image representation RGB
 - 3 numbers per pixel
 - O Stored as **n x m x 3** matrix









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Matrix Operations

Addition

$$\begin{bmatrix} a & b \\ c & d \end{bmatrix} + \begin{bmatrix} e & f \\ g & h \end{bmatrix} = \begin{bmatrix} a+e & b+f \\ c+g & d+h \end{bmatrix}$$

• Both matrices should have same shape, except with a scalar

$$\begin{bmatrix} a & b \\ c & d \end{bmatrix} + 2 = \begin{bmatrix} a+2 & b+2 \\ c+2 & d+2 \end{bmatrix}$$

Same with subtraction

Scaling

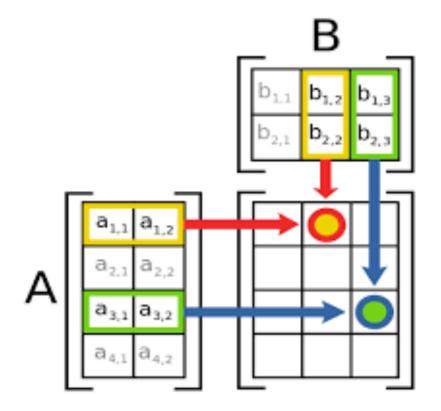
$$s \times \begin{bmatrix} a & b \\ c & d \end{bmatrix} = \begin{bmatrix} sxa & sxb \\ sxc & sxd \end{bmatrix}$$

Hadamard product

$$\begin{bmatrix} a & b \\ c & d \end{bmatrix} \odot \begin{bmatrix} e & f \\ g & h \end{bmatrix} = \begin{bmatrix} axe & bxf \\ cxg & dxh \end{bmatrix}$$

Matrix Operations

- Matrix Multiplication
 - o Complexity?
 - o mxn and nxp
 - Results in **m x p** matrix



- Matrix multiplication
- Let a_i denote the *i*-th column of the matrix A, and
- b_i denote the j-th column of the matrix B

$$\pmb{A} = [\pmb{a}_1 \quad \pmb{a}_2 \quad \cdots \quad \pmb{a}_n]$$
, and $\pmb{B} = [\pmb{b}_1 \quad \pmb{b}_2 \quad \cdots \quad \pmb{b}_m]$

• The product of the two matrices is defined as

$$\boldsymbol{A}^T \boldsymbol{B} = \begin{bmatrix} \boldsymbol{a}_1^T \boldsymbol{b}_1 & \boldsymbol{a}_1^T \boldsymbol{b}_2 & \cdots & \boldsymbol{a}_1^T \boldsymbol{b}_m \\ \boldsymbol{a}_2^T \boldsymbol{b}_1 & \boldsymbol{a}_2^T \boldsymbol{b}_2 & \cdots & \boldsymbol{a}_2^T \boldsymbol{b}_m \\ \vdots & \vdots & \ddots & \vdots \\ \boldsymbol{a}_n^T \boldsymbol{b}_1 & \boldsymbol{a}_n^T \boldsymbol{b}_2 & \cdots & \boldsymbol{a}_n^T \boldsymbol{b}_m \end{bmatrix}$$

Matrix Operations

- Matrix Multiplication another interpretation (very intuity)
- The first of column of AB
 - Linear combination of all the columns in A

$$[a_1b_{11} + a_2b_{12} + ... + a_nb_{1n}]$$

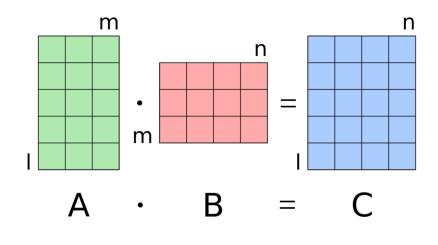
Similarly we can get other columns...

$$A \cdot B = C$$

$$[a_{1}b_{11} + a_{2}b_{12} + ... + a_{n}b_{1n}]$$

$$[a_1b_{11} + a_2b_{12} + ... + a_nb_{1n}]$$

- How about linear combination of all rows of **B**?
 - Each row of C = AB is a linear combination of rows of B



$$[\boldsymbol{b_1}a_{11} + \boldsymbol{b_2}a_{21} + ... + \boldsymbol{b_n}a_{n1}]$$

Matrix Operations

Transpose

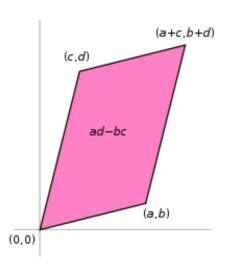
$$\mathbf{A} = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{bmatrix}$$

$$A^{T} = \begin{bmatrix} a_{11} & a_{21} & \cdots & a_{m1} \\ a_{12} & a_{22} & \cdots & a_{m2} \\ \vdots & \vdots & \ddots & \vdots \\ a_{1n} & a_{2n} & \cdots & a_{mn} \end{bmatrix}$$

