

Q1

June 17, 2020

```
In [41]: import numpy as np
```

```
In [16]: # part 1
```

```
lambda = np.array([39.288, 10.676])
```

```
u1 = np.array([[0.5719, 0.1469],  
               [0.5885, 0.9817],  
               [0.5715, -0.1210]])
```

```
u2 = np.array([[0.5121, -0.4042],  
               [0.6284, 0.5877],  
               [0.5856, 0.7009]])
```

```
u3 = np.array([[0.5605, -0.3179 ],  
               [0.4921, -0.3682 ],  
               [0.6661, 0.8737]])
```

```
u4 = np.array([[0.7502, -0.9201],  
               [0.6612, 0.3917]])
```

```
In [36]: u11 = u1[:,0]
```

```
u21 = u2[:,0]
```

```
u31 = u3[:,0]
```

```
u41 = u4[:,0]
```

```
u12 = u1[:,1]
```

```
u22 = u2[:,1]
```

```
u32 = u3[:,1]
```

```
u42 = u4[:,1]
```

```
In [34]: # part a
```

```
np.einsum('i,j',u11,u21)
```

```
Out[34]: array([[0.29286999, 0.35938196, 0.33490464],  
                [0.30137085, 0.3698134 , 0.3446256 ],  
                [0.29266515, 0.3591306 , 0.3346704 ]])
```

$$U_{1,1} \circ U_{2,1} = \begin{bmatrix} .5719x.5121 & .5719x.6284 & .5719x.5856 \\ .5885x.5121 & .5885x.6284 & .5885x.5856 \\ .5715x.5121 & .5715x.6284 & .5715x.5856 \end{bmatrix} = \begin{bmatrix} 0.29286999 & 0.35938196 & 0.33490464 \\ 0.30137085 & 0.3698134 & 0.3446256 \\ 0.29266515 & 0.3591306 & 0.3346704 \end{bmatrix}$$

```
In [37]: #part b
         b1=lmbda[0] * np.einsum('i,j,k,l',u11,u21,u31, u41)
         b2=lmbda[1] * np.einsum('i,j,k,l',u12,u22,u32, u42)
         b3= b1+b2
```

```
In [38]: b1
```

```
Out[38]: array([[[[4.8382407 , 4.26425586],
                  [4.24781132, 3.7438721 ],
                  [5.74978078, 5.06765536]],

                [[5.93702491, 5.23268577],
                  [5.21250661, 4.59412073],
                  [7.05557946, 6.21854058]],

                [[5.5326572 , 4.87629024],
                  [4.85748547, 4.28121754],
                  [6.57502758, 5.79499899]]],

               [[4.97867573, 4.38803038],
                  [4.37110852, 3.85254193],
                  [5.91667423, 5.2147494 ]],

               [[6.10935331, 5.38456999],
                  [5.36380511, 4.72746992],
                  [7.26037509, 6.39904027]],

               [[5.6932484 , 5.01782971],
                  [4.99847911, 4.40548439],
                  [6.76587469, 5.96320494]]],

               [[4.83485672, 4.26127335],
                  [4.24484031, 3.74125355],
                  [5.74575925, 5.06411093]],

               [[5.93287241, 5.22902591],
                  [5.20886087, 4.5909075 ],
                  [7.05064463, 6.21419119]],

               [[5.52878753, 4.87287965],
                  [4.85408804, 4.27822315],
                  [6.57042886, 5.79094583]]]])
```

