**Hermitian matrix** For  $A \in \mathbb{C}^{n \times n}$ , we define  $A^*$  to be its con-

jugate transpose. A is called Hermitian if

$$A = A^*.$$

## Properties of a Hermitian matrix

1. Its eigenvalues are all real.

2. If all of its entries are real, it is symmetric.

3. It is normal matrix. (i.e.  $AA^* = A^*A$ .) Therefore, it is diagonalizable, and its eigenvectors form orthogonal basis of  $\mathbb{C}^n$ . **Rayleigh Quotient** If A is Hermitian, we define

$$r(x) = \frac{x^* A x}{x^* x},$$

for all x in  $\mathbb{C}^n$ .

**Note** If x is an eigenvector of A, we have that  $r(x) = \dots$ 

**Restriction to Real Symmetric Matrices** In most application, we are interested in matrices with real entries. Therefore, we restrict ourselves to real symmetric matrices.

For real symmetric  $A \in \mathbb{R}^{n \times n}$ , the Rayleigh quotient is defined by

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

for  $x \in \mathbb{R}^n$ .

Note that if  $||x||_2 = 1$ , r(x) reduces to  $x^T A x$ .

**Stationary point** We can consider r as a real-valued function of multivariable. An eigenvector of A is a stationary point of r.

## **Power Iteration**

Choose  $u^{(0)}$  such that  $||u^{(0)}||_2 = 1$ . Then, we repeat the following steps.

1. 
$$w^{(k)} = Au^{(k-1)}$$
  
2.  $u^{(k)} = \frac{w^{(k)}}{\|w^{(k)}\|_2}$ 

**Theorem** Suppose  $|\lambda_1| > |\lambda_2| \ge \ldots \ge |\lambda_n|$  and  $v_1^T u^{(0)} \ne 0$ .

Then the power iteration satisfies

1. the difference between  $u^{(k)}$  and  $|v_1|$  is of order  $O\left(\left|\frac{\lambda_2}{\lambda_1}\right|^k\right)$ 

2. 
$$|\lambda^{(k)} - \lambda_1| = O\left(\left|\frac{\lambda_2}{\lambda_1}\right|^{2k}\right)$$
. Here, we define  $\lambda^{(k)} = (u^{(k)})^T A u^{(k)}$ .

Assume  $\mu$  is not an eigenvalue of A. We have that if  $\lambda$  is an eigenvalue of A, then  $(\lambda - \mu)^{-1}$  is an eigenvalue of  $(A - \mu I)^{-1}$ . If  $\lambda_i$  is closer to  $\mu$  than any other eigenvalues of A, we have that  $(\lambda_i - \mu)^{-1}$  is much lager than any other  $(\lambda - \mu)^{-1}$ .

If we apply the power iteration to the matrix  $(A - \mu I)^{-1}$ , the process will quickly converge to  $v_i$ , a unit eigenvector corresponding to  $\lambda_i$ . We call this...

## **Inverse Iteration**

Choose  $u^{(0)}$  such that  $||u^{(0)}||_2 = 1$ .

1. Solve for  $w^{(k)}$  from  $(A - \mu I)w^{(k)} = u^{(k-1)}$ .

2. 
$$u^{(k)} = \frac{w^{(k)}}{\|w^{(k)}\|_2}$$

**Rayleigh Quotient Iteration** We incorporate the Rayleigh quotient to the inverse iteration to approximate both eigenvector and eigenvalue of A.

Choose  $u^{(0)}$  such that  $||u^{(0)}||_2 = 1$ . Then, we have that  $\lambda^{(0)} = (u^{(0)})^T A u^{(0)}$ . Then, we repeat the following steps.

1. Solve for  $w^{(k)}$  from  $(A - \mu I)w^{(k)} = u^{(k-1)}$ .

2. 
$$u^{(k)} = \frac{w^{(k)}}{\|w^{(k)}\|_2}$$
  
3.  $\lambda^{(k)} = (u^{(k)})^T A u^{(k)}$ 

**Theorem** When  $u^{(0)}$  is chosen so that the Rayleigh quotient iteration converges, the convergence rate is cubic. i.e.

$$\|\lambda^{(k+1)} - \lambda_i\|_2 = O(|\lambda^{(k)} - \lambda_i|^3).$$