

Overview

- Determinants
 - geometric interpretation
 - computation rules
- Characteristic polynomial
 - repeated roots, algebraic multiplicity
- Eigenspaces
 - null space dimension, geometric multiplicity
- Eigendecomposition
 - possible if algebraic multiplicity equals geometric multiplicity for each eigenvalues
 - simple, meaning, an orthogonal or unitary decomposition for normal matrices
- Computing the SVD reduces to computing two eigenproblems

Definition. The determinant of a square matrix $\mathbf{A} = [\mathbf{a}_1 \ \dots \ \mathbf{a}_m] \in \mathbb{R}^{m \times m}$

$$\det(\mathbf{A}) = \begin{vmatrix} a_{11} & a_{12} & \dots & a_{1m} \\ a_{21} & a_{22} & \dots & a_{2m} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mm} \end{vmatrix} \in \mathbb{R}$$

is a real number giving the (oriented) volume of the parallelepiped spanned by matrix column vectors.

- $m = 2$

$$\mathbf{A} = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix}, \det(\mathbf{A}) = \begin{vmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{vmatrix}.$$

- $m = 3$

$$\mathbf{A} = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix}, \det(\mathbf{A}) = \begin{vmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{vmatrix}.$$



- Computation of a determinant with $m = 2$

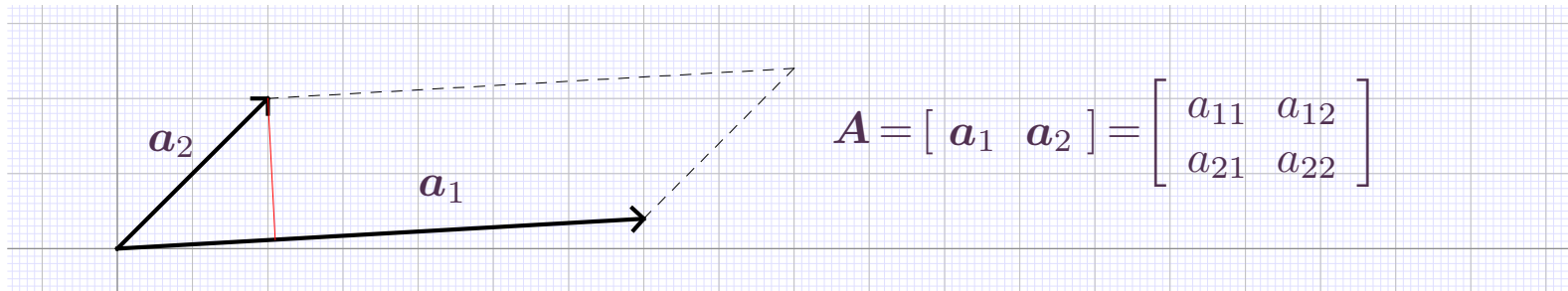
$$\begin{vmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{vmatrix} = a_{11}a_{22} - a_{12}a_{21}$$

- Computation of a determinant with $m = 3$

$$\begin{vmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{vmatrix} = a_{11}a_{22}a_{33} + a_{21}a_{32}a_{13} + a_{31}a_{12}a_{23} \\ - a_{13}a_{22}a_{31} - a_{23}a_{32}a_{11} - a_{33}a_{12}a_{21}$$

- Where do these determinant computation rules come from? Two viewpoints
 - *Geometric viewpoint*: determinants express parallelepiped volumes
 - *Algebraic viewpoint*: determinants are computed from all possible products that can be formed from choosing a factor from each row and each column

- $m = 2$



- In two dimensions a “parallelepiped” becomes a parallelogram with area given as

$$(\text{Area}) = (\text{Length of Base}) \times (\text{Length of Height})$$

- Take \mathbf{a}_1 as the base, with length $b = \|\mathbf{a}_1\|$. Vector \mathbf{a}_1 is at angle φ_1 to x_1 -axis, \mathbf{a}_2 is at angle φ_2 to x_2 -axis, and the angle between \mathbf{a}_1 , \mathbf{a}_2 is $\theta = \varphi_2 - \varphi_1$. The height has length

$$h = \|\mathbf{a}_2\| \sin \theta = \|\mathbf{a}_2\| \sin(\varphi_2 - \varphi_1) = \|\mathbf{a}_2\| (\sin \varphi_2 \cos \varphi_1 - \sin \varphi_1 \cos \varphi_2)$$