In [39]: b2

```
Out[39]: array([[[[-0.18541814,  0.07893521],
                  [-0.21475609,  0.0914248 ],
                  [ 0.50959368, -0.21694147]],

                [[ 0.26959486, -0.11477047],
                  [ 0.31225174, -0.13293012],
                  [-0.74094064,  0.31542925]],

                [[ 0.32152295, -0.13687701],
                  [ 0.3723962 , -0.1585345 ],
                  [-0.88365713,  0.37618574]]],

               [[[-1.23910818,  0.52750644],
                  [-1.43516713,  0.6109716 ],
                  [ 3.40550115, -1.44977155]],

                [[ 1.80164245, -0.76698549],
                  [ 2.08670887, -0.88834242],
                  [-4.95154139,  2.10794344]],

                [[ 2.14866631, -0.91471861],
                  [ 2.48864088, -1.05945075],
                  [-5.90528391,  2.51396556]]],

               [[[ 0.15272699, -0.06501811],
                  [ 0.17689235, -0.07530566],
                  [-0.41974701,  0.17869243]],

                [[-0.22206248,  0.09453524],
                  [-0.25719851,  0.10949316],
                  [ 0.61030509, -0.25981579]],

                [[-0.26483511,  0.11274417],
                  [-0.30673887,  0.13058321],
                  [ 0.72785918, -0.30986028]]]])
```

In [40]: b3

```
Out[40]: array([[[[4.65282255, 4.34319107],
                  [4.03305524, 3.8352969 ],
                  [6.25937447, 4.85071389]],

                [[6.20661977, 5.11791531],
                  [5.52475835, 4.46119061],
```

```

        [6.31463882, 6.53396982]],

        [[5.85418015, 4.73941323],
         [5.22988167, 4.12268304],
         [5.69137045, 6.17118473]]],

        [[3.73956755, 4.91553682],
         [2.93594139, 4.46351353],
         [9.32217538, 3.76497785]],

        [[7.91099576, 4.6175845 ],
         [7.45051398, 3.8391275 ],
         [2.30883371, 8.50698371]]],

        [[7.84191472, 4.10311109],
         [7.48711999, 3.34603364],
         [0.86059077, 8.47717049]]],

        [[4.98758371, 4.19625523],
         [4.42173266, 3.66594789],
         [5.32601224, 5.24280336]],

        [[5.71080993, 5.32356115],
         [4.95166236, 4.70040065],
         [7.66094972, 5.9543754 ]],

        [[5.26395243, 4.98562382],
         [4.54734917, 4.40880637],
         [7.29828804, 5.48108555]]])

```

In [69]: # *part 2*

```

g11 = np.array([[38.946, 0.8653],
                [0.9666, -4.8832]])

g21 = np.array([[-0.4799, -0.0792],
                [-1.7302, -4.3675]])

g12 = np.array([[0.7059, -1.6496],
                [0.7553, -1.1648]])

g22 = np.array([[5.7493, -3.3204],
                [-2.0019, 7.6587]])

u1 = np.array([[0.5661, -0.1945],
                [0.6005, -0.5685],
                [0.5648, 0.7994]])

```

```

u2 = np.array([[0.5031, 0.8331],
               [0.6345, -0.1755],
               [0.5867, -0.5246]])

u3 = np.array([[0.5773, -0.3364],
               [0.5013, -0.5733],
               [0.6445, 0.7471]])

u4 = np.array([[0.7524, -0.658 ],
               [0.6587, 0.7524]])

In [133]: c1 = np.array([g11,g21])
          c2 = np.array([g12,g22])
          g = np.array([c1,c2])
          g.shape

Out[133]: (2, 2, 2, 2)

In [134]: tucker = tensorly.tucker_to_tensor(g,[u1,u2,u3,u4])

In [136]: tucker.shape

Out[136]: (3, 3, 3, 2)

In [135]: tucker

Out[135]: array([[[[ 3.00005471,  5.3689007 ],
                   [ 1.81956239,  5.68430554],
                   [ 6.48561691,  1.91262153]],

                  [[ 5.69605257,  5.70916416],
                   [ 4.701262   ,  5.08490697],
                   [ 7.33696029,  5.86532227]],

                  [[ 5.83203974,  4.96526222],
                   [ 5.05816843,  4.08603546],
                   [ 6.53526251,  6.4438178 ]]],

                [[[ 1.2956372 ,  5.98013323],
                  [-0.19517582,  6.63479675],
                  [ 6.71764191,  0.91921501]],

                  [[ 6.30977072,  6.08204755],
                   [ 5.4225399 ,  5.31253386],
                   [ 7.27009883,  6.66556679]],

                  [[ 7.21603762,  5.19245382],

