```
import numpy as np
import cv2
import scipy.io
import os
from numpy.linalg import norm
from matplotlib import pyplot as plt
from numpy.linalg import det
from numpy.linalg import inv
from scipy.linalg import rq
from numpy.linalg import svd
import matplotlib.pyplot as plt
import numpy as np
import math
import random
import sys
from scipy import ndimage, spatial
from tqdm.notebook import tqdm, trange
```

▼ Importing Drive (Dataset-University)

```
from google.colab import drive
# This will prompt for authorization.
drive.mount('/content/drive')
     Mounted at /content/drive
plt.figure(figsize=(20,10))
     <Figure size 1440x720 with 0 Axes>
     <Figure size 1440x720 with 0 Axes>
class Image:
    def __init__(self, img, position):
        self.img = img
        self.position = position
inlier_matchset = []
def features_matching(a,keypointlength,threshold):
  #threshold=0.2
  bestmatch=np.empty((keypointlength),dtype= np.int16)
  img1index=np.empty((keypointlength),dtype=np.int16)
  distance=np.empty((keypointlength))
  index=0
  for j in range(0,keypointlength):
    #For a descriptor fa in Ia, take the two closest descriptors fb1 and fb2 in Ib
    x=a[j]
    listx=x.tolist()
    x.sort()
    minval1=x[0]
                                               # min
    minval2=x[1]
                                               # 2nd min
    itemindex1 = listx.index(minval1)
                                               #index of min val
```

```
#index of second min value
   itemindex2 = listx.index(minval2)
   ratio=minval1/minval2
                                               #Ratio Test
   if ratio<threshold:
     #Low distance ratio: fb1 can be a good match
     bestmatch[index]=itemindex1
     distance[index]=minval1
     img1index[index]=j
     index=index+1
  return [cv2.DMatch(imglindex[i],bestmatch[i].astype(int),distance[i]) for i in range(0,index)]
def compute_Homography(im1_pts,im2_pts):
  im1 pts and im2 pts are 2×n matrices with
  4 point correspondences from the two images
  num_matches=len(im1_pts)
  num rows = 2 * num matches
  num_cols = 9
  A_matrix_shape = (num_rows,num_cols)
  A = np.zeros(A_matrix_shape)
  a index = 0
  for i in range(0,num_matches):
   (a_x, a_y) = im1_pts[i]
   (b_x, b_y) = im2_pts[i]
   row1 = [a_x, a_y, 1, 0, 0, 0, -b_x*a_x, -b_x*a_y, -b_x] # First row
   row2 = [0, 0, 0, a_x, a_y, 1, -b_y*a_x, -b_y*a_y, -b_y] # Second row
   # place the rows in the matrix
   A[a\_index] = row1
   A[a_index+1] = row2
   a_index += 2
  U, s, Vt = np.linalg.svd(A)
  #s is a 1-D array of singular values sorted in descending order
  #U, Vt are unitary matrices
  #Rows of Vt are the eigenvectors of A^TA.
  #Columns of U are the eigenvectors of AA^T.
  H = np.eye(3)
  H = Vt[-1].reshape(3,3) # take the last row of the Vt matrix
  return H
def displayplot(img,title):
  plt.figure(figsize=(15,15))
  plt.title(title)
  plt.imshow(cv2.cvtColor(img, cv2.COLOR_BGR2RGB))
  plt.show()
```

