

Connected to pytorch (Python 3.9.18)

```
In [ ]: import warnings
warnings.filterwarnings("ignore")
import numpy as np
```

## 1.1 Power method

```
In [ ]: def power_method(A, v, max_iter=500, tol=1e-8):

    for _ in range(max_iter):
        v_next = A @ v
        entry_pos = np.argmax(np.abs(v_next))
        entry_max = v_next[entry_pos,0]
        v_next = v_next/entry_max
        diff_norm_v = np.linalg.norm(v_next-v)

        v = v_next.copy()

        if diff_norm_v <= tol:
            break

    return entry_max, v/np.linalg.norm(v)

# verification
n = 4
A = np.random.random([n,n])+ np.eye(n)
x = np.random.random([n,1]) + 1.j*np.random.random([n,1])
a, v = power_method(A, x)

eigvals, eigvecs = np.linalg.eig(A)

print(f'found a = {np.round(a,3)}')
print(f'\nfound v = \n', np.round(v.ravel(),3))
print(f'\nAv = \n', np.round((A@v).ravel(),3))
print(f'\nAv / v = (all = a) \n', np.round((A@v / v).ravel(),3))
```

```
print('\ncompare to numpy v ratio (unqiue value):\n', np.round(v.ravel()/eigvecs[:,0].ravel(),3))
```

found a = (3.003-0j)

found v =

```
[0.322-0.j 0.596-0.j 0.674-0.j 0.296-0.j]
```

Av =

```
[0.968-0.j 1.789-0.j 2.023-0.j 0.889-0.j]
```

Av / v = (all = a)

```
[3.003-0.j 3.003-0.j 3.003-0.j 3.003+0.j]
```

compare to numpy v ratio (unqiue value):

```
[1.-0.j 1.-0.j 1.-0.j 1.-0.j]
```

## 1.2 QR decomposition

```
In [ ]: def myQR(A):
    n = len(A)
    Q = A.copy()
    R = np.zeros_like(A)

    #i=0
    q = A[:,0]
    b = np.linalg.norm(q)

    R[0,0] = b
    Q[:,0] = q/b

    for i in range(1,n):

        q = Q[:,i:i+1]
        a = np.sum(q*q[:,i].conj(),0)
        q = q.ravel() - np.sum(Q[:,i]*a,1)
        q = q/np.linalg.norm(q)

        R[:,i] = a
        R[i,i] = q.conj() @ A[:,i]
```

```

        Q[:,i] = q

    return Q, R

n = 4
A = np.random.random([n,n]) + 1.j * np.random.random([n,n]) + np.diag(np.arange(n))

Q, R = myQR(A)
print(f'residual of A-QR = {np.sum(np.abs(A-Q@R)):.02e}')
print(f'residual of QQ*-I = {np.sum(np.abs(Q@Q.conj().T-np.eye(n)):.02e}')
print(f'R is upper triangular = {np.all(np.triu(R)==R)}')

```

```

residual of A-QR = 1.76e-15
residual of QQ*-I = 1.32e-15
R is upper triangular = True

```

## 1.3 Permutation function

```

In [ ]: def permutation(v):
        nn = len(v.ravel())
        idxOrder = np.argsort(np.abs(v))
        M = np.zeros((nn, nn))
        for i in range(nn):
            M[i, idxOrder[nn-1-i]] = 1
        return M

v = np.arange(10)
np.random.shuffle(v)
P = permutation(v)
print('random v :      ', v)
print('Pv =          :      ', P@v)

```

```

random v :      [9 3 2 0 4 5 7 6 1 8]
Pv =          : [9. 8. 7. 6. 5. 4. 3. 2. 1. 0.]

```

## 1.4 QR algorithm

```

In [ ]: def QR_algo(A, tol=1e-4):
    diff = 1
    Ak = A.copy()
    nn = A.shape[0]
    V = np.eye(nn)
    i = 0

    while diff > tol:
        i+=1
        shift = complex(np.random.rand(1), np.random.rand(1))/(n+1) * np.eye(nn)
        A_tmp = Ak.copy()
        # Q,R = np.linalg.qr(Ak - shift)
        Q,R = myQR(Ak - shift)
        Ak = R@Q + shift
        Perm = permutation(np.diag(Ak))
        Ak = Perm.T@Ak@Perm
        V = V@Q@Perm
        diff = np.sum(np.abs(np.sort(np.diag(Ak))-np.sort(np.diag(A_tmp))))
    return Ak, V

n = 4
A = np.random.random([n,n])+1.j*np.random.random([n,n]) + np.diag(np.arange(n))
A = A+A.T.conj()

Ak,V = QR_algo(A)
print('Ak = \n', np.round(Ak,2))
print(f'\nresidual of VV*-I = {np.sum(np.abs(V@V.conj().T-np.eye(n))):.02e}')
print(f'\nresidual of V*DV-A = {np.sum(np.abs(V@Ak@V.conj().T - A)):.02e}' )
print(f'\nv is first col of V, Av / v = {np.round(A@V[:,0] / V[:,0],2)}')

