

Power Method

Note Title

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$$\left\{ \begin{array}{l} Au_j = \lambda_j u_j, \quad j = 1, \dots, n \\ \{u_1, \dots, u_n\} : \text{basis for } \mathbb{C}^n \\ |\lambda_1| > |\lambda_2| \geq \dots \geq |\lambda_n| \\ z \neq \overline{0}, \quad z = \alpha_1 u_1 + \dots + \alpha_n u_n, \quad \alpha_1 \neq 0 \\ \lambda_1 > 0 \\ \frac{A^k z}{\|A^k z\|} \rightarrow \frac{\alpha_1 u_1}{\|\alpha_1 u_1\|} : \text{unit ev}^* \text{ assoc.} \end{array} \right.$$

Inverse Power Method

A has n linearly independent eigenvectors $A u_j = \lambda_j u_j$,
 $j = 1, \dots, n$

μ : approximation to λ_d .

$$|\mu - \lambda_d| < |\mu - \lambda_j|, j \neq d$$

$$A u_j = \lambda_j u_j, j=1, \dots, n$$

$$|\mu - \lambda_l| < |\mu - \lambda_j|, j \neq l$$

$$(A - \mu I) u_j = (\lambda_j - \mu) u_j$$

$$(A - \mu I)^{-1} u_j = \frac{1}{(\lambda_j - \mu)} u_j$$

The matrix $(A - \mu I)^{-1}$ has

$\frac{1}{\lambda_l - \mu}$ as the dominant eigenvalue.

Choose $z \neq \bar{0}$.

Define $z^{(0)} = \frac{z}{\|z\|}$,

$$z^{(k)} = \frac{(A - \mu I)^{-1} z^{(k-1)}}{\|(A - \mu I)^{-1} z^{(k-1)}\|}$$

Then $z^{(k)} \rightarrow \omega$, where

$$(A - \mu I)^{-1} \omega = \frac{1}{\lambda_1 - \mu} \omega$$

$$(A - \mu I)^{-1} w = \frac{1}{\lambda_1 - \mu} w$$

$$\Rightarrow (A - \mu I) w = (\lambda_1 - \mu) w$$

$$\Rightarrow A w = \lambda_1 w$$

w : eigenvector associated with
 λ_1 .

$$z^{(k)} = \frac{(A - \mu I)^{-1} z^{(k-1)}}{\|(A - \mu I)^{-1} z^{(k-1)}\|}$$

We need to find

$$(A - \mu I)^{-1} z^{(k-1)} = r^{(k-1)}$$

or equivalently solve

$$(A - \mu I) r^{(k-1)} = z^{(k-1)}$$

A invertible matrix.

λ, μ eigenvalues of A

$$|\lambda| \leq \|A\|, \frac{1}{|\mu|} \leq \|A^{-1}\|$$

$$\Rightarrow \frac{|\lambda|}{|\mu|} \leq \|A\| \|A^{-1}\|$$

$$\Rightarrow \frac{|\lambda_1|}{|\lambda_n|} \leq \|A\| \|A^{-1}\|$$

μ : approximation to λ_1 .

$$\| (A - \mu I)^{-1} \| \geq \frac{1}{|\mu - \lambda_1|}$$

$A - \mu I$ is ill-conditioned.

Recall :

$$A \alpha = b, \quad A(\alpha + \delta\alpha) = b + \delta b$$

$$\frac{\|\delta\alpha\|}{\|\alpha\|} \leq (\|A\| \|A^{-1}\|) \frac{\|\delta b\|}{\|b\|}$$

sensitive to perturbation

We need to solve

$$(A - \mu I) r = q$$

Instead we solve

$$(A - \mu I) \hat{r} = \hat{q}$$

Even though $\frac{\|q - \hat{q}\|}{\|q\|}$ is small,

$\frac{\|r - \hat{r}\|}{\|r\|}$ can be big.

$$q = c_1 u_1 + c_2 u_2 + \cdots + c_n u_n$$

$$\hat{q} = \hat{c}_1 u_1 + \hat{c}_2 u_2 + \cdots + \hat{c}_n u_n$$

$$r = (A - \mu I)^{-1} q = \frac{c_1}{\lambda_1 - \mu} u_1 + \cdots + \frac{c_l}{\lambda_l - \mu} u_l \\ \cdots + \frac{c_n}{\lambda_n - \mu} u_n$$

$$\hat{r} = (A - \mu I)^{-1} \hat{q} = \frac{\hat{c}_1}{\lambda_1 - \mu} u_1 + \cdots + \frac{\hat{c}_l}{\lambda_l - \mu} u_l \\ \cdots + \frac{c_n}{\lambda_n - \mu} u_n$$

$$r \simeq \frac{c_\ell}{\lambda_\ell - \mu} u_\ell, \quad \hat{r} \simeq \frac{\hat{c}_\ell}{\lambda_\ell - \mu} u_\ell$$

$$\|r - \hat{r}\| \simeq \frac{|c_\ell - \hat{c}_\ell|}{|\lambda_\ell - \mu|}, \quad \text{can be big}$$

We are only interested in
eigenvector, that is direction,
hence it is fine



q : approximate eigenvector

What should be chosen as
an approximate eigenvalue?

Rayleigh quotient: $\frac{q^* A q}{q^* q} = \eta$

$$\eta = \frac{q^* A q}{q^* q}$$

For $z \in \mathbb{C}$, consider

$$\|Aq - zq\|_2^2 = \|\underbrace{Aq - \eta q}_{\text{underbrace}} + \underbrace{(\eta - z)q}_{\text{underbrace}}\|_2^2$$

$$\langle Aq - \eta q, q \rangle = q^* A q - \eta q^* q = 0$$

$$\|Aq - zq\|_2^2 = \|Aq - \eta q\|_2^2 + \|(\eta - z)q\|_2^2$$

$$\eta = \frac{q^* A q}{q^* q}$$

For any $z \in \mathbb{C}$,

$$\|Aq - \eta q\|_2 \leq \|Aq - zq\|_2$$

q : approximate eigenvector

Choose η : approximate eigenvalue



QR decomposition

A : $n \times n$ real invertible matrix

Aim : To write

$$A = QR,$$

where

Q : orthogonal, $Q^t Q = Q Q^t = I$,

R : upper triangular

$$Q = [q_1 \ q_2 \ \cdots \ q_n]$$

q_j : jth column of Q

$$\begin{aligned} Q^t Q &= \begin{bmatrix} q_1^t \\ q_2^t \\ \vdots \\ q_n^t \end{bmatrix} [q_1 \ q_2 \ \cdots \ q_n] \\ &= [q_i^t \ q_j] \\ &= [\langle q_j, q_i \rangle] \\ &= I \end{aligned}$$

$$Q^T Q = I$$

$$\Rightarrow \langle q_j, q_i \rangle = q_i^T q_j = \begin{cases} 1, & i=j \\ 0, & i \neq j \end{cases}$$

The columns of Q are orthonormal.

$Q Q^T = I \Rightarrow$ The rows of Q are
Orthonormal

$A = Q R$, Q orthogonal,

R : upper triangular

$$[c_1 \ c_2 \ \dots \ c_n] = [q_1 \ q_2 \ \dots \ q_n] \begin{bmatrix} r_{11} & \dots & r_{1n} \\ 0 & r_{22} & \dots & r_{2n} \\ \vdots & & & \\ 0 & 0 & \dots & r_{nn} \end{bmatrix}$$

$$c_1 = r_{11} q_1$$

$$c_2 = r_{12} q_1 + r_{22} q_2$$

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