

Eigenvalue Problem

Recall: Convergence of iterative scheme:

$N\vec{x}^{k+1} = P\vec{x}^k + \vec{f}$ depends on the spectral radius $\rho(N^{-1}P)$.

\therefore Need: numerical method to compute eigenvalues.

Computation of Spectral radius

Goal: Find eigenvalues with largest magnitude \leftarrow Spectral radius

Two methods =

1. Power method
2. QR method

1. Power method

Let $A \in M_{n \times n}(\mathbb{C})$ with n eigenvalues $\lambda_1, \lambda_2, \dots, \lambda_n$ with eigenvectors $\vec{x}_1, \vec{x}_2, \dots, \vec{x}_n$.

Let $X = \begin{pmatrix} \frac{1}{|\vec{x}_1|} & \frac{1}{|\vec{x}_2|} & \dots & \frac{1}{|\vec{x}_n|} \\ | & | & & | \end{pmatrix} \in M_{n \times n}(\mathbb{C})$. Then, we know:

$$AX = X \begin{pmatrix} \lambda_1 & & & \\ & \lambda_2 & & \\ & & \dots & \\ & & & \lambda_n \end{pmatrix}.$$

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

We'll use Power method to compute $|\lambda_1|$.

Observation: Start with an initial vector $\vec{x}^{(0)}$.

Consider the iterative scheme: $\vec{x}^{(k+1)} = \frac{A \vec{x}^{(k)}}{\|A \vec{x}^{(k)}\|_\infty}$ for $k=0, 1, \dots$

Suppose A is diagonalizable. That's, we can assume $\vec{x}_1, \vec{x}_2, \dots, \vec{x}_n$ form a basis for \mathbb{C}^n .

Take $\vec{x}^{(0)} = a_1 \vec{x}_1 + a_2 \vec{x}_2 + \dots + a_n \vec{x}_n$ (assuming $a_1 \neq 0$)

Note that $A^k \vec{x}^{(0)} = a_1 \lambda_1^k \left[\vec{x}_1 + \sum_{j=2}^n \frac{a_j}{a_1} \left(\frac{\lambda_j}{\lambda_1} \right)^k \vec{x}_j \right]$.

$$\begin{aligned} \text{Hence, } \vec{x}^{(k)} &= \frac{A \vec{x}^{(k-1)}}{\|A \vec{x}^{(k-1)}\|_\infty} = \frac{A \left(A \vec{x}^{(k-2)} / \|A \vec{x}^{(k-2)}\|_\infty \right)}{\|A \left(A \vec{x}^{(k-2)} / \|A \vec{x}^{(k-2)}\|_\infty \right)\|_\infty} \\ &= \frac{A^2 \vec{x}^{(k-2)}}{\|A^2 \vec{x}^{(k-2)}\|_\infty} = \dots = \frac{A^k \vec{x}^{(0)}}{\|A^k \vec{x}^{(0)}\|_\infty} \end{aligned}$$

$$= \frac{a_1 \lambda_1^k [\vec{x}_1 + \sum_{j=2}^n \frac{a_j}{a_1} (\frac{\lambda_j}{\lambda_1})^k \vec{x}_j]}{\| a_1 \lambda_1^k [\vec{x}_1 + \sum_{j=2}^n \frac{a_j}{a_1} (\frac{\lambda_j}{\lambda_1})^k \vec{x}_j] \|_\infty} \approx \frac{a_1 \vec{x}_1 \lambda_1^k}{|a_1| \|\vec{x}_1\|_\infty |\lambda_1|^k}$$

Note: \vec{v} is an eigenvector associated to λ_1 .