- Use $\cos \varphi_1 = a_{11} / \|\mathbf{a}_1\|$, $\sin \varphi_1 = a_{12} / \|\mathbf{a}_1\|$, $\cos \varphi_2 = a_{21} / \|\mathbf{a}_2\|$, $\sin \varphi_2 = a_{22} / \|\mathbf{a}_2\|$

$$(\text{Area}) = \|\mathbf{a}_1\| \|\mathbf{a}_2\| (\sin \varphi_2 \cos \varphi_1 - \sin \varphi_1 \cos \varphi_2) = a_{11}a_{22} - a_{12}a_{21}$$



- The geometric interpretation of a determinant as an oriented volume is useful in establishing rules for calculation with determinants:
 - Determinant of matrix with repeated columns is zero (since two edges of the parallelepiped are identical).
Example for $m = 3$

$$\Delta = \begin{vmatrix} a & a & u \\ b & b & v \\ c & c & w \end{vmatrix} = abw + bcu + cav - ubc - vca - wab = 0$$

This is more easily seen using the column notation

$$\Delta = \det(\mathbf{a}_1 \ \mathbf{a}_1 \ \mathbf{a}_3 \ \dots) = 0$$

- Determinant of matrix with linearly dependent columns is zero (since one edge lies in the 'hyperplane' formed by all the others)

- Separating sums in a column (similar for rows)

$$\det(\mathbf{a}_1 + \mathbf{b}_1 \quad \mathbf{a}_2 \quad \dots \quad \mathbf{a}_m) = \det(\mathbf{a}_1 \quad \mathbf{a}_2 \quad \dots \quad \mathbf{a}_m) + \det(\mathbf{b}_1 \quad \mathbf{a}_2 \quad \dots \quad \mathbf{a}_m)$$

with $\mathbf{a}_i, \mathbf{b}_1 \in \mathbb{R}^m$

- Scalar product in a column (similar for rows)

$$\det(\alpha \mathbf{a}_1 \quad \mathbf{a}_2 \quad \dots \quad \mathbf{a}_m) = \alpha \det(\mathbf{a}_1 \quad \mathbf{a}_2 \quad \dots \quad \mathbf{a}_m)$$

with $\alpha \in \mathbb{R}$

- Linear combinations of columns (similar for rows)

$$\det(\mathbf{a}_1 \quad \mathbf{a}_2 \quad \dots \quad \mathbf{a}_m) = \det(\mathbf{a}_1 \quad \alpha \mathbf{a}_1 + \mathbf{a}_2 \quad \dots \quad \mathbf{a}_m)$$

with $\alpha \in \mathbb{R}$.



- $\det(\mathbf{A}) = \det(\mathbf{A}^T)$
- $\det(\mathbf{A}\mathbf{B}) = \det(\mathbf{A}) \det(\mathbf{B})$
- $\mathbf{A} \in \mathbb{R}^{m \times m}$, if $\text{rank}(\mathbf{A}) < m$ then $\det(\mathbf{A}) = 0$

- For square matrix $\mathbf{A} \in \mathbb{R}^{m \times m}$ find *non-zero* vectors whose *directions* are not changed by multiplication by \mathbf{A} , $\mathbf{A}\mathbf{x} = \lambda\mathbf{x}$, λ is scalar, the *eigenvalue problem*.
- Consider the eigenproblem $\mathbf{A}\mathbf{x} = \lambda\mathbf{x}$ for $\mathbf{A} \in \mathbb{R}^{m \times m}$. Rewrite as

$$\mathbf{A}\mathbf{x} = \lambda\mathbf{x} \Rightarrow (\mathbf{A} - \lambda\mathbf{I})\mathbf{x} = \mathbf{0}.$$

Since $\mathbf{x} \neq \mathbf{0}$, a solution to eigenproblem exists only if $\mathbf{A} - \lambda\mathbf{I}$ is singular.