def RANSAC_alg(f1, f2, matches, nRANSAC, RANSACthresh):

```
minMatches = 4
nBest = 0
best_inliers = []
H = np.eye(3,3)
global inlier_matchset
inlier_matchset=[]
for iteration in range(nRANSAC):
    #Choose a minimal set of feature matches.
    matchSample = random.sample(matches, minMatches)
    #Estimate the Homography implied by these matches
    im1_pts=np.empty((minMatches,2))
    im2_pts=np.empty((minMatches,2))
    for i in range(0,minMatches):
     m = matchSample[i]
     im1_pts[i] = f1[m.queryIdx].pt
     im2_pts[i] = f2[m.trainIdx].pt
      #im1_pts[i] = f1[m[0]].pt
      #im2_pts[i] = f2[m[1]].pt
    H_estimate=compute_Homography(im1_pts,im2_pts)
    # Calculate the inliers for the H
    inliers = get_inliers(f1, f2, matches, H_estimate, RANSACthresh)
    # if the number of inliers is higher than previous iterations, update the best estimates
    if len(inliers) > nBest:
        nBest= len(inliers)
        best inliers = inliers
print("Number of best inliers",len(best_inliers))
for i in range(len(best_inliers)):
 inlier_matchset.append(matches[best_inliers[i]])
# compute a homography given this set of matches
im1_pts=np.empty((len(best_inliers),2))
im2_pts=np.empty((len(best_inliers),2))
for i in range(0,len(best_inliers)):
 m = inlier_matchset[i]
 im1_pts[i] = f1[m.queryIdx].pt
 im2_pts[i] = f2[m.trainIdx].pt
  #im1_pts[i] = f1[m[0]].pt
 #im2_pts[i] = f2[m[1]].pt
M=compute_Homography(im1_pts,im2_pts)
return M
```

```
def get_inliers(f1, f2, matches, H, RANSACthresh):
  inlier_indices = []
  for i in range(len(matches)):
    queryInd = matches[i].queryIdx
    trainInd = matches[i].trainIdx
    #queryInd = matches[i][0]
```

```
#trainInd = matches[i][1]
    queryPoint = np.array([f1[queryInd].pt[0], f1[queryInd].pt[1], 1]).T
    trans_query = H.dot(queryPoint)
    comp1 = [trans_query[0]/trans_query[2], trans_query[1]/trans_query[2]] # normalize with respect to z
    comp2 = np.array(f2[trainInd].pt)[:2]
   if(np.linalg.norm(comp1-comp2) <= RANSACthresh): # check against threshold</pre>
     inlier_indices.append(i)
  return inlier_indices
def ImageBounds(img, H):
   h, w= img.shape[0], img.shape[1]
    p1 = np.dot(H, np.array([0, 0, 1]))
    p2 = np.dot(H, np.array([0, h - 1, 1]))
    p3 = np.dot(H, np.array([w - 1, 0, 1]))
    p4 = np.dot(H, np.array([w - 1, h - 1, 1]))
    x1 = p1[0] / p1[2]
   y1 = p1[1] / p1[2]
    x2 = p2[0] / p2[2]
   y2 = p2[1] / p2[2]
   x3 = p3[0] / p3[2]
   y3 = p3[1] / p3[2]
   x4 = p4[0] / p4[2]
   y4 = p4[1] / p4[2]
   minX = math.ceil(min(x1, x2, x3, x4))
    minY = math.ceil(min(y1, y2, y3, y4))
    maxX = math.ceil(max(x1, x2, x3, x4))
    maxY = math.ceil(max(y1, y2, y3, y4))
    return int(minX), int(minY), int(maxX), int(maxY)
def Populate_Images(img, accumulator, H, bw):
    h, w = img.shape[0], img.shape[1]
    minX, minY, maxX, maxY = ImageBounds(img, H)
    for i in range(minX, maxX + 1):
        for j in range(minY, maxY + 1):
            p = np.dot(np.linalg.inv(H), np.array([i, j, 1]))
            x = p[0]
            y = p[1]
           z = p[2]
            _x = int(x / z)
            _y = int(y / z)
            if _x < 0 or _x >= w - 1 or _y < 0 or _y >= h - 1:
               continue
            if img[_y, _x, 0] == 0 and img[_y, _x, 1] == 0 and img[_y, _x, 2] == 0:
                continue
```

```
if _x >= minX and _x < minX + bw:
                wt = float(_x - minX) / bw
            if _x <= maxX and _x > maxX -bw:
                wt = float(maxX - _x) /bw
            accumulator[j, i, 3] += wt
            for c in range(3):
                accumulator[j, i, c] += img[_y, _x, c] *wt
def Image_Stitch(Imagesall, blendWidth, accWidth, accHeight, translation):
    channels=3
    #width=720
   acc = np.zeros((accHeight, accWidth, channels + 1))
   M = np.identity(3)
    for count, i in enumerate(Imagesall):
        M = i.position
        img = i.img
        M_trans = translation.dot(M)
        Populate_Images(img, acc, M_trans, blendWidth)
    height, width = acc.shape[0], acc.shape[1]
   img = np.zeros((height, width, 3))
    for i in range(height):
        for j in range(width):
            weights = acc[i, j, 3]
            if weights > 0:
               for c in range(3):
                   img[i, j, c] = int(acc[i, j, c] / weights)
    Imagefull = np.uint8(img)
   M = np.identity(3)
    for count, i in enumerate(Imagesall):
        if count != 0 and count != (len(Imagesall) - 1):
            continue
        M = i.position
        M_trans = translation.dot(M)
        p = np.array([0.5 * width, 0, 1])
        p = M_trans.dot(p)
        if count == 0:
            x_{init}, y_{init} = p[:2] / p[2]
        if count == (len(Imagesall) - 1):
            x_{final}, y_{final} = p[:2] / p[2]
```

