

**CS434a/541a: Pattern Recognition**  
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**Lecture 2**

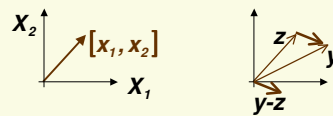
**Why Linear Algebra?**

- For each example (e.g. a fish image), we will extract a set of features (e.g. length, width, color)
- This set of features we will represent as a *feature vector*
  - [length, width, color,...]
- All collected examples will be represented as collection of (feature) vectors
  - $[l_1, w_1, c_1, \dots]$ ,  $[l_2, w_2, c_2, \dots]$ ,  $[l_3, w_3, c_3, \dots]$ , ...  
     example 1           example 2           example 3
- Besides representation, we will often use linear models since they are simple and computationally feasible

**Outline**

- Review of Linear Algebra
  - vectors and matrices
  - products and norms
  - vector spaces and linear transformations
  - eigenvalues and eigenvectors
- Introduction to Matlab




**Vectors**



- n-dimensional row vector  $\mathbf{x} = [x_1 \ x_2 \ \dots \ x_n]$
- Transpose of row vector is column vector  $\mathbf{x}^T = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}$
- *Vector* product (or *inner* or *dot* product)  
 $\langle \mathbf{x}, \mathbf{y} \rangle = \mathbf{x} \cdot \mathbf{y} = \mathbf{x}^T \mathbf{y} = x_1 y_1 + x_2 y_2 + \dots + x_n y_n = \sum_{i=1..k} x_i y_i$

### More on Vectors

- Euclidian norm or length  $|\mathbf{x}| = \sqrt{\langle \mathbf{x}, \mathbf{x} \rangle} = \sqrt{\sum_{i=1..n} x_i^2}$
- If  $|\mathbf{x}|=1$  we say  $\mathbf{x}$  is *normalized* or *unit* length
- Angle  $\theta$  between vectors  $\mathbf{x}$  and  $\mathbf{y}$   $\cos \theta = \frac{\mathbf{x}^T \mathbf{y}}{|\mathbf{x}| |\mathbf{y}|}$

 <p><math>\cos \theta = 0</math> <math>\mathbf{x}^T \mathbf{y} = 0</math> x orthogonal to y <math>\mathbf{x} \perp \mathbf{y}</math></p>	 <p><math>\cos \theta = 1</math> <math>\mathbf{x}^T \mathbf{y} =  \mathbf{x}   \mathbf{y}  &gt; 0</math></p>	 <p><math>\cos \theta = -1</math> <math>\mathbf{x}^T \mathbf{y} = - \mathbf{x}   \mathbf{y}  &lt; 0</math></p>
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- Thus inner product captures direction relationship between  $\mathbf{x}$  and  $\mathbf{y}$

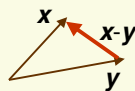
### Linear Dependence and Independence

- Vectors  $\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n$  are linearly **dependent** if there exist constants  $\alpha_1, \alpha_2, \dots, \alpha_n$  s.t.
  1.  $\alpha_1 \mathbf{x}_1 + \alpha_2 \mathbf{x}_2 + \dots + \alpha_n \mathbf{x}_n = \mathbf{0}$
  2. at least one  $\alpha_i \neq 0$
- Vectors  $\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n$  are linearly **independent** if  $\alpha_1 \mathbf{x}_1 + \alpha_2 \mathbf{x}_2 + \dots + \alpha_n \mathbf{x}_n = \mathbf{0} \Rightarrow \alpha_1 = \dots = \alpha_n = 0$