```

```
[ 6.6890497 ,  4.00956391],
[ 6.36721486,  7.79044013]]],
```

```
[[[ 8.16839446,  4.57487163],
   [ 7.64504902,  4.01154602],
   [ 6.91531628,  4.95191776]]],
```

```
[[ 4.94898436,  5.62485519],
 [ 3.49563136,  5.29641449],
 [ 8.72644606,  4.6344422 ]],
```

```
[[ 2.99446728,  5.15829595],
 [ 1.41618797,  4.96747811],
 [ 8.07051237,  3.8093327 ]]]])
```

In [94]: *# part 3*

```
x11 = np.array([[4, 0, 9],
                 [7, 9, 9],
                 [4, 8, 5]])
```

```
x21 = np.array([[7, 8, 2],
                 [1, 5, 8],
                 [7, 9, 2]])
```

```
x31 = np.array([[7, 9, 4],
                 [10, 1, 2],
                 [1, 5, 8]])
```

```
x12 = np.array([[6, 5, 1],
                 [3, 3, 5],
                 [1, 8, 7]])
```

```
x22 = np.array([[8, 2, 3],
                 [4, 3, 3],
                 [2, 4, 6]])
```

```
x23 = np.array([[6, 6, 8],
                 [5, 9, 8],
                 [3, 9, 5]])
```

```
In [156]: c1 = np.array([x11,x21,x31])
          c2 = np.array([x12,x22,x23])
          X = np.array([c1,c2]).transpose(2,3,1,0)
```

```
In [159]: np.mean((X - tucker)**2)
```

```
Out[159]: 8.97580143460948
```

```
In [ ]: np.mean((X - b3)**2)
```

The cp composition resulted in a greater reduction of features (30 vs 38) and had a lower MSE (5 vs 9).

Q2

June 17, 2020

```
In [2]: import tensorly
import numpy as np
```

```
In [35]: ab = np.array([1,2,3,4]).reshape(2,2)
ac = np.array([5,6,7,8]).reshape(2,2)
cdd = np.array([6,4,16,10]).reshape(2,2)
y = np.array([1,2,3,4]).reshape(4,1)
```

Given:

$$\hat{\beta} = \operatorname{argmin}_{\beta} \|y - \{[(A \otimes C)^T * (B^T \otimes A^T)][(B \odot C) * (A \odot D) + A * B \odot D]\}\beta\|_2^2$$

Reduces to:

$$\hat{\beta} = \operatorname{argmin}_{\beta} \|y - \{[(A * B) \otimes (A * C)]^T [A * B \odot ((C * D) + D)]\}\beta\|_2^2$$

```
In [20]: first_bracket = tensorly.tenalg.kronecker([ab,ac]).T
second_bracket = tensorly.tenalg.khatri_rao([ab,cdd])
m = np.dot(first_bracket, second_bracket)
```

```
In [66]: # beta
np.linalg.lstsq(m, y, rcond=None)[0]
```

```
Out[66]: array([[ -0.0309884 ],
               [ 0.03603101]])
```


