

```
In [2]: #importing necessary libraries
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
%matplotlib inline
from skimage.io import imread, imshow
```

```
In [3]: img_1 = imread('EP-00-00012_0119_0001.JPG')
imshow(img_1);

#Shape of image
print(img_1.shape)

#Pixel of image
pixmap1 = np.reshape(img_1, (3648*5472*3))
print(pixmap1)
```

(3648, 5472, 3)

[127 138 142 ... 56 51 47]



```
In [5]: greym1 = imread('EP-00-00012_0119_0001.JPG', as_gray=True)
        imshow(greym1);

        #Shape of image
        print(greym1.shape)

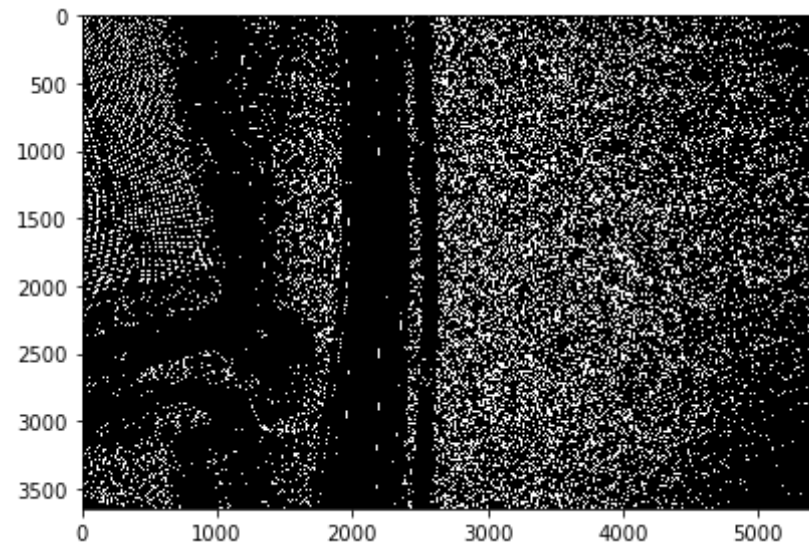
        #Pixel of image
        greypixm_1 = np.reshape(greym1, (3648*5472))
        print(greypixm_1)

(3648, 5472)
[0.53314078 0.54490549 0.5684349 ... 0.19127098 0.19911412 0.20303569]
```

