

- Dropping exam, 1 hw, 2 quizzes
- Exam 2 ready to schedule (online + in person)
- HW 8: short
- Recitation section: Mondays @ 2:30 in  
Loomis 151 ↪ -3:30
  - ↪ BYOP
  - ↪ replaces 1 office hour
  - ↪ see class calendar
- HW poll

## Goals:

- methods to compute one eigenvector  
↳ "power method" / "power iteration"  
 $A^{1,000} \cdot x_0$
- eigenvalues
- orthogonal iteration  $\rightarrow$  QR iteration
- Schur form / Schur factorization

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

$$A^3 = X^{-1} D X \cancel{X^{-1} D X} \cancel{X^{-1} D X} = X^{-1} D^3 X$$

$$y = A(A(A(A(Ax))))$$

# Power Iteration

## Demo: Motivating Power Iteration [cleared]

Assume  $|\lambda_1| > |\lambda_2| > \dots > |\lambda_n|$  with eigenvectors  $\mathbf{x}_1, \dots, \mathbf{x}_n$ .

Further assume  $\|\mathbf{x}_i\| = 1$ .

$$A \mathbf{x}_i = \lambda_i \mathbf{x}_i$$

$\vec{x}_0 =$  "something random"

$$\vec{x}_i = A \vec{x}_{i-1} / \|A \vec{x}_{i-1}\| \quad (i=1, 2, 3, \dots)$$

Assume  $\vec{x}_0 = \alpha \vec{v}_1 + \beta \vec{v}_2 + \dots$

$$\vec{y} = A^{1000} \vec{x}_0 = A^{1000} (\alpha \vec{v}_1 + \beta \vec{v}_2) = \alpha \lambda_1^{1000} \vec{v}_1 + \beta \lambda_2^{1000} \vec{v}_2$$

$$\begin{aligned} A \vec{v}_1 &= \lambda_1 \vec{v}_1 \\ A \vec{v}_2 &= \lambda_2 \vec{v}_2 \end{aligned}$$

$$\frac{\vec{y}}{\lambda_1^{1000}} = \alpha \vec{v}_1 + \beta \frac{\lambda_2^{1000}}{\lambda_1^{1000}} \vec{v}_2 = \alpha \vec{v}_1 + \rho \left( \frac{\lambda_2}{\lambda_1} \right)^{1000} \vec{v}_2$$

$$\vec{e}_i = \vec{x}_i - \vec{v}_1$$

$\uparrow$   
 $< 1$

$$\vec{x}_0 = \sum_j \alpha_j \vec{v}_j$$

Assume  $\|\vec{x}\|_4 \leq 1$

Assume  $\|\vec{v}_j\| = 1$

Moss,  
Sorry

$$\vec{e}_i = \underbrace{\vec{x}_i - \beta \vec{v}_1}_{\lambda_2} = \frac{A x_{i:1}}{\lambda_2} = \frac{A^i}{\lambda_2} x_0 - \beta v_1$$

$\beta$ : unknown  
scalar factor

$$= \sum_{j=1}^n \alpha_j \frac{\lambda_j}{\lambda_2} \vec{v}_j - \beta v_1$$

$$= \alpha_1 \vec{v}_1 - \beta v_1 + \sum_{j=2}^n \alpha_j \underbrace{\left(\frac{\lambda_j}{\lambda_2}\right)^i}_{< \left(\frac{\lambda_2}{\lambda_1}\right)} \vec{v}_j$$

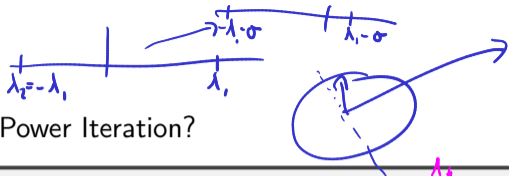
$$\|\vec{e}_i\| \leq \left(\frac{\lambda_2}{\lambda_1}\right)^i \|\vec{e}_0\| \approx 1$$

and  $\beta = \alpha_1$

$$\leq \left(\frac{\lambda_2}{\lambda_1}\right)^i$$

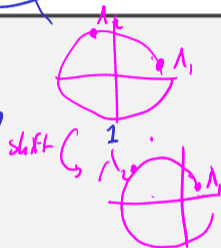
error

## Power Iteration: Issues?



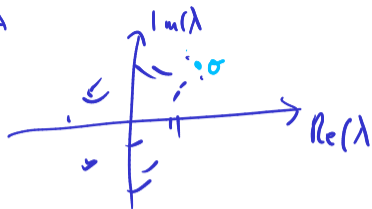
What could go wrong with Power Iteration?

- What if  $|\lambda_1| = |\lambda_2|$ ?
  - ↳ multiplicity
  - ↳ allow  $a$  to be complex!
- What about complex eigenvalues?
- $\vec{x}_0$  does not contain a component in  $\vec{v}_1$ ,



↳ roundoff error will give you one, with time.

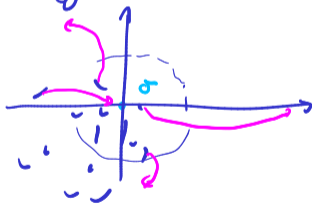
Eigenvalues of  $A$



Eigenvalues of  $A - \sigma I$



Eigenvalues of  $(A - \sigma I)^T$



## What about Eigenvalues?

 $\vec{x}$ 

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

$$\left\| \frac{A\vec{x}}{\vec{x}} \right\|$$

$$\left\| \frac{A\vec{x}}{\|\vec{x}\|} \right\|$$

Power Iteration generates eigenvectors. What if we would like to know eigenvalues?

$$\frac{\vec{x}^T A \vec{x}}{\vec{x}^T \vec{x}} = \lambda$$

"Rayleigh quotient"

→ Rayleigh quotient iteration



## Convergence of Power Iteration

What can you say about the convergence of the power method?

Say  $\mathbf{v}_1^{(k)}$  is the  $k$ th estimate of the eigenvector  $\mathbf{x}_1$ , and

$$e_k = \left\| \mathbf{x}_1 - \mathbf{v}_1^{(k)} \right\|.$$