### More on Vectors

- Vectors  $\mathbf{x}$  and  $\mathbf{y}$  are orthonormal if they are orthogonal and  $|\mathbf{x}|=|\mathbf{y}|=1$
- Euclidian distance between vectors  $\mathbf{x}$  and  $\mathbf{y}$

$$|\mathbf{x} - \mathbf{y}| = \sqrt{\sum_{i=1..n} (x_i - y_i)^2}$$



### Vector Spaces and Basis

- The set of all n-dimensional vectors is called a **vector space V**
- A set of vectors  $\{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_n\}$  are called a basis for vector space if any  $\mathbf{v}$  in  $\mathbf{V}$  can be written as  $\mathbf{v} = \alpha_1 \mathbf{u}_1 + \alpha_2 \mathbf{u}_2 + \dots + \alpha_n \mathbf{u}_n$
- $\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_n$  are independent implies they form a basis, and vice versa
- $\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_n$  give an orthonormal basis if
  1.  $|\mathbf{u}_i| = 1 \quad \forall i$
  2.  $\mathbf{u}_i \perp \mathbf{u}_j \quad \forall i \neq j$

## Matrices

- n by m matrix A and its m by n transpose  $A^T$

$$A = \begin{bmatrix} x_{11} & x_{12} & \cdots & x_{1m} \\ x_{21} & x_{22} & \cdots & x_{2m} \\ \vdots & \vdots & \cdots & \vdots \\ x_{n1} & x_{n2} & \cdots & x_{nm} \end{bmatrix} \quad A^T = \begin{bmatrix} x_{11} & x_{12} & \cdots & x_{n1} \\ x_{12} & x_{22} & \cdots & x_{n2} \\ \vdots & \vdots & \cdots & \vdots \\ x_{1m} & x_{2m} & \cdots & x_{nm} \end{bmatrix}$$

## Matrices

- Rank** of a matrix is the number of linearly independent rows (or equivalently columns)
- A square matrix is **non-singular** if its rank equal to the number of rows. If its rank is less than number of rows it is **singular**.

- Identity matrix**  $I = \begin{bmatrix} 1 & 0 & \cdots & 0 \\ 0 & 1 & \cdots & 0 \\ 0 & 0 & \ddots & \vdots \\ 0 & 0 & \cdots & 1 \end{bmatrix}$   
 $AI=IA=A$

- Matrix **A** is **symmetric** if  $A=A^T$

$$\begin{bmatrix} 1 & 2 & 9 & 5 \\ 2 & 7 & 4 & 8 \\ 9 & 4 & 3 & 6 \\ 5 & 8 & 6 & 4 \end{bmatrix}$$

## Matrix Product

$$AB = \begin{bmatrix} a_{11} & a_{12} & a_{13} & \cdots & a_{1d} \\ \vdots & \vdots & \vdots & \cdots & \vdots \\ a_{n1} & a_{n2} & a_{n3} & \cdots & a_{nd} \end{bmatrix} \begin{bmatrix} b_{11} & \cdots & b_{1m} \\ b_{21} & \cdots & b_{2m} \\ b_{31} & \cdots & b_{3m} \\ \vdots & \cdots & \vdots \\ b_{d1} & \cdots & b_{dm} \end{bmatrix} = \begin{bmatrix} c_{1j} \\ \vdots \\ c_{ij} \\ \vdots \end{bmatrix} = C$$

$$c_{ij} = \langle a^i, b_j \rangle$$

$a^i$  is row  $i$  of  $A$   
 $b_j$  is column  $j$  of  $B$

- # of columns of A = # of rows of B
- even if defined, in general  $AB \neq BA$

## Matrices

- Matrix **A** is **positive definite** if  $x^T A x = \sum_{i,j} A_{i,j} x_i x_j > 0$
- Matrix **A** is **positive semi-definite** if  $x^T A x = \sum_{i,j} A_{i,j} x_i x_j \geq 0$
- Trace of a square matrix **A** is sum on the elements on the diagonal

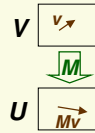
$$tr[A] = \sum_{i=1}^n a_{ii}$$

## Matrices

- **Inverse** of a square matrix  $\mathbf{A}$  is matrix  $\mathbf{A}^{-1}$  s.t.  $\mathbf{A}\mathbf{A}^{-1} = \mathbf{I}$
- If  $\mathbf{A}$  is singular or not square, inverse does not exist. **Pseudo-inverse**  $\mathbf{A}^\dagger$  is defined whenever  $\mathbf{A}^\dagger\mathbf{A}$  is not singular (it is square)
  - $\mathbf{A}^\dagger = (\mathbf{A}^\dagger\mathbf{A})^{-1}\mathbf{A}^\dagger$
  - $\mathbf{A}\mathbf{A}^\dagger = (\mathbf{A}^\dagger\mathbf{A})^{-1}\mathbf{A}\mathbf{A}^\dagger = \mathbf{I}$

