

# THE POWER METHOD

Assume that

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

are the eigenvalues of  $A \in \mathbb{C}^{n \times n}$

The spectral gap between  $\lambda_1$  & the rest of the spectrum is essential for this method to work properly.

The matrix does not have to be non-defective, but for easy of presentation we'll assume this is the case.

Let  $\{q_1, q_2, \dots, q_m\}$  be the e'vectors associated with  $\lambda_1, \lambda_2, \dots, \lambda_m$ .

Hence  $(\neq) \quad Aq_i = \lambda_i q_i \quad 1 \leq i \leq m$

let  $x_k$  be a set of  $\mathbb{C}^m$  vectors  
where  $x_0$  is a known initial  
guess.

Write  $x_0 = \sum_{i=1}^m a_i q_i$ , then let

$$x_1 = Ax_0$$

$$x_2 = Ax_1$$

$$\vdots$$

$$x_k = A^k x_0$$

(we'll fold into the  $q_i$ 's the coefficients  
 $a_i$  : that is let  $\tilde{q}_i \equiv a_i q_i$   
from now on)

$$x_k = \lambda_1^k q_1 + \lambda_2^k q_2 + \dots + \lambda_m^k q_m$$

(using  $\neq$ ) .

Factoring  $\lambda_1^k$

$$x_k = \lambda_1^k \left[ q_1 + \left( \frac{\lambda_2}{\lambda_1} \right)^k q_2 + \dots + \left( \frac{\lambda_m}{\lambda_1} \right)^k q_m \right]$$

Since  $|\lambda_1| > |\lambda_i| \quad i=2, \dots, m$

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

So  $x_k$  will be "changing direction"  
and lining up with the direction  
of  $q_1$

For finite  $k$ :

$$(\star) \quad x_k = \lambda_1^k [q_1 + \varepsilon_k]$$

$$\text{where } e_k = G \left[ \left( \frac{\lambda_2}{\lambda_1} \right)^k \right] //$$

## Implementation

Write an iteration process ... however, we need to be careful to keep the iterates balanced in magnitude.

One way to do this is to take ratios ... of what? let's formally rewrite (★) as

$$\mathcal{O}(x_k) = \lambda_1^k [\mathcal{O}(q_1) + \mathcal{O}(e_k)]$$

A reasonable choice for  $\mathcal{O}$  is

a norm:

$$r_k = \frac{\mathcal{O}(x_{k+1})}{\mathcal{O}(x_k)} = \lambda_1 \left[ \frac{\mathcal{O}(q_1) + \mathcal{O}(e_k)}{\mathcal{O}(q_1) + \mathcal{O}(e_k)} \right]$$

so  $r_k \rightarrow \lambda_1$  as  $k \rightarrow \infty$

### ALGORITHM:

provide  $x_0$ , an initial guess

for  $k=1, 2, \dots$

$$y = Ax$$

$$r = \theta(y)/\theta(x)$$

$$x = y/\|y\|$$

output:  $r \sim \lambda_1$

$$x \sim q_1$$

Relative Error: since  $\lim_{k \rightarrow \infty} r_k = \lambda_1$

$$\text{then } \frac{r_k - \lambda_1}{\lambda_1} = \left( \frac{\lambda_2}{\lambda_1} \right)^k C_k$$

$C_k$  is some  $O(1)$  constant

which forms a bounded sequence.

## Inverse Power Method (Shifted Power Method)

Power Method can be used to find  
smallest eigenvalue

$$|\lambda_m^{-1}| > |\lambda_{m-1}^{-1}| \geq |\lambda_{m-2}^{-1}| \geq \dots \geq |\lambda_1^{-1}| > 0$$

We know that if  $\lambda$  is an e'value  
of  $A$  and  $A$  is nonsingular then  
 $\lambda^{-1}$  is an e'value of  $A^{-1}$

So power method could  
be formulated so that

$$x_k = (A^{-1})^k x_0$$

will then converge

to  $\lambda_m$ .

We don't know, or cannot safely  
compute  $A^{-1}$  so we introduce  
a **SHIFT**

The trick is to propose

$$\hat{A} = (A - \mu I) \text{ where } \mu \in \mathbb{C}$$

Then do power method on  
 $\hat{A}$ :

$$x_k = \hat{A}^k x_0.$$

So long as  $\mu$  is reasonable guess  
the iterative process will yield  
 $\lambda_m$  &  $q_m$

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