assignment1

September 20, 2021

1 Common imports

```
[1]: %matplotlib inline
  import os
  import sys
  import glob
  import re
  import numpy as np
  import matplotlib
  import matplotlib.pyplot as plt
  from mpl_toolkits.mplot3d import Axes3D
  from PIL import Image
```

2 Provided functions

2.0.1 Image loading and saving

```
[2]: def LoadFaceImages(pathname, subject_name, num_images):
         Load the set of face images.
         The routine returns
             ambimage: image illuminated under the ambient lighting
             imarray: a 3-D array of images, h x w x Nimages
             lightdirs: Nimages x 3 array of light source directions
         11 11 11
         def load_image(fname):
             return np.asarray(Image.open(fname))
         def fname_to_ang(fname):
             yale_name = os.path.basename(fname)
             return int(yale_name[12:16]), int(yale_name[17:20])
         def sph2cart(az, el, r):
             rcos_theta = r * np.cos(el)
             x = rcos theta * np.cos(az)
             y = rcos_theta * np.sin(az)
```

```
z = r * np.sin(el)
             return x, y, z
         ambimage = load_image(
             os.path.join(pathname, subject_name + '_POO_Ambient.pgm'))
         im_list = glob.glob(os.path.join(pathname, subject_name + '_POOA*.pgm'))
         if num images <= len(im list):</pre>
             im_sub_list = np.random.choice(im_list, num_images, replace=False)
         else:
             print(
                  'Total available images is less than specified.\nProceeding with \( \dagger d_{1} \)
      \rightarrowimages.\n'
                  % len(im_list))
             im_sub_list = im_list
         # filter image
         # filters = ["yaleB07_P00A-110E+40.pgm", "yaleB07_P00A+130E+20","
      \rightarrow "yaleB07_P00A+110E+65", "yaleB07_P00A-120E+00.pqm", "yaleB07_P00A+095E+00.
      →pgm", "yaleB07_P00A+110E+15.pgm", "yaleB07_P00A-110E-20.pgm", "
      → "yaleB07 P00A-085E-20.pgm", "yaleB07 P00A-110E+15.pgm",
      → "yaleB07_P00A+085E+20.pgm", "yaleB07_P00A-095E+00.pgm", "
      → "yaleB07 P00A+085E-20.pqm", "yaleB07 P00A+000E+90.pqm",
      → "yaleB07 P00A+110E-20.pgm", "yaleB07 P00A+110E+40.pgm",
      \rightarrow "yaleB07\_P00A+120E+00.pgm", "yaleB07\_P00A-110E+65.pgm", \square 
      → "yaleB07_P00A-130E+20.pgm", ]
         filters = ["yaleB05_P00A-095E+00.pgm", "yaleB05_P00A-110E+15.pgm", "

¬"yaleB05_P00A+110E-20.pgm", "yaleB05_P00A+000E+90.pgm",

¬"yaleB05 P00A+110E+40.pgm", "yaleB05 P00A-130E+20.pgm", "]

¬"yaleB05_P00A-110E+65.pgm", "yaleB05_P00A+120E+00.pgm",

      _{\hookrightarrow} "yaleB05_P00A-110E+40.pgm", "yaleB05_P00A-120E+00.pgm", _{\sqcup}
      →"yaleB05_P00A+110E+65.pgm", "yaleB05_P00A+130E+20.pgm", "

¬"yaleB05_P00A+110E+15.pgm", "yaleB05_P00A-110E-20.pgm",]

         filtered_list = []
         for img in im_sub_list:
             if not os.path.basename(img) in filters:
                  filtered_list.append(img)
         im_sub_list = filtered_list
         im_sub_list.sort()
         imarray = np.stack([load_image(fname) for fname in im_sub_list], axis=-1)
         Ang = np.array([fname_to_ang(fname) for fname in im_sub_list])
         x, y, z = sph2cart(Ang[:, 0] / 180.0 * np.pi, Ang[:, 1] / 180.0 * np.pi, 1)
         lightdirs = np.stack([y, z, x], axis=-1)
         return ambimage, imarray, lightdirs
[3]: def save_outputs(subject_name, albedo_image, surface_normals):
```

```
im = Image.fromarray((albedo_image*255).astype(np.uint8))
im.save("%s_albedo.jpg" % subject_name)
```

```
im = Image.fromarray((surface_normals[:,:,0]*128+128).astype(np.uint8))
im.save("%s_normals_x.jpg" % subject_name)
im = Image.fromarray((surface_normals[:,:,1]*128+128).astype(np.uint8))
im.save("%s_normals_y.jpg" % subject_name)
im = Image.fromarray((surface_normals[:,:,2]*128+128).astype(np.uint8))
im.save("%s_normals_z.jpg" % subject_name)
```

