

CS434a/541a: Pattern Recognition
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Lecture 2

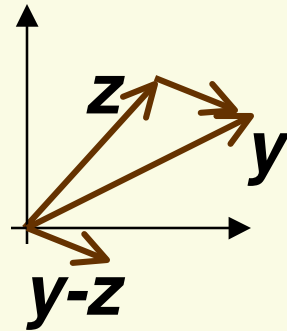
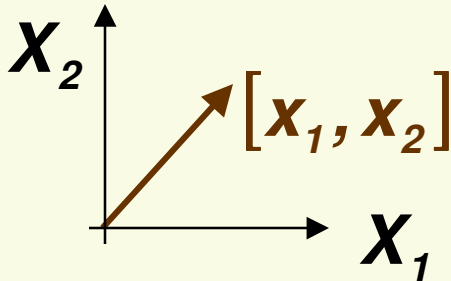
Outline

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

Why Linear Algebra?

- For each data point, we will represent a set of features as feature vector
 - [length, weight, color, ...]
- Collected data will be represented as collection of (feature) vectors
 - $[l_1, w_1, c_1, \dots]$ $[l_2, w_2, c_2, \dots]$ $[l_3, w_3, c_3, \dots]$...
- Linear models are simple and computationally feasible

Vectors



- n-dimensional row vector $\mathbf{x} = [\mathbf{x}_1 \ \mathbf{x}_2 \ \dots \ \mathbf{x}_n]$



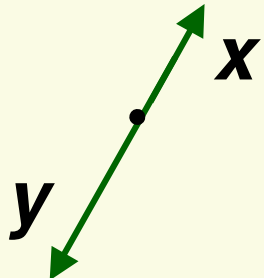
- Transpose of row vector is column vector $\mathbf{x}^T = \begin{bmatrix} \mathbf{x}_1 \\ \mathbf{x}_2 \\ \vdots \\ \mathbf{x}_n \end{bmatrix}$

- *Vector* product (or *inner* or *dot* product)

$$\langle \mathbf{x}, \mathbf{y} \rangle = \mathbf{x} \cdot \mathbf{y} = \mathbf{x}^T \mathbf{y} = \mathbf{x}_1 \mathbf{y}_1 + \mathbf{x}_2 \mathbf{y}_2 + \dots + \mathbf{x}_n \mathbf{y}_n = \sum_{i=1..k} \mathbf{x}_i \mathbf{y}_i$$

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 θ between vectors \mathbf{x} and \mathbf{y} $\cos \theta = \frac{\mathbf{x}^T \mathbf{y}}{|\mathbf{x}| |\mathbf{y}|}$

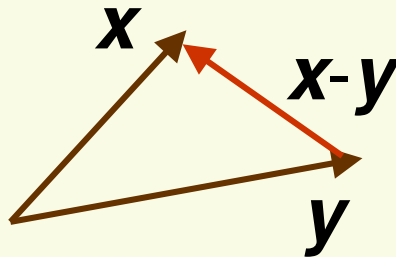
 <p>$\cos \theta = 0$ $\mathbf{x}^T \mathbf{y} = 0$ x orthogonal to y $\mathbf{x} \perp \mathbf{y}$</p>	 <p>$\cos \theta = 1$ $\mathbf{x}^T \mathbf{y} = \mathbf{x} \mathbf{y} > 0$</p>	 <p>$\cos \theta = -1$ $\mathbf{x}^T \mathbf{y} = - \mathbf{x} \mathbf{y} < 0$</p>
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-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

More on Vectors

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

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



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 $\alpha_i \neq \mathbf{0}$
- Vectors $\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n$ are linearly **independent** if $\alpha_1 \mathbf{x}_1 + \alpha_2 \mathbf{x}_2 + \dots + \alpha_n \mathbf{x}_n = \mathbf{0} \Rightarrow \alpha_1 = \dots = \alpha_n = \mathbf{0}$