- $\mathbf{A} - \lambda\mathbf{I}$ singular implies $\det(\mathbf{A} - \lambda\mathbf{I}) = 0$.
- Investigate form of $\det(\mathbf{A} - \lambda\mathbf{I}) = 0$

$$\det(\mathbf{A} - \lambda\mathbf{I}) = \begin{vmatrix} a_{11} - \lambda & a_{12} & a_{13} & \dots & a_{1m} \\ a_{21} & a_{22} - \lambda & a_{23} & \dots & a_{2m} \\ a_{31} & a_{32} & a_{33} - \lambda & \dots & a_{3m} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & a_{m3} & \dots & a_{mm} - \lambda \end{vmatrix}$$

- $p_m(\lambda) = \det(\lambda\mathbf{I} - \mathbf{A})$, an m^{th} -degree polynomial in λ , *characteristic polynomial* of \mathbf{A} , with m roots, $\lambda_1, \lambda_2, \dots, \lambda_m$, the eigenvalues of \mathbf{A}

- $\mathbf{A} \in \mathbb{R}^{m \times m}$, eigenvalue problem $\mathbf{A}\mathbf{x} = \lambda\mathbf{x}$ ($\mathbf{x} \neq \mathbf{0}$) in matrix form:

$$\mathbf{A}\mathbf{X} = \mathbf{X}\mathbf{\Lambda}$$

$$\mathbf{X} = [\mathbf{x}_1 \ \dots \ \mathbf{x}_m], \mathbf{\Lambda} = \text{diag}(\lambda_1, \dots, \lambda_m) = \begin{bmatrix} \lambda_1 & 0 & \dots & 0 \\ 0 & \lambda_2 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & \lambda_m \end{bmatrix}.$$

- \mathbf{X} is the *eigenvector matrix*, $\mathbf{\Lambda}$ is the (diagonal) *eigenvalue matrix*
- If column vectors of \mathbf{X} are linearly independent, then \mathbf{X} is invertible

$$\mathbf{A} = \mathbf{X}\mathbf{\Lambda}\mathbf{X}^{-1},$$

the *eigendecomposition* of \mathbf{A} (compare to $\mathbf{A} = \mathbf{L}\mathbf{U}$, $\mathbf{A} = \mathbf{Q}\mathbf{R}$)

- Rule “determinant of product = product of determinants” implies

$$\det(\mathbf{A}\mathbf{X}) = \det(\mathbf{X}\mathbf{\Lambda}) \Rightarrow \det(\mathbf{A}) = \det(\mathbf{\Lambda}) \text{ (for } \det(\mathbf{X}) \neq 0\text{)}.$$

- Eigendecomposition of $I \in \mathbb{R}^{m \times m}$. Compare $AX = X\Lambda$

$$II = II, A = I, X = I, \Lambda = I$$

to find eigenvalues $\lambda_1 = 1, \dots, \lambda_m = 1$, eigenvectors $\mathbf{x}_1 = \mathbf{e}_1, \dots, \mathbf{x}_m = \mathbf{e}_m$.

- Eigendecomposition of $A = \text{diag}(s_1, s_2, \dots, s_m)$. Compare $AX = X\Lambda$

$$AI = IA$$

to find eigenvalues $\lambda_1 = s_1, \dots, \lambda_m = s_m$, eigenvectors $\mathbf{x}_1 = \mathbf{e}_1, \dots, \mathbf{x}_m = \mathbf{e}_m$.

- Reflection across x_1 -axis in \mathbb{R}^2

$$A = \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}$$

is a diagonal matrix, $\lambda_1 = 1, \lambda_2 = -1, \mathbf{x}_1 = \mathbf{e}_1, \mathbf{x}_2 = \mathbf{e}_2$



- Rotate by θ around x_3 axis in \mathbb{R}^3

$$\mathbf{A} = \begin{bmatrix} \cos \theta & -\sin \theta & 0 \\ \sin \theta & \cos \theta & 0 \\ 0 & 0 & 1 \end{bmatrix} \in \mathbb{R}^{3 \times 3}, m = 3$$

- One direction not change by rotation is $\mathbf{x}_3 = \mathbf{e}_3$ with $\lambda_3 = 1$
- Where are the other two directions?
 - Compute characteristic polynomial $p_3(\lambda) = \det(\lambda \mathbf{I} - \mathbf{A})$

$$p_3(\lambda) = \begin{vmatrix} \lambda - \cos \theta & \sin \theta & 0 \\ -\sin \theta & \lambda - \cos \theta & 0 \\ 0 & 0 & \lambda - 1 \end{vmatrix} = (\lambda - 1)(\lambda^2 - 2\lambda \cos \theta + 1)$$

