week09-00

March 24, 2025

1 Sorting...

[1]:
```
import numpy as np

L = np.random.rand(20)
L
```

[1]:
```
array([0.15496462, 0.43771374, 0.7691895 , 0.07377782, 0.79910591,
       0.56433127, 0.73494644, 0.13545665, 0.66473965, 0.19518556,
       0.21220215, 0.33412276, 0.94886737, 0.4219934 , 0.98005206,
       0.42314489, 0.67667685, 0.610882  , 0.28401761, 0.12808084])
```

[2]:
```
L.sort()
L
```

[2]:
```
array([0.07377782, 0.12808084, 0.13545665, 0.15496462, 0.19518556,
       0.21220215, 0.28401761, 0.33412276, 0.4219934 , 0.42314489,
       0.43771374, 0.56433127, 0.610882  , 0.66473965, 0.67667685,
       0.73494644, 0.7691895 , 0.79910591, 0.94886737, 0.98005206])
```

[3]:
```
np.flip(L)
```

[3]:
```
array([0.98005206, 0.94886737, 0.79910591, 0.7691895 , 0.73494644,
       0.67667685, 0.66473965, 0.610882  , 0.56433127, 0.43771374,
       0.42314489, 0.4219934 , 0.33412276, 0.28401761, 0.21220215,
       0.19518556, 0.15496462, 0.13545665, 0.12808084, 0.07377782])
```

[5]:
```
rng = np.random.default_rng()

import pprint
float_formatter = "{:.5f}".format
np.set_printoptions(formatter={'float_kind':float_formatter})

colors = [ "Gainsboro",
           "Gamboge",
           "Glossy grape",
           "Gold (metallic)",
           "Gold (Crayola)",
```

```
            "Golden poppy",
            "Golden yellow",
            "Goldenrod",
            "Gotham green",
            "Granite gray",
            "Granny Smith apple",
            "Gray (web)",
            "Gray (X11 gray)",
            "Green",
            "Green (Crayola)",
            "Green (web)",
            "Green (Munsell)",
            "Green (pigment)",
            "Green-blue",
            "Green Lizard"
            ]

vals = rng.random(20)

{ colors[i]:float(vals[i]) for i in range(20) }
```

[5]:
```
{'Gainsboro': 0.06757705361621236,
 'Gamboge': 0.6814114044289292,
 'Glossy grape': 0.7277477398996349,
 'Gold (metallic)': 0.2668064570821498,
 'Gold (Crayola)': 0.7745500036614242,
 'Golden poppy': 0.19907674132921627,
 'Golden yellow': 0.3086028613701629,
 'Goldenrod': 0.43478402025095453,
 'Gotham green': 0.30007057477468146,
 'Granite gray': 0.9523168993121307,
 'Granny Smith apple': 0.1460670657546972,
 'Gray (web)': 0.7686009347303981,
 'Gray (X11 gray)': 0.0034087779127690565,
 'Green': 0.25618679306680814,
 'Green (Crayola)': 0.5852818538285638,
 'Green (web)': 0.7002443464296474,
 'Green (Munsell)': 0.24841718303543836,
 'Green (pigment)': 0.026560789524888828,
 'Green-blue': 0.3865936880142321,
 'Green Lizard': 0.6988164796183741}
```

[7]:
```
vals
```

[7]:
```
array([0.06758, 0.68141, 0.72775, 0.26681, 0.77455, 0.19908, 0.30860,
       0.43478, 0.30007, 0.95232, 0.14607, 0.76860, 0.00341, 0.25619,
       0.58528, 0.70024, 0.24842, 0.02656, 0.38659, 0.69882])
```

[6]:
```
np.argsort(vals)
```

[6]:
```
array([12, 17,  0, 10,  5, 16, 13,  3,  8,  6, 18,  7, 14,  1, 19, 15,  2,
       11,  4,  9])
```

[10]:
```
np.array([ vals[i] for i in np.argsort(-vals)])[0:3]
```

[10]:
```
array([0.95232, 0.77455, 0.76860])
```

[ ]:
```
{ colors[i]:vals[i] for i in np.argsort( - vals) }
```

[ ]:
```
np.array([ colors[i] for i in np.argsort(vals) ])
```

[12]:
```
np.array([ colors[i] for i in np.argsort( -vals) ])[0:3]
```

[12]:
```
array(['Granite gray', 'Gold (Crayola)', 'Gray (web)'], dtype='<U18')
```

2 Example: financial markets, again briefly

Recall our model:

	bull	*bear*	*recession*
followed by bull	0.90	0.15	0.25
followed by bear	0.075	0.80	0.25
followed by recession	0.025	0.05	0.50

[13]:
```
import numpy.linalg as npl

states = [ "bull", "bear", "recession" ]
1
A = np.array([[0.90 , 0.15 , 0.25],
              [0.075, 0.80 , 0.25],
              [0.025, 0.05 , 0.50]])
```

[14]:
```
vals, vecs = npl.eig(A)
vals
```

[14]:
```
array([1.00000, 0.74142, 0.45858])
```

Now, recall that the long term behavior of the market is predicted by the *normalized* 1-eigenvector.

[15]:
```
w = vecs[:,0]

# not normalized!
(w, sum(w))
```

[15]:
```
(array([0.89087, 0.44544, 0.08909]), np.float64(1.4253932901995967))
```

[16]:
```
# normalized
nw = 1/sum(w) * w
(nw,sum(nw))
```

[16]:
```
(array([0.62500, 0.31250, 0.06250]), np.float64(1.0))
```

Now we can display the market states in descending order of their long-term probabilities:

[18]:
```
{ states[i] : float(nw[i])  for i in np.argsort(nw) }
```

[18]:
```
{'recession': 0.0625, 'bear': 0.3125, 'bull': 0.625}
```

[20]:
```
def market_probabilities(A):
    vals, vecs = npl.eig(A)

    # get the 1-eigenvectors,as a  list (which should be of length 1)
    oe = [ vecs[:,i] for i in range(len(vals)) if vals[i] - 1 < 1E-5 ]
    w = oe[0]

    nw = 1/sum(w) * w

    return { states[i] : float(nw[i]) for i in np.argsort(-nw) }
```

[21]:
```
market_probabilities(A)
```

[21]:
```
{'bull': 0.625, 'bear': 0.3125, 'recession': 0.0625}
```

[22]:
```
B = np.array([[0.80 , 0.15 , 0.25],
              [0.075, 0.80 , 0.25],
              [0.035, 0.05 , 0.50]])
```

[23]:
```
market_probabilities(B)
```

[23]:
```
{'bull': 0.5228037465286232,
 'bear': 0.3933745222671231,
 'recession': 0.0838217312042538}
```

[ ]:
```

```