2.0.2 Plot the height map

```
[4]: def set_aspect_equal_3d(ax):
         """https://stackoverflow.com/questions/13685386"""
         """Fix equal aspect bug for 3D plots."""
         xlim = ax.get_xlim3d()
         ylim = ax.get_ylim3d()
         zlim = ax.get_zlim3d()
         from numpy import mean
         xmean = mean(xlim)
         ymean = mean(ylim)
         zmean = mean(zlim)
         plot_radius = max([
             abs(lim - mean_)
             for lims, mean_ in ((xlim, xmean), (ylim, ymean), (zlim, zmean))
             for lim in lims
         ])
         ax.set_xlim3d([xmean - plot_radius, xmean + plot_radius])
         ax.set_ylim3d([ymean - plot_radius, ymean + plot_radius])
         ax.set_zlim3d([zmean - plot_radius, zmean + plot_radius])
     def display_output(albedo_image, height_map):
         fig = plt.figure()
         plt.imshow(albedo_image, cmap='gray')
         plt.axis('off')
         fig = plt.figure(figsize=(10, 10))
         ax = fig.gca(projection='3d')
         ax.view init(40, 20)
         X = np.arange(albedo_image.shape[0])
         Y = np.arange(albedo_image.shape[1])
         X, Y = np.meshgrid(Y, X)
         H = np.flipud(np.fliplr(height_map))
         A = np.flipud(np.fliplr(albedo_image))
         A = np.stack([A, A, A], axis=-1)
         ax.xaxis.set_ticks([])
         ax.xaxis.set_label_text('Z')
         ax.yaxis.set_ticks([])
         ax.yaxis.set_label_text('X')
```

```
ax.zaxis.set_ticks([])
ax.yaxis.set_label_text('Y')
surf = ax.plot_surface(
    H, X, Y, cmap='gray', facecolors=A, linewidth=0, antialiased=False)
set_aspect_equal_3d(ax)
```

2.0.3 Plot the surface norms.

```
[5]: def plot_surface_normals(surface_normals):
         surface\_normals: h x w x 3 matrix.
         11 11 11
         fig = plt.figure()
         ax = plt.subplot(1, 3, 1)
         ax.axis('off')
         ax.set_title('X')
         im = ax.imshow(surface_normals[:,:,0])
         ax = plt.subplot(1, 3, 2)
         ax.axis('off')
         ax.set_title('Y')
         im = ax.imshow(surface_normals[:,:,1])
         ax = plt.subplot(1, 3, 3)
         ax.axis('off')
         ax.set title('Z')
         im = ax.imshow(surface_normals[:,:,2])
```