In fact,

$$\| A \vec{x}^{(k)} \|_\infty \rightarrow \| A \left(\frac{a_1 \vec{x}_1}{|a_1| \|\vec{x}_1\|_\infty} \right) \|_\infty = \left\| \frac{a_1}{|a_1|} \lambda_1 \frac{\vec{x}_1}{\|\vec{x}_1\|_\infty} \right\|_\infty = |\lambda_1|$$

when k is big
 $(\because |\frac{\lambda_j}{\lambda_1}| < 1 \text{ for } j=2, \dots, n)$

$\frac{\lambda_1^k}{|\lambda_1|^k}$ can be removed
 under $\| \cdot \|_\infty$

Power method:

Step 1: Initialize $\vec{x}^{(0)}$ (such that $\vec{x}^{(0)} = a_1 \vec{x}_1 + \dots + a_n \vec{x}_n$
with $a_1 \neq 0$)

Step 2: Compute $\vec{x}^{(k+1)} = A \vec{x}^{(k)} / \|A \vec{x}^{(k)}\|_\infty$

Step 3: Compute $\lambda^{(k+1)} = \|A \vec{x}^{(k+1)}\|_\infty$

Then: $\lambda^{(k)} \rightarrow |\lambda_1|$ as $k \rightarrow \infty$

How about if A is NOT diagonalizable?

(Jordan Canonical Form)

Let $A = V J V^{-1}$, $J =$ Jordan Canonical Form

Since the dominant eigenvalue, λ_1 , has multiplicity 1, the first Jordan block of A must be a 1×1 matrix.

$$\therefore J = \begin{pmatrix} \lambda_1 & & & \\ & [J(\lambda_{i_2})] & & \\ & & \ddots & \\ & & & [J(\lambda_{i_k})] \end{pmatrix}$$

Recall:

$$J(\lambda_j) = \begin{pmatrix} \lambda_j & & & 0 \\ & \lambda_j & & \\ & & \ddots & \\ 0 & & & \lambda_j \end{pmatrix}$$

Assume $c_1 \neq 0$, then:

$$\vec{x}^{(k)} = \frac{A^k \vec{x}^{(0)}}{\|A^k \vec{x}^{(0)}\|_\infty} = \frac{(VJV^{-1})^k \vec{x}^{(0)}}{\|(VJV^{-1})^k \vec{x}^{(0)}\|_\infty}$$

$$= \frac{(VJ^k V^{-1}) (c_1 \vec{v}_1 + c_2 \vec{v}_2 + \dots + c_n \vec{v}_n)}{\|(VJ^k V^{-1}) (c_1 \vec{v}_1 + c_2 \vec{v}_2 + \dots + c_n \vec{v}_n)\|_\infty}$$

$$= \frac{VJ^k (c_1 \vec{e}_1 + c_2 \vec{e}_2 + \dots + c_n \vec{e}_n)}{\|VJ^k (c_1 \vec{e}_1 + c_2 \vec{e}_2 + \dots + c_n \vec{e}_n)\|_\infty}$$

$$\left(\because V^{-1} \vec{v}_j = \vec{e}_j = \begin{pmatrix} 0 \\ \vdots \\ 1 \\ \vdots \\ 0 \end{pmatrix} \right)$$

$$= \frac{c_1 \lambda_1^k (\vec{v}_1 + \frac{1}{c_1} V (\frac{1}{\lambda_1} J)^k (c_2 \vec{e}_2 + \dots + c_n \vec{e}_n))}{\|c_1 \lambda_1^k (\vec{v}_1 + \frac{1}{c_1} V (\frac{1}{\lambda_1} J)^k (c_2 \vec{e}_2 + \dots + c_n \vec{e}_n))\|_\infty} \quad \left(\text{Note: } J^k \vec{e}_1 = \lambda_1^k \vec{e}_1 \right)$$

$\Rightarrow VJ^k \vec{e}_1 = \lambda_1^k \vec{v}_1$

$$\text{Now, } \left(\frac{1}{\lambda_1} J\right)^k = \begin{pmatrix} \left[\frac{1}{\lambda_1} J(\lambda_{i_1})\right]^k & & \\ & \dots & \\ & & \left[\frac{1}{\lambda_1} J(\lambda_{i_k})\right]^k \end{pmatrix} \rightarrow \begin{pmatrix} 1 & & \\ & 0 & \\ & & \dots \\ & & & 0 \end{pmatrix} \text{ as } k \rightarrow \infty$$