```

```

Ak =
[[ 8.25-0.j  0.  +0.j  0.  -0.j  0.  -0.j]
 [ 0.  -0.j  4.54+0.j -0.  -0.j  0.  -0.j]
 [ 0.  +0.j -0.  +0.j  2.44-0.j -0.  +0.j]
 [ 0.  +0.j  0.  +0.j -0.  -0.j  0.74-0.j]]

```

residual of  $VV^*-I = 6.56e-15$

residual of  $V*DV-A = 2.89e-14$

$v$  is first col of  $V$ ,  $Av / v = [8.25+0.j \ 8.25+0.j \ 8.24+0.j \ 8.25-0.j]$

## 2. Page Rank

```

In [ ]: A = np.array([[0, 0, 1, 0, 0],
                      [0, 0, 0, 1/2, 1/2],
                      [1, 0, 0, 0, 0],
                      [0, 1/2, 0, 0, 1/2],
                      [0, 1/2, 0, 1/2, 0]], dtype='float')

```

### 2.1 10 trials with different initial guesses

**Observation :** Power Method outputs different outputs each time (reached max\_iter). We did NOT get eigen vectors.

```

In [ ]: for i in range(10):
        a, v = power_method(A, np.random.random([5,1]) + 1.j*np.random.random([5,1]), max_iter=int(1e5))
        print(f'\n{i+1}th trial:\na = {np.round(a,2)}, v = {np.round(v.ravel(),2)}')
        print(f'(A * v) / v = {np.round((A@v).ravel()/v.ravel(),2)}')

```

1th trial:

$$a = (1+0j), v = [0.67+0.j \quad 0.3 +0.14j \quad 0.47+0.02j \quad 0.3 +0.14j \quad 0.3 +0.14j] \\ (A * v) / v = [0.7 +0.03j \quad 1. \quad +0.j \quad 1.42-0.05j \quad 1. \quad +0.j \quad 1. \quad +0.j \quad ]$$

2th trial:

$$a = (1+0j), v = [0.49+0.j \quad 0.39+0.18j \quad 0.23+0.4j \quad 0.39+0.18j \quad 0.39+0.18j] \\ (A * v) / v = [0.47+0.81j \quad 1. \quad +0.j \quad 0.53-0.92j \quad 1. \quad +0.j \quad 1. \quad +0.j \quad ]$$

3th trial:

$$a = (1+0j), v = [0.17-0.08j \quad 0.35-0.09j \quad 0.76+0.j \quad 0.35-0.09j \quad 0.35-0.09j] \\ (A * v) / v = [3.64+1.81j \quad 1. \quad +0.j \quad 0.22-0.11j \quad 1. \quad -0.j \quad 1. \quad +0.j \quad ]$$

4th trial:

$$a = (1+0j), v = [0.28+0.17j \quad 0.48+0.j \quad 0.39-0.2j \quad 0.48+0.j \quad 0.48+0.j \quad ] \\ (A * v) / v = [0.68-1.13j \quad 1. \quad +0.j \quad 0.39+0.65j \quad 1. \quad +0.j \quad 1. \quad +0.j \quad ]$$

5th trial:

$$a = (1+0j), v = [0.3 -0.32j \quad 0.46+0.j \quad 0.39+0.15j \quad 0.46+0.j \quad 0.46+0.j \quad ] \\ (A * v) / v = [0.37+0.9j \quad 1. \quad +0.j \quad 0.39-0.95j \quad 1. \quad +0.j \quad 1. \quad +0.j \quad ]$$

6th trial:

$$a = (1+0j), v = [0.42+0.14j \quad 0.35-0.23j \quad 0.52+0.j \quad 0.35-0.23j \quad 0.35-0.23j] \\ (A * v) / v = [1.1 -0.37j \quad 1. \quad +0.j \quad 0.82+0.28j \quad 1. \quad +0.j \quad 1. \quad -0.j \quad ]$$

