Exercise 4: Stereo Matching and DLT

Due date: 7/06/2021

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 50 pt

1.1 SSD 25 pt

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, k_size: int) -> np.ndarray
"""
  img_l: Left image
  img_r: Right image
  range: The Maximum disparity range. Ex. 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:

$$S_{LR}(r,c) = \sum_{i \in P(r,c)} (L_i - R_i)^2$$

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

P(r,c) is the area around the location (r,c))

Try that on images pair0-L.png and pair0-R.png.

1.2 Normalized Correlation 25 pt

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, k_size: int) -> np.ndarray
    """
    img_l: Left image
    img_r: Right image
    range: The Maximum disparity range. Ex. 80
    k_size: Kernel size for computing the NormCorolation, kernel.shape = (k_size*2+1,k_size*2+1)
    return: Disparity map, disp_map.shape = Left.shape
    """
```

Normalized Corolation is defined by:

$$NC_{LR}(r,c) = \frac{1}{N} \sum_{i \in P(r,c)} \frac{(L_i - \bar{L})(R_i - \bar{R})}{\sigma_L \sigma_R}$$

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 50 pt

2.1 DLT 10 pt

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:n) -> (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 destenation image. Shape:[4+,2]
```

```
return: (Homography matrix shape:[3,3],
Homography error)
```

Comments:

11 11 11

- 1. The bottom left 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 = \begin{bmatrix} -0.0026660762 & -0.00010110695 & 0.7429984733 \\ -0.0013069813 & -0.00297205775 & 0.6692794374 \\ -0.000003331 & -0.00000113885 & -0.0008023928 \end{bmatrix}
```

MAE: 5.993266638223986e-11

2.2 Warping 40 pt

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 or more points on each image. Then calculate

```
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.
```

Comments:

- 1. To make things easy, the src image will be projected fully on the dst image (like the billboard example in class)
- 2. For the warping look at the example on 'Warping'
- 3. For the stitching (pasting) use a mask.

```
mask = proj_src==0
canvas = dst_img*mask+(1-mask)*proj_src
```

3 Submission

Good luck and enjoy the exercise!

