STAT 714 fa 2023

Linear algebra review 5/6 Eigenvalues and eigenvectors

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These slides are an instructional aid; their sole purpose is to display, during the lecture, definitions, plots, results, etc. which take too much time to write by hand on the blackboard. They are not intended to explain or expound on any material.

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These notes include scanned excerpts from Lay (2003):



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2 Determinants

3 Diagonalization

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Eigenvectors and eigenvalues

Let **A** be an $n \times n$ matrix.

 An eigenvector of A is a nonzero vec. x such that Ax = λx for some scalar λ.
 A scalar λ is an eigenvector of A if there is a nontrivial solution to Ax = λx. Such an x is called an eigenvector corresponding to λ.

Interpretation: The magnitudes of the eigenvalues of **A** represent the amount by which **A** stretches or shrinks certain vectors.

$$A_{\mathcal{X}} = \lambda_{\mathcal{X}}^{*}$$

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 $\begin{cases} \begin{pmatrix} 1 & 6 \\ 5 & 2 \end{pmatrix} - \begin{pmatrix} -\epsilon_1 \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \end{pmatrix} = \begin{bmatrix} 5 & 6 \\ 5 & 6 \end{bmatrix}$ cheal if $\begin{pmatrix} 5 & 6 \\ 5 & 6 \end{pmatrix} = 2$ has an trial solution. The weight $\sqrt{2} = \begin{bmatrix} 6 \\ -5 \end{bmatrix}$ is a solution so $Y \equiv 5, -4$ is an eigenvalue.



Every verter in Spen $\left\{ \begin{pmatrix} 6 \\ -5 \end{pmatrix} \right\}$ is else en eigenverter of A corresponding to the eigenvolve -4.

Eigenspaces

If λ is an eigenvalue of **A**, the set of all solutions to $(\mathbf{A} - \lambda \mathbf{I}_n)\mathbf{x} = \mathbf{0}$ is called the *eigenspace* of **A** corresponding to λ .

Exercise: An eigenvalue of the matrix below is **2**. Find a basis for the corresponding eigenspace:

$$\mathbf{A} = \begin{bmatrix} \mathbf{4} & -1 & \mathbf{6} \\ 2 & \mathbf{1} & \mathbf{6} \\ 2 & -1 & \mathbf{8} \end{bmatrix}$$

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:
 $\begin{cases} \chi : A \chi = 2 \chi \end{cases} = \begin{cases} \chi : (A - 2I) \chi = 0 \end{cases} = Nul(A - 2I) \end{cases}$

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$$\begin{bmatrix} A - 2 I & 0 \end{bmatrix} = \begin{pmatrix} 2 & -1 & 6 & 0 \\ 2 & -1 & 6 & 0 \\ 2 & -1 & 6 & 0 \end{pmatrix}$$

$$\sim \begin{pmatrix} 2 & -1 & 6 & 0 \\ 0 & 6 & 0 & 0 \\ 0 & 6 & 0 & 6 \end{pmatrix}$$

2 x - x + 6 x - 0



Theorem (Linear independence of eigenspaces)

If $\mathbf{v}_1, \ldots, \mathbf{v}_r$ are eigenvectors corresponding to distinct eigenvalues $\lambda_1, \ldots, \lambda_r$ of a matrix, then $\{\mathbf{v}_1, \ldots, \mathbf{v}_r\}$ is linearly independent

Prove the result.
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$$\lambda_{1},...,\lambda_{r}$$
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$$A_{\mu\rho\tau_{1}} = A\left(c_{1}\chi_{1} + \dots + c_{\mu}\chi_{\mu}\right)$$

$$= c_{1}A_{\chi_{1}} + \dots + c_{\mu}A_{\chi\mu}$$

$$= c_{1}\lambda_{1}\chi_{1} + \dots + c_{\mu}\lambda_{\mu}\chi_{\mu}$$

$$\lambda_{\mu\tau_{1}} = \lambda_{\mu\tau_{1}}\left(c_{1}\chi_{1} + \dots + c_{\mu}\lambda_{\mu}\chi_{\mu}\right)$$

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The Invertible Matrix Theorem (continued)

Let A be an $n \times n$ matrix. Then A is invertible if and only if:

- s. The number 0 is not an eigenvalue of A.
- t. The determinant of A is not zero.

