PCA Best

April 18, 2023

```
[]: import rasterio
     import numpy as np
     import matplotlib.pyplot as plt
[]: import rasterio
     import numpy as np
     from sklearn.decomposition import PCA
     # Input file path
     file_path = "H:/Term Paper/VINIR1_stacked.tif"
     # Read the image data and metadata
     with rasterio.open(file_path) as src:
         img = src.read()
         meta = src.meta
     # Reshape the image array to a 2D array
     n_pixels = img.shape[1] * img.shape[2]
     n_bands = img.shape[0]
     img_2d = img.transpose(1, 2, 0).reshape(n_pixels, n_bands)
     # Apply PCA to the image data
     pca = PCA(n_components=n_bands)
     PC = pca.fit_transform(img_2d)
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     PC = pca.fit_transform(img_2d)
     # Print the explained variance ratios for each principal component
     print(pca.explained_variance_ratio_)
    [9.83385356e-01 1.24758329e-02 3.51407477e-03 1.76412751e-04
     1.42104890e-04 9.22144741e-05 5.64793531e-05 4.65735198e-05
     2.23032357e-05 1.42569023e-05 1.01021618e-05 7.25545519e-06
     5.82354326e-06 5.20475982e-06 4.12998174e-06 3.94785745e-06
     3.69156752e-06 3.18678237e-06 2.84958329e-06 2.65582771e-06
     2.64178564e-06 2.51192455e-06 2.33935720e-06 2.18873932e-06
     2.14563906e-06 2.03137548e-06 1.93725890e-06 1.86670914e-06
     1.78804500e-06 1.70436231e-06 1.64388577e-06 1.55554447e-06
     1.18898851e-06]
[]: # Plot the graph
     plt.figure(figsize=(8,6))
     plt.plot(range(1, n_bands+1), pca.explained_variance_ratio_.cumsum(), 'bo-', u
      →linewidth=2)
     plt.xlabel('Number of Principal Components')
     plt.ylabel('Variance Retained (%)')
     plt.title('Variance Retained by Principal Components')
     plt.xticks(range(1, n_bands+1))
     plt.grid()
     plt.show()
```

