Power Iteration and its Variants

```
In [2]:
```

```
#keep
import numpy as np
import numpy.linalg as la
```

Let's prepare a matrix with some random or deliberately chosen eigenvalues:

```
In [6]:
```

```
#keep
n = 6

if 1:
    np.random.seed(70)
    eigvecs = np.random.randn(n, n)
    eigvals = np.sort(np.random.randn(n))
    # Uncomment for near-duplicate largest-magnitude eigenvalue
    # eigvals[1] = eigvals[0] + 1e-3

A = eigvecs.dot(np.diag(eigvals)).dot(la.inv(eigvecs))
    print(eigvals)

else:
    # Complex eigenvalues
    np.random.seed(40)
    A = np.random.randn(n, n)
    print(la.eig(A)[0])
```

```
\begin{bmatrix} -2.667651 \\ 8093 \end{bmatrix} \begin{bmatrix} -0.95797093 \\ -0.33019549 \\ -0.29151942 \\ -0.18635343 \\ -0.1441 \\ -0.18635343 \\ -0.1441 \\ -0.18635343 \\ -0.1441 \\ -0.18635343 \\ -0.1441 \\ -0.18635343 \\ -0.1441 \\ -0.18635343 \\ -0.1441 \\ -0.18635343 \\ -0.1441 \\ -0.18635343 \\ -0.1441 \\ -0.18635343 \\ -0.1441 \\ -0.18635343 \\ -0.1441 \\ -0.18635343 \\ -0.1441 \\ -0.18635343 \\ -0.1441 \\ -0.18635343 \\ -0.1441 \\ -0.18635343 \\ -0.1441 \\ -0.18635343 \\ -0.1441 \\ -0.18635343 \\ -0.1441 \\ -0.18635343 \\ -0.1441 \\ -0.1863534 \\ -0.1863534 \\ -0.1863534 \\ -0.1863534 \\ -0.1863534 \\ -0.1863534 \\ -0.1863534 \\ -0.1863534 \\ -0.1863534 \\ -0.1863534 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\ -0.186354 \\
```

Let's also pick an initial vector:

```
In [4]:
x0 = np.random.randn(n)
x0
Out[4]:
```

```
array([ 2.26930477, 0.66356156, 0.8991019 , -0.36580094, 0.462690 04, 0.079874 ])
```

Power iteration

```
In [5]:
```

Now implement plain power iteration.

Run the below cell in-place (Ctrl-Enter) many times.

```
In [6]:
```

```
x = np.dot(A, x)
x
```

```
Out[6]:
```

```
array([-6.40496816, 7.1851682 , 1.97149585, 4.77787616, 3.090991 27, 4.33054803])
```

- What's the problem with this method?
- Does anything useful come of this?
- How do we fix it?

Normalized power iteration

Back to the beginning: Reset to the initial vector.

```
In [7]:
```

```
#keep x = x0/la.norm(x0)
```

Implement normalized power iteration.

Run this cell in-place (Ctrl-Enter) many times.

```
In [8]:

x = np.dot(A, x)
nrm = la.norm(x)
x = x/nrm

print(nrm)
print(x)
```

```
4.67639723405
[-0.52706768 0.59127069 0.16223527 0.39317355 0.25435905 0.3563
6272]
```

- What do you observe about the norm?
- What about the sign?
- What is the vector *x* now?

Extensions:

- Now try the matrix variants above.
- Suggest a better way of estimating the eigenvalue. <u>Hint</u> (https://en.wikipedia.org/wiki/Rayleigh_quotient)

What if we want the smallest eigenvalue (by magnitude)?

Once again, reset to the beginning.

```
In [9]:
```

```
#keep x = x0/la.norm(x0)
```

Run the cell below in-place many times.

```
In [10]:
```

```
x = la.solve(A, x)
nrm = la.norm(x)
x = x/nrm
print(1/nrm)
print(x)
```

- What's the computational cost per iteration?
- Can we make this method search for a specific eigenvalue?
- What is this method (https://en.wikipedia.org/wiki/Inverse_iteration) called?

Can we feed an estimate of the current approximate eigenvalue back into the calculation? (Hint: Rayleigh quotient)

Reset once more.

```
In [11]:
```

```
#keep x = x0/la.norm(x0)
```

Run this cell in-place (Ctrl-Enter) many times.

```
In [12]:
```

```
sigma = np.dot(x, np.dot(A, x))/np.dot(x, x)
x = la.solve(A-sigma*np.eye(n), x)
x = x/la.norm(x)
print(sigma)
print(x)
```

```
-1.17969302281
[-0.06324034 -0.54447197 0.4628917 -0.37622458 0.41482856 0.4143
1213]
```

- What's this method (https://en.wikipedia.org/wiki/Rayleigh_quotient_iteration) called?
- What's a reasonable stopping criterion?
- Computational downside of this iteration?