Prove the first of the above results.

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Recall det
$$\begin{bmatrix} a & b \\ c & d \end{bmatrix} = ad - bc.$$

For larger matrices the determinant is defined as follows:

Definition of determinant by co-factor expansion

Let **A** be an $n \times n$ matrix with *i*, *j* entry a_{ij} and let $A_{(i,j)}$ be the matrix **A** with row *i* and column *j* removed. Then, for i, j = 1, ..., n, define the (i, j)-cofactor as

$$C_{ij} = (-1)^{i+j} \det \mathbf{A}_{(i,j)}.$$

Then for any *i* and *j* we have det
$$\mathbf{A} = \sum_{k=1}^{n} a_{ik} C_{ik} = \sum_{k=1}^{n} a_{kj} C_{kj}$$
.

Often write det **A** as $|\mathbf{A}|$.

This requires over *n*! multiplications, so computers use a different method.

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Exercise: Compute the determinant of

$$A = \begin{bmatrix} 1 & 5 & 0 \\ 2 & 4 & -1 \\ 0 & -2 & 0 \end{bmatrix}.$$

= $2 \begin{bmatrix} 1 & 5 & 0 \\ 2 & 4 & -1 \\ 0 & -2 & 0 \end{bmatrix} + 0 \begin{bmatrix} 5 & 0 \\ -2 & 0 \end{bmatrix} + 0 \begin{bmatrix} 5 & 0 \\ -2 & 0 \end{bmatrix}$

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The determinant and cofactors give a formula for a matrix inverse:

Theorem (An inverse formula using cofactors) If **A** is an invertible $n \times n$ matrix, we have $\begin{bmatrix} C_{11} & C_{12} & \cdots & C_{1n} \\ C & C & C & C \end{bmatrix}^{\mathsf{T}}$

$$\mathbf{A}^{-1} = \frac{1}{\det \mathbf{A}} \begin{bmatrix} C_{21} & C_{22} & \dots & C_{2n} \\ \vdots & \vdots & & \vdots \\ C_{n1} & C_{n2} & \dots & C_{nn} \end{bmatrix}$$

Exercise: Compute the inverse of the matrix on the previous slide.

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$$\mathbf{A} = \begin{bmatrix} 1 & 5 & 0 \\ 2 & 4 & -1 \\ 0 & -2 & 0 \end{bmatrix}.$$

$$\begin{bmatrix} + & - & + \\ - & + & - \\ + & - & + \end{bmatrix}$$







Discuss (no time to prove) the above results.

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Theorem (Finding eigenvalues with the characteristic equation) A scalar λ is an eigenvalue of an $n \times n$ matrix **A** iff λ satisfies $det(\mathbf{A} - \lambda \mathbf{I}) = 0$.

The equation det($\mathbf{A} - \lambda \mathbf{I}$) = 0 is called the *characteristic equation*.

\squareHS an *n*-degree polynomial, which has *n* roots (some roots may be complex).





Prove the results.





The *trace* of a square matrix A, denoted tr(A), is the sum of its diagonal entries.

Theorem (Properties of the trace)For any matrices A and B, we haveImage: tr(AB) = tr(BA).Image: tr(A^TA) = $\sum_i \sum_j a_{ij}^2$, where a_{ij} are the entries of A.

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The function $p_{A}(t) = determined is called the$ *characteristic polynomial*of**A**.

Theorem (Expansion of characteristic polynomial)

The characteristic polynomial of an $n \times n$ matrix **A** has the terms

$$p_{\mathbf{A}}(t) = t^{n} - (\operatorname{tr} \mathbf{A})t^{n-1} + \cdots + (-1)^{n} \operatorname{det} \mathbf{A}.$$

Exercise: For an $n \times n$ matrix **A** with eigenvalues $\lambda_1, \ldots, \lambda_n$, use above to show

1 tr
$$A = \sum_{i=1}^{n} \lambda_i$$

2 det $A = \prod_{i=1}^{n} \lambda_i$
IP $\lambda_1, \dots, \lambda_n$ on eign velues of A , then we can write
 $P_A(t) = (t - \lambda_1)(t - \lambda_2) \cdot \dots \cdot (t - \lambda_n) = t^n - (\lambda_1 + \lambda_2 + \dots + \lambda_n) t^{-1} + \dots + (-1)^n \lambda_1 \cdot \lambda_2 \cdot \dots \cdot \lambda_n$

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Theorem (Further properties of the determinant)

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See Res A.18 of Monahan (2008).

