

Trilogy on Computing Maximal Eigenpair

Mu-Fa Chen

(Beijing Normal University)

School of Math, Sichuan Univ

2017 年 7 月 6 日

0 Introduction

Theorem (O. Perron, 1907; G.Frobenius 1912)

For $A \geq 0$, irreducible, “Trace (A) > 0 ”,

$\exists!$ maximal eigenvalue $\rho(A) > 0$ with

unique left-eigenvector $u > 0$ and

right-eigenvector $g > 0$, up to constant.

$$uA = \lambda u, \quad Ag = \lambda g, \quad \lambda = \rho(A)$$

Example (L.K. Hua, 1984)

$$A = \frac{1}{100} \begin{pmatrix} 25 & 14 \\ 40 & 12 \end{pmatrix},$$
$$\rho(A) = (37 + \sqrt{2409})/200.$$

$$u = (5(13 + \sqrt{2409})/7, 20),$$
$$\approx 44.34397483$$

$$g = ((13 + \sqrt{2409})/4, 20)^*$$

Let $A \geq 0$ be irreducible and invertible.

Input-output method:

$$x_n = x_0 A^{-n}, \quad n \geq 1.$$

$x_n = (x_n^{(0)}, \dots, x_n^{(d)})$: products in n th year

Theorem (Hua's Fundamental Theorem, 1984)

- The optimal choice is $x_0 = u$, it has the **fastest grow**: $x_n = x_0 \rho(A)^{-n}$.
- Unless $A^{-1} \geq 0$, if $x_0 \neq u$, then **collapse**: $\exists n, j$ such that $x_n^{(j)} \leq 0$.

Importance of $u \approx (44.34397483, 20)$

Table Input and collapse time

| \mathbf{x}_0 [decimals] | Collapse time n |
|---------------------------|-------------------|
| (44, 20) | 3 |
| (44.344, 20) | 8 |
| (44.34397483, 20) | 13 |

Importance of $(\rho(A), u)$:
Need high precision, large system

Remarks

1) Need study right-eigenvector only:

$$uA = \lambda u \iff A^*u^* = \lambda u^*,$$

2) Diagonals are free:

$$(A + mI)g = \lambda g \iff Ag = (\lambda - m)g$$

The eigenvector is kept, eigenvalues are shifted

I Tridiagonal case. Example $Q =$

$$\begin{pmatrix}
 -1 & 1^2 & 0 & 0 & 0 & 0 & 0 & 0 \\
 1^2 & -5 & 2^2 & 0 & 0 & 0 & 0 & 0 \\
 0 & 2^2 & -13 & 3^2 & 0 & 0 & 0 & 0 \\
 0 & 0 & 3^2 & -25 & 4^2 & 0 & 0 & 0 \\
 0 & 0 & 0 & 4^2 & -41 & 5^2 & 0 & 0 \\
 0 & 0 & 0 & 0 & 5^2 & -61 & 6^2 & 0 \\
 0 & 0 & 0 & 0 & 0 & 6^2 & -85 & 7^2 \\
 0 & 0 & 0 & 0 & 0 & 0 & 7^2 & -113
 \end{pmatrix}$$

$$\rho(Q) \approx -0.525268, \quad \rho(\text{Infinite } Q) = -1/4$$

Power iteration/method, 1929

Given $\mathbf{v}_0 \in \mathbb{R}^{N+1}$, $\mathbf{v}_0 \not\perp \mathbf{g}$ with $\|\mathbf{v}_0\| = 1$,
define

$$\mathbf{v}_k = \frac{A\mathbf{v}_{k-1}}{\|A\mathbf{v}_{k-1}\|}, \quad z_k = \|A\mathbf{v}_k\|, \quad k \geq 1,$$

Then $\mathbf{v}_k \rightarrow \mathbf{g}$ and $z_k \rightarrow \rho(A)$ as $k \rightarrow \infty$.

$$\mathbf{v}_k = \frac{A^k \mathbf{v}_0}{\|A^k \mathbf{v}_0\|} \quad \text{“Power”}$$

