

```
In [1]: import numpy as np
        from scipy.io import loadmat
        import matplotlib.pyplot as plt
        from mpl_toolkits.mplot3d import Axes3D

        # Load data for activity
        #
        in_data = loadmat('bucky.mat')
        A = in_data['A']
        ##

        # Load data for activity: Another option
        # A = imageio.imread("Whateveryoulike.png")
        # A = np.average(A[:, :, 0:3], axis=2)/256

        rows, cols = np.array(A.shape)
```

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In [2]: # Display image
        fig = plt.figure()
        ax = fig.add_subplot(111)

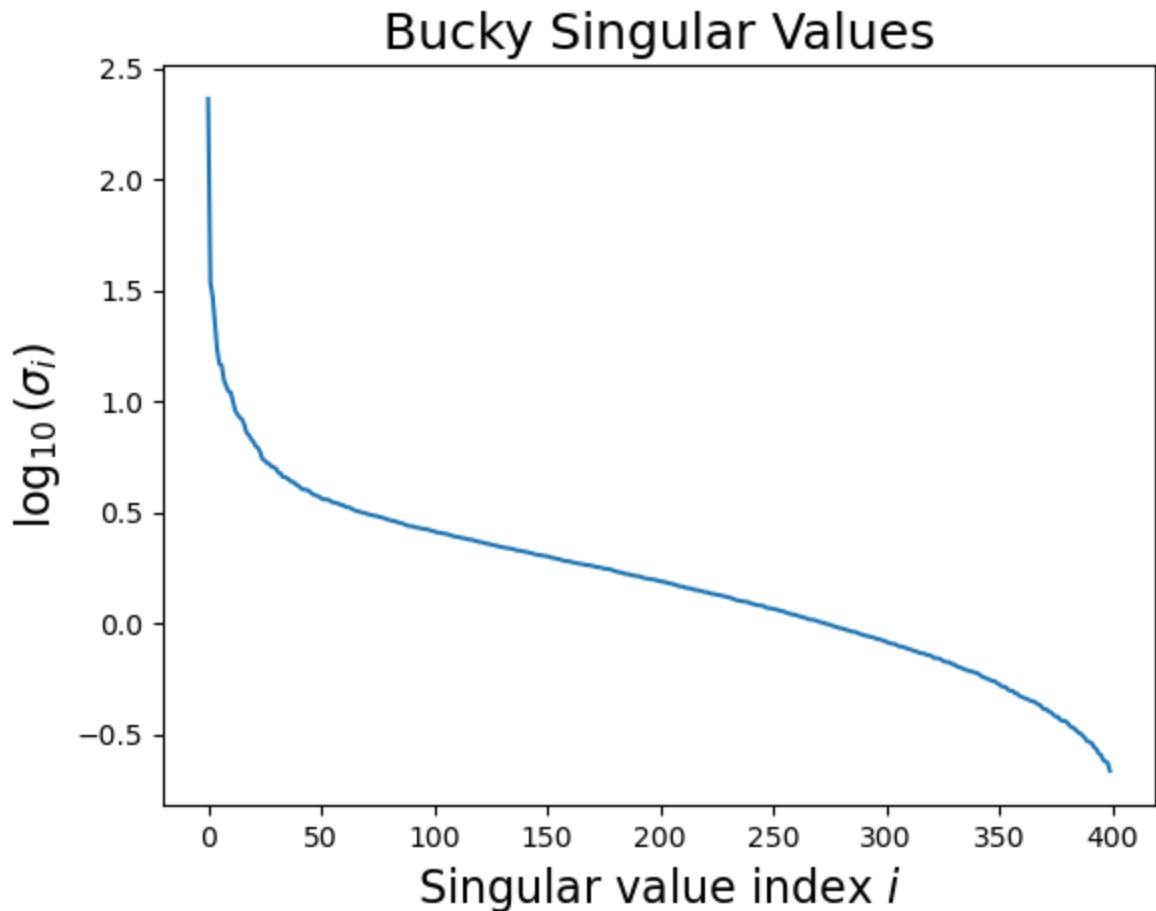
        ax.imshow(A, cmap='gray')
        ax.set_axis_off()
        plt.show()
```