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Eigenvalues and eigenvectors



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A square matrix **A** is *diagonalizable* if $\mathbf{A} = \mathbf{P}\mathbf{D}\mathbf{P}^{-1}$ with **P** invertible, **D** diagonal.

P is invertible

Theorem (Sufficient and necessary conditions for diagonalizability)

 $A = PDP^{-1}$ with D diagonal iff

- the columns of **P** are <u>n</u> linearly independent eigenvectors of **A**, and
- **2** the diagonal entries of **D** are the corresponding eigenvalues of **A**.

Prove the result.

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$$= \begin{bmatrix} A_{y_1} \cdots A_{y_n} \end{bmatrix}$$

$$= \begin{bmatrix} a_{iy_1} \cdots a_n y_n \end{bmatrix}$$

$$= \begin{bmatrix} y_1 \cdots y_n \end{bmatrix} \begin{bmatrix} a_i & & \\ \vdots & a_n \end{bmatrix}$$

$$= \begin{bmatrix} p & p \\ man \end{bmatrix}$$
Since y_1, \dots, y_n are transfy indep. P is involute.
Therefore we have

$$A = P D P^{-1}$$

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$$= A = P D P^{-1}$$

$$= \begin{bmatrix} A_{y_1} \cdots y_n \end{bmatrix}$$

$$A = P D P^{-1}$$

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$$= \begin{bmatrix} A_{y_1} \cdots y_n \end{bmatrix}$$

$$= \begin{bmatrix} A_{y_1} \cdots A_{y_n} \end{bmatrix}$$

$$= \begin{bmatrix} A_{y_1} \cdots A_{y_n} \end{bmatrix}$$

$$= \begin{bmatrix} A_{y_1} \cdots A_{y_n} \end{bmatrix}$$

1,

=>
$$\left[A_{p_1} \cdots A_{p_n}\right] = \left[d_{p_1} \cdots d_{p_n}\right]$$

Exercise: If possible, diagonalize the matrix

$$\mathbf{A} = \begin{bmatrix} 1 & 3 & 3 \\ -3 & -5 & -3 \\ 3 & 3 & 1 \end{bmatrix}$$

Steps:

- Find the eigenvalues of A. $\lambda_1 = 1$ $\lambda_2 = -2$, $\lambda_3 = -2$.
- Find three linearly indep. eigenvectors (if not possible, A not diagonalizable).
- Give (if possible) the diagonalization $\mathbf{A} = \mathbf{P}\mathbf{D}\mathbf{P}^{-1}$.

company to 2=-2, 23=-2. Fad eigenverten $\begin{bmatrix} \mathbf{A} - (-2)\mathbf{I} & \mathbf{0} \end{bmatrix} = \begin{bmatrix} 1+2 & 3 & 3 & 0 \\ -3 & -5+2 & -3 & 0 \\ -3 & 3 & 1+2 & 0 \end{bmatrix}$ $\begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} -1 \\ i \\ 0 \end{pmatrix} + \begin{pmatrix} -1 \\ i \\ 0 \end{pmatrix} = \begin{pmatrix} -1 \\ 0 \\ i \end{pmatrix}$ 2:1 linearly andypendet.

$$A = P D P^{-1} = \begin{bmatrix} 1 & -1 & -1 \\ -1 & 1 & 0 \\ 1 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & -2 & 0 \\ 0 & 0 & -2 \end{bmatrix} \begin{bmatrix} 1 & -1 & -1 \\ -1 & 1 & 0 \\ 1 & 0 & 1 \end{bmatrix}^{-1}$$

Lay, D. C. (2003). *Linear algebra and its applications. Third edition.* Pearson Education.

Monahan, J. F. (2008). A primer on linear models. CRC Press.

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