Power iteration/method, 1929

$$\mathbf{g} \approx (55.878, 26.5271, 15.7059, \\ 9.97983, 6.43129, 4.0251, 2.2954, 1)^*$$

$$\tilde{\mathbf{v}}_0 = (1, 0.587624, 0.426178, \\ 0.329975, 0.260701, 0.204394, \\ 0.153593, 0.101142)^*,$$

$$\mathbf{v}_0 = \frac{\tilde{\mathbf{v}}_0}{\|\tilde{\mathbf{v}}_0\|}, \quad \|\mathbf{v}\| = \sum_k |\mathbf{v}_k|, \quad \ell^1\text{-norm}$$

Power iteration/method, 1929

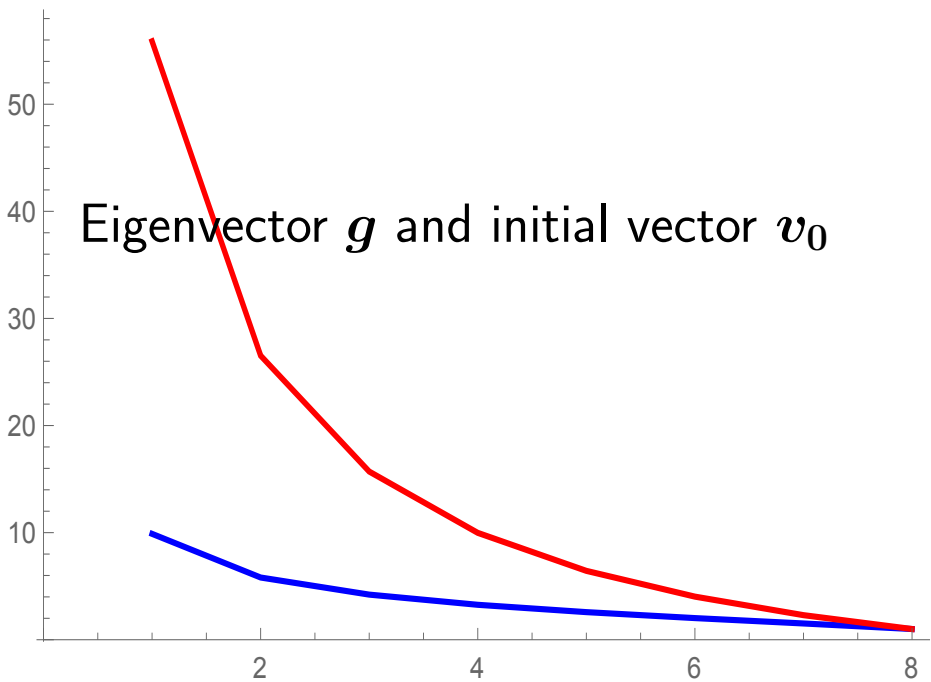
$$\mathbf{g} \approx (55.878, 26.5271, 15.7059, \\ 9.97983, 6.43129, 4.0251, 2.2954, 1)^*$$

$$\tilde{\mathbf{v}}_0 = (1, 0.587624, 0.426178, \\ 0.329975, 0.260701, 0.204394, \\ 0.153593, 0.101142)^*,$$

