Exercise 4: Stereo Matching and DLT

Due date: 9/6/2022

In the lectures we discussed creating a depth image from two images and warping images to align them to each other. In this exercise you will implement stereo matching and build a panorama pipeline.

1 Stereo Matching

1.1 SSD

Write a function which takes two images, Left and Right, and finds outputs the disparity map using SSD.

```
def disparitySSD(img_l: np.ndarray, img_r: np.ndarray, disp_range: (int, int), k_size: int)
-> np.ndarray:
```

" " "

img_l: Left image

img_r: Right image

range: Minimum and Maximum disparity range. Ex. (10,80)

k_size: Kernel size for computing the SSD, kernel.shape = (k_size*2+1,k_size*2+1)

return: Disparity map, disp_map.shape = Left.shape
"""

SSD is defined by:

$$S_{LR}(r,c) = \sum_{i=0}^{krows} \sum_{j=0}^{kcols} [(L(i,j) - R(r+i - \frac{krows}{2}, c+j - \frac{kcols}{2})]^2$$

For each pixel(r,c), you should return the shift in x, that gives you the minimum SSD response. Try that on images pair0-L.png and pair0-R.png.

1.2 Normalized Correlation

Write a function which takes two images, Left and Right, and finds outputs the disparity map using Normalized Correlation.

def disparityNC(img_l: np.ndarray, img_r: np.ndarray, disp_range: (int, int), k_size: int)
-> np.ndarray:

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img_l: Left image

img_r: Right image

range: Minimum and Maximum disparity range. Ex. (10,80)

k_size: Kernel size for computing the SSD, kernel.shape = (k_size*2+1,k_size*2+1)

return: Disparity map, disp_map.shape = Left.shape

Normalized Corolation is defined by:

$$R_{LR}(r,c) = \sum_{i=0}^{krows} \sum_{j=0}^{kcols} (L(i,j) \cdot R(r+i - \frac{krows}{2}, c+j - \frac{kcols}{2}))$$

The result is then normalized:

$$NC_{LR} = (r, c) = \frac{R_{LR}(r, c)}{\sqrt{R_{RR}(r, c)R_{LL}(\frac{krows}{2}, \frac{kcols}{2})}}$$

For each pixel(r,c), you should return the shift in x, that gives you the **Maximum** response.

Try that on images pair1-L.png and pair1-R.png.

2 Homography and Warping

2.1 DLT

Write a function which takes four, or more, pairs of matching keypoints and returns the homography matrix H. Calculate the homography using DLT as shown in the presentation (Hint: SVD).

def computeHomography(src_pnt: np.ndarray, dst_pnt: np.ndarray) -> (np.ndarray, float):
 """

Finds the homography matrix, M, that transforms points from src_pnt to dst_pnt.

```
returns the homography and the error between the transformed points to their
      destination (matched) points. Error = np.sqrt(sum((M.dot(src_pnt)-dst_pnt)**2))
      src_pnt: 4+ keypoints locations (x,y) on the original image. Shape:[4+,2]
      dst_pnt: 4+ keypoints locations (x,y) on the destination image. Shape: [4+,2]
      return: (Homography matrix shape: [3,3],
               Homography error)
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Comments:
1. The last number in the M matrix (M[2,2]) should be 1.0!
2. Don't forget to convert the coordinates to homogeneous coordinates for calculating the error.
3. To test, try this:
input:
src_pnt = np.array([[279, 552], [372, 559], [362, 472], [277, 469]])
dst_pnt = np.array([[24, 566], [114, 552], [106, 474], [19, 481]])
output:
M =
[[\ 1.84928168e + 00\ 6.29286954e - 01\ 2.00796141e + 02]
[\ 9.37137905 e{-} 02\ 3.12582854 e{+} 00\ -5.89220129 e{+} 02]
[-5.32787017e-04\ 2.02578161e-03\ 1.000000000e+00]]
```

2.2 Warping

Error: 1.815947868067344

Write a function which takes two images (eg. a poster and a billboard), and warps the poster on to the billboard. The user will mark the source points and their matches on the destination image.

```
def warpImag(src_img: np.ndarray, dst_img: np.ndarray) -> None:
```

Displays both images, and lets the user mark $4\ \mathrm{or}$ more points on each image.

Then transforms the source image onto the destination image and displays the result.

Then calculates the homography and transforms the source image on to the destination image

```
src_img: The image that will be 'pasted' onto the destination image.
    dst_img: The image that the source image will be 'pasted' on.
    output: None.
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Comments:
1. For the clicking, use one for each window (src/dst):
     def onclick_1(event):
         global fig1, fig2, locations_1
         print('onclick_1')
         x = event.xdata
         y = event.ydata
         plt.plot(x, y, '*r')
         plt.show()
         locations_1.append([x, y])
     # display image 1
     fig1 = plt.figure()
     cid = fig1.canvas.mpl\_connect('button\_press\_event', onclick\_1)
     plt.imshow(img1)
2. For the warping look at the example on 'Warping' in week: 4
3. For the stitching (pasting) use a mask.
```

3 Submission

mask = proj_src == 0

canvas = dst_img*mask+(1-mask)*proj_src

Good luck!