- Determinant
 - A scalar

• For
$$A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$$

$$, det(A) = ad-bc$$



• Represents area of the parallelogram described by the vectors in the rows of the matrix

Determinant

$$|A| = \sum_{i=1}^K a_{ij} C_{ij}$$

where $oldsymbol{\mathcal{C}_{ij}}$ is the cofactor of $oldsymbol{a_{ij}}$ defined by

$$C_{ii} = (-1)^{i+1} |\mathbf{M}_{ii}|$$
, and

 \mathbf{M}_{ij} is the minor of matrix $oldsymbol{A}$ formed by eliminating row i and column j of $oldsymbol{A}$

Some Properties

- \circ |AB| = |A||B|
- \circ |AB| = |BA|
- \circ $|A^T| = |A|$

$$|A| = \begin{vmatrix} a & b & c \\ d & e & f \\ g & h & i \end{vmatrix} = a \begin{vmatrix} e & f \\ h & i \end{vmatrix} - b \begin{vmatrix} d & f \\ g & i \end{vmatrix} + c \begin{vmatrix} d & e \\ g & h \end{vmatrix}$$

Matrix Operations

- Trace
 - Tr(A) = Sum of diagonal elements

$$Tr \begin{bmatrix} a_{00} & a_{01} & a_{02} \\ a_{10} & a_{11} & a_{12} \\ a_{20} & a_{21} & a_{22} \end{bmatrix} = a_{00} + a_{11} + a_{22}$$

- Properties
 - \circ tr(AB) = tr(BA)
 - $\circ tr(A+B) = tr(A) + tr(B)$

- Inverse
 - Given a matrix **A**, its inverse A⁻¹ is a matrix such that

$$AA^{-1} = A^{-1}A = I$$

- Inverse does not always exist
 - o Singular vs non-singular

- Properties
 - \circ $(A^{-1})^{-1} = A$
 - \circ (AB)⁻¹ = B⁻¹A⁻¹

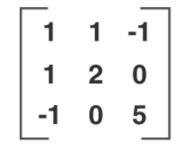
Special matrices

• Symmetric matrix

$$A^T = A$$

Skew- symmetric matrix

$$A^T = -A$$



0	1	-2
-1	0	3
2	-3	0

- Diagonal matrix
 - Used for row scaling

- Identity matrix
 - Special diagonal matrix
 - 1 along diagonals

$$I.A = A$$

$$A=egin{bmatrix} A_1 & 0 & \cdots & 0 \ 0 & A_2 & \cdots & 0 \ dots & dots & \ddots & dots \ 0 & 0 & \cdots & A_n \end{bmatrix}$$

$$\left[\begin{array}{cccc} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{array}\right]$$

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Transformation - scaling

- Matrices are useful for vector transformations
- Matrix multiplication

$$X' = AX$$

$$\begin{bmatrix} s_x & 0 \\ 0 & s_y \end{bmatrix} \times \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} s_x x \\ s_y y \end{bmatrix}$$

Linear combination of columns

Transformation

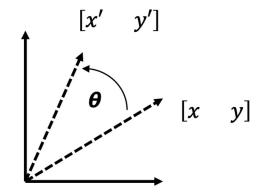
- Rotation
 - Matrix multiplication to rotate a vector

Rotation matrix

$$\begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} \cos \boldsymbol{\theta} & -\sin \boldsymbol{\theta} \\ \sin \boldsymbol{\theta} & \cos \boldsymbol{\theta} \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$$

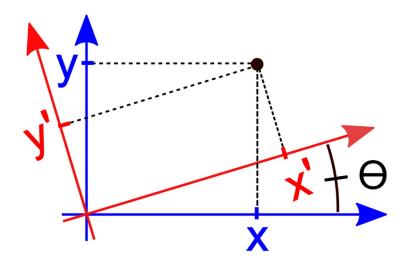
$$x' = \cos \theta x - \sin \theta y$$

 $y' = \sin \theta x + \cos \theta y$



Transformation - rotation

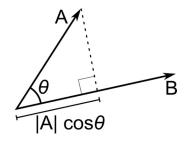
- Rotating axis first
- Vector v = [x y]
 - x projection of v on x axis
 - y- projection of v on y axis

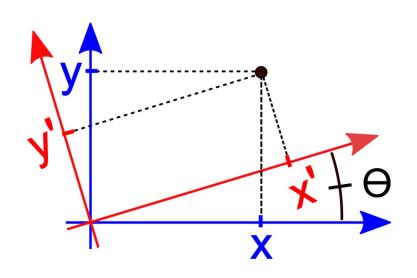


Transformation - rotation

- Rotating axis first
- Vector v = [x y]
 - x projection of v on x axis
 - y- projection of v on y axis

Remember vector dot product?

