```
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']

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

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

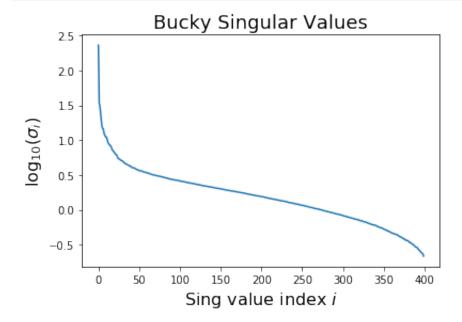
ax.imshow(A,cmap='gray')
ax.set_axis_off()
plt.show()
```



```
In [5]: # 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('Sing value index $i$', fontsize=16)
ax.set_ylabel('$\log_{10}(\sigma_i)$', fontsize=16)
ax.set_title('Bucky Singular Values', fontsize=18)
plt.show()
```



2a

It appears that rank-50 approximation would be a good fit. Becasue it compresses the vertical axis.

```
In [14]: # 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 in range(len(r_vals)):

    ind = int(r_vals[i]-1)
    # Complete and uncomment two lines below
    Ar = U[:, :ind] @ np.diag(s[:ind]) @ VT[0:ind, :]
    Fr = A - Ar
```

```
err_fro[i] = np.linalg.norm(Er,ord='fro')

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()
```

['Bucky Rank =', '10']



['Bucky Rank =', '20']



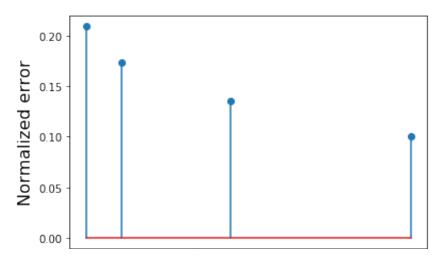
['Bucky Rank =', '50']



['Bucky Rank =', '100']



/usr/local/lib/python3.7/site-packages/ipykernel_launcher.py:27: User Warning: In Matplotlib 3.3 individual lines on a stem plot will be ad ded as a LineCollection instead of individual lines. This significant ly improves the performance of a stem plot. To remove this warning an d switch to the new behaviour, set the "use_line_collection" keyword argument to True.





2c

As r increases from 10 to 50, the improvement in quality is significant.

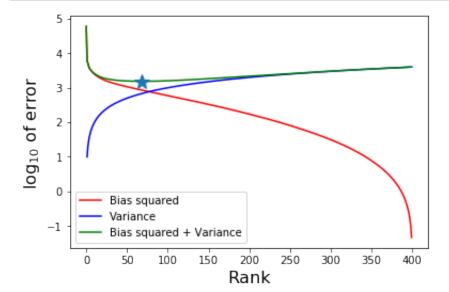
As r incrasses from 50 to 100, the improvement in quality is marginal.

2d

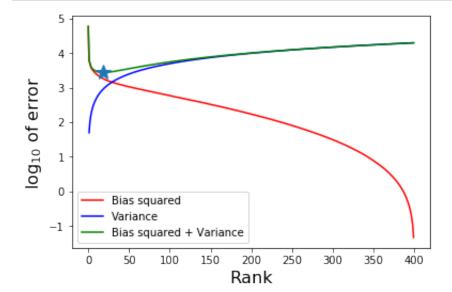
```
In [19]: full_rank_sq = 400**2;
    r_vals_sq = np.square(r_vals)
    storage_perc = r_vals_sq * (full_rank_sq**-1)
    print("Percentage of storage: ", storage_perc)
```

Percentage of storage: [0.000625 0.0025 0.015625 0.0625]

```
In [22]:
         # 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 = 10
         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]
         ax.set_xlabel('Rank', fontsize=16)
         ax.set ylabel('$\log {10}$ of error', fontsize=16)
         ax.legend()
         plt.show()
```



```
In [23]:
         # 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]
         ax.set_xlabel('Rank', fontsize=16)
         ax.set ylabel('$\log {10}$ of error', fontsize=16)
         ax.legend()
         plt.show()
```



```
In []:
```