

Eigenvalues and eigenvectors approximation



approximation of some eigenvalues/eigenvectors

approximation of all the eigenvalues/eigenvectors

$$Ax = \lambda x$$

$$A \in \mathbb{R}^{n \times n}$$

$$\lambda \in \mathbb{C} \quad x \in \mathbb{C}^n$$

$$x \neq 0$$

$$|\lambda_1| > |\lambda_2| \geq |\lambda_3| \geq \dots \geq |\lambda_n|$$

$$\lambda_i \in \mathbb{R}$$

Power method $\Rightarrow \lambda_1, x_1$

$$Ax_1 = \lambda_1 x_1$$

$$v_0 \in \mathbb{R}^n$$

$$v_0 = \alpha_1 x_1 + \alpha_2 x_2 + \alpha_3 x_3 + \dots + \alpha_n x_n \quad \alpha_1 \neq 0$$

$$v_1 = Av_0 = \alpha_1 Ax_1 + \alpha_2 Ax_2 + \dots + \alpha_n Ax_n$$

x_1, \dots, x_n eigenvectors
corresponding to eigenvalues
 $\lambda_1, \dots, \lambda_n$

$$= \alpha_1 \lambda_1 x_1 + \alpha_2 \lambda_2 x_2 + \dots + \alpha_n \lambda_n x_n$$

$$= \lambda_1 \left(\alpha_1 x_1 + \alpha_2 \frac{\lambda_2}{\lambda_1} x_2 + \dots + \alpha_n \frac{\lambda_n}{\lambda_1} x_n \right)$$

$$v_2 = Av_1 = \lambda_1 \left(\alpha_1 \lambda_1 x_1 + \alpha_2 \frac{\lambda_2}{\lambda_1} \lambda_2 x_2 + \dots + \alpha_n \frac{\lambda_n}{\lambda_1} \lambda_n x_n \right)$$

$$= \lambda_1^2 \left(\alpha_1 x_1 + \alpha_2 \left(\frac{\lambda_2}{\lambda_1} \right)^2 x_2 + \dots + \alpha_n \left(\frac{\lambda_n}{\lambda_1} \right)^2 x_n \right)$$

$$v_k = Av_{k-1} = A^k v_0 = \lambda_1^k \left(\alpha_1 x_1 + \alpha_2 \left(\frac{\lambda_2}{\lambda_1} \right)^k x_2 + \dots + \alpha_n \left(\frac{\lambda_n}{\lambda_1} \right)^k x_n \right)$$

$$\frac{1}{\lambda_1^k} v_k = \alpha_1 x_1 + \alpha_2 \left(\frac{\lambda_2}{\lambda_1} \right)^k x_2 + \dots + \alpha_n \left(\frac{\lambda_n}{\lambda_1} \right)^k x_n$$

$$\lim_{k \rightarrow \infty} \frac{1}{\lambda_1^k} v_k = \lim_{k \rightarrow \infty} \alpha_1 x_1 + \lim_{k \rightarrow \infty} \alpha_2 \left(\frac{\lambda_2}{\lambda_1} \right)^k x_2 + \dots + \lim_{k \rightarrow \infty} \alpha_n \left(\frac{\lambda_n}{\lambda_1} \right)^k x_n$$

$$= \alpha_1 x_1$$

$$\lim_{k \rightarrow \infty} \left(\frac{\lambda_i}{\lambda_1} \right)^k = 0$$

$$|\lambda_i| < |\lambda_1|$$

$$\lim_{k \rightarrow \infty} \frac{\frac{1}{\lambda_1^{k+2}} v_{k+2}(i)}{\frac{1}{\lambda_1^k} v_k(i)} = \frac{\lim_{k \rightarrow \infty} \frac{1}{\lambda_1^{k+2}} v_{k+2}(i)}{\lim_{k \rightarrow \infty} \frac{1}{\lambda_1^k} v_k(i)} = \frac{\alpha_2 x_2(i)}{\alpha_2 x_2(i)} \quad i = 1, \dots, n$$

$$\lim_{k \rightarrow \infty} \frac{v_{k+2}(i)}{v_k(i)} = \lambda_1$$

Algorithm Power method (by components)

$$v_0 \neq 0$$

$$v_0 = \alpha_1 x_1 + \alpha_2 x_2 + \dots + \alpha_n x_n$$

$$\alpha_1 \neq 0$$

do

$$v_{k+1} = A v_k$$

$$\lambda_1^{(k+1)} = \frac{v_{k+1}(i)}{v_k(i)}$$

$$\text{while } (k \leq \text{MaxIter} \text{ \& \& } \frac{|\lambda_1^{(k+1)} - \lambda_1^{(k)}|}{|\lambda_1^{(k+1)}|} > \text{tol})$$