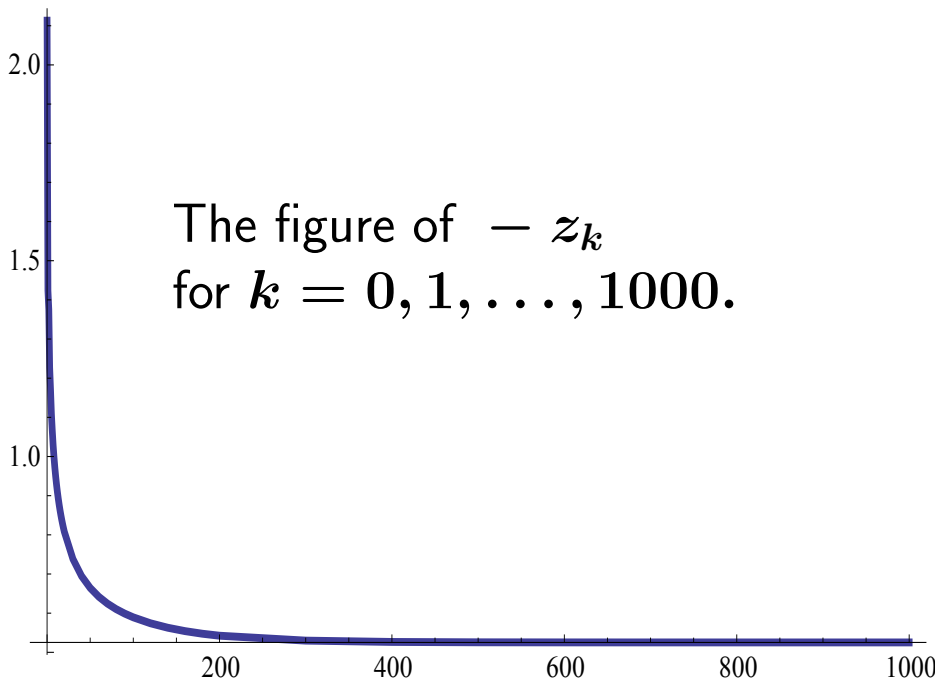
$$\mathbf{v}_0 = \frac{\tilde{\mathbf{v}}_0}{\|\tilde{\mathbf{v}}_0\|}, \quad \|\mathbf{v}\| = \sum_k |\mathbf{v}_k|, \quad \ell^1\text{-norm}$$

180 times, 10^3 iterations, 64 pages

Eigenvector g and initial vector v_0



The figure of $-z_k$
for $k = 0, 1, \dots, 1000$.



Convergence speed of Power Md

(λ_j, g_j) . Write $v_0 = \sum_{j=0}^N c_j g_j$, $c_0 \neq 0$

$$A^k v_0 = \sum_{j=0}^N c_j \lambda_j^k g_j$$

$$= c_0 \lambda_0^k \left[g_0 + \sum_{j=1}^N \frac{c_j}{c_0} \left(\frac{\lambda_j}{\lambda_0} \right)^k g_j \right]$$

$$\frac{A^k v_0}{\|A^k v_0\|} = \frac{c_0}{|c_0|} g_0 + O\left(\left| \frac{\lambda_1}{\lambda_0} \right|^k \right)$$

where $|\lambda_1| := \max\{|\lambda_j| : j > 0\}$.

(Shift) Inverse iteration, 1944

For given v_0 , at the k th step, define

$$v_k = \frac{A^{-k} v_0}{\|A^{-k} v_0\|}, \quad k \geq 1$$

i.e., the input-output method in economy

(Shift) Inverse iteration, 1944

For given v_0 , at the k th step, define

$$v_k = \frac{A^{-k}v_0}{\|A^{-k}v_0\|}, \quad k \geq 1$$

i.e., the input-output method in economy

\iff Double integral approx (C. 2001)

$$\text{Shift: } v_k = \frac{(zI - A)^{-k}v_0}{\|(zI - A)^{-k}v_0\|}, \quad k \geq 1$$

Rayleigh quotient iteration, 1944

Choose $(z_0, v_0) \approx (\rho(A), g)$ with $v_0^* v_0 = \mathbf{1}$, where v^* = transpose of v . L^2 -norm. Particular, $z_0 = v_0^* A v_0$ for given v_0 . At the k th step ($k \geq 1$), def

$$v_k = \frac{(z_{k-1} I - A)^{-1} v_{k-1}}{\| (z_{k-1} I - A)^{-1} v_{k-1} \|},$$
$$z_k = v_k^* A v_k \left[= \frac{v_k^* A v_k}{v_k^* v_k} \leftarrow \text{RQ} \right]$$

Rayleigh quotient iteration, 1944

where $I = \text{identity}$. Then

$$v_k \rightarrow g, \quad z_k \rightarrow \rho(A) \quad \text{as } k \rightarrow \infty$$

provided (z_0, v_0) is closed enough to $(\rho(A), g)$.

What can we expect for RQI?

100 iterations?

What can we expect for RQI?

100 iterations?

Example

The same matrix Q and \tilde{v}_0 , need 2 steps only:

$$z_1 \approx -0.528215, \quad z_2 \approx -0.525268.$$

\tilde{v}_0 efficient!