Q3

June 17, 2020

```
In [1]: import glob
        from scipy.io import loadmat
        from PIL import Image
        import numpy as np
        import matplotlib.pyplot as plt
        import matplotlib.image as mpimg
        import tensorly
        from sklearn.metrics import accuracy_score
        from sklearn.ensemble import RandomForestClassifier
        from tensorly.decomposition import tucker
        %matplotlib inline

In [2]: y_train = [x[0] for x in loadmat('train_lab.mat')['train']]
        y_test = [x[0] for x in loadmat('test_lab.mat')['test']]

        train_files = glob.glob('CatsBirds/train*')
        test_files = glob.glob('CatsBirds/Test*')

In [3]: np.array(Image.open(train_files[0]).convert('L')).shape

Out[3]: (500, 500)

In [4]: # part 1
        tnsr_train = np.zeros((500,500,len(train_files)))
        tnsr_test = np.zeros((500,500,len(test_files)))

        for tf in train_files:
            idx = int(tf.split("train")[1].split(".")[0]) - 1
            tnsr_train[:, :, idx] = np.array(Image.open(tf).convert('L'))

        for tf in test_files:
            idx = int(tf.split("Test")[1].split(".")[0]) - 1
            tnsr_test[:, :, idx] = np.array(Image.open(tf).convert('L'))

In [5]: #partial decomposition
        G, factors = tensorly.decomposition.partial_tucker(tnsr_train, modes = [0,1], ranks=[10
```

```

A,B =factors

G_f = np.zeros((28,100))
for i in range(28):
    G_f[i] = G[:, :, i].flatten()

G_test = tensorly.tenalg.multi_mode_dot(tnsr_test, [x.T for x in factors], modes=[0,1])

G_test_f = np.zeros((12,100))
for i in range(12):
    G_test_f[i] = G_test[:, :, i].flatten()

In [6]: clf = RandomForestClassifier(max_depth=2,n_estimators=100)
        clf.fit(G_f, y_train)
        y_hat = clf.predict(G_test_f)
        # error rate
        1-accuracy_score(y_test, y_hat)

Out[6]: 0.08333333333333337

In [7]: # part 2
        tnsr_train = np.zeros((500,500,3,len(train_files)))
        tnsr_test = np.zeros((500,500,3,len(test_files)))

        for tf in train_files:
            idx = int(tf.split("train")[1].split(".")[0]) - 1
            tnsr_train[:, :, :, idx] = np.array(Image.open(tf))

        for tf in test_files:
            idx = int(tf.split("Test")[1].split(".")[0]) - 1
            tnsr_test[:, :, :, idx] = np.array(Image.open(tf))

In [8]: #partial decomposition
        G, factors = tensorly.decomposition.partial_tucker(tnsr_train, modes = [0,1,2],ranks=[
        A,B,C =factors

        G_f = np.zeros((28,300))
        for i in range(28):
            G_f[i] = G[:, :, :, i].flatten()

        G_test = tensorly.tenalg.multi_mode_dot(tnsr_test, [x.T for x in factors], modes=[0,1,2])

        G_test_f = np.zeros((12,300))
        for i in range(12):
            G_test_f[i] = G_test[:, :, :, i].flatten()

In [9]: clf = RandomForestClassifier(max_depth=2,n_estimators=100)
        clf.fit(G_f, y_train)

```

```
y_hat = clf.predict(G_test_f)
# error rate
1-accuracy_score(y_test, y_hat)
```

Out[9]: 0.16666666666666663

1 Discussion

The error rate for the grayscale and color images was 0.08 and 0.17 respectively with a random forest using 100 trees and a max depth of 2.