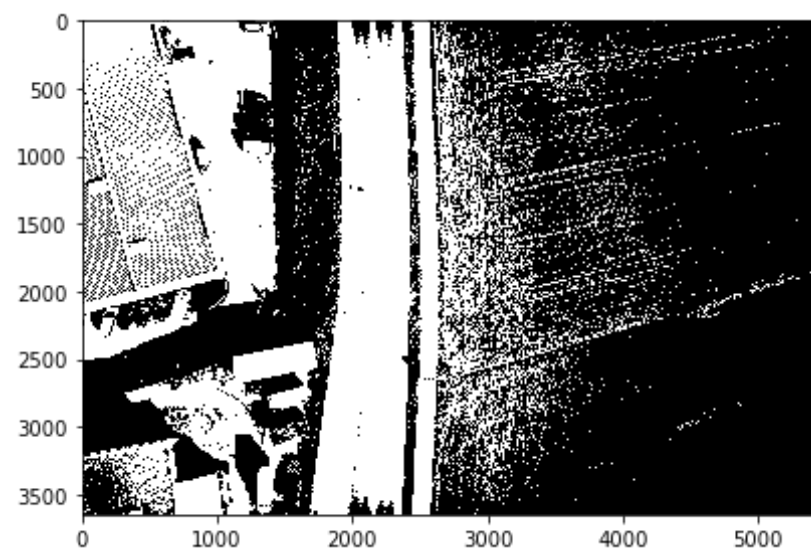


```
In [9]: #Applying Edge feature detection, canny algorithm as it helps to find
        the edges of the image in a wide range.

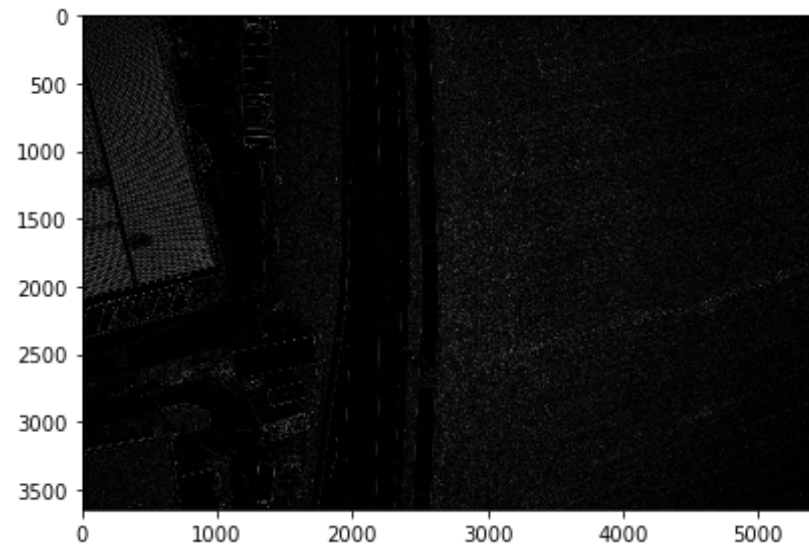
        from skimage import feature
        from skimage.feature import canny
        can = feature.canny(greym1)
        imshow(can, cmap='gray');
```



```
In [22]: #Here we are going to find the threshold value for the image so that we  
can understand the features of the image more clearly.  
#Threshold is the image segmentation method.  
#Selecting a wrong threshold values can distort the images.  
  
from skimage.filters import threshold_otsu  
  
thresh_val = threshold_otsu(greyimg1)  
  
img_threshold = greyimg1 > thresh_val  
  
# Show the original image  
imshow(greyimg1)  
plt.show()  
  
# Show the thresholded image  
imshow(img_threshold)  
plt.show()
```

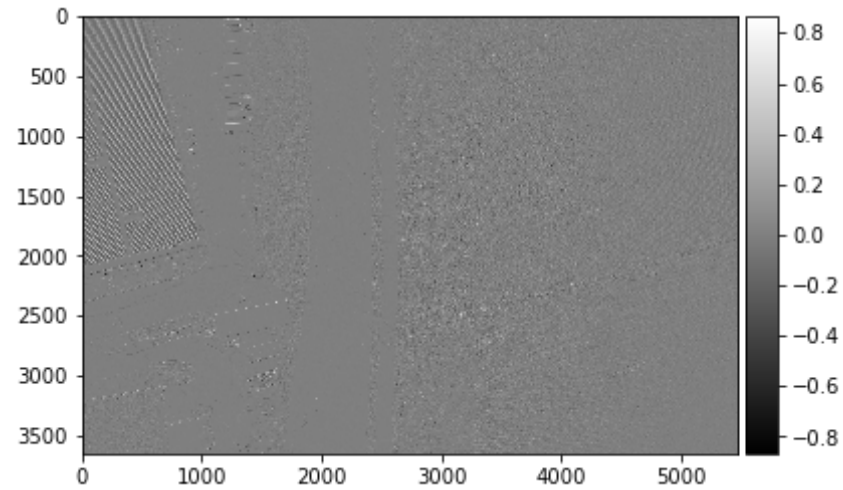


```
In [23]: #Applying Edge feature detection, Sobel Kernel. It helps to create images which have more underlining on edges.  
from skimage import filters  
sobel = filters.sobel(greyimg1)  
imshow(sobel, cmap='gray');
```