$$v_k = \lambda_1^k \alpha_1 x_1 + \lambda_2^k \left(\alpha_2 \left(\frac{\lambda_2}{\lambda_1} \right)^k x_2 + \dots \right)$$

$$|\lambda_1| > 1 \quad \lambda_1^k \alpha_1 \rightarrow \infty \quad k \rightarrow \infty$$

$$|\lambda_1| < 1 \quad \lambda_1^k \alpha_1 \rightarrow 0 \quad k \rightarrow \infty$$

$$v_0 \neq 0$$

$$\alpha_1 \neq 0$$

$$v_{old} = v_0 \quad k=0$$

$$\lambda_1^{new} = \infty$$

$$\frac{v_{old}}{\|v_{old}\|_\infty} = v_{old} / \|v_{old}\|_\infty$$

$$\lambda_1^{old} = \lambda_1^{new}$$

$$v_{new} = A v_{old}$$

$$\lambda_1^{new} = \frac{v_{new}(i)}{v_{old}(i)}$$

$$(\pm v_{new}(i))$$

$$v_{old} = v_{new}$$

$$[\|v_{old}\|_\infty, i] = \max(|v_{old}|)$$

$$v_{old} = \frac{v_{old}}{\|v_{old}\|_\infty}$$

$$k++$$

$$\text{while } (k < \text{MaxIter} \text{ \& \& } \frac{|\lambda_1^{new} - \lambda_1^{old}|}{|\lambda_1^{new}|} > \text{tol})$$

$$\mathcal{O}(n^2)$$

matrix vector product

$$\mathcal{O}(n)$$

max

$$\mathcal{O}(n)$$

normalization

$$\sim \mathcal{O}(n^2)$$

Rayleigh quotient

$$R_1^{(k)} = \frac{v_k^T A v_k}{v_k^T v_k} = \frac{v_k^T \boxed{v_{k+1}}}{v_k^T v_k} \rightarrow \lambda_1$$

$$v_k = \lambda_1^k \left(\alpha_1 x_1 + \alpha_2 \left(\frac{\lambda_2}{\lambda_1} \right)^k x_2 + \alpha_3 \left(\frac{\lambda_3}{\lambda_1} \right)^k x_3 \right) \quad \|x_i\|=1 \quad i=1, \dots, n \quad (3)$$

$$v_{k+1} = \lambda_1^{k+1} \left(\alpha_1 x_1 + \alpha_2 \left(\frac{\lambda_2}{\lambda_1} \right)^{k+1} x_2 + \alpha_3 \left(\frac{\lambda_3}{\lambda_1} \right)^{k+1} x_3 \right)$$

$$\begin{aligned} v_k^T v_k = \lambda_1^{2k} & \left(\alpha_1^2 x_1^T x_1 + \alpha_1 \alpha_2 \left(\frac{\lambda_2}{\lambda_1} \right)^k x_1^T x_2 + \alpha_1 \alpha_3 \left(\frac{\lambda_3}{\lambda_1} \right)^k x_1^T x_3 + \right. \\ & + \alpha_2 \alpha_2 \left(\frac{\lambda_2}{\lambda_1} \right)^k x_2^T x_1 + \alpha_2^2 \left(\frac{\lambda_2}{\lambda_1} \right)^{2k} x_2^T x_2 + \alpha_2 \alpha_3 \left(\frac{\lambda_2}{\lambda_1} \right)^k \left(\frac{\lambda_3}{\lambda_1} \right)^k x_2^T x_3 + \\ & \left. + \alpha_3 \alpha_1 \left(\frac{\lambda_3}{\lambda_1} \right)^k x_3^T x_1 + \alpha_3 \alpha_2 \left(\frac{\lambda_3}{\lambda_1} \right)^k \left(\frac{\lambda_2}{\lambda_1} \right)^k x_3^T x_2 + \alpha_3^2 \left(\frac{\lambda_3}{\lambda_1} \right)^{2k} x_3^T x_3 \right) \end{aligned}$$