3 Your implementation

```
[6]: def preprocess(ambimage, imarray):
         HHHH
         preprocess the data:
             1. subtract ambient image from each image in imarray.
             2. make sure no pixel is less than zero.
             3. rescale values in imarray to be between 0 and 1.
         Inputs:
             ambimage: h x w
             imarray: h x w x Nimages
         Outputs:
             processed_imarray: h x w x Nimages
         processed_imarray = imarray - np.expand_dims(ambimage, axis=-1)
         # remove 0
         processed_imarray = np.clip(processed_imarray, 0, 255)
         # rescale
         processed_imarray = processed_imarray / 255
         return processed_imarray
```

```
[40]: def photometric_stereo(imarray, light_dirs):
           Inputs:
               imarray: h x w x Nimages
               light_dirs: Nimages x 3
           Outputs:
               albedo_image: h x w
               surface_norms: h x w x 3
           imarray_shape = imarray.shape
           # reshape imarray to (hw, Nimages)
           imarray = np.reshape(imarray, (-1, imarray_shape[-1]))
           # compute inverse of V
           # inv_V = np.linalq.pinv(light_dirs)
           # compute q for each pixel (hw, 3)
           # q = np.matmul(inv_V, imarray.transpose()) # (3, hw)
           g = np.linalg.lstsq(light_dirs, imarray.transpose())[0] # as website
           g = np.reshape(g.transpose(), (imarray_shape[0], imarray_shape[1], 3))
           # compute norm
           fx = np.true_divide(g[:, :, 0], g[:, :, 2], where=g[:,:,2]!=0)
           fy = np.true_divide(g[:, :, 1], g[:, :, 2], where=g[:,:,2]!=0)
           norm = 1/np.sqrt(fx**2+fy**2+1)
           surface_normals = np.stack([fx, fy, np.ones_like(fx)], axis=-1)
           surface_normals = surface_normals * np.expand_dims(norm, axis=-1)
           # compute albedo
           albedo image = np.linalg.norm(g, axis=-1)
           # check
           # print(surface_normals * np.expand_dims(albedo_image, axis=-1)-g)
           return albedo_image, surface_normals
[107]: def get_surface(surface_normals, integration_method):
           11 11 11
           Inputs:
               surface normals:h x w x 3
               integration_method: string in ['average', 'column', 'row', 'random']
           Outputs:
               height_map: h x w
           def _row_integral(fx, fy):
               tmp = fy.copy()
               tmp[0, :] = np.cumsum(fx[0, :])
               hm = np.cumsum(tmp, axis=0)
               return hm
           def _col_integral(fx, fy):
               tmp = fx.copy()
               tmp[:, 0] = np.cumsum(fy[:, 0])
```

```
hm = np.cumsum(tmp, axis=1)
       return hm
   def random_integral(inputs):
       fx, fy = inputs
       hm = np.zeros_like(fx)
       h, w = fx.shape
       for i in range(h):
           for j in range(w):
               steps = ['v'] * i + ['h'] * j
               # randomize
               np.random.shuffle(steps)
               xy_sum = fx[0][0] + fy[0][0] # initialize
               x, y = 0, 0
               for s in steps:
                   if s == "v":
                       y += 1
                       xy_sum += fy[y][x]
                   elif s == "h":
                       x += 1
                       xy_sum += fx[y][x]
               hm[i][j] = xy_sum
       return hm
   fx = surface_normals[:, :, 0]
   fy = surface_normals[:, :, 1]
   if integration_method == 'column':
       height_map = _row_integral(fx, fy)
   elif integration_method == "row":
       height_map = _col_integral(fx, fy)
   elif integration_method == 'average':
       hm_row = _row_integral(fx, fy)
       hm_col = _row_integral(fx, fy)
       height_map = (hm_row + hm_col) / 2
   elif integration_method == 'random':
       # from multiprocessing import Pool
       # pool = Pool(8)
       num_trials = 10
       height maps = np.zeros like(fx)
       for _ in range(num_trials):
           height maps += random integral([fx, fy])
       # height_maps = pool.map(random_integral, [(fx, fy) for _ in_{\bot}])
→range(num_trials)])
       # pool.close()
       # pool.join()
       # height_map = np.sum(height_maps, axis=0) / num_trials
       height_map = height_maps / num_trials
```

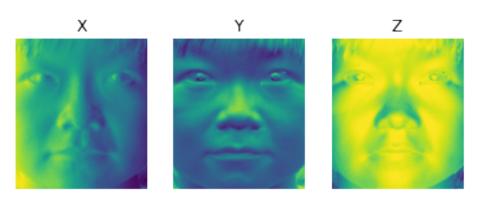
4 Main function

(192, 168) (192, 168, 50) (50, 3)

/Users/zongfan/opt/miniconda3/envs/pytorch/lib/python3.7/site-packages/ipykernel_launcher.py:17: FutureWarning: `rcond` parameter will change to the default of machine precision times ``max(M, N)`` where M and N are the input matrix dimensions.

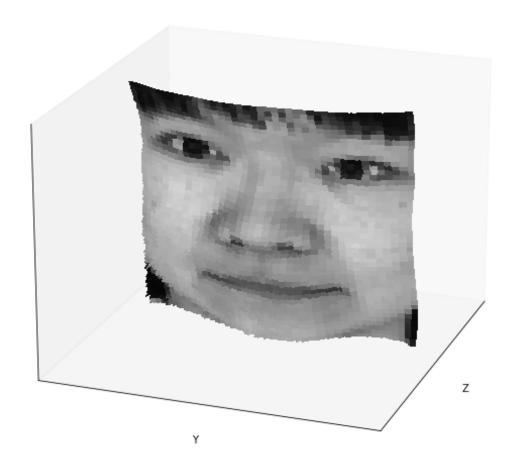
To use the future default and silence this warning we advise to pass `rcond=None`, to keep using the old, explicitly pass `rcond=-1`.

