

MA232 Linear Algebra

Alex Myasnikov

Stevens Institute of Technology

November 7, 2011

Eigenvectors and eigenvalues.

- Suppose we can *diagonalize* A , i.e. for some diagonal matrix D

$$A = SDS^{-1}$$

- Algebraic operations can be performed very efficiently

$$A^m = (SDS^{-1})^m = (SDS^{-1})(SDS^{-1}) \cdots (SDS^{-1}) = SD^m S^{-1}$$

$$A + B = SD_1 S^{-1} + SD_2 S^{-1} = S(D_1 + D_2)S^{-1}$$

$$cA = c(SDS^{-1}) = ScDS^{-1}$$

- In fact, for any polynomial function $f(t)$

$$f(A) = Sf(D)S^{-1} = S [\text{diag}(f(s_{11}) \ f(s_{22}) \ \cdots \ f(s_{nn}))] S^{-1}$$

Eigenvectors and eigenvalues.

A nonzero vector $\bar{\mathbf{x}} \neq \bar{\mathbf{0}}$ is called an **eigenvector** of matrix A if

$$A\bar{\mathbf{x}} = \lambda\bar{\mathbf{x}}$$

for some scalar λ , called **eigenvalue** of A corresponding to $\bar{\mathbf{x}}$

– Multiplication by A does not change the direction of $\bar{\mathbf{x}}$.

Eigenvectors and eigenvalues.

- $A\bar{x} = \lambda\bar{x} \Rightarrow (A - \lambda I)\bar{x} = \bar{0}$
- $\bar{x} \in N(A - \lambda I)$ null space of $A - \lambda I$
- If we know λ , then we can solve for \bar{x}

Eigenvectors and eigenvalues.

- Eigenvector $\bar{\mathbf{x}} \neq \bar{\mathbf{0}}$
- $(A - \lambda I)\bar{\mathbf{x}} = \bar{\mathbf{0}}$ has a non-zero solution if $A - \lambda I$ is singular
- Therefore if λ is eigenvalue then

$$\det(A - \lambda I) = 0$$

Eigenvectors and eigenvalues.

- $\Delta(\lambda) = \det(A - \lambda I)$ is a *characteristic polynomial* of matrix A
- For $n \times n$ matrix A characteristic polynomial has degree n
- Characteristic polynomial of triangular matrix U

$$\Delta(\lambda) = \det(U - \lambda I) = (u_{11} - \lambda)(u_{22} - \lambda) \cdots (u_{nn} - \lambda)$$

Warning! If U is triangular form of A , then $\Delta(A)$ and $\Delta(U)$ may be different

Eigenvectors and eigenvalues.

- Characteristic polynomial of degree 2:

$$A = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \quad \text{Then}$$

$$\begin{aligned} \Delta(A) &= \lambda^2 - \lambda(a_{11} + a_{22}) + (a_{11}a_{22} - a_{12}a_{21}) \\ &= \lambda^2 - \text{trace}(A)\lambda + \det(A) \end{aligned}$$

Eigenvectors and eigenvalues.

- Characteristic polynomial of degree 3:

$$A = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix} \quad \text{Then}$$

$$\Delta(A) = \lambda^3 - \text{trace}(A)\lambda^2 + (C_{11} + C_{22} + C_{33})\lambda - \det(A)$$

- C_{ij} are cofactors of A

Eigenvectors and eigenvalues.

- Characteristic polynomial of degree n . Let A is $n \times n$ then

$$\Delta(A) = \lambda^n - S_1\lambda^{n-1} + S_2\lambda^{n-2} + \cdots + (-1)^n S_n,$$

where S_k is the sum of principal minors of order k .

Eigenvectors and eigenvalues.

Computing eigenvalues and eigenvectors

- 1 Find characteristic polynomial $\Delta(A)$ of A
- 2 Obtain eigenvalues of A by finding roots of $\Delta(A)$
- 3 For each eigenvalue λ of A
 - a Form matrix $(A - \lambda I)$
 - b Find a basis of the nullspace $N(A - \lambda I)$
 - c This basis are eigenvectors of A belonging to λ

Eigenvectors and eigenvalues.

- Identity Matrix: $I\bar{\mathbf{x}} = 1 \cdot \bar{\mathbf{x}}$, $\lambda = 1$, $\bar{\mathbf{x}}$ - any

- Rotation matrix:

$$\begin{bmatrix} \cos \varphi & \sin \varphi \\ -\sin \varphi & \cos \varphi \end{bmatrix} : \text{no real eigenvalues unless } \varphi = k\pi$$

- Reflexion matrix:

$$\begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} : \lambda = \pm 1$$

- Projection matrix:

$$\begin{bmatrix} 0 & 1 \\ 0 & 1 \end{bmatrix} : \lambda_1 = 0, \lambda_2 = 1$$

- Singular matrix always has $\lambda = 0$

Eigenvectors and eigenvalues.

- Computing eigenvalues cannot be reduced to triangular matrix. If $A = LU$, then eigenvalues of A may be different from eigenvalues of U :

$$A = \begin{bmatrix} 1 & 2 \\ 2 & 6 \end{bmatrix}, \lambda_1 = \frac{7 + \sqrt{41}}{2}, \lambda_2 = \frac{7 - \sqrt{41}}{2};$$

$$U = \begin{bmatrix} 1 & 2 \\ 0 & 2 \end{bmatrix}, \lambda_1 = 1, \lambda_2 = 2.$$

Eigenvectors and eigenvalues.

