Assignment 2

```
In [113... # Importing the libraries
    import numpy as np
    import matplotlib.pyplot as plt
    import cv2
```

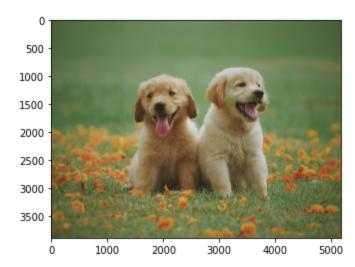
Question 1

A file Dog.jpg is provided with this assignment. Convert the RGB image to Greyscale image.

```
In [2]: # Reading the image
   image = cv2.imread('Dog.jpg')

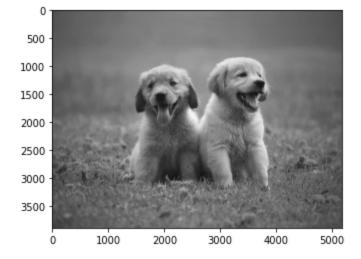
In [4]: image.shape
Out[4]: (3888, 5184, 3)

In [5]: # show the image
   plt.imshow(image[...,::-1])
Out[5]: <matplotlib.image.AxesImage at 0x2013af0fd60>
```



```
In [18]: # Using the formula Y = 0.299R + 0.587G + 0.114B
    def rgb2gray(rgbImage):
        return np.dot(image[...,:3], [0.2989, 0.5870, 0.1140])

In [23]: gray_img = rgb2gray(image)
    plt.imshow(gray_img, cmap=plt.get_cmap('gray'))
    plt.show()
```



```
In [25]: gray_img.shape
Out[25]: (3888, 5184)
```

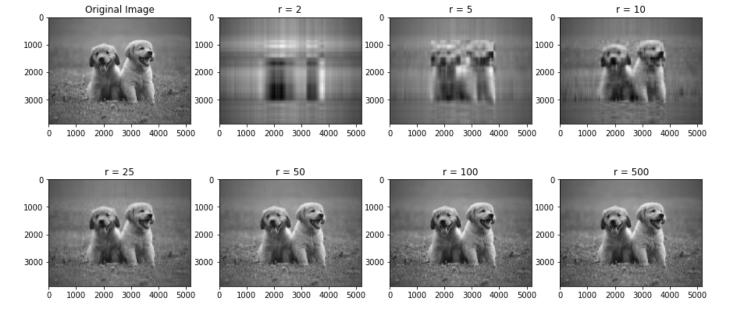
Question 2

plt.show()

Write code to perform SVD image compression with varying numbers of singular values and plot the reconstructed images. Explore how the image quality changes with different numbers of singular values. Take singular values(r) as 2, 5, 10, 25, 50, 100, 500.

```
# Function to perform SVD image compression
In [35]:
         def svd image compression(image, r values):
             U, s, Vt = np.linalg.svd(image, full matrices=False)
             compressed images = []
             for r in r values:
                 compressed image = np.dot(U[:, :r], np.dot(np.diag(s[:r]), Vt[:r, :]))
                 compressed images.append(compressed image)
             return compressed images, U, s, Vt
         # r values
In [36]:
         r \text{ values} = [2, 5, 10, 25, 50, 100, 500]
         # Compressing the image
         compressed images, U, s, Vt = svd image compression(gray img, r values)
In [37]: # Plotting the compressed images
         fig, axes = plt.subplots(2, 4, figsize=(15, 7))
         ax = axes.ravel()
         ax[0].imshow(gray img, cmap=plt.get cmap('gray'))
         ax[0].set title('Original Image')
         for i in range (1, 8):
             ax[i].imshow(compressed images[i-1], cmap=plt.get cmap('gray'))
```

ax[i].set title('r = %d' % r values[i-1])



The image quality increases with the increase in the number of singular values.

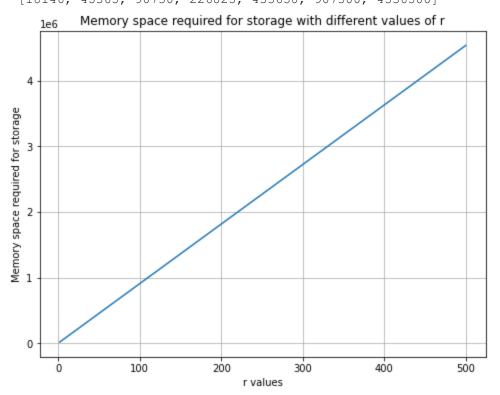
Question 3

How does the memory space required for storage change with different values of 'r'?

```
In [76]: # Memory space required for storage
    memory_space = [np.prod(U[:,:r].shape) + np.prod(s[:r].shape) + np.prod(Vt[:r,:].shape)
    print("Memory space required for storage with different values of 'r': \n", memory_space

    plt.figure(figsize=(8, 6))
    plt.plot(r_values, memory_space)
    plt.ylabel('Memory space required for storage')
    plt.xlabel('r values')
    plt.grid(True)
    plt.title('Memory space required for storage with different values of r')
    plt.show()
```