[273]: plot_surface_normals(surface_normals)



[109]: display_output(albedo_image, height_map)





```
[125]: # Phantom 3D model data to test the accuracy of surface normal estimation
def LoadDiLiGenTImages(full_path, subject_name, num_images=96):
    # crop background
    h_up, h_bottom = 70, 380
    w_left, w_right = 200, 500
    def load_image(fname, ints):
        image = np.asarray(Image.open(fname))
        # each color channel should be normalized with intensity
        ints = np.array([[ints]])
        image = image / ints
        image = image / ints
        image = image.astype(np.uint8)
        gray = np.asarray(Image.fromarray(image).convert("L"))
        # gray = np.asarray(Image.open(fname).convert("L"))
```

```
gray = gray[h_up:h_bottom, w_left:w_right]
       return gray
   def load_light_direction_intensity(fname):
       data = []
       with open(fname, "r") as f:
           for 1 in f:
               dirs = [float(f) for f in l.strip().split(" ")]
               data.append(dirs)
       f.close()
       return np.array(data)
   def load normal(mat file):
       from scipy.io import loadmat
       data = loadmat(mat file)
       normal = data["Normal_gt"]
       normal = normal[h_up:h_bottom, w_left:w_right]
       normal[:, :, 2] = 1
       return normal
   # load intensity for normalizing image values
   ints_file = os.path.join(full_path, subject_name, "light_intensities.txt")
   lightints = load_light_direction_intensity(ints_file)
   # load mask which is used to mask out non-object regions
   mask_file = os.path.join(full_path, subject_name, "mask.png")
   mask = np.asarray(Image.open(mask file).convert("L")) / 255
   mask = mask[h_up:h_bottom, w_left:w_right]
   # load images
   im_list = [os.path.join(full_path, subject_name, "{:03d}.png".format(i))__
→for i in range(1, num_images+1)]
   imarray = np.stack([load_image(im_list[i], lightints[i])*mask for i in_
→range(num images)], axis=-1)
   # load directions
   dir file = os.path.join(full path, subject name, "light directions.txt")
   light_dirs = load_light_direction_intensity(dir_file)
   # change xyz to yzx
   # light_dirs_cp = light_dirs.copy()
   # light_dirs = light_dirs_cp[:, [0, 2, 1]]
   # load normal_qt
   gt_file = os.path.join(full_path, subject_name, "Normal_gt.mat")
   normal_gt = load_normal(gt_file)
   # set ambient bg as black
   img_h, img_w, _ = imarray.shape
   ambient = np.zeros((img_h, img_w)).astype(np.float)
   return ambient, imarray, light_dirs, normal_gt
```

```
[126]: root_path = '/Users/zongfan/Downloads/DiLiGenT/pmsData'
       subject_name = 'catPNG'
       integration_method = 'average'
       save_flag = True
       ambient_image, imarray, light_dirs, normal_gt = LoadDiLiGenTImages(root_path,__
       ⇒subject name, 96)
       # print(imarray.shape, light_dirs.shape, normal_gt.shape)
       processed_imarray = preprocess(ambient_image, imarray)
       albedo_image, surface_normals = photometric_stereo(processed_imarray,
                                                          light_dirs)
       # compute loss between prediction and gt with L2
       diff = np.square(surface_normals - normal_gt).sum()
       print("squared difference between predicted Normal and GT Normal: ", diff)
       height_map = get_surface(surface_normals, integration_method)
       if save flag:
           save_outputs(subject_name, albedo_image, surface_normals)
       plot_surface_normals(surface_normals)
       display_output(albedo_image, height_map)
```

/Users/zongfan/opt/miniconda3/envs/pytorch/lib/python3.7/site-packages/ipykernel_launcher.py:17: FutureWarning: `rcond` parameter will change to the default of machine precision times ``max(M, N)`` where M and N are the input matrix dimensions.

To use the future default and silence this warning we advise to pass `rcond=None`, to keep using the old, explicitly pass `rcond=-1`.

squared difference between predicted Normal and GT Normal: 6921.343864143864