- One root of $p_3(\lambda)$ is $\lambda_3 = 1$, as expected.
- Solve $\lambda^2 - 2\lambda \cos \theta + 1 = 0$ to find remaining eigenvalues to be *complex*

$$\lambda_{1,2} = \cos \theta \pm \sqrt{\cos^2 \theta - 1} = \cos \theta \pm i \sin \theta = e^{\pm i\theta} \in \mathbb{C}, i^2 = -1.$$



- $z \in \mathbb{C}$ can be represented in
 - Cartesian form $z = x + iy$
 - Polar form $z = re^{i\theta} = r(\cos \theta + i \sin \theta)$
- Complex conjugate of $z \in \mathbb{C}$ negates imaginary part $\bar{z} = x - iy = re^{-i\theta}$
- Absolute value of $z \in \mathbb{C}$ is $|z| = (x^2 + y^2)^{1/2} = r$
- Argument of z is angle θ from polar form $z = re^{i\theta}$
- Absolute value can be expressed as $|z| = (\bar{z}z)^{1/2}$
- Recall for $\mathbf{x} \in \mathbb{R}^m$

$$\|\mathbf{x}\|_2^2 = \mathbf{x}^T \mathbf{x} = x_1^2 + \dots + x_m^2,$$

stating that squared 2-norm of real vector \mathbf{x} is sum of squares of components.

- Extend above to vector of complex numbers $\mathbf{u} \in \mathbb{C}^m$ by

$$\|\mathbf{u}\|_2^2 = |u_1|^2 + \dots + |u_m|^2 = (\bar{\mathbf{u}})^T \mathbf{u}.$$

- Taking the complex conjugate and transposing arises frequently, notation

$$\mathbf{u}^* = (\bar{\mathbf{u}})^T, \text{ adjoint of } \mathbf{u}$$



- Consider $\lambda_2 = e^{i\theta} = \exp(i\theta) = \cos \theta + i \sin \theta$. Eigenvector \mathbf{x}_2 satisfies

$$(\mathbf{A} - \lambda_2 \mathbf{I})\mathbf{x}_2 = \mathbf{0},$$

which implies $\mathbf{x}_2 \in N(\mathbf{A} - \lambda_2 \mathbf{I})$

- Compute basis vector for $N(\mathbf{A} - \lambda_2 \mathbf{I})$

$$\mathbf{A} - \lambda_2 \mathbf{I} = \begin{bmatrix} -i \sin \theta & -\sin \theta & 0 \\ \sin \theta & -i \sin \theta & 0 \\ 0 & 0 & -e^{i\theta} \end{bmatrix} \sim \begin{bmatrix} -i \sin \theta & -\sin \theta & 0 \\ 0 & 0 & 0 \\ 0 & 0 & -e^{i\theta} \end{bmatrix}.$$

- Find eigenvector

$$\mathbf{x}_2 = \begin{bmatrix} i \\ 1 \\ 0 \end{bmatrix}$$

- Repeat for $\lambda_2 = e^{-i\theta}$, find $\mathbf{x}_3 = \begin{bmatrix} -i \\ 1 \\ 0 \end{bmatrix}$



- Compute $A\mathbf{x}_2 - \lambda_2\mathbf{x}_2 = (A - \lambda_2\mathbf{I})\mathbf{x}_2$

$$(A - \lambda_2\mathbf{I})\mathbf{x}_2 = \begin{bmatrix} -i \sin \theta & -\sin \theta & 0 \\ \sin \theta & -i \sin \theta & 0 \\ 0 & 0 & -e^{i\theta} \end{bmatrix} \begin{bmatrix} i \\ 1 \\ 0 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} = \mathbf{0}. \checkmark$$

- In general a polynomial of degree m $p_m(\lambda)$ with real coefficients has m complex roots $\lambda_1, \dots, \lambda_m \in \mathbb{C}$



- Consider

$$\mathbf{A} = \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix}$$

- Eigenvalues $\lambda_1 = \lambda_2 = 1$, a *repeated root*, since

$$\det(\lambda \mathbf{I} - \mathbf{A}) = \begin{vmatrix} \lambda - 1 & 1 \\ 0 & \lambda - 1 \end{vmatrix} = (\lambda - 1)^2.$$

- However

$$\mathbf{A} - \lambda_1 \mathbf{I} = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}$$

