CS 445: Computational Photography

Programming Project #3: Gradient Domain Fusion

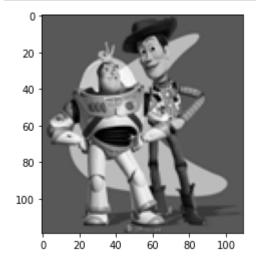
```
In [ ]: from google.colab import drive
        drive.mount('/content/drive')
In [1]: import cv2
        import numpy as np
        import matplotlib.pyplot as plt
        %matplotlib inline
        import os
        from random import random
        import time
        import scipy
        import scipy.sparse.linalg
        # modify to where you store your project data including utils.py
        datadir = "C:/Users/jackt/Desktop/CS445/cs445fusion"
        #utilfn = datadir + "utils.py"
        #!cp "$utilfn" .
        samplesfn = datadir + "samples"
        #!cp -r "$samplesfn" .
        import utils
```

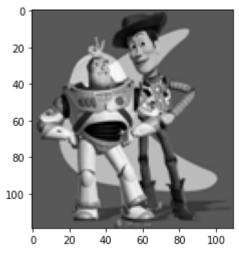
Part 1 Toy Problem (20 pts)

```
In [2]: def toy_reconstruct(img):
            The implementation for gradient domain processing is not complicated, but
         it is easy to make a mistake,
            so let's start with a toy example. Reconstruct this image from its gradien
        t values, plus one pixel intensity.
            Denote the intensity of the source image at (x, y) as s(x, y) and the value
        to solve for as v(x,y).
            For each pixel, then, we have two objectives:
            1. minimize (v(x+1,y)-v(x,y) - (s(x+1,y)-s(x,y)))^2
            2. minimize (v(x,y+1)-v(x,y) - (s(x,y+1)-s(x,y)))^2
            Note that these could be solved while adding any constant value to v, so w
        e will add one more objective:
            3. minimize (v(1,1)-s(1,1))^2
             :param toy_img: numpy.ndarray
            im h, im w = img.shape
            im2var = np.arange(im h * im w).reshape(im h, im w)
            neq = (im_h - 1) * im_w + (im_w - 1) * im_h + 1
            # construct equation matrix A, solution vector b
            A = scipy.sparse.lil_matrix((neq, im_h*im_w), dtype='double')
            b = np.zeros((neq, 1), dtype='double')
            for e in range(neq):
                 if e < (im_h-1) * im_w:</pre>
                     # vertical edges
                     x = e \% im_w
                     y = e // im_w
                     A[e, im2var[y+1][x]] = 1
                     A[e, im2var[y][x]] = -1
                     b[e] = img[y+1][x] - img[y][x]
                 elif e < neq-1:</pre>
                     # horizontal edges
                     temp = e - ((im_h-1)*im_w)
                     y = temp % im_h
                     x = temp // im h
                     A[e, im2var[y][x+1]] = 1
                     A[e, im2var[y][x]] = -1
                     b[e] = img[y][x+1] - img[y][x]
                 else:
                     # objective 3
                     x = im w - 1
                     y = im h - 1
                     A[e, im2var[y][x]] = 1
                     b[e] = img[y][x]
            # solve least squares system
            x = scipy.sparse.linalg.lsqr(A.tocsr(), b)
            return x[0].reshape(im h, im w)
```

```
In [3]: toy_img = cv2.cvtColor(cv2.imread('samples/toy_problem.png'), cv2.COLOR_BGR2GR
AY).astype('double') / 255.0
plt.imshow(toy_img, cmap="gray")
plt.show()

im_out = toy_reconstruct(toy_img)
plt.imshow(im_out, cmap="gray")
plt.show()
print("Max error is: ", np.sqrt(((im_out - toy_img)**2).max()))
```

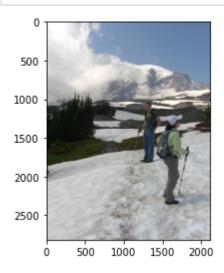


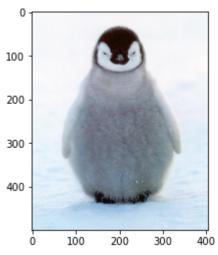


Max error is: 7.5332676048867064e-06

Preparation

```
In [4]:
        background img = cv2.cvtColor(cv2.imread('samples/im2.JPG'), cv2.COLOR BGR2RGB
        ).astype('double') / 255.0
        plt.figure()
        plt.imshow(background img)
        plt.show()
        object_img = cv2.cvtColor(cv2.imread('samples/penguin-chick.jpeg'), cv2.COLOR_
        BGR2RGB).astype('double') / 255.0
        plt.imshow(object_img)
        plt.show()
        use_interface = True # set to true if you want to use the interface to choose
        points (might not work in Colab)
        if not use_interface:
          xs = (65, 359, 359, 65)
          ys = (24, 24, 457, 457)
          object_mask = utils.get_mask(ys, xs, object_img)
          bottom_center = (500, 2500) \# (x,y)
          object_img, object_mask = utils.crop_object_img(object_img, object_mask)
          bg_ul = utils.upper_left_background_rc(object_mask, bottom_center)
          plt.imshow(utils.get_combined_img(background_img, object_img, object_mask, b
        g_ul))
```