```
In [3]: # Bucky's singular values

        # Complete and uncomment line below
        U, s, VT = np.linalg.svd(A, full_matrices=True)
```

```
fig = plt.figure()
ax = fig.add_subplot(111)
ax.plot(np.log10(s))
ax.set_xlabel('Singular value index $i$', fontsize=16)
ax.set_ylabel('$\log_{10}(\sigma_i)$', fontsize=16)
ax.set_title('Bucky Singular Values', fontsize=18)
plt.show()
```



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In [4]: # Find and display low-rank approximations

r_vals = np.array([10, 20, 50, 100 ])
err_fro = np.zeros(len(r_vals))

# display images of various rank approximations
for i, r in enumerate(r_vals):

    # Complete and uncomment two lines below
    # Ar = A[:, :r]

    Ar = U[:, :r] @ np.diag(s[:r]) @ VT[:, :r]
    print(U[:, :r].shape)
    print(np.diag(s[:r]).shape)
    print(VT[:, :r].shape)
    print(Ar.shape)
    Er = A - Ar
    err_fro[i] = np.linalg.norm(Er, ord='fro')
```

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fig = plt.figure()
ax = fig.add_subplot(111)
ax.imshow(Ar,cmap='gray',interpolation='none')
ax.set_axis_off()
ax.set_title(['Bucky Rank =', str(r_vals[i])], fontsize=18)
plt.show()

# plot normalized error versus rank
norm_err = err_fro/np.linalg.norm(A,ord='fro')

fig = plt.figure()
ax = fig.add_subplot(111)
ax.stem(r_vals,norm_err)
ax.set_xlabel('Rank', fontsize=16)
ax.set_ylabel('Normalized error', fontsize=16)
plt.show()

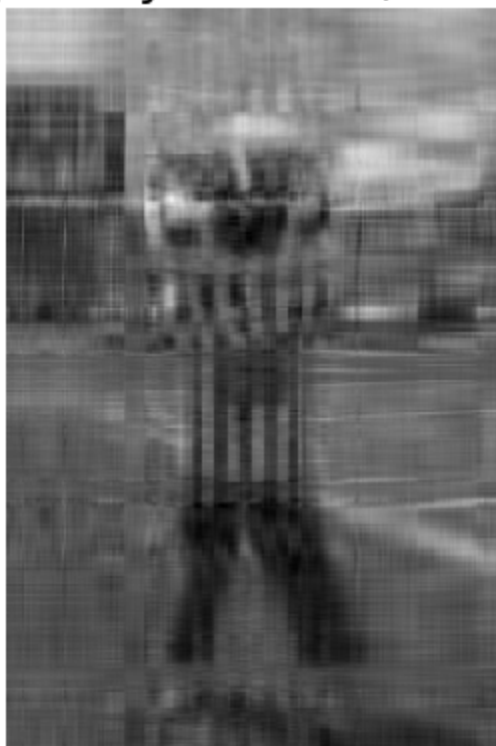
```

```

(600, 10)
(10, 10)
(10, 400)
(600, 400)

```

['Bucky Rank =', '10']