wt = 1.0

```
Imagefull, A, (accWidth, accHeight), flags=cv2.INTER_LINEAR
     displayplot(croppedImage, 'Final Stitched Image')
  #!pip uninstall opencv-python
  #!pip install opencv-contrib-python===4.4.0.44
  #!pip install opencv-python==4.4.0.44
  #!pip install opency-contrib-python==4.4.0.44
 import cv2
 print(cv2.__version__)
      4.1.2

    Reading GPS and Metdata information

  Georeferencing through the data (Incomplete)
  from PIL import Image, ExifTags
 img = Image.open(f"{left_files_path[0]}")
 exif = { ExifTags.TAGS[k]: v for k, v in img._getexif().items() if k in ExifTags.TAGS }
 from PIL.ExifTags import TAGS
  def get exif(filename):
     image = Image.open(filename)
     image.verify()
     return image._getexif()
  def get_labeled_exif(exif):
     labeled = {}
     for (key, val) in exif.items():
         labeled[TAGS.get(key)] = val
     return labeled
  exif = get_exif(f"{left_files_path[0]}")
 labeled = get_labeled_exif(exif)
 print(labeled)
       {'ExifVersion': b'0230', 'ApertureValue': (497, 100), 'DateTimeOriginal': '2018:09:02 05:27:46', 'ExposureBiasValue': (0, 10), 'MaxApertureValue': (297, 100), 'SubjectDistance': (4294967295, 1000), 'MeteringMode': 1, 'LightSource': 9, 'Flash
 print(TAGS)
      {11: 'ProcessingSoftware', 254: 'NewSubfileType', 255: 'SubfileType', 256: 'ImageWidth', 257: 'ImageLength', 258: 'BitsPerSample', 259: 'Compression', 262: 'PhotometricInterpretation', 263: 'Thresholding', 264: 'CellWidth', 265: 'CellLength'
 from PIL.ExifTags import GPSTAGS
  def get_geotagging(exif):
```

A = np.identity(3)

if not exif:

croppedImage = cv2.warpPerspective(

▼ Extracting Geotags from each image

```
all_files_path = left_files_path[::-1] + right_files_path[1:]
for file1 in all_files_path:
  exif = get_exif(f"{file1}")
  geotags = get_geotagging(exif)
  #print(geotags)
def get_decimal_from_dms(dms, ref):
    degrees = dms[0][0] / dms[0][1]
    minutes = dms[1][0] / dms[1][1] / 60.0
    seconds = dms[2][0] / dms[2][1] / 3600.0
    if ref in ['S', 'W']:
        degrees = -degrees
        minutes = -minutes
        seconds = -seconds
    return round(degrees + minutes + seconds, 5)
def get_coordinates(geotags):
    lat = get_decimal_from_dms(geotags['GPSLatitude'], geotags['GPSLatitudeRef'])
    lon = get_decimal_from_dms(geotags['GPSLongitude'], geotags['GPSLongitudeRef'])
    return (lat,lon)
all_geocoords = []
plt.figure(figsize = (20,10))
for file1 in all_files_path:
  exif = get_exif(f"{file1}")
  geotags = get_geotagging(exif)
  #print(get_coordinates(geotags))
  geocoord = get_coordinates(geotags)
  all_geocoords.append(geocoord)
  plt.scatter(x=geocoord[0], y=geocoord[1])
```

```
+1.0061e2
      0.0080
      0.0075
      0.0070
      0.0065
      0.0060
      0.0055
!pip install gmplot
     Requirement already satisfied: gmplot in /usr/local/lib/python3.7/dist-packages (1.4.1)
     Requirement already satisfied: requests in /usr/local/lib/python3.7/dist-packages (from gmplot) (2.23.0)
     Requirement already satisfied: idna<3,>=2.5 in /usr/local/lib/python3.7/dist-packages (from requests->gmplot) (2.10)
     Requirement already satisfied: urllib3!=1.25.0,!=1.25.1,<1.26,>=1.21.1 in /usr/local/lib/python3.7/dist-packages (from requests->gmplot) (1.24.3)
     Requirement already satisfied: chardet<4,>=3.0.2 in /usr/local/lib/python3.7/dist-packages (from requests->gmplot) (3.0.4)
     Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.7/dist-packages (from requests->gmplot) (2020.12.5)
print(all_geocoords[0])
     (14.06462, 100.61807)
import gmplot
# Create the map plotter:
apikey = 'AIzaSyDSvwNbHrZIKZSwuQkdAgxc6ONmJ_k5Q20' # (your API key here)
gmap = gmplot.GoogleMapPlotter(all_geocoords[0][0], all_geocoords[0][1], 14, apikey=apikey)
attractions = zip(*[
    (37.769901, -122.498331),
    (37.768645, -122.475328),
    (37.771478, -122.468677),
    (37.769867, -122.466102),
    (37.767187, -122.467496),
    (37.770104, -122.470436)
])
gmap.scatter(
    *attractions,
    color=['red', 'orange', 'yellow', 'green', 'blue', 'purple'],
    s=60,
    ew=2,
```