“Too Good” is dangerous. Pitfall

$$\lambda_j := \lambda_j(-Q). \quad \lambda_0 = -\rho(Q) > 0.$$

Example

Let Q be the same as above. Choose

$$v_0 = (1, 1, 1, 1, 1, 1, 1, 1)^* / \sqrt{8}.$$

Then

$$\lambda_2 \approx 5.91867$$

$$(z_1, z_2, z_3, z_4) \approx$$

$$(4.78557, 5.67061, 5.91766, 5.91867)$$

Google's PageRank 1998

Langville, A.N., Meyer, C. D. (2006).
Google's PageRank and Beyond:
The Science of Search Engine Rankings.
Princeton University Press.

Power Iteration, included.

Inverse Iteration & RQI, not touched!
Need large system, fast algorithm

- **Google's search**–PageRank
- **Input–output method** in economy.
Eigenvector
- **Stability speed** in stochastic systems.
Stationary distribution: eigenvector
Stability rate: eigenvalue
- **Principal component analysis**
–BigData, five-diagonal

- **Image recognition**, Poisson, Toeplitz, Block-tridiagonal matrices. PDE
- **Random algorithm**, MCMC
- **Phase transitions**. Next eigenpair

Estimation of $(g, \rho(A))$, central (hard) problem in many branches of math!

- **Image recognition**, Poisson, Toeplitz, Block-tridiagonal matrices. PDE
- **Random algorithm**, MCMC
- **Phase transitions**. Next eigenpair

Estimation of $(g, \rho(A))$, central (hard) problem in many branches of math!

For large N , guess: # of iterations $\sim N^\alpha$.
 $N = 10^4$? Subvert/overtun

Large N . $\lambda_0 = 1/4$ if $N = \infty$. ≤ 30 Sec

Use \tilde{v}_0 and δ_1 . Let $z_0 = 7/(8\delta_1) + v_0^*(-Q)v_0/8$.

| $N+1$ | z_0 | z_1 | $z_2 = \lambda_0$ |
|--------|---------|----------|-------------------|
| 10^4 | 0.31437 | 0.302586 | 0.302561 |

Examples for $N > 8$ computed by Yue-Shuang Li

Large N . $\lambda_0 = 1/4$ if $N = \infty$. ≤ 30 Sec

Use \tilde{v}_0 and δ_1 . Let $z_0 = 7/(8\delta_1) + v_0^*(-Q)v_0/8$

| $N+1$ | z_0 | z_1 | $z_2 = \lambda_0$ | upper/lower |
|--------|----------|----------|-------------------|----------------|
| 8 | 0.523309 | 0.525268 | 0.525268 | $1 + 10^{-11}$ |
| 100 | 0.387333 | 0.376393 | 0.376383 | $1 + 10^{-8}$ |
| 500 | 0.349147 | 0.338342 | 0.338329 | $1 + 10^{-7}$ |
| 1000 | 0.338027 | 0.327254 | 0.32724 | $1 + 10^{-7}$ |
| 5000 | 0.319895 | 0.30855 | 0.308529 | $1 + 10^{-7}$ |
| 7500 | 0.316529 | 0.304942 | 0.304918 | $1 + 10^{-7}$ |
| 10^4 | 0.31437 | 0.302586 | 0.302561 | $1 + 10^{-7}$ |

Examples for $N > 8$ computed by Yue-Shuang Li

I Efficient initials. Tridiagonal case

$$E = \{0, 1, \dots, N\}.$$

$$Q^c = A - mI, \quad m := \max_{i \in E} \sum_{j \in E} a_{ij}.$$

$$Q^c = (q_{ij}) =$$

$$\begin{pmatrix} -b_0 - c_0 & b_0 & 0 & 0 & \dots \\ a_1 & -a_1 - b_1 - c_1 & b_1 & 0 & \dots \\ 0 & a_2 & -a_2 - b_2 - c_2 & b_2 & \dots \\ \vdots & \vdots & \ddots & \ddots & \ddots \\ 0 & 0 & 0 & \dots & a_N & -a_N - c_N \end{pmatrix}$$

$$a_i > 0, \quad b_i > 0, \quad c_i \geq 0, \quad \neq 0.$$

$$q_{k,k+1} = b_k, \quad q_{k,k-1} = a_k, \quad q_{kk} = -a_k - b_k - c_k$$

Define $\{h_n\}$. Explicit

$$h_0 = 1, \quad h_n = h_{n-1} r_{n-1}, \quad 1 \leq n \leq N,$$

where $r_0 = 1 + c_0/b_0$,

$$r_n = 1 + \frac{a_n + c_n}{b_n} - \frac{a_n}{b_n r_{n-1}}, \quad 1 \leq n < N,$$

$$c_i \equiv 0 (i < N) \implies h_i \equiv 1.$$

$Q^c \setminus$ the last row $h = 0$.