$$\begin{aligned} v_k^T v_{k+1} = \lambda_1^{2k+1} & \left(\alpha_1^2 x_1^T x_1 + \alpha_1 \alpha_2 \left(\frac{\lambda_2}{\lambda_1} \right)^{k+1} x_1^T x_2 + \alpha_1 \alpha_3 \left(\frac{\lambda_3}{\lambda_1} \right)^{k+1} x_1^T x_3 + \right. \\ & + \alpha_2 \alpha_1 \left(\frac{\lambda_2}{\lambda_1} \right)^k x_2^T x_1 + \alpha_2^2 \left(\frac{\lambda_2}{\lambda_1} \right)^{2k+1} x_2^T x_2 + \alpha_2 \alpha_3 \left(\frac{\lambda_2}{\lambda_1} \right)^k \left(\frac{\lambda_3}{\lambda_1} \right)^{k+1} x_2^T x_3 + \\ & \left. + \alpha_3 \alpha_1 \left(\frac{\lambda_3}{\lambda_1} \right)^k x_3^T x_1 + \alpha_3 \alpha_2 \left(\frac{\lambda_3}{\lambda_1} \right)^k \left(\frac{\lambda_2}{\lambda_1} \right)^{k+1} x_3^T x_2 + \alpha_3^2 \left(\frac{\lambda_3}{\lambda_1} \right)^{2k+1} x_3^T x_3 \right) \end{aligned}$$

$$\frac{v_k^T v_{k+1}}{v_k^T v_k} = \frac{\lambda_1^{2k+1} \left(\alpha_1^2 x_1^T x_1 + \dots \right)}{\lambda_1^{2k} \left(\alpha_1^2 x_1^T x_1 + \dots \right)}$$

$$\lim_{k \rightarrow \infty} \frac{v_k^T v_{k+1}}{v_k^T v_k} = \lambda_1$$

$$1 > \left| \frac{\lambda_2}{\lambda_1} \right| \geq \left| \frac{\lambda_3}{\lambda_1} \right| \geq \left| \frac{\lambda_4}{\lambda_1} \right| > \dots \left| \frac{\lambda_n}{\lambda_1} \right|$$

$\left| \frac{\lambda_2}{\lambda_1} \right|$ is small the iterative methods converge more fast

↓
0

↓
0

↓
0

$$v_0 \neq 0 \quad \alpha_1 \neq 0 \quad v_{old} = \frac{v_0}{\|v_0\|_2} \quad (v_{old}^T v_{old} = \|v_{old}\|_2^2 = 1) \quad \lambda_1^{new} = \infty$$

do

$$\lambda_1^{old} = \lambda_1^{new}$$

$$v_{new} = A v_{old}$$

$$\lambda_1 = v_{old}^T v_{new}$$

$$v_{old} = \frac{v_{new}}{\|v_{new}\|_2}$$

$$\longrightarrow x_1^{(k)} = v_{old}$$

$$\frac{v_{old}^T A v_{old}}{v_{old}^T v_{old}} = \frac{v_{old}^T A v_{old}}{1}$$

$$\text{while } \left(\begin{array}{l} k \leq \text{MaxIter} \text{ \& } \\ \left| \frac{\lambda_1^{new} - \lambda_1^{old}}{\lambda_1^{new}} \right| > \text{tol} \end{array} \right)$$

$$\text{output } \begin{array}{l} \lambda_1^{new} \rightarrow \lambda_1 \\ v_{old} \rightarrow x_1 \end{array}$$

$$\mathcal{O}(n^2) \quad \text{matrix vector}$$

$$\mathcal{O}(n) \quad \lambda_1$$

$$\mathcal{O}(n) \quad \|v_{new}\|_2$$

$$\mathcal{O}(n) \quad v_{new} / \|v_{new}\|_2$$

$$\left. \begin{array}{l} \mathcal{O}(n^2) \\ \mathcal{O}(n) \\ \mathcal{O}(n) \\ \mathcal{O}(n) \end{array} \right\} \mathcal{O}(n^2)$$

Speed of convergence

$$|\lambda_1 - \lambda_1^k| \leq C_1 \left(\frac{\lambda_2}{\lambda_1} \right)^k$$

$$\|x_1 - x_1^k\| \leq C_2 \left(\frac{\lambda_2}{\lambda_1} \right)^k$$

A symmetric

$$x_i \perp x_j$$

$$x_i^T x_j = \delta_{ij}$$

λ_i real

$$v_k^T v_k = \lambda_1^2 \left(\alpha_1^2 + \alpha_2^2 \left(\frac{\lambda_2}{\lambda_1} \right)^{2k} + \alpha_3^2 \left(\frac{\lambda_3}{\lambda_1} \right)^{2k} \right)$$