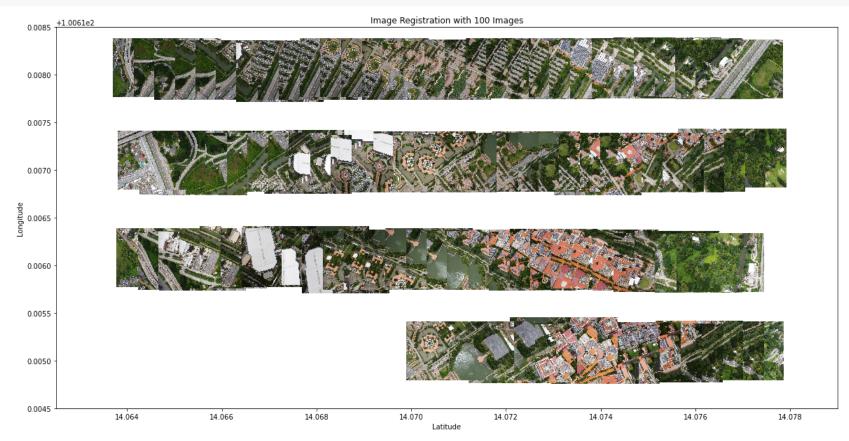
```
marker=[True, True, False, True, False, False],
    symbol=[None, None, 'o', None, 'x', '+'],
    title=['First', 'Second', None, 'Third', None, None],
    label=['A', 'B', 'C', 'D', 'E', 'F']
# Mark a hidden gem:
#gmap.marker(all_geocoords[0][0], all_geocoords[0][1], color='cornflowerblue')
\#len1 = 10
#gmap.scatter(np.array(all_geocoords)[:len1,0], np.array(all_geocoords)[:len1,1], color='#3B0B39', size=40, marker=False)
print(gmap.get())
gmap.draw('drive/MyDrive/map.html')
from matplotlib.offsetbox import OffsetImage, AnnotationBbox
print(len(all_geocoords))
print(len(all_files_path))
     101
     101
print(np.min(np.array(all_geocoords)[:30,1]))
     100.61712
import matplotlib as mpl
mpl.rcParams['figure.dpi'] = 72
print(np.min(np.array(all_geocoords)[:len1,1]))
     100.61506
print(np.max(np.array(all_geocoords)[:len1,1]))
     100.61808
```

Image Registration (100 imgs)

```
fig, ax = plt.subplots()
fig.set_size_inches(20,10)
ax.set_xlabel('Latitude')
ax.set_ylabel('Longitude')
ax.set_ylim(100.6145,100.6185)
len1 = 100
ax.set_title(f'Image Registration with {len1} Images')
ax.set_xlim(14.0625,14.079)
ax.plot(np.array(all_geocoords)[:len1,0], np.array(all_geocoords)[:len1,1],linestyle='None')
```

```
def aerial_images_register(x, y,ax=None):
    ax = ax or plt.gca()
    for count,points in enumerate(zip(x,y)):
        lat,lon = points
        image = plt.imread(all_files_path[count])
        #print(ax.figure.dpi)
        im = OffsetImage(image, zoom=1.5/ax.figure.dpi)
        im.image.axes = ax
        ab = AnnotationBbox(im, (lat,lon), frameon=False, pad=0.0,)
        ax.add_artist(ab)

aerial_images_register(np.array(all_geocoords)[:len1,0], np.array(all_geocoords)[:len1,1], ax=ax)
```