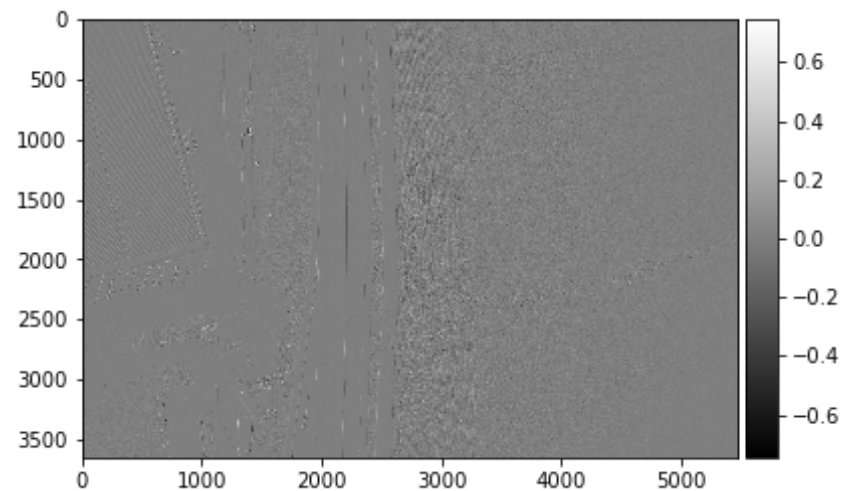
```
In [24]: #Applying Edge feature detection, prewitt kernel. The advantage of this algorithm is that it detects edges both horizontally and vertically  
from skimage.filters import prewitt_h, prewitt_v  
  
#calculating horizontal edges using prewitt kernel  
pkh = prewitt_h(greyimg1)  
  
imshow(pkh, cmap='gray')
```

```
Out[24]: <matplotlib.image.AxesImage at 0x1e20e4c3438>
```



```
In [52]: #calculating vertical edges using prewitt kernel  
pkv = prewitt_v(greyimg1)  
imshow(pkv, cmap='gray')
```

```
Out[52]: <matplotlib.image.AxesImage at 0x1e200091b70>
```

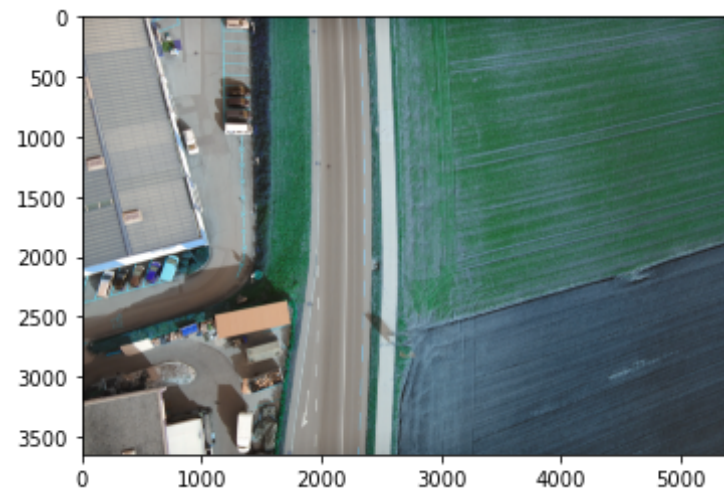


PCA Dimensionality Reduction method


```
In [37]: # Importing necessary libraries
import cv2
import numpy as np
import matplotlib.pyplot as plt
from sklearn.decomposition import PCA
```

```
In [38]: # Reading the image
img_1 = cv2.imread('EP-00-00012_0119_0001.JPG')
plt.imshow(img_1)
```

Out[38]: <matplotlib.image.AxesImage at 0x1e201a85978>



```
In [39]: blue,green,red = cv2.split(img_1) # Splits the image in R,G,B arrays.
```

```
In [49]: #Applying PCA to each array which we got after splitting the image.
#We are also applying inverse transform to transformed array. As we are
#just keeping some components, inverse transform is important to recreat
#e the original dimensions of the base image.

pca = PCA(200) # we are initializing PCA with first 30 principal compon
ents
```