Q4

June 17, 2020

```
In [1]: import glob
        from itertools import combinations

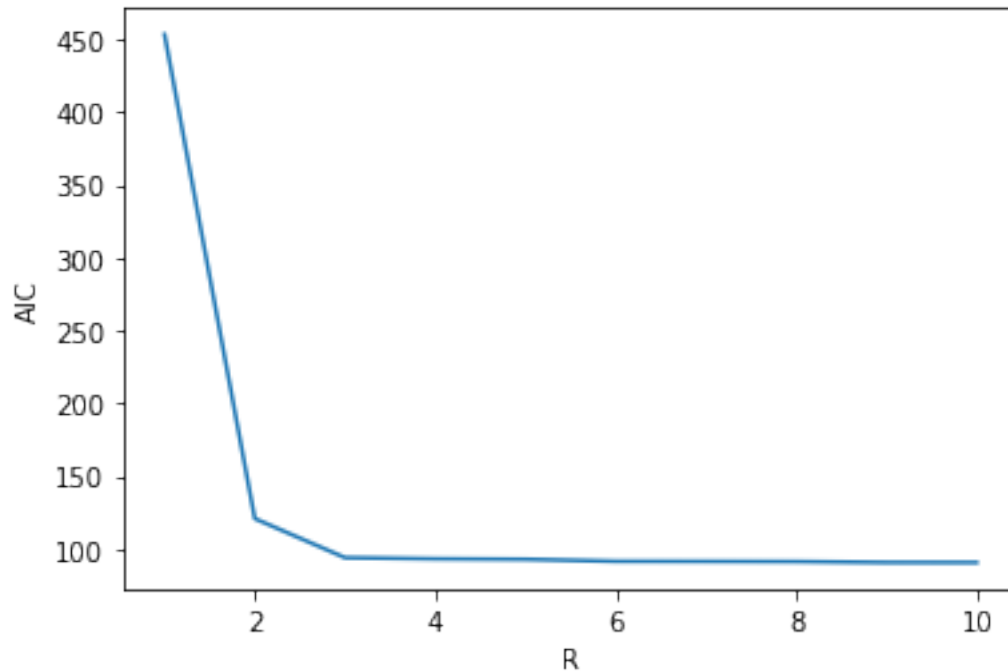
        import numpy as np
        import matplotlib.pyplot as plt
        import matplotlib.image as mpimg
        import tensorly
        from tensorly.decomposition import parafac as cp
        from scipy.io import loadmat
        import tensorly as tl
        %matplotlib inline

In [2]: mat = loadmat('heatT.mat')
        t1 = mat["T1"][0][0][0]
        t2 = mat["T2"][0][0][0]
        t3 = mat["T3"][0][0][0]

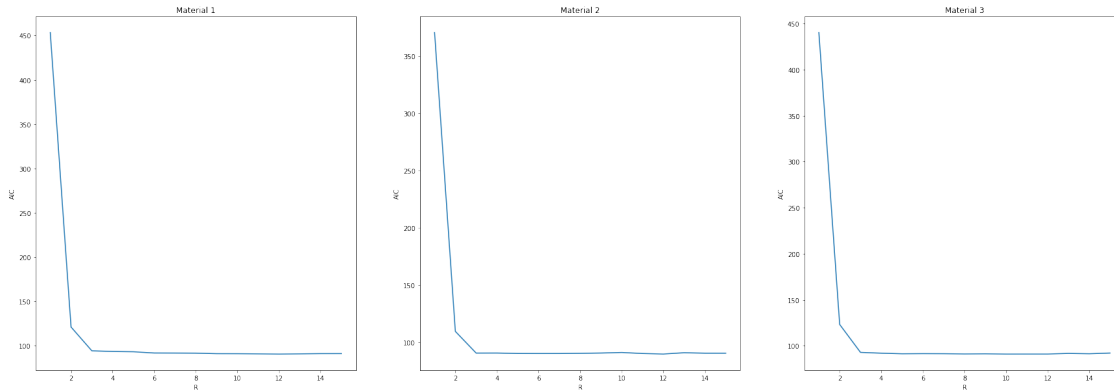
In [3]: max_rank = np.min([np.multiply(*x) for x in combinations(t1.shape, 2)])

In [4]: aic = []
        for k in range(1,11):
            kt,e = cp(t1,k)
            reconstructed = tensorly.kruskal_to_tensor((kt,e))
            err = ((t1-reconstructed)**2).sum()
            aic.append(2*err + 2*k)

        plt.plot(np.arange(1,11),aic)
        plt.xlabel('R')
        plt.ylabel('AIC')
        plt.show()
```



```
In [5]: fig, ax = plt.subplots(1, 3, figsize=(30,10))
        images = []
        min_aic = []
        max_rank = 16
        for i,t in enumerate([t1,t2,t3]):
            aic = []
            for k in range(1,max_rank):
                wf = cp(t,k)
                reconstructed = tensorly.kruskal_to_tensor(wf)
                err = ((t-reconstructed)**2).sum()
                aic.append(2*err + 2*k)
            min_aic.append(np.argmin(aic)+1)
            ax[i].plot(np.arange(1,max_rank),aic)
            ax[i].set(xlabel='R', ylabel='AIC')
            ax[i].set_title(f'Material {i+1}')
```