## Linear Transformations

- A linear transformation from vector space  $\mathbf{V}$  to vector space  $\mathbf{U}$  is a mapping which can be represented by a matrix  $\mathbf{M}$ :
  - $\mathbf{u} = \mathbf{M}\mathbf{v}$
- If  $\mathbf{U}$  and  $\mathbf{V}$  have the same dimension,  $\mathbf{M}$  is a square matrix
- In pattern recognition, often  $\mathbf{U}$  has smaller dimensionality than  $\mathbf{V}$ , i.e. transformation  $\mathbf{M}$  is used to reduce the number of features.



$$\mathbf{M} \begin{bmatrix} \mathbf{v} \end{bmatrix} = \begin{bmatrix} \mathbf{u} \end{bmatrix}$$

## Matrices

- Determinant of  $n$  by  $n$  matrix  $\mathbf{A}$  is
 
$$\det(\mathbf{A}) = \sum_{k=1}^n (-1)^{k+i} a_{ik} \det(\mathbf{A}_{ik})$$
  - Where  $\mathbf{A}_{ik}$  obtained from  $\mathbf{A}$  by removing the  $i$ th row and  $k$ th column
- Absolute value of determinant gives the volume of parallelepiped spanned by the matrix rows
 
$$\{\beta_1 \mathbf{a}^1 + \beta_2 \mathbf{a}^2 + \dots + \beta_n \mathbf{a}^n\}$$

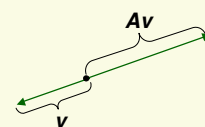
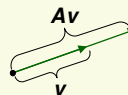
$$\beta_i \in [0, 1] \quad \forall i$$



## Eigenvectors and Eigenvalues

Note:  $\mathbf{A}\mathbf{0} = \lambda\mathbf{0}$  for any  $\lambda$ , not interesting

- Given  $n$  by  $n$  matrix  $\mathbf{A}$ , and nonzero vector  $\mathbf{x}$ . Suppose there is  $\lambda$  which satisfies  $\mathbf{A}\mathbf{x} = \lambda\mathbf{x}$ 
  - $\mathbf{x}$  is called an eigenvector of  $\mathbf{A}$
  - $\lambda$  is called an eigenvalue of  $\mathbf{A}$
- Linear transformation  $\mathbf{A}$  maps an eigenvector  $\mathbf{v}$  in a simple way. Magnitude changes by  $\lambda$ , direction
  - If  $\lambda > 0$
  - If  $\lambda < 0$



## ***Eigenvectors and Eigenvalues***

- If  $\mathbf{A}$  is real and symmetric, then all eigenvalues are real (not complex)
- If  $\mathbf{A}$  is non singular, all eigenvalues are non zero
- If  $\mathbf{A}$  is positive definite, all eigenvalues are positive

- Starting matlab
  - xterm -fn 12X24
  - matlab
- Basic Navigation
  - quit
  - more
  - help general
- Scalars, variables, basic arithmetic
  - Clear
  - + - \* / ^
  - help arith
- Relational operators
  - ==, &, |, ~, xor
  - help relop
- Lists, vectors, matrices
  - A=[2 3;4 5]
  - A'
- Matrix and vector operations
  - find(A>3), colon operator
  - \* / ^ .\* ./ .^
  - eye(n), norm(A), det(A), eig(A)
  - max,min,std
  - help matfun
- Elementary functions
  - help elfun
- Data types
  - double
  - Char
- Programming in Matlab
  - .m files
  - scripts
  - function y=square(x)
  - help lang
- Flow control
  - if i== 1else end, if else if end
  - for i=1:0.5:2 ... end
  - while i == 1 ... end
  - Return
  - help lang
- Graphics
  - help graphics
  - help graph3d
- File I/O
  - load,save
  - fopen, fclose, fprintf, fscanf

***MATLAB***