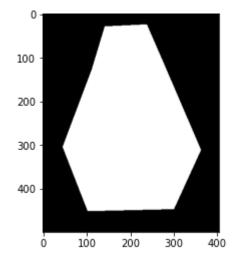
```
In [6]: if use_interface:
    import matplotlib.pyplot as plt
    %matplotlib notebook
    mask_coords = utils.specify_mask(object_img)
```

If it doesn't get you to the drawing mode, then rerun this function again.



```
In [7]: if use_interface:
    xs = mask_coords[0]
    ys = mask_coords[1]
    %matplotlib inline
    import matplotlib.pyplot as plt
    plt.figure()
    object_mask = utils.get_mask(ys, xs, object_img)
```

<Figure size 432x288 with 0 Axes>

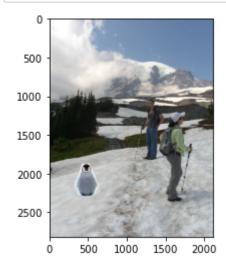


If it doesn't get you to the drawing mode, then rerun this function again. Al so, make sure the object fill fit into the background image. Otherwise it wil l crash



```
In [9]: if use_interface:
    %matplotlib inline
    import matplotlib.pyplot as plt

    object_img, object_mask = utils.crop_object_img(object_img, object_mask)
    bg_ul = utils.upper_left_background_rc(object_mask, bottom_center)
    plt.imshow(utils.get_combined_img(background_img, object_img, object_mask, b
    g_ul))
```



Part 2 Poisson Blending (50 pts)

```
In [10]: def combine_img_bw(bg_img, object_img, object_mask, bg_ul):
    """Same as utils.combine_img, but only intensity channel"""
    combined_img = bg_img.copy()
    (nr, nc) = object_img.shape

    combined_patch = combined_img[bg_ul[0]:bg_ul[0]+nr, bg_ul[1]:bg_ul[1]+nc]
    combined_patch = combined_patch*(1-object_mask) + object_img[:,:]*object_m
    ask
    combined_img[bg_ul[0]:bg_ul[0]+nr, bg_ul[1]:bg_ul[1]+nc] = combined_patch
    return combined_img
```

```
In [11]:
         def poisson_blend(object_img, object_mask, bg_img, bg_ul):
             Returns a Poisson blended image with masked object imag over the bg imag at
          position specified by bg ul.
             Can be implemented to operate on a single channel or multiple channels
              :param object_img: the image containing the foreground object
              :param object_mask: the mask of the foreground object in object_img
              :param background_img: the background image
             :param bg_ul: position (row, col) in background image corresponding to (0,
         0) of object_img
             im_h, im_w = object_img.shape
             im2var = np.arange(im_h * im_w).reshape(im_h, im_w)
             # find neg
             neq = 0
             for y in range(im_h-1):
                 for x in range(im_w-1):
                      if object_mask[y][x] == 1:
                          neq += 2
                          continue
                      if object_mask[y+1][x] == 1:
                          neq += 1
                      if object_mask[y][x+1] == 1:
                          neq += 1
             # construct equation matrix A, solution vector b
             A = scipy.sparse.lil_matrix((neq, im_h*im_w), dtype='double')
             b = np.zeros((neq, 1), dtype='double')
             count = 0
             for y in range(im_h-1):
                 for x in range(im w-1):
                      # right
                      if object_mask[y][x] == 1:
                          if object_mask[y][x+1] == 1:
                              A[count, im2var[y][x]] = 1
                              A[count, im2var[y][x+1]] = -1
                              b[count] = object img[y][x] - object img[y][x+1]
                              count += 1
                          else:
                              A[count, im2var[y][x]] = 1
                              b[count] = object_img[y][x] - object_img[y][x+1] + bg_img[
         bg_ul[0]+y][bg_ul[1]+x+1]
                              count += 1
                      else:
                          if object_mask[y][x+1] == 1:
                              A[count, im2var[y][x+1]] = 1
                              b[count] = object_img[y][x+1] - object_img[y][x] + bg_img[
         bg_ul[0]+y][bg_ul[1]+x]
                              count += 1
                      # down
                      if object_mask[y][x] == 1:
                          if object_mask[y+1][x] == 1:
                              A[count, im2var[y][x]] = 1
                              A[count, im2var[y+1][x]] = -1
```

```
b[count] = object_img[y][x] - object_img[y+1][x]
                    count += 1
                else:
                    A[count, im2var[y][x]] = 1
                    b[count] = object_img[y][x] - object_img[y+1][x] + bg_img[
bg_ul[0]+y+1][bg_ul[1]+x]
                    count += 1
            else:
                if object_mask[y+1][x] == 1:
                    A[count, im2var[y+1][x]] = 1
                    b[count] = object\_img[y+1][x] - object\_img[y][x] + bg\_img[
bg_ul[0]+y][bg_ul[1]+x]
                    count += 1
            if count >= neq:
                break
        if count >= neq:
            break
   # solve least squares system
   v = scipy.sparse.linalg.lsqr(A.tocsr(), b)
   blend = v[0].reshape(im_h, im_w)
    #plt.imshow(blend)
   #plt.show()
   return combine_img_bw(bg_img, blend, object_mask, bg_ul)
```

```
In [12]: im_blend = np.zeros(background_img.shape)
    for b in np.arange(3):
        im_blend[:,:,b] = poisson_blend(object_img[:,:,b], object_mask, background_i
        mg[:,:,b].copy(), bg_ul)

    plt.figure(figsize=(15,15))
    plt.imshow(im_blend)
```

Clipping input data to the valid range for imshow with RGB data ([0..1] for f loats or [0..255] for integers).

