

# Numeric Linear Algebra

Eigenvalues and eigenvectors

# Matrix decompositions

1. Lower-upper(LU)
2. Cholesky
3. QR
4. Eigen decomposition (Schur)
5. Singular value decomposition(SVD)

# Eigenvalue Problem

Let  $\mathbf{A}$  is a real-valued square  $m \times m$  matrix.

$\lambda$  is an *eigenvalue* of  $\mathbf{A}$  and  $\mathbf{x} \neq \mathbf{0}$  is an *eigenvector* if

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

Equivalently,

$$(\mathbf{A} - \lambda\hat{\mathbf{1}})\mathbf{x} = \mathbf{0}$$

This (homogenous) system has a non-null solution iff

$$\det(\mathbf{A} - \lambda\hat{\mathbf{1}}) = 0$$

# Eigenvalue Problem

The l.h.s. of the *secular equation*

$$\det(\mathbf{A} - \lambda \hat{\mathbf{1}}) = 0$$

is a *characteristic polynomial* (C.P.) of  $\mathbf{A}$ .

- ▶ The eigenvalues are the roots of the C.P.
- ▶ C.P. has degree  $m$  and therefore has  $m$  roots
- ▶ For  $m > 4$  there are no explicit formulas for  $\lambda$  in terms of  $a_{ij}$   
 $\implies$  need numerics.

# Eigenvalue Problem

Formally, the eigenvalue problem is equivalent to the root-finding problem for the C.P.

In practice, the latter is poorly conditioned. E.g. compare the roots of

$$(x - 1)^{32} = 0 \quad \text{and} \quad (x - 1)^{32} = 10^{-16}$$

Need specialized methods.

## Detour: a companion matrix of a polynomial

Consider an  $n$ -by- $n$  matrix

$$\mathbf{A}_n = \begin{pmatrix} 0 & & & -a_0 \\ 1 & 0 & & -a_1 \\ 0 & 1 & \ddots & -a_2 \\ & \ddots & \ddots & 0 & \vdots \\ & & 0 & 1 & -a_{n-1} \end{pmatrix}$$

Its secular equation

$$0 = \det(x\mathbf{1} - \mathbf{A}_n) = x^n + a_{n-1}x^{n-1} + \cdots + a_1x + a_0$$

Root-finding  $\iff$  eigenvalue problem for  $\mathbf{A}_n$

# Numerical eigenvalue problem

Two classes of approaches

- ▶ All eigenvalues  
e.g. eigenmodes of a system
- ▶ A select subset  
e.g. PageRank, effective low-energy theory etc

## Localization of eigenvalues



# Localization of eigenvalues

Given a matrix, what is the distribution of  $\lambda_k$  in  $\mathbb{R}$ ?

Let  $r_k$  is the sum of the off-diag elements of the  $k$ -th row of  $\mathbf{A}$ :

$$r_k = \sum_{j=1, j \neq k}^m |a_{kj}|$$

*Gershgorin circle*: a circle with center  $a_{kk}$  & the radius  $r_k$ :

$$\mathcal{S}_k = \{z \in \mathbb{C} : |z - a_{kk}| \leq r_k\}$$

# Gerschgorin theorem

## Theorem: (Gerschgorin)

*All eigenvalues of  $\mathbf{A}$  belong to the union of  $\mathcal{S}_k, k = 1, \dots, m$ .*

## Proof

Consider an arbitrary eigenvalue  $\lambda$  of  $\mathbf{A}$ , and its eigenvector  $\vec{x}$ .

Let  $x_k$  is the max abs component of  $\vec{x} = (x_1, x_2, \dots, x_k, \dots, x_m)$ :

$$|x_k| \geq |x_j|, \quad j \neq k$$

# Gerschgorin theorem

Consider the linear system  $(\mathbf{A} - \lambda \hat{\mathbf{1}})\vec{x} = 0$ , write out its  $k$ -th equation:

$$(a_{kk} - \lambda)x_k = - \sum_{j \neq k} a_{kj}x_j$$

Then,

$$|a_{kk} - \lambda| \leq \frac{\left| \sum_{j \neq k} a_{kj}x_j \right|}{|x_k|} \leq \sum_{j \neq k} |a_{kj}| \cdot \underbrace{\left| \frac{x_j}{x_k} \right|}_{\leq 1} \leq r_k$$

by assumption

# Gerschgorin 2nd theorem

## Theorem: (Gerschgorin 2)

*If  $s$  Gerschgorin circles form a closed area  $\overline{G}$ , which is isolated from other Gerschgorin circles, then  $\overline{G}$  has exactly  $s$  eigenvalues (including multiplicity).*

Corollary: An isolated Gerschgorin circle contains exactly one eigenvalue.