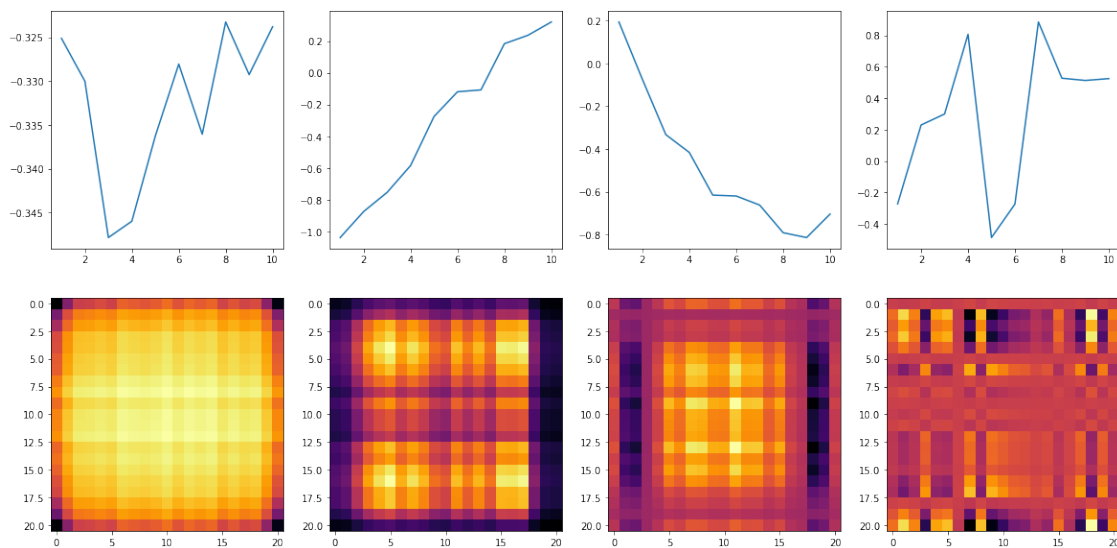


```
In [6]: # rank with min aic
min_aic
```

```
Out[6]: [12, 12, 10]
```

```
In [7]: # material 1
wf = cp(t1,min_aic[0])
```

```
fig, ax = plt.subplots(2, 4, figsize=(20,10))
a,b,c = wf[1]
for i in range(4):
    ax[0,i].plot(np.arange(1,11),c[:,i])
    A = a[:,i]
    B = b[:,i]
    XY = np.outer(A,B)
    ax[1,i].imshow(XY,cmap='inferno')
```



In [8]: # material 2

```
wf = cp(t2,min_aic[1])
```

```
fig, ax = plt.subplots(2, 4, figsize=(20,10))
```

```
a,b,c = wf[1]
```

```
for i in range(4):
```

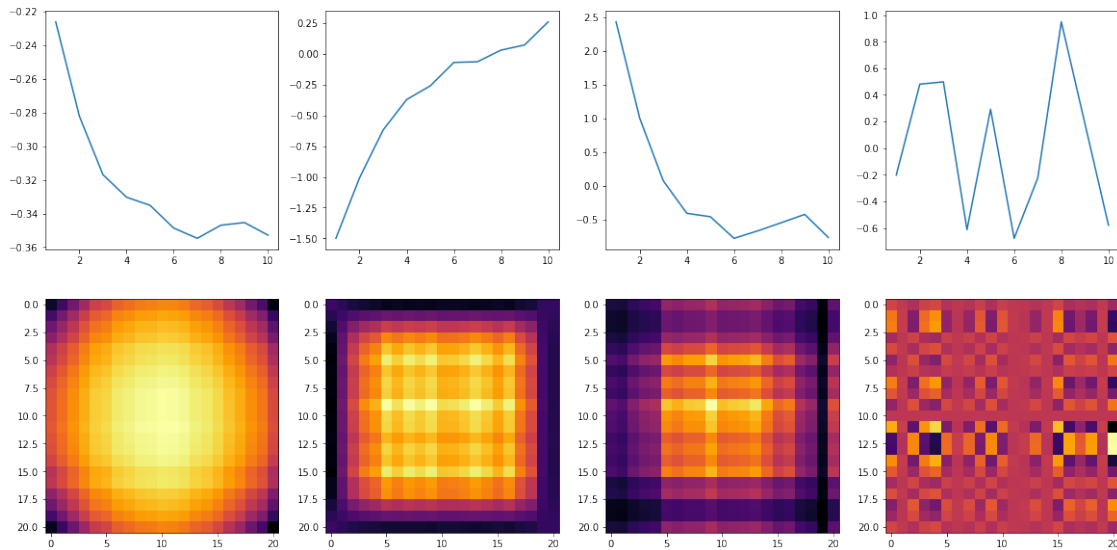
```
    ax[0,i].plot(np.arange(1,11),c[:,i])
```

```
    A = a[:,i]
```

```
    B = b[:,i]
```

```
    XY = np.outer(A,B)
```

```
    ax[1,i].imshow(XY,cmap='inferno')
```