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 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

$$\mathbf{A} = \begin{bmatrix} \mathbf{x}_{11} & \mathbf{x}_{12} & \cdots & \mathbf{x}_{1m} \\ \mathbf{x}_{21} & \mathbf{x}_{22} & \cdots & \mathbf{x}_{2m} \\ \vdots & \vdots & \cdots & \vdots \\ \mathbf{x}_{n1} & \mathbf{x}_{n2} & \cdots & \mathbf{x}_{nm} \end{bmatrix}$$

$$\mathbf{A}^T = \begin{bmatrix} \mathbf{x}_{11} & \mathbf{x}_{12} & \cdots & \mathbf{x}_{n1} \\ \mathbf{x}_{12} & \mathbf{x}_{22} & \cdots & \mathbf{x}_{n2} \\ \vdots & \vdots & \cdots & \vdots \\ \mathbf{x}_{1m} & \mathbf{x}_{2m} & \cdots & \mathbf{x}_{nm} \end{bmatrix}$$

Matrix Product

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

$$\mathbf{c}_{ij} = \langle \mathbf{a}^i, \mathbf{b}_j \rangle$$

\mathbf{a}^i is row i of \mathbf{A}
 \mathbf{b}_j is column j of \mathbf{B}

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

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** $I = \begin{bmatrix} 1 & 0 & \dots & 0 \\ 0 & 1 & \dots & 0 \\ 0 & 0 & \ddots & \vdots \\ 0 & 0 & \dots & 1 \end{bmatrix}$
 $AI=IA=A$

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

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

Matrices

- Matrix **A** is **positive definite** if

$$\mathbf{x}^T \mathbf{A} \mathbf{x} = \sum_{i,j} \mathbf{A}_{i,j} \mathbf{x}_i \mathbf{x}_j > 0$$

- Matrix **A** is **positive semi-definite** if

$$\mathbf{x}^T \mathbf{A} \mathbf{x} = \sum_{i,j} \mathbf{A}_{i,j} \mathbf{x}_i \mathbf{x}_j \geq 0$$

- Trace of a square matrix **A** is sum on the elements on the diagonal

$$\text{tr}[\mathbf{A}] = \sum_{i=1}^n \mathbf{a}_{ii}$$

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 $\mathbf{A}^\top \mathbf{A}$ is not singular (it is square)
 - $\mathbf{A}^\dagger = (\mathbf{A}^\top \mathbf{A})^{-1} \mathbf{A}^\top$
 - $\mathbf{A}^\dagger \mathbf{A} = (\mathbf{A}^\top \mathbf{A})^{-1} \mathbf{A}^\top \mathbf{A} = \mathbf{I}$

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

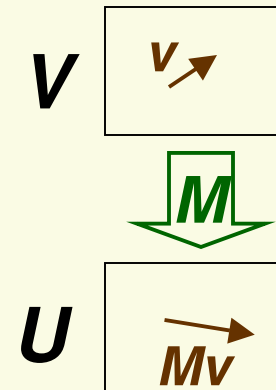
$$\{\beta_1 \mathbf{a}^1 + \beta_2 \mathbf{a}^2 + \dots + \beta_n \mathbf{a}^n\}$$

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



Linear Transformations

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

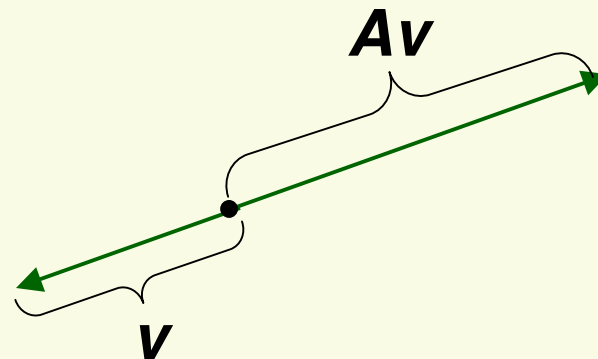
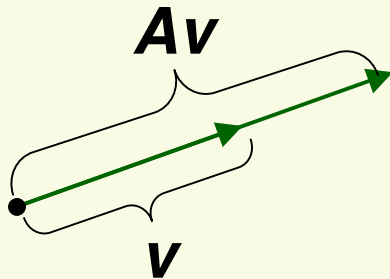


$$M \begin{bmatrix} v \end{bmatrix} = \begin{bmatrix} u \end{bmatrix}$$

Eigenvectors and Eigenvalues

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

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

MATLAB

- 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== 1 else 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