$$v_k^T v_{k+2} = \lambda_1^{2k+2} \left(\alpha_1^2 + \alpha_2^2 \left(\frac{\lambda_2}{\lambda_1} \right)^{2k+2} + \alpha_3^2 \left(\frac{\lambda_3}{\lambda_1} \right)^{2k+2} \right)$$

$$\lim_{k \rightarrow \infty} \frac{v_k^T v_{k+2}}{v_k^T v_k} = \frac{\lim_{k \rightarrow \infty} \lambda_1^{2k+2} \left(\alpha_1^2 + \alpha_2^2 \left(\frac{\lambda_2}{\lambda_1} \right)^{2k+2} + \dots \right)}{\lim_{k \rightarrow \infty} \lambda_1^{2k} \left(\alpha_1^2 + \alpha_2^2 \left(\frac{\lambda_2}{\lambda_1} \right)^{2k} + \dots \right)} = \lambda_1$$

$$|\lambda_2 - \lambda_1^k| \leq C_1 \left(\frac{\lambda_2}{\lambda_1} \right)^{2k}$$

$$\|x_1 - x_1^k\| \leq C_2 \left(\frac{\lambda_2}{\lambda_1} \right)^{2k}$$

Inverse power method

$$|\lambda_1| \geq |\lambda_2| \geq \dots \geq |\lambda_{n-1}| > |\lambda_n|$$

x_n

↑

$$\boxed{Ax = \lambda x}$$

A invertible

$\lambda = 0$ is not an eigenvalue

$$A^{-1}Ax = \lambda A^{-1}x$$

$$\frac{1}{\lambda}x = A^{-1}x$$

A^{-1} has eigenvalues $\frac{1}{\lambda}$ and eigenvectors x

$$A^{-1}x = \sigma x$$

$$\sigma = \frac{1}{\lambda}$$

$$v_0 \neq 0$$

$$\alpha_n \neq 0$$

$$v^{\text{old}} = v_0 / \|v_0\|$$

$$\sigma^{\text{new}} = \infty$$

$$\underline{A = LU}$$

do

$$\sigma^{\text{old}} = \sigma^{\text{new}}$$

$$v^{\text{new}} = A^{-1}v^{\text{old}}$$

$$\uparrow A v^{\text{new}} = v^{\text{old}}$$

$$\sigma = v^{\text{old}T} v^{\text{new}}$$

$$v^{\text{new}} = v^{\text{new}} / \text{norm}(v^{\text{new}})$$

$$v^{\text{old}} = v^{\text{new}}$$

$k++$

$$\text{while } (k < \text{MaxIter} \text{ and } \frac{|\sigma^{\text{new}} - \sigma^{\text{old}}|}{|\sigma^{\text{new}}|} > \text{tol})$$

$$\lambda_n = \frac{1}{\sigma}$$

μ find the closest eigenvalue of A to μ

$$Ax = \lambda x$$

$$Ax - \mu x = \lambda x - \mu x$$

$$(A - \mu I)x = (\lambda - \mu)x$$

$$|\lambda_1| \geq |\lambda_2| \geq |\lambda_3| \geq \underbrace{|\lambda_4|}_{\mu} \geq |\lambda_5| \geq |\lambda_6|$$

λ_4

$$> |\lambda_4 - \mu| \quad A - \mu I$$

smallest eigenvalue of

$$\sigma = \frac{1}{\lambda - \mu}$$

Apply inverse power method to $A - \mu I$

$$v_0 \neq 0$$

$$\underline{\alpha_0 \neq 0}$$

$$v^{new} = v_0 / \|v_0\|_2^2$$

$$\sigma^{new} = \infty$$

$$A - \mu I = LU$$

do

$$\sigma^{old} = \sigma^{new}$$

$$v^{new} = (A - \mu I)^{-1} v^{old}$$

$$(A - \mu I) v^{new} = v^{old}$$

$$\sigma = v^{old} v^{new}$$

$$v^{new} = v^{new} / \|v^{new}\|$$

$$v^{old} = v^{new}$$

$k++$

$$\text{while } \left\{ \begin{array}{l} k \leq \text{MAXITER} \text{ and } \frac{|\sigma^{new} - \sigma^{old}|}{\|\sigma^{new}\|} > \text{tol} \end{array} \right\}$$

$$\sigma = \frac{1}{\lambda - \mu}$$

$$\lambda - \mu = \frac{1}{\sigma}$$

$$\lambda = \frac{1}{\sigma} + \mu$$

v^{new} : approx of corresp
eigenvector