- $\lambda_1 \cdot \lambda_2 \cdots \lambda_n = \det(A)$
- $\lambda_1 + \lambda_2 + \cdots + \lambda_n = A_{11} + A_{22} + \cdots + A_{nn}$
- Eigenvalues of real matrix may be not real numbers

Diagonalization

- Let $\bar{\mathbf{x}}_1, \dots, \bar{\mathbf{x}}_n$ be eigenvectors and $\lambda_1, \dots, \lambda_n$ be corresponding eigenvalues of A
- $S = [\bar{\mathbf{x}}_1 \ \bar{\mathbf{x}}_2 \ \cdots \ \bar{\mathbf{x}}_n]$ is the *eigenvector* matrix of A
- $D = \text{diag}(\lambda_1 \ \lambda_2 \ \cdots \ \lambda_n)$ is the *eigenvalue* matrix of A

A is diagonalizable if and only if A has n independent eigenvectors.
Then

$$A = SDS^{-1} \text{ and } D = S^{-1}AS$$

Diagonalization

Proof:

$$\begin{aligned} AS &= A \begin{bmatrix} | & | & & | \\ \bar{\mathbf{x}}_1 & \bar{\mathbf{x}}_2 & \cdots & \bar{\mathbf{x}}_n \\ | & | & & | \end{bmatrix} = \begin{bmatrix} | & | & & | \\ \lambda_1 \bar{\mathbf{x}}_1 & \lambda_2 \bar{\mathbf{x}}_2 & \cdots & \lambda_n \bar{\mathbf{x}}_n \\ | & | & & | \end{bmatrix} \\ &= \begin{bmatrix} | & | & & | \\ \bar{\mathbf{x}}_1 & \bar{\mathbf{x}}_2 & \cdots & \bar{\mathbf{x}}_n \\ | & | & & | \end{bmatrix} \begin{bmatrix} \lambda_1 & & & \\ & \lambda_2 & & \\ & & \ddots & \\ & & & \lambda_n \end{bmatrix} = SD \end{aligned}$$

– Since S has n independent columns it is invertible and

$$AS = SD \Rightarrow A = SDS^{-1} \text{ or } D = S^{-1}AS$$

Diagonalization.

Suppose $\bar{\mathbf{x}}_1, \dots, \bar{\mathbf{x}}_k$ are nonzero eigenvectors of A corresponding to distinct eigenvalues $\lambda_1, \dots, \lambda_k$. Then $\bar{\mathbf{x}}_1, \dots, \bar{\mathbf{x}}_k$ are linearly independent.

Proof:

- Prove by contradiction: Suppose $\bar{\mathbf{x}}_1, \dots, \bar{\mathbf{x}}_n$ are linearly dependent and $\bar{\mathbf{x}}_1, \dots, \bar{\mathbf{x}}_s$ is the minimal dependent set, i.e. $\bar{\mathbf{x}}_2, \dots, \bar{\mathbf{x}}_s$ are independent but

$$\bar{\mathbf{x}}_1 = a_2\bar{\mathbf{x}}_2 + a_3\bar{\mathbf{x}}_3 + \cdots + a_s\bar{\mathbf{x}}_s \quad (\star)$$

- Note order can be taken arbitrary

Diagonalization.

Proof (continue):

- Multiply (\star) by A . Using eigenvector property:

$$A\bar{\mathbf{x}}_1 = A(a_2\bar{\mathbf{x}}_2 + \cdots + a_s\bar{\mathbf{x}}_s) \Rightarrow \lambda_1\bar{\mathbf{x}}_1 = a_2\lambda_2\bar{\mathbf{x}}_2 + \cdots + a_s\lambda_s\bar{\mathbf{x}}_s$$

- Multiply (\star) by λ_1

$$\lambda_1\bar{\mathbf{x}}_1 = a_2\lambda_1\bar{\mathbf{x}}_2 + \cdots + a_s\lambda_1\bar{\mathbf{x}}_s$$

- Subtract one from another we get

$$a_2(\lambda_1 - \lambda_2)\bar{\mathbf{x}}_2 + a_3(\lambda_1 - \lambda_3)\bar{\mathbf{x}}_3 + \cdots + a_s(\lambda_1 - \lambda_s)\bar{\mathbf{x}}_s = 0$$

- Since all λ_i are distinct, then all a_j , $j = 2, \dots, s$ must be zero which contradicts assumption that $\bar{\mathbf{x}}_2, \dots, \bar{\mathbf{x}}_s$ are linearly independent.

Diagonalization.

Matrix Diagonalization

- Let A be $n \times n$ matrix
- Find eigenvectors $\bar{\mathbf{x}}_1, \dots, \bar{\mathbf{x}}_n$ and corresponding eigenvalues of A
- Consider the collection of *distinct* vectors $\bar{\mathbf{x}}_1, \dots, \bar{\mathbf{x}}_m$
- If $m \neq n$ then matrix is **not diagonalizable**
- If $m = n$ then

$$A = SDS^{-1},$$

where S is the eigenvector matrix and D is the diagonal eigenvalue matrix