```

(600, 20)
(20, 20)
(20, 400)
(600, 400)

```

['Bucky Rank =', '20']



(600, 50)  
(50, 50)  
(50, 400)  
(600, 400)

['Bucky Rank =', '50']



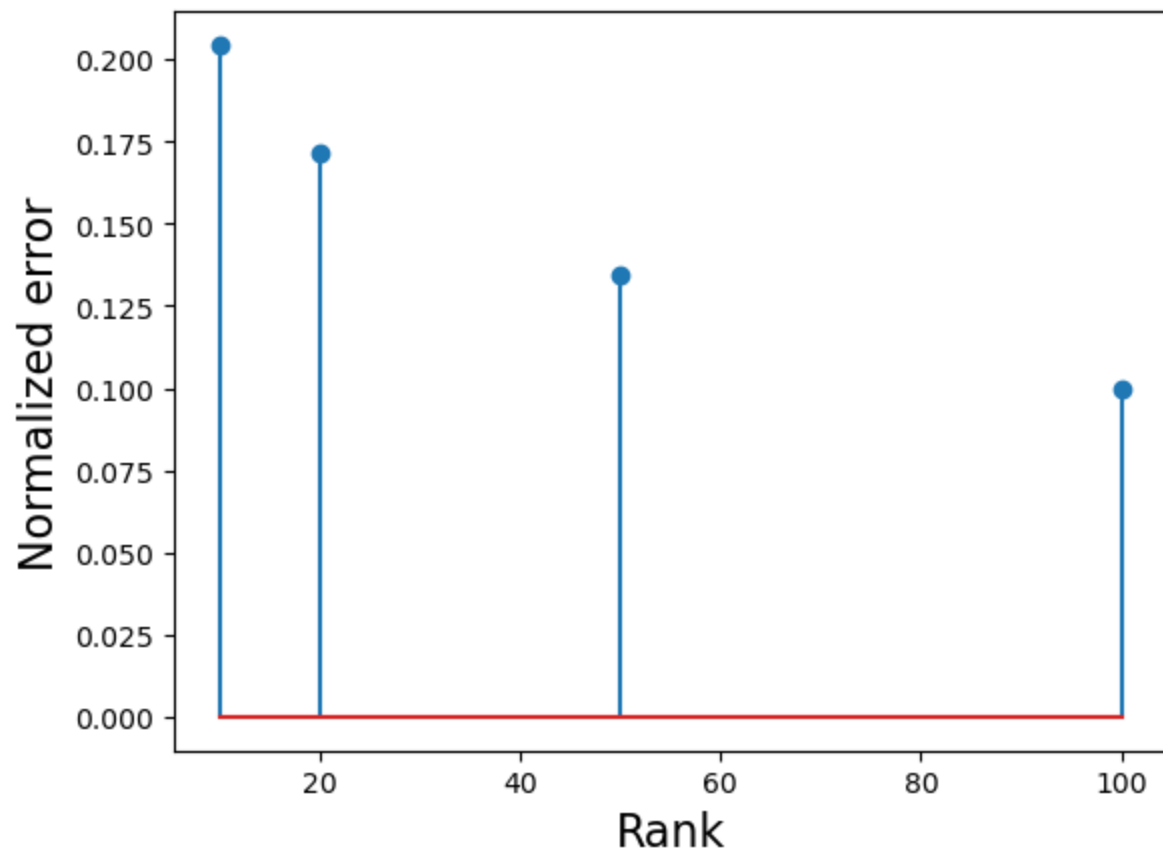
(600, 100)

(100, 100)

(100, 400)

(600, 400)

['Bucky Rank =', '100']