Memory space required for storage with different values of 'r': [18146, 45365, 90730, 226825, 453650, 907300, 4536500]

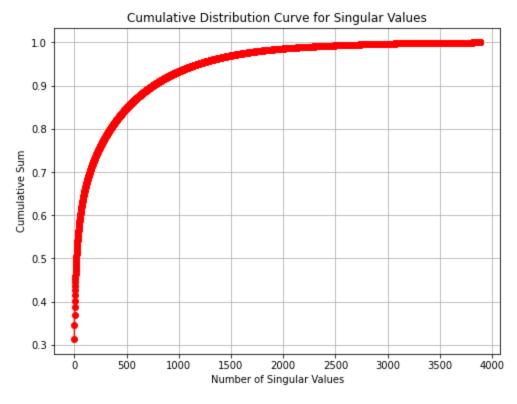


Memory space required for storage is directly proportional to the number of singular values. As the number of singular values increases, the memory space required for storage increases linearly.

Question 4

Plot cumulative distribution curve change with different values of singular values.

```
In [114... # Plot cumulative distribution curve
    plt.figure(figsize=(8, 6))
        cumulative_sum = np.cumsum(s)
        plt.plot(cumulative_sum / np.sum(s), 'ro-')
        plt.title('Cumulative Distribution Curve for Singular Values')
        plt.xlabel('Number of Singular Values')
        plt.ylabel('Cumulative Sum')
        plt.grid(True)
        plt.show()
```



Question 5

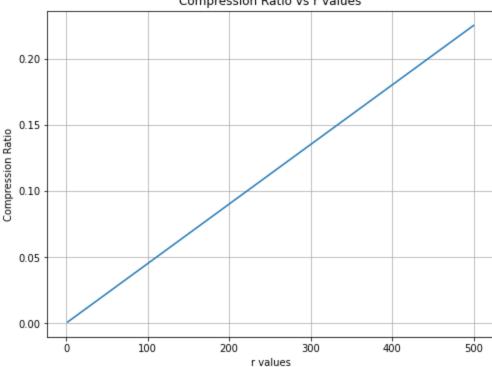
Calculate the compression ratio and plot with respect to singular values.

```
In [73]: # Compression ratio
    original_size = gray_img.size
    compression_ratios = [memory_req/original_size for memory_req in memory_space]
    print("Compression ratios with different values of 'r': \n", compression_ratios)

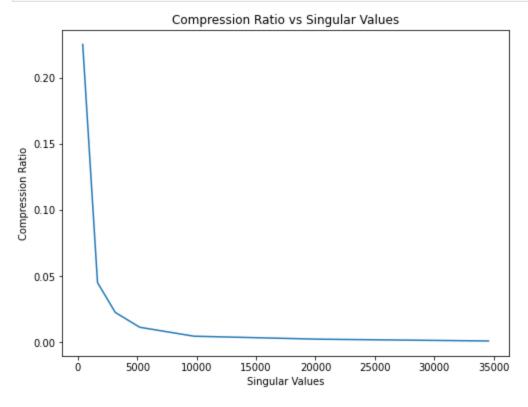
plt.figure(figsize=(8, 6))
    plt.plot(r_values, compression_ratios)
    plt.ylabel('Compression Ratio')
    plt.xlabel('r values')
    plt.grid(True)
    plt.title('Compression Ratio vs r values')
    plt.show()
```

Compression ratios with different values of 'r': [0.0009003049903469999, 0.0022507624758675, 0.004501524951735, 0.011253812379337499, 0.022507624758674998, 0.045015249517349996, 0.22507624758675]