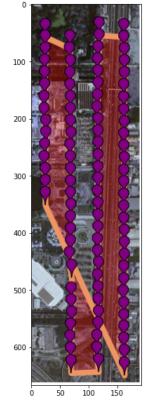
Overlaying on Maps (Incomplete)

```
!pip install gmplot

import gmplot

# Create the map plotter:
apikey = '' # (your API key here) (Hid the Key because it's a private key)
gmap = gmplot.GoogleMapPlotter(all_geocoords[0][0], all_geocoords[0][1], 14, apikey=apikey)
```

```
# Mark a hidden gem:
\verb|#gmap.marker(all\_geocoords[0][0], all\_geocoords[0][1], color='cornflowerblue'|)
len1 = 100
gmap.scatter(np.array(all_geocoords)[:len1,0], np.array(all_geocoords)[:len1,1], color='purple', size=40, marker=[True]*len1)
gmap.polygon(np.array(all_geocoords)[:len1,0],np.array(all_geocoords)[:len1,1], face_color='blue', edge_color='cornflowerblue', edge_width=10)
gmap.draw('drive/MyDrive/map14.html')
fig.savefig('drive/MyDrive/uni_dtset_regist_100_chnged.jpg', dpi=300)
!pip install pyproj
     Collecting pyproj
       Downloading https://files.pythonhosted.org/packages/11/1d/1c54c672c2faf08d28fe78e15d664c048f786225bef95ad87b6c435cf69e/pyproj-3.1.0-cp37-cp37m-manylinux2010_x86_64.whl (6.6MB)
                        6.6MB 1.1MB/s
     Requirement already satisfied: certifi in /usr/local/lib/python3.7/dist-packages (from pyproj) (2020.12.5)
     Installing collected packages: pyproj
     Successfully installed pyproj-3.1.0
import pyproj
from pyproj import Proj
!pip install GDAL
     Requirement already satisfied: GDAL in /usr/local/lib/python3.7/dist-packages (2.2.2)
import gdal
# open the dataset and get the geo transform matrix
ds = gdal.Open((f"{all_files_path[0]}"))
xoff, pix_width, rotatonal, yoff, px_height, rotation_second = ds.GetGeoTransform()
# Describe source image size
x_height = ds.RasterXSize
y_width = ds.RasterYSize
p = pyproj.Proj(proj='utm', zone=47, ellps='WGS84')
lat_file,long_file = get_coordinates(get_geotagging(get_exif(f"{all_files_path[0]}")))
UTM_east, UTM_north = p(long_file, lat_file)
upper_pix = x_height/2
left_pix = y_width/2
print(ds.GetMetadata_Dict())
     {'EXIF_ApertureValue': '(4.97)', 'EXIF_DateTimeOriginal': '2018:09:02 05:23:42', 'EXIF_ExifVersion': '0230', 'EXIF_ExposureBiasValue': '(0)', 'EXIF_ExposureProgram': '4', 'EXIF_ExposureTime': '(0.0005)', 'EXIF_Flash': '16', 'EXIF_FlashEnergy
```



Getting incorrect results when overlaying images onto the geolocations, will fix tomorrow

Reading images and Extracting SuperPoint Keypoints and Descriptors from each image
■ 1. 17 cells hidden
▶ Loading and Initialing the SuperPoint Pretrained Network
[] L, 1 cell hidden
Now Extracting Keypoints and Descriptors from all images and storing them
[] Ļ7 cells hidden
► Image Matching (Robust) through RANSAC and Homography Matrix computation
[] ц 8 cells hidden
Auto-Selection/Ordering of Images (Complete)
[] L, 20 cells hidden
▶ Perspective Transformation b/w consecutive pairs through the computed Homography Matrices
[] L, 6 cells hidden
Final Mosaiced Image (with 24 images)
[] L, 1 cell hidden
> To-Do Tasks
 Seam Removal Improve On this Enhancement Extend to 50 images
[] L, 1 cell hidden

✓ 0s completed at 8:58 PM

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