$\text{rank}(\mathbf{A} - \lambda_1 \mathbf{I}) = 1 = \dim C((\mathbf{A} - \lambda_1 \mathbf{I})^T)$, FT LA $\Rightarrow \dim N(\mathbf{A} - \lambda_1 \mathbf{I}) = 1$, only one non-zero eigenvector



Definition 1. The *algebraic multiplicity* of an eigenvalue λ is the number of times it appears as a repeated root of the characteristic polynomial $p(\lambda) = \det(A - \lambda I)$

Example. $p(\lambda) = \lambda(\lambda - 1)(\lambda - 2)^2$ has two single roots $\lambda_1 = 0$, $\lambda_2 = 1$ and a repeated root $\lambda_{3,4} = 2$. The eigenvalue $\lambda = 2$ has an algebraic multiplicity of 2

Definition 2. The *geometric multiplicity* of an eigenvalue λ is the dimension of the null space of $A - \lambda I$

Definition 3. An eigenvalue for which the geometric multiplicity is less than the algebraic multiplicity is said to be *defective*

Theorem. A matrix is diagonalizable if the geometric multiplicity of each eigenvalue is equal to the algebraic multiplicity of that eigenvalue.

- Find eigenvectors as non-trivial solutions of system $(\mathbf{A} - \lambda\mathbf{I})\mathbf{x} = 0$, e.g., $\lambda_1 = 1$

$$\mathbf{A} - \lambda_1\mathbf{I} = \begin{pmatrix} 4 & -4 & 2 \\ 5 & -5 & 1 \\ -2 & 2 & -4 \end{pmatrix} \sim \begin{pmatrix} -2 & 2 & -4 \\ 0 & 0 & -6 \\ 5 & -5 & 1 \end{pmatrix} \sim \begin{pmatrix} -2 & 2 & -4 \\ 0 & 0 & -6 \\ 0 & 0 & 0 \end{pmatrix}$$

Note convenient choice of row operations to reduce amount of arithmetic, and use of knowledge that $\mathbf{A} - \lambda_1\mathbf{I}$ is singular to deduce that last row must be null

- In traditional form the above row-echelon reduced system corresponds to

$$\begin{cases} -2x_1 + 2x_2 - 4x_3 = 0 \\ 0x_1 + 0x_2 - 6x_3 = 0 \\ 0x_1 + 0x_2 + 0x_3 = 0 \end{cases} \Rightarrow \mathbf{x} = \alpha \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}, \|\mathbf{x}\| = 1 \Rightarrow \alpha = 1/\sqrt{2}$$

- Suppose $\mathbf{A} \in \mathbb{R}^{m \times m}$ diagonalizable, $\mathbf{A} = \mathbf{X} \mathbf{\Lambda} \mathbf{X}^{-1}$
- Repeated application of \mathbf{A}

$$\mathbf{A}^2 = (\mathbf{X} \mathbf{\Lambda} \mathbf{X}^{-1})(\mathbf{X} \mathbf{\Lambda} \mathbf{X}^{-1}) = \mathbf{X} \mathbf{\Lambda}^2 \mathbf{X}^{-1}$$

$$\mathbf{A}^k = (\mathbf{X} \mathbf{\Lambda} \mathbf{X}^{-1}) \cdot \dots \cdot (\mathbf{X} \mathbf{\Lambda} \mathbf{X}^{-1}) = \mathbf{X} \mathbf{\Lambda}^k \mathbf{X}^{-1}$$

- Above allows definition of $e^{\mathbf{A}}$, $\sin(\mathbf{A})$, $\cos(\mathbf{A})$, for example

$$e^x = \frac{1}{0!} x^0 + \frac{1}{1!} x + \frac{1}{2!} x^2 + \dots + \frac{1}{k!} x^k + \dots \Rightarrow$$

$$e^{\mathbf{A}} = \mathbf{X} \left(\frac{1}{0!} \mathbf{\Lambda}^0 + \frac{1}{1!} \mathbf{\Lambda} + \frac{1}{2!} \mathbf{\Lambda}^2 + \dots + \frac{1}{k!} \mathbf{\Lambda}^k + \dots \right) \mathbf{X}^{-1}$$

- The differential system $\mathbf{y}' = \mathbf{A} \mathbf{y}$ has solution $\mathbf{y}(t) = e^{\mathbf{A}t} \mathbf{y}(0)$.