$$\tilde{Q} = \text{Diag}(h_i)^{-1} Q^c \text{Diag}(h_i).$$

Define $\{\mu_n\}$ and $\{\varphi_n\}$.

Explicit

$$Q = \tilde{Q}$$

$$\begin{pmatrix} -b_0 & b_0 & 0 & 0 & \cdots \\ a_1 & -a_1 - b_1 & b_1 & 0 & \cdots \\ 0 & a_2 & -a_2 - b_2 & b_2 & \cdots \\ \vdots & \vdots & \ddots & \ddots & \ddots \\ 0 & 0 & 0 & \cdots & a_N & -a_N - b_N \end{pmatrix},$$

where $a_i, b_i > 0$. (c_i) erased!

$$\mu_0 = 1, \quad \mu_n = \mu_{n-1} b_{n-1} / a_n, \quad 1 \leq n \leq N.$$

$$\varphi_n = \sum_{k=n}^N (\mu_k b_k)^{-1}, \quad 0 \leq n \leq N.$$

RQI: tridiagonal case

$$\tilde{v}_0(i) = \sqrt{\varphi_i}, \quad i \leq N; \quad v_0 = \tilde{v}_0 / \|\tilde{v}_0\|_{\mu,2};$$

$$\delta_1 = \max_{0 \leq n \leq N} \left[\sqrt{\varphi_n} \sum_{k=0}^n \mu_k \sqrt{\varphi_k} + \frac{1}{\sqrt{\varphi_n}} \sum_{n+1 \leq j \leq N} \mu_j \varphi_j^{3/2} \right] =: z_0^{-1}.$$

Solve w_k : $(-Q - z_{k-1}I)w_k = v_{k-1}$,

$$v_k = \frac{w_k}{\|w_k\|_{\mu}} \rightarrow g, \quad z_k = (v_k, -Qv_k)_{\mu} \rightarrow \lambda_0$$

II Efficient initials. General case

Let $A = (a_{ij} : i, j \in E)$ be irreducible, $a_{ij} \geq 0$, $\forall i \neq j$.

Define $A_i = \sum_{j \in E} a_{ij}$ and

$$Q^c = A - \left(\max_{i \in E} A_i \right) I.$$

Let $h = (h_0, h_1, \dots, h_N)^*$ (with $h_0 = 1$) solve $Q^c \setminus \text{the last row } h = 0$ and define $\tilde{Q} = \text{Diag}(h_i)^{-1} Q^c \text{Diag}(h_i)$.

$$c_0 = \dots = c_{N-1} = 0, c_N =: q_{N, N+1} > 0$$

$Q := \tilde{Q}$. Let $\varphi = (\varphi_0, \varphi_1, \dots, \varphi_N)^*$
(with $\varphi_0 = 1$) solve

$\varphi \setminus \text{the first row} = P \setminus \text{the first row} \varphi$, where

$$P = \text{Diag}((-q_{ii})^{-1})Q + I.$$

$\mu := (\mu_0, \mu_1, \dots, \mu_N)$ with $\mu_0 = 1$
solves $Q^* \setminus \text{the last row} \mu^* = 0$.

Algorithm. Def $\tilde{v}_0(i) = \sqrt{\varphi_i}$, $i \leq N$;

$$v_0 = \tilde{v}_0 / \|\tilde{v}_0\|_\mu; \quad z_0 = (v_0, -Qv_0)_\mu.$$

Alternative z_0

For $k \geq 1$, let w_k solve

$$(-Q - z_{k-1}I)w_k = v_{k-1} \text{ and set}$$

$$v_k = w_k / \|w_k\|_\mu \rightarrow g,$$

$$z_k = (v_k, -Qv_k)_\mu \rightarrow \lambda_0 \text{ as } k \rightarrow \infty.$$