\uparrow $\left|\frac{\lambda_{i_1}}{\lambda_1}\right| < 1$ \uparrow $\left|\frac{\lambda_{i_k}}{\lambda_1}\right| < 1$

$$\therefore \vec{x}^{(k)} \approx \frac{c_1 \lambda_1^k \vec{v}_1}{\|c_1 \lambda_1^k \vec{v}_1\|_\infty} \text{ when } k \text{ is large.}$$

$$\therefore \|A \vec{x}^{(k)}\|_\infty \rightarrow |\lambda_1| \text{ as } k \rightarrow \infty \text{ as before!}$$

Example: Consider: $A = \begin{pmatrix} 0 & 11 & -5 \\ -2 & 17 & -7 \\ -4 & 26 & 10 \end{pmatrix}$. Using power method with $\vec{x}^{(0)} = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$, find the spectral radius of A .

Solution: $\vec{x}^{(1)} = \frac{A\vec{x}^{(0)}}{\|A\vec{x}^{(0)}\|_\infty} = \frac{(6, 8, 12)^T}{\|(6, 8, 12)^T\|_\infty} = \begin{pmatrix} \frac{1}{2} \\ \frac{2}{3} \\ 1 \end{pmatrix}$.

$\vec{x}^{(2)} = \frac{A\vec{x}^{(1)}}{\|A\vec{x}^{(1)}\|_\infty} = \frac{(\frac{7}{3}, \frac{10}{3}, \frac{16}{3})^T}{\|(\frac{7}{3}, \frac{10}{3}, \frac{16}{3})^T\|_\infty} = \begin{pmatrix} \frac{7}{16} \\ \frac{5}{8} \\ 1 \end{pmatrix}$.

Compute $\|A\vec{x}^{(k)}\|_\infty$, we have:

k	6	8	10
$\ A\vec{x}^{(k)}\ _\infty$	4.02536	4.066270	4.001564

The dominant eigenvalue ≈ 4

Generalization of Power method

Consider an invertible matrix A . Suppose A has eigenvalues:

$$|\lambda_1| > |\lambda_2| > \dots > |\lambda_n| (> 0)$$

Consider A^{-1} (exist as all eigenvalues are non-zero). Then A^{-1} has eigenvalues:

$$\frac{1}{\lambda_1}, \frac{1}{\lambda_2}, \dots, \frac{1}{\lambda_n} \text{ with } \left| \frac{1}{\lambda_n} \right| > \left| \frac{1}{\lambda_{n-1}} \right| > \dots > \left| \frac{1}{\lambda_1} \right|$$

Extension: Apply Power's method on A^{-1} to obtain $\left| \frac{1}{\lambda_n} \right|$.

\therefore the minimal eigenvalue can be determined! (Inverse Power method)

Remark: Computing A^{-1} is difficult! We solve: $A\vec{y} = \vec{x}^{(n)}$ in each iteration to determine $A^{-1}\vec{x}^{(n)}$.

Finding A^{-1} is equivalent to solving:

$$A\vec{y} = \begin{pmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix}, A\vec{y} = \begin{pmatrix} 0 \\ 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix}, \dots, A\vec{y} = \begin{pmatrix} 0 \\ 0 \\ \vdots \\ 0 \\ 1 \end{pmatrix}.$$

$$A \underbrace{\begin{pmatrix} \frac{1}{v_1} & \frac{1}{v_2} & \dots & \frac{1}{v_n} \\ | & | & & | \\ 1 & 1 & & 1 \end{pmatrix}}_{A^{-1}} = \begin{pmatrix} 1 & 0 & \dots & 0 \\ 0 & 1 & & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & & 1 \end{pmatrix}$$

Algorithm: (Inverse Power method)

Step 1: Pick $\vec{x}^{(0)}$ with $\|\vec{x}^{(0)}\|_\infty = 1$

Step 2: For $k = 1, 2, \dots$, solve $A\vec{w} = \vec{x}^{(k-1)}$.

$$\text{Let } \vec{x}^{(k)} = \frac{\vec{w}}{\|\vec{w}\|_\infty}.$$

$$\text{Let } \rho_k = \|A\vec{x}^{(k)}\|_\infty.$$

Remark: Again, $\rho_k \rightarrow |\lambda_n|$ as $k \rightarrow \infty$. ($\vec{x}^{(k)} \approx$ eigenvector of eigenvalue λ_n)