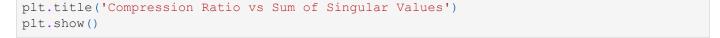
Compression Ratio vs r values

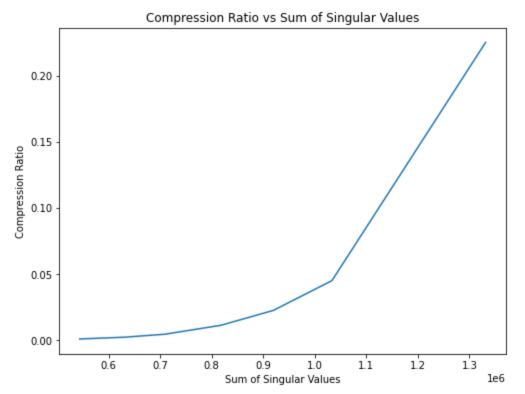


```
In [82]:
         # Plotting compression ratio vs singular values
         singular_valuess = [ s[r] for r in r_values]
         plt.figure(figsize=(8, 6))
        plt.plot(singular valuess, compression ratios)
         plt.ylabel('Compression Ratio')
         plt.xlabel('Singular Values')
        plt.title('Compression Ratio vs Singular Values')
         plt.show()
```



```
In [81]:
         # Plotting compression ratio vs sum of singular values
         singular valuess = [ np.sum(s[:r]) for r in r values]
         plt.figure(figsize=(8, 6))
         plt.plot(singular valuess, compression ratios)
         plt.ylabel('Compression Ratio')
         plt.xlabel('Sum of Singular Values')
```



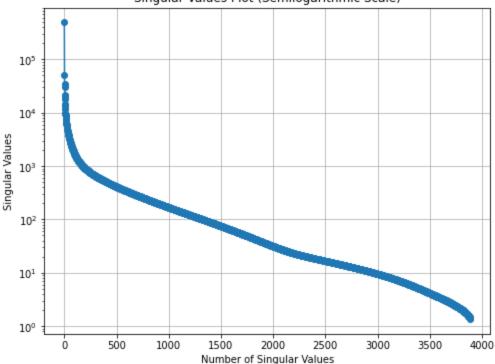


Question 6

Plot the Singular Values Plot (Semilogarithmic Scale) between Number of singular values and Singular values.

```
In [116... # Singular Values Plot (Semilogarithmic Scale)
   plt.figure(figsize=(8, 6))
   plt.semilogy(range(len(s)), s, marker='o', linestyle='-')
   plt.title('Singular Values Plot (Semilogarithmic Scale)')
   plt.xlabel('Number of Singular Values')
   plt.ylabel('Singular Values')
   plt.grid(True)
   plt.show()
```

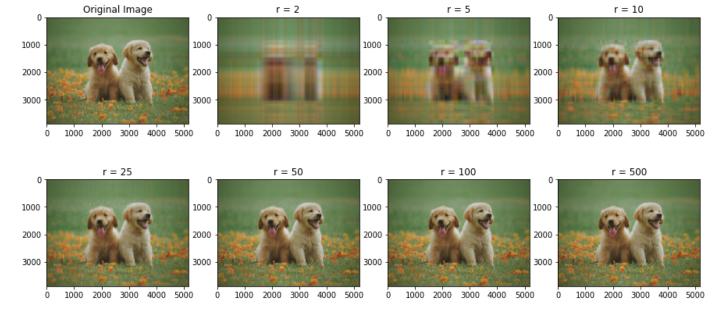
Singular Values Plot (Semilogarithmic Scale)



Question 7

Repeat Q. 2 for the original RGB image. Show three channels of RGB image along with the original images

```
image.shape
In [84]:
         (3888, 5184, 3)
Out[84]:
         R = image[..., 2]
In [94]:
         G = image[..., 1]
         B = image[..., 0]
         R_compressed_images, _, _, = svd_image_compression(R, r_values)
         G_compressed_images, _, _, _ = svd_image_compression(G, r_values)
         B_compressed_images, _, _, = svd_image_compression(B, r_values)
In [109...
         compressed images = []
         for i in range(len(r values)):
             print(f'Processing {i+1}/{len(r values)}')
             compressed image = np.stack([R compressed images[i], G compressed images[i], B compr
             compressed images.append(compressed image)
         Processing 1/7
         Processing 2/7
         Processing 3/7
         Processing 4/7
         Processing 5/7
         Processing 6/7
         Processing 7/7
In [111...
         # Plotting the compressed images
         fig, axes = plt.subplots(2, 4, figsize=(15, 7))
         ax = axes.ravel()
         ax[0].imshow(image[...,::-1])
         ax[0].set title('Original Image')
         for i in range (1, 8):
             ax[i].imshow(compressed images[i-1])
             ax[i].set title('r = %d' % r values[i-1])
         plt.show()
```



We can see that the image quality increases with the increase in the number of singular values.

The End