Alter choice of $z_0 = (v_0, -Qv_0)_\mu$:

$$z_0^{-1} = (1 - \varphi_1)^{-1} \max_{0 \leq n \leq N} \left[\sqrt{\varphi_n} \times \right.$$

$$\left. \sum_{k=0}^n \mu_k \sqrt{\varphi_k} + \frac{1}{\sqrt{\varphi_n}} \sum_{n+1 \leq j \leq N} \mu_j \varphi_j^{3/2} \right]$$

Example. Block-tridiagonal matrix

$$Q = \begin{pmatrix} A_0 & B_0 & 0 & 0 & \cdots \\ C_1 & A_1 & B_1 & 0 & \cdots \\ 0 & C_2 & A_2 & B_2 & \cdots \\ \vdots & \vdots & \ddots & \ddots & \ddots \\ 0 & 0 & 0 & C_N & A_N \end{pmatrix}$$

A_k, B_k, C_k :
 40×40
 B, C 's: Identity
 A 's: tridiagonal

| $N+1$ | z_0 | z_1 | $z_2 = \lambda_0$ |
|-------------|----------|----------|-------------------|
| 1600 | 7.985026 | 7.988219 | 7.988263 |
| 3600 | 7.993232 | 7.994676 | 7.994696 |
| 6400 | 7.996161 | 7.988256 | 7.987972 |

Example. Toeplitz Matrix [Xu Zhu]

$$A = \begin{pmatrix} 1 & \mathbf{2} & 3 & \cdots & n-1 & n \\ 2 & 1 & \mathbf{2} & \cdots & n-2 & n-1 \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ \mathbf{n-1} & n-2 & n-3 & \cdots & 1 & \mathbf{2} \\ n & \mathbf{n-1} & n-2 & \cdots & 2 & 1 \end{pmatrix}$$

| $N+1$ | $z_0 \times 10^6$ | $z_1 \times 10^6$ | $z_2 \times 10^6$ | $z_3 = \lambda_0$ |
|-------------|-------------------|-------------------|-------------------|-------------------|
| 1600 | 0.156992 | 0.451326 | 0.390252 | 0.389890 |
| 3600 | 0.157398 | 2.30731 | 1.97816 | 1.97591 |
| 6400 | 0.157450 | 7.32791 | 6.25506 | 6.24718 |

III Global algorithms

Tridiagonally dominant matrices.

More negative? complex?

Analytic: Block-tri, lower triangular +1,

Upper triangular + one line, ... AlphaGo

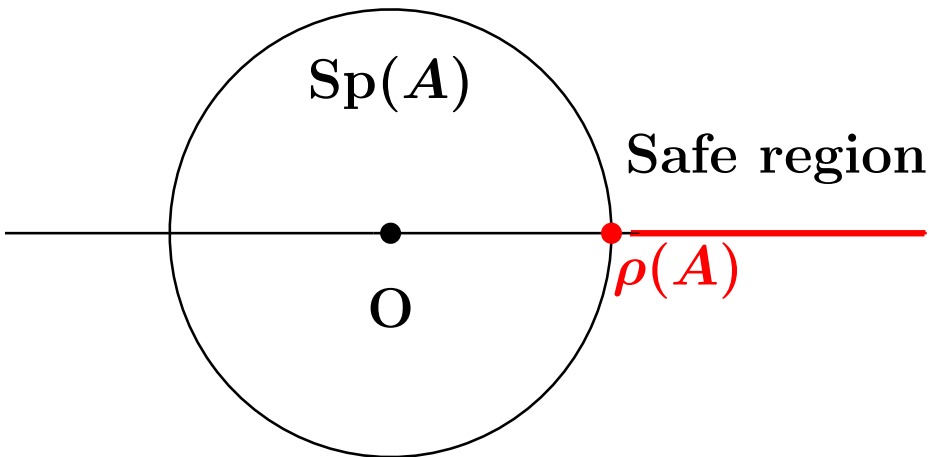
Ask computer to find efficient initials?