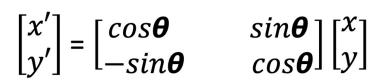


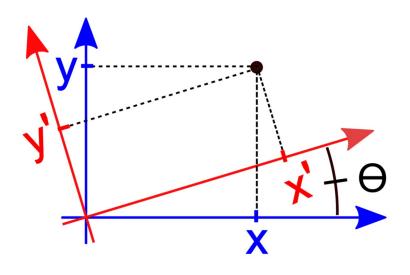


Transformation - rotation

- Now we need new x and y axis
 - \circ x' = v **dot** new x-axis
 - \circ y' = v **dot** new y-axis

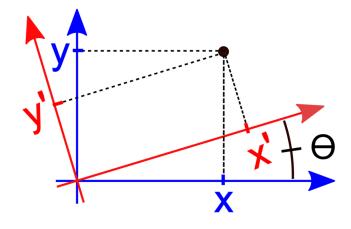
- ullet For rotation of $oldsymbol{ heta}$
 - New x-axis = $[\cos \theta, \sin \theta]$
 - New y-axis = $[-\sin \theta, \cos \theta]$
- We can form a matrix using new axis





- When we rotate the axis to left
- We are rotating the vector to right
- We can use rotation matrix to rotate the vector
- We need new x y axis coordinates
 - When we rotate the axis right
- Updated rotation matrix

$$\begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} \cos \boldsymbol{\theta} & -\sin \boldsymbol{\theta} \\ \sin \boldsymbol{\theta} & \cos \boldsymbol{\theta} \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$$



Transformation

Linear combination

$$\begin{bmatrix} a & b \\ c & d \end{bmatrix} \times \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} ax + by \\ cx + dy \end{bmatrix}$$

- Sufficient for
 - Scaling
 - Rotating and
 - Skew transformations
- But no shifting

Linear combination

$$\begin{bmatrix} a & b & 0 \\ c & d & 0 \\ 0 & 0 & 1 \end{bmatrix} \times \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \begin{bmatrix} ax + by \\ cx + dy \\ 1 \end{bmatrix}$$

- Still be able to:
 - Scaling
 - Rotating and
 - Skew transformations

Transformation

Homogeneous System

$$\begin{bmatrix} a & b & c \\ d & e & f \\ 0 & 0 & 1 \end{bmatrix} x \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \begin{bmatrix} ax + by + c \\ dx + ey + f \\ 1 \end{bmatrix}$$

- We can also perform shifting now
- This is called homogeneous coordinates

Scaling + rotation + translation

Careful about the order

$$\circ V' = (TRS)V$$

$$\begin{bmatrix} 1 & 0 & t_y \\ 0 & 1 & t_x \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} \cos \boldsymbol{\theta} & -\sin \boldsymbol{\theta} & 0 \\ \sin \boldsymbol{\theta} & \cos \boldsymbol{\theta} & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} s_x & 0 & 0 \\ 0 & s_y & 0 \\ 0 & 0 & 1 \end{bmatrix} \mathbf{x} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$$

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Linear independence

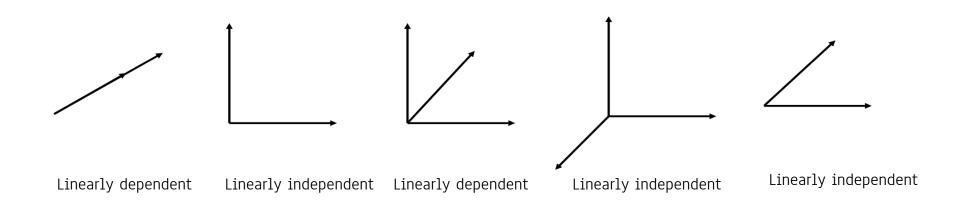
- Linear Independence
 - $\circ \{x_1, x_2 \cdots, x_M\}$ is a set of linearly independent vectors provided

$$a_1 x_1 + a_2 x_2 + \cdots + a_M x_M = 0 \Rightarrow a_1 = 0 = a_2 = a_M$$

- In other words, none of the vectors can be expressed as a linear combination of the the other vectors
 - Each vector is perpendicular to every other vector
 - o For example axis in cartesian coordinate system

Intuition

- In terms of features
 - Person recognition [height, hair color, weight, specs, eye color, etc.]



Matrix factorization

Singular value decomposition (SVD)

 $A = U\Sigma V^{T}$

- If A is mxn matrix, then
 - U will be m x m,
 - \circ Σ will be m x n, and
 - O VT will be n x n

- U and V are unitary matrices
 - Each Column is a unit Vector

 \bullet Σ is a diagonal matrix

Singular value decomposition

Interpretation

$$A = U\Sigma V^{T}$$

- ullet Columns of U are scaled by values in Σ
- The resultant columns are linearly combined by V
- A is formed as a linear combination of columns of U
- If we use all column, we will get original A
- We can just use few columns of U and we get an approximation
 - We call these columns principal components

Application