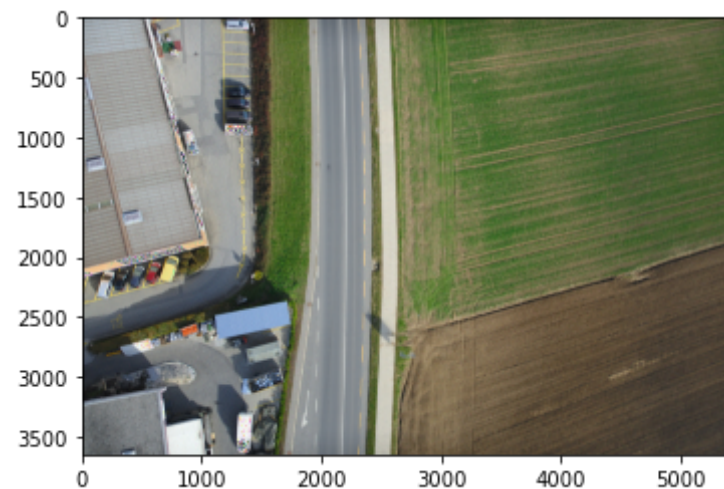
```
red_transformed = pca.fit_transform(red)
red_inverted = pca.inverse_transform(red_transformed)

green_transformed = pca.fit_transform(green)
green_inverted = pca.inverse_transform(green_transformed)

blue_transformed = pca.fit_transform(blue)
blue_inverted = pca.inverse_transform(blue_transformed)
```

```
In [50]: # compressing the image
img_compress= (np.dstack((red_inverted, green_inverted, blue_inverted
)).astype(np.uint8)
plt.imshow(img_compress)
```

Out[50]: <matplotlib.image.AxesImage at 0x1e20e898f60>



```
In [51]: pca = PCA(400) # we are increasing the number of components

red_transformed = pca.fit_transform(red)
red_inverted = pca.inverse_transform(red_transformed)
```

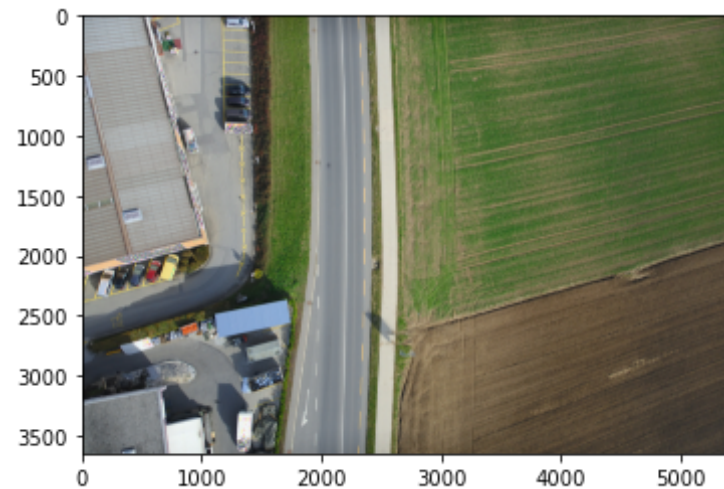


```
green_transformed = pca.fit_transform(green)
green_inverted = pca.inverse_transform(green_transformed)

blue_transformed = pca.fit_transform(blue)
blue_inverted = pca.inverse_transform(blue_transformed)

img_compress= (np.dstack((red_inverted, green_inverted, blue_inverted
))).astype(np.uint8)
plt.imshow(img_compress)
```

Out[51]: <matplotlib.image.AxesImage at 0x1e2009542e8>



```
In [53]: import cv2
from sklearn.decomposition import FastICA
from pylab import *
import matplotlib.pyplot as plt
```

References

<https://towardsdatascience.com/feature-extraction-techniques-d619b56e31be>

<https://www.pluralsight.com/guides/building-features-from-image-data-in-python>

<https://analyticsindiamag.com/image-feature-extraction-using-scikit-image-a-hands-on-guide/>