```

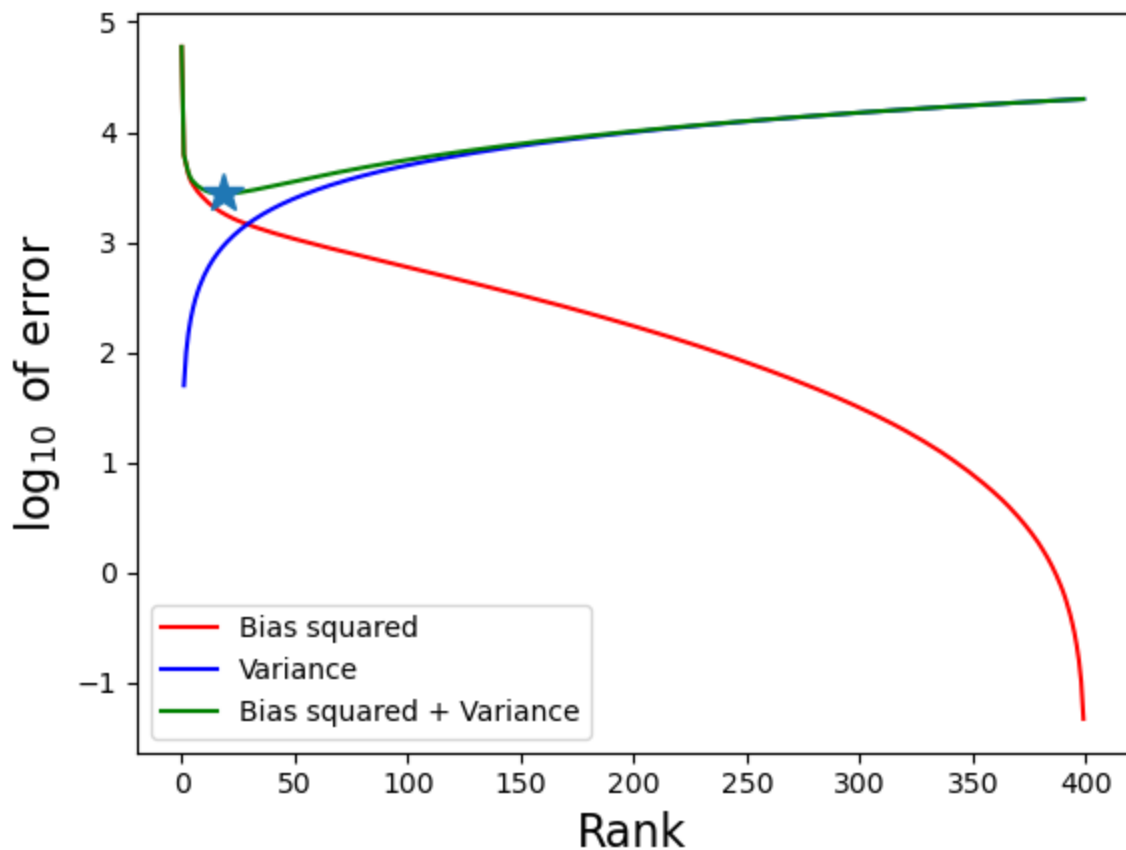
In [5]: # bias-variance tradeoff
num_sv = min(rows, cols)
bias_2 = np.zeros(num_sv)
ranks = np.arange(num_sv)

for r in range(num_sv):
    bias_2[r] = np.linalg.norm(s[r:num_sv])**2

sigma2 = 50
var = sigma2*ranks
#print(var)

fig = plt.figure()
ax = fig.add_subplot(111)
ax.plot(ranks, np.log10(bias_2), 'r', label='Bias squared')
ax.plot(ranks[1:], np.log10(var[1:]), 'b', label='Variance')
ax.plot(ranks, np.log10(bias_2+var), 'g', label='Bias squared + Variance')
min_bias_plus_variance_index = np.argmin(np.log10(bias_2+var))
ax.plot(ranks[min_bias_plus_variance_index], np.log10(bias_2+var)[min_bias_plus_variance_index], 'b', marker='*')
ax.set_xlabel('Rank', fontsize=16)
ax.set_ylabel('$\log_{10}$ of error', fontsize=16)
ax.legend()
plt.show()

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In [6]: Anoise = A + np.sqrt(sigma2/600)*np.random.randn(np.shape(A)[0], np.shape(A)[1])
Anoise

```

```
Out[6]: array([[ 1.17499756,  1.36972229,  0.93419817, ...,  0.55393274,
                0.30611841,  0.60007574],
               [ 0.73926491,  0.85861056,  1.062526 , ...,  0.70923174,
                0.33014406,  0.83035018],
               [ 1.12926001,  0.66152898,  1.1292177 , ...,  0.93548155,
                0.64496376,  0.87052681],
               ...,
               [ 0.18041518,  0.16425375,  0.33564449, ...,  0.05687059,
                -0.24820276,  1.00275999],
               [ 0.25111453, -0.09212523, -0.43009124, ..., -0.21575601,
                0.38428501,  0.35657473],
               [ 0.5349397 ,  0.36562045,  0.32224568, ...,  0.97987537,
                0.96737879, -0.21214737]])
```

In [ ]: