

Rayleigh quotient

- Assume A is real and symmetric. Thus A has real eigenvalues and a complete set of orthogonal eigenvectors.

$$\{\lambda_1, \dots, \lambda_n\}, \{q_1, \dots, q_n\} \quad ||q_j|| = 1$$

Def: The Rayleigh quotient of a vector x is:

$$r(x) \equiv \frac{x^T A x}{x^T x}$$

Notes

- If x is an eigenvector, then $r(x)$ is an eigenvalue.
- Given x , find α such that

$$\min_{\alpha} \left\| \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} \alpha - Ax \right\|_2 \quad (n \times 1 \text{ least squares})$$

The normal equations: $(x^T x) \alpha = x^T (Ax)$

$$\alpha = r(x)$$

- Theorem: Let q_j be an eigenvector and $x \approx q_j$. Then

$$r(x) - r(q_j) = O(||x - q_j||^2) \quad \text{as } x \rightarrow q_j$$

Power iteration

Let $v^{(0)}$ = approx. eigenvector, $\|v^{(0)}\| = 1$, and $\{q_i\}$ = set of eigenvectors.

Then
$$v^{(0)} = c_1 q_1 + c_2 q_2 + \dots + c_n q_n$$

$$A v^{(0)} = c_1 \lambda_1 q_1 + c_2 \lambda_2 q_2 + \dots + c_n \lambda_n q_n$$

Similarly,
$$A^k v^{(0)} = c_1 \lambda_1^k q_1 + c_2 \lambda_2^k q_2 + \dots + c_n \lambda_n^k q_n$$
$$= \lambda_1^k (c_1 q_1 + c_2 (\lambda_2/\lambda_1)^k q_2 + \dots + c_n (\lambda_n/\lambda_1)^k q_n)$$

Suppose $|\lambda_1| > |\lambda_2| \geq \dots \geq |\lambda_n|$. Then $|\lambda_i/\lambda_1|^k \rightarrow 0$ as $k \rightarrow \infty$.

$\therefore A^k v^{(0)} \sim c_1 \lambda_1^k q_1$ for large k

i.e.

$$q_1 \sim \frac{A^k v^{(0)}}{\|A^k v^{(0)}\|}$$

Example

$$A = \begin{bmatrix} 21 & 7 & -1 \\ 5 & 7 & 7 \\ 4 & -4 & 20 \end{bmatrix}, \quad v^{(0)} = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$$

$$w = A v^{(0)} = (27, 19, 20)^T$$

$$v^{(1)} = w / ||w|| = (0.70, 0.49, 0.52)^T$$

$$\lambda^{(1)} = r(v^{(1)}) = 23.3235$$

$$w = A v^{(1)} = (17.62, 10.57, 11.19)^T$$

$$v^{(2)} = w / ||w|| = (0.75, 0.45, 0.48)^T$$

$$\lambda^{(2)} = r(v^{(2)}) = 23.7250$$

$$w = A v^{(2)} = (18.50, 10.28, 10.71)^T$$

$$v^{(3)} = w / ||w|| = (0.78, 0.43, 0.45)^T$$

$$\lambda^{(3)} = r(v^{(3)}) = 23.8670$$

\vdots
 \vdots

$$q_1 = (0.8165, 0.4082, 0.4082)^T, \quad \lambda_1 = 24.$$

Algorithm

$\mathbf{v}^{(0)}$ = initial guess, $||\mathbf{v}^{(0)}|| = 1$

for $k = 1, 2, \dots$

$$\mathbf{w} = \mathbf{A} \mathbf{v}^{(k-1)}$$

$$\mathbf{v}^{(k)} = \mathbf{w} / ||\mathbf{w}||$$

$$\lambda^{(k)} = (\mathbf{v}^{(k)})^T \mathbf{A} \mathbf{v}^{(k)} \quad (\text{Rayleigh quotient})$$

end

Notes

- 1) We normalize $\mathbf{A} \mathbf{v}^{(k-1)}$ in each computation of $\mathbf{v}^{(k)}$.
- 2) Theorem: Suppose $|\lambda_1| > |\lambda_2| \geq |\lambda_3| \geq \dots \geq |\lambda_n|$, $\mathbf{q}_1^T \mathbf{v}^{(0)} \neq 0$.
Then

$$\left\| \mathbf{v}^{(k)} - (\pm \mathbf{q}_1) \right\| = O\left(\left|\frac{\lambda_2}{\lambda_1}\right|^k\right), \quad \left| \lambda^{(k)} - \lambda_1 \right| = O\left(\left|\frac{\lambda_2}{\lambda_1}\right|^{2k}\right)$$

as $k \rightarrow \infty$

- 3) It only computes \mathbf{q}_1 .
- 4) The convergence is linear, the convergence rate = $|\lambda_2|/|\lambda_1|$.
- 5) The convergence can be slow if $|\lambda_1| \sim |\lambda_2|$.

Inverse iteration

Idea 1: Use A^{-1} to compute the smallest eigenvalue.

(Note: $\Lambda(A^{-1}) = \{ 1/\lambda_1, 1/\lambda_2, \dots, 1/\lambda_n \}$.)

$$\text{Thus } v^{(0)} = c_1 q_1 + c_2 q_2 + \dots + c_n q_n$$

$$A^{-1} v^{(0)} = c_1 1/\lambda_1 q_1 + \dots + c_n 1/\lambda_n q_n$$

:

$$A^{-k} v^{(0)} = c_1 (1/\lambda_1)^k q_1 + \dots + c_n (1/\lambda_n)^k q_n$$

$$= (1/\lambda_n)^k [c_1 (\lambda_n/\lambda_1)^k q_1 + \dots + c_{n-1} (\lambda_n/\lambda_{n-1})^k q_{n-1} + c_n q_n]$$

$$\therefore A^{-k} v^{(0)} \sim c_n (1/\lambda_n)^k q_n \quad \text{for large } k$$

Idea 2: Shifting.

Consider $B = A - \mu I$, μ is not an eigenvalue of A . Then B has the same eigenvectors of A and its eigenvalues are $\{ \lambda_j - \mu \}$, $\lambda_j \in \Lambda(A)$.

If μ is close to λ_j , $\lambda_j - \mu$ would be the smallest eigenvalue of B .

We can apply idea 1 to compute $\lambda_j - \mu$.

Example

$$A = \begin{bmatrix} 21 & 7 & -1 \\ 5 & 7 & 7 \\ 4 & -4 & 20 \end{bmatrix}, \quad \Lambda(A) = \{8, 16, 24\}, \quad \mu = 15$$

$$\mathbf{v}^{(0)} = (1, 1, 1)^T$$

$$\mathbf{w} = (A - \mu I)^{-1} \mathbf{v}^{(0)} = (0.032, 0.16, 0.30)^T$$

$$\mathbf{v}^{(1)} = \mathbf{w} / \|\mathbf{w}\| = (0.093, 0.46, 0.88)^T$$

$$\lambda^{(1)} = r(\mathbf{v}^{(1)}) = 19.2000$$

$$\mathbf{w} = (A - \mu I)^{-1} \mathbf{v}^{(1)} = (-0.33, 0.40, 0.76)^T$$

$$\mathbf{v}^{(2)} = \mathbf{w} / \|\mathbf{w}\| = (-0.36, 0.44, 0.83)^T$$

$$\lambda^{(2)} = r(\mathbf{v}^{(2)}) = 15.9749$$

$$\mathbf{w} = (A - \mu I)^{-1} \mathbf{v}^{(2)} = (-0.39, 0.40, 0.79)^T$$

$$\mathbf{v}^{(3)} = \mathbf{w} / \|\mathbf{w}\| = (-0.40, 0.41, 0.82)^T$$

$$\lambda^{(3)} = r(\mathbf{v}^{(3)}) = 16.0290$$

\vdots
 \vdots

$$\mathbf{q}_2 = (-0.4082, 0.4082, 0.8165)^T, \quad \lambda_2 = 16.$$