Out[12]: <matplotlib.image.AxesImage at 0x1fc15924ef0>



Part 3 Mixed Gradients (20 pts)

```
In [13]:
         def mixed_blend(object_img, object_mask, bg_img, bg_ul):
             Returns a mixed gradient blended image with masked object_img over the bg_
         img at position specified by bg ul.
             Can be implemented to operate on a single channel or multiple channels
              :param object_img: the image containing the foreground object
              :param object_mask: the mask of the foreground object in object_img
              :param background_img: the background image
             :param bg_ul: position (row, col) in background image corresponding to (0,
         0) of object_img
             im_h, im_w = object_img.shape
             im2var = np.arange(im_h * im_w).reshape(im_h, im_w)
             # find neg
             neq = 0
             for y in range(im_h-1):
                 for x in range(im_w-1):
                      if object_mask[y][x] == 1:
                          neq += 2
                          continue
                      if object_mask[y+1][x] == 1:
                          neq += 1
                      if object_mask[y][x+1] == 1:
                          neq += 1
             # construct equation matrix A, solution vector b
             A = scipy.sparse.lil_matrix((neq, im_h*im_w), dtype='double')
             b = np.zeros((neq, 1), dtype='double')
             count = 0
             for y in range(im_h-1):
                 for x in range(im w-1):
                      # right
                      if object_mask[y][x] == 1:
                          if object_mask[y][x+1] == 1:
                              A[count, im2var[y][x]] = 1
                              A[count, im2var[y][x+1]] = -1
                              grad s = object img[y][x] - object img[y][x+1]
                              grad_t = bg_img[bg_ul[0]+y][bg_ul[1]+x] - bg_img[bg_ul[0]+y]
         y][bg_ul[1]+x+1]
                              b[count] = grad_s if abs(grad_s) > abs(grad_t) else grad_t
                              count += 1
                          else:
                              A[count, im2var[y][x]] = 1
                              b[count] = object_img[y][x] - object_img[y][x+1] + bg_img[
         bg_ul[0]+y][bg_ul[1]+x+1]
                              count += 1
                      else:
                          if object_mask[y][x+1] == 1:
                              A[count, im2var[y][x+1]] = 1
                              b[count] = object_img[y][x+1] - object_img[y][x] + bg_img[
         bg_ul[0]+y][bg_ul[1]+x]
                              count += 1
                      # down
                      if object_mask[y][x] == 1:
```

```
if object_mask[y+1][x] == 1:
                    A[count, im2var[y][x]] = 1
                    A[count, im2var[y+1][x]] = -1
                    b[count] = object_img[y][x] - object_img[y+1][x]
                    count += 1
                else:
                    A[count, im2var[y][x]] = 1
                    b[count] = object_img[y][x] - object_img[y+1][x] + bg_img[
bg_ul[0]+y+1][bg_ul[1]+x]
                    count += 1
            else:
                if object_mask[y+1][x] == 1:
                    A[count, im2var[y+1][x]] = 1
                    b[count] = object_img[y+1][x] - object_img[y][x] + bg_img[
bg_ul[0]+y][bg_ul[1]+x]
                    count += 1
            if count >= neq:
                break
        if count >= neq:
            break
   # solve least squares system
   v = scipy.sparse.linalg.lsqr(A.tocsr(), b)
   blend = v[0].reshape(im_h, im_w)
    #plt.imshow(blend)
    #plt.show()
    return combine_img_bw(bg_img, blend, object_mask, bg_ul)
```

```
In [14]: im_mix = np.zeros(background_img.shape)
    for b in np.arange(3):
        im_mix[:,:,b] = mixed_blend(object_img[:,:,b], object_mask, background_img
        [:,:,b].copy(), bg_ul)

    plt.figure(figsize=(15,15))
    plt.imshow(im_mix)
```

Clipping input data to the valid range for imshow with RGB data ([0..1] for f loats or [0..255] for integers).

Out[14]: <matplotlib.image.AxesImage at 0x1fc15778978>



Bells & Whistles (Extra Points)

Color2Gray (20 pts)

```
In [ ]: def color2gray(img):
    pass
```

Laplacian pyramid blending (20 pts)

```
In [ ]: def laplacian_blend(object_img, object_mask, bg_img, bg_ul):
    # feel free to change input parameters
    pass
```

More gradient domain processing (up to 20 pts)

In []:	