v_0 = uniform distr. $[(1, \dots, 1)^* / \sqrt{N+1}]$

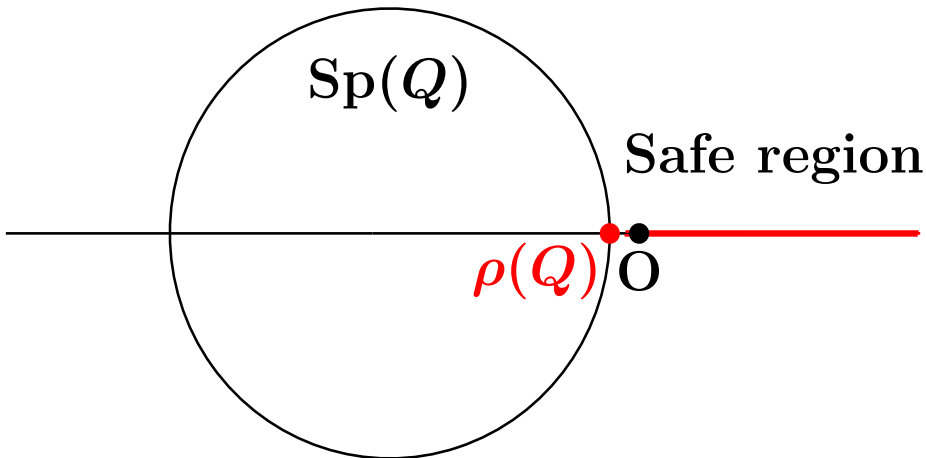
How to choose z_0 (z_n) in unified way?

$$z_0 \neq \frac{v_0^* A v_0}{v_0^* v_0} \text{ [RQ!]}, \quad z_n = ?$$

Complex plane



Complex plane



III Global algorithms: A1 & A2

$v_0 = \text{unif. distri.}, \quad z_0 = \max_{0 \leq i \leq N} \frac{Av_0}{v_0}(i).$

A1. Specific Rayleigh quotient iteration

(C. 2016: §4.1 with Choice I)

For $k \geq 1$, $v := v_{k-1}$, $z := z_{k-1}$,

let w solve equation $(zI - A)w = v$.

Set $v_k = w / \|w\|$, $z_k = v_k^* A v_k$.

A2. Shifted inverse iteration

$$z_k = \max_{0 \leq i \leq N} \frac{Av_k}{v_k}(i).$$

Lower triangular + upper-diagonal

$Q =$

$$\begin{pmatrix} -1 & \mathbf{1} & 0 & 0 & \dots & 0 & 0 \\ a_1 & -a_1 - 2 & \mathbf{2} & 0 & \dots & 0 & 0 \\ a_2 & 0 & -a_2 - 3 & \mathbf{3} & \dots & 0 & 0 \\ \vdots & \vdots & \vdots & \vdots & \dots & \mathbf{N-1} & 0 \\ a_{N-1} & 0 & 0 & 0 & \dots & -a_{N-1} - N & \mathbf{N} \\ a_N & 0 & 0 & 0 & \dots & 0 & -a_N - \mathbf{N-1} \end{pmatrix}$$

$$q_{k,k+1} = k + 1, \quad q_{k0} = 1/(k + 1).$$

We have computed 4 cases:

$$a_k = \mathbf{1/(k + 1)}, \quad \equiv 1, \quad = k, \quad = k^2.$$

The outputs for different N by Algorithm 2

| $N+1$ | z_1 | z_2 | z_3 |
|--------|-----------|----------|----------|
| 8 | 0.276727 | 0.427307 | 0.451902 |
| 16 | 0.222132 | 0.367827 | 0.399959 |
| 32 | 0.187826 | 0.329646 | 0.370364 |
| 50 | 0.171657 | 0.311197 | 0.357814 |
| 100 | 0.152106 | 0.287996 | 0.343847 |
| 500 | 0.121403 | 0.247450 | 0.321751 |
| 1000 | 0.111879 | 0.233257 | 0.313274 |
| 5000 | 0.0947429 | 0.205212 | 0.293025 |
| 10^4 | 0.0888963 | 0.194859 | 0.284064 |