7th trial:

$$a = (1+0j), v = [0.42+0.19j \quad 0.29-0.21j \quad 0.65+0.j \quad 0.29-0.21j \quad 0.29-0.21j] \\ (A * v) / v = [1.29-0.59j \quad 1. \quad +0.j \quad 0.64+0.29j \quad 1. \quad +0.j \quad 1. \quad +0.j \quad ]$$

8th trial:

$$a = (1+0j), v = [0.15-0.42j \quad 0.32-0.23j \quad 0.58+0.j \quad 0.32-0.23j \quad 0.32-0.23j] \\ (A * v) / v = [0.43+1.23j \quad 1. \quad +0.j \quad 0.25-0.73j \quad 1. \quad +0.j \quad 1. \quad +0.j \quad ]$$

9th trial:

$$a = (1+0j), v = [0.12-0.19j \quad 0.44-0.19j \quad 0.51+0.j \quad 0.44-0.19j \quad 0.44-0.19j] \\ (A * v) / v = [1.16+1.94j \quad 1. \quad +0.j \quad 0.23-0.38j \quad 1. \quad +0.j \quad 1. \quad +0.j \quad ]$$

10th trial:

$$a = (1+0j), v = [0.54+0.j \quad 0.45-0.11j \quad 0.26+0.08j \quad 0.45-0.11j \quad 0.45-0.11j] \\ (A * v) / v = [0.49+0.15j \quad 1. \quad +0.j \quad 1.85-0.57j \quad 1. \quad +0.j \quad 1. \quad +0.j \quad ]$$

## 2.2 Given initial guess

**Observation :** Both time we converged to some vector which is close to the initial guess and the outputs are indeed eigenvectors associated to eigenvalue 1.

```
In [ ]: x1 = np.array([0. , 0.57, 0. , 0.57, 0.57])[:,None]
x2 = np.array([0.7, 0., 0.7, 0., 0.])[:,None]

for i in [1,2]:
    print(f'\ninitial guess = x{i} : \n')
    exec(f"a, v = power_method(A, x{i}, max_iter=int(1e10))")
    print(f'a = {np.round(a,3)}, v = {np.round(v.ravel(),3)}')
    print(f'A * v = {np.round((A@v).ravel(),3)}')
    # print(f'(A * v) / v = {np.round((A@v).ravel()/v.ravel(),3)}')
```

initial guess = x1 :

```
a = 1.0, v = [0.    0.577 0.    0.577 0.577]
A * v = [0.    0.577 0.    0.577 0.577]
```

initial guess = x2 :

```
a = 1.0, v = [0.707 0.    0.707 0.    0.    ]
A * v = [0.707 0.    0.707 0.    0.    ]
```

## 2.3 Why?

**Discussion :** The Power Method is unstable with this problem because A's principal eigenspace is of dimension 2, which violates the hypothesis which guarantees the convergence of Power Method.

## 2.4 Perturbed PageRank

**Answer :** With a small lambda, we found a vector which is in the 1-eigenspace of A.

```
In [ ]: alpha = 1/20
Aprime = (1-alpha)*A + alpha * np.ones([5,5])/5
```

```
a, v = power_method(Aprime, np.random.random([5,1])+1.j*np.random.random([5,1]), max_iter=int(1e15))
print(f'perturbed A:\na = {np.round(a,3)}, v = {np.round(v.ravel(),3)}')
print(f'(A * v) / v = {np.round((A@v).ravel()/v.ravel(),3)}')
```

perturbed A:

```
a = (1-0j), v = [0.447+0.j 0.447+0.j 0.447-0.j 0.447+0.j 0.447+0.j]
(A * v) / v = [1.-0.j 1.+0.j 1.+0.j 1.+0.j 1.+0.j]
```

We also note that with this perturbation, the dimension of the eigenspace of A' associated to 1 is reduced to 1. Due to this, Power Method can converge.

```
In [ ]: a, v = np.linalg.eig(Aprime)
a
```

```
Out[ ]: array([-0.95 ,  0.95 ,  1.   , -0.475, -0.475])
```

## 2.5 Eigenvalues of A

```
In [ ]: Ak,V = QR_algo(A, tol=1e-8)
print(np.round(np.diag(Ak),2))
```

```
[ 1. +0.j -1. -0.j  1. -0.j -0.5-0.j -0.5+0.j]
```

```
In [ ]: eigval, _ = np.linalg.eig(A)
print(eigval)
```

```
[ 1. -1. -0.5  1. -0.5]
```