In [9]: # material 3

```
wf = cp(t3,min_aic[2])
```

```
fig, ax = plt.subplots(2, 4, figsize=(20,10))
```

```
a,b,c = wf[1]
```

```
for i in range(4):
```

```
    A = a[:,i]
```

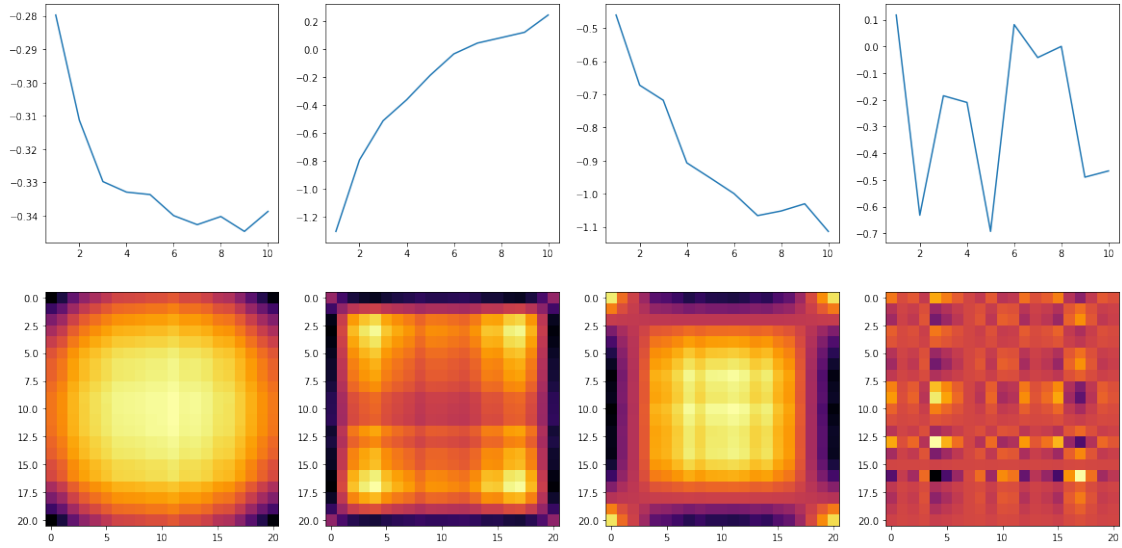
```
    B = b[:,i]
```

```
    C = c[:,i]
```

```
    ax[0,i].plot(np.arange(1,11),C)
```

```
    XY = np.outer(A,B)
```

```
    ax[1,i].imshow(XY,cmap='inferno')
```



```
In [14]: # just for fun
fig, ax = plt.subplots(1, 10, figsize=(20,10))
a,b,c = wf[1]
A = a[:,0]
B = b[:,0]
C = c[:,0]
XY = np.outer(A,B)
for i in range(10):
    t = XY*C[i]
    ax[i].imshow(t,cmap='inferno')
```