The outputs for different N by Algorithm 2

| $N+1$ | z_4 | z_5 | z_6 |
|--------|----------|----------|----------|
| 8 | 0.452339 | | |
| 16 | 0.400910 | | |
| 32 | 0.372308 | 0.372311 | |
| 50 | 0.360776 | 0.360784 | |
| 100 | 0.349166 | 0.349197 | |
| 500 | 0.336811 | 0.337186 | |
| 1000 | 0.334155 | 0.335009 | 0.335010 |
| 5000 | 0.328961 | 0.332609 | 0.332635 |
| 10^4 | 0.326285 | 0.332113 | 0.332188 |

Upper triangular + lower-diagonal

$Q =$

$$\begin{pmatrix} -1 & p_2 & p_3 & p_4 & \cdots & p_{N-1} & \sum_{k \geq N} p_k \\ 2p_0 & -2 & 2p_2 & 2p_3 & \cdots & 2p_{N-2} & 2 \sum_{k \geq N-1} p_k \\ 0 & 3p_0 & -3 & 3p_2 & \cdots & 3p_{N-3} & 3 \sum_{k \geq N-2} p_k \\ \vdots & \vdots & \vdots & \ddots & \ddots & \vdots & \vdots \\ \vdots & \vdots & \vdots & \ddots & \ddots & -N+1 & (N-1) \sum_{k \geq 2} p_k \\ 0 & 0 & 0 & 0 & \cdots & Np_0 & -Np_0 \end{pmatrix}$$

$$p_0 = \alpha/2, \quad p_1 = 0, \quad p_n = (2 - \alpha)/2^n, \\ \alpha \in (4/3, 2). \quad \alpha = 7/4.$$

The outputs in the subcritical case

| N | z_1 | z_2 | z_3 | z_4 |
|------------------|----------|----------|----------|-----------------|
| 8 | 0.637800 | 0.638153 | | |
| 16 | 0.621430 | 0.625490 | 0.625539 | |
| 50 | 0.609976 | 0.624052 | 0.624997 | 0.625000 |
| 100 | 0.606948 | 0.623377 | 0.624991 | 0.625000 |
| 500 | 0.604409 | 0.622116 | 0.624962 | 0.625000 |
| 10^3 | 0.604082 | 0.621688 | 0.624944 | 0.625000 |
| $\frac{10^4}{2}$ | 0.603817 | 0.620838 | 0.62489 | 0.625000 |
| 10^4 | 0.603784 | 0.620511 | 0.624861 | 0.625000 |

Matrix with more “-” elements

$$A = \begin{pmatrix} -1 & 8 & -1 \\ 8 & 8 & 8 \\ -1 & 8 & 8 \end{pmatrix}.$$

Eigenvalues of A : **17.5124**, -7.4675 , 4.95513 .

| n | z_n : Algorithm 1 | z_n : Algorithm 2 |
|-----|---------------------|---------------------|
| 1 | 17.3772 | 18.5316 |
| 2 | 17.5124 | 17.5416 |
| 3 | | 17.5124 |

Complex matrix

 $A =$

$$\begin{pmatrix} 0.75 - 1.125i & 0.5882 - 0.1471i & 1.0735 + 1.4191i \\ -0.5 - i & 2.1765 + 0.7059i & 2.1471 - 0.4118i \\ 2.75 - 0.125i & 0.5882 - 0.1471i & -0.9265 + 0.4191i \end{pmatrix}$$


Eigenvalues of A : **3**, $-2 - i$, $1 + i$.

Outputs of “Algorithm 2”

| | |
|-------|---|
| y_1 | $3.03949 - 0.0451599i$ |
| y_2 | $3.00471 - 0.0015769i$ |
| y_3 | 3 Fast cubic |



$\operatorname{Re}(A^n) > 0 \quad \forall n \gg 1$ up to mI .

For Further Reading I

-  C. (2016): Efficient initials for computing the maximal eigenpair, Front. Math. China 11(6): 1379–1418.

A package based on the paper is available on CRAN now [by X.J. Mao]:
<https://cran.r-project.org/web/packages/EfficientMaxEigenpair/index.html>

For Further Reading II

-  C. (2017a). Global algorithms for maximal eigenpair. Preprint
-  C. (2017b). Efficient algorithm for principal eigenpair of discrete p -Laplacian. Preprint

<http://math0.bnu.edu.cn/~chenmf>

4 volumes, 19 popularizing papers, 8 videos

The end!

Thank you, everybody!

谢谢大家!

49