Project2

February 28, 2020

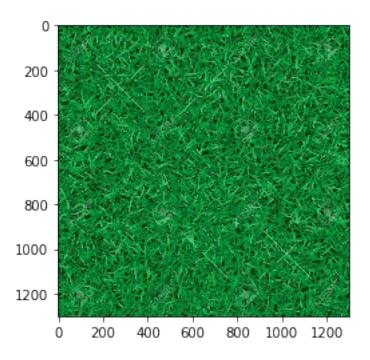
1 Programming Project #2: Image Quilting

1.1 CS445: Computational Photography - Fall 2019

```
[]: import cv2
import numpy as np
import matplotlib.pyplot as plt
%matplotlib notebook
import utils
import os
[2]: from utils import cut # default cut function for seam finding section
```

1.1.1 Part I: Randomly Sampled Texture (10 pts)

```
[3]: # sample_img_dir = 'samples/bricks_small.jpg' # feel free to change sample_img_dir = 'samples/ggrass.jpg' # feel free to change sample_img = None if os.path.exists(sample_img_dir):
    sample_img = cv2.imread(sample_img_dir)
    plt.imshow(sample_img)
```



```
[4]: def quilt_random(sample, out_size, patch_size):
        Randomly samples square patches of size patchsize from sample in order to_{\sqcup}
     ⇒create an output image of size outsize.
        :param sample: numpy.ndarray
                                       The image you read from sample directory
                                       The width of the square output image
        :param out_size: int
        :param patch_size: int
                                        The width of the square sample patch
        :return: numpy.ndarray
        11 11 11
        if(patch_size > sample.shape[0] or patch_size > sample.shape[1]):
            print("patch size is bigger than sample image size. fix it")
            return
        # zero initialize our resultant image with int type
        res = np.zeros((out_size, out_size, 3), np.uint8)
        # print("res type and shape ", type(res), res.shape)
        # print("out_size and patch_size ", out_size, patch_size)
        # loop in the steps of patch_size row, coloumns
        # borders that doesn't fit in the patch_size are left untouched
        for i in range(0, out_size, patch_size):
            if(i+patch_size > out_size): break
            for j in range(0, out_size, patch_size):
                if (j+patch_size > out_size): break
                noOfrow = sample.shape[0]
```

```
noOfcol = sample.shape[1]

# get a random row number between 0 to noOfrow in the sample_img

# print("error here ", noOfrow, patch_size, noOfrow - patch_size)

rowrandom = np.random.randint(low = 0, high = noOfrow - patch_size, u

⇒size = 1)[0]

# get a random col number between 0 to noOfcol in the sample_img

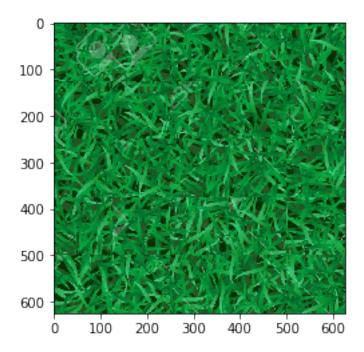
colrandom = np.random.randint(low = 0, high = noOfcol - patch_size, u

⇒size = 1)[0]

res[i:i+patch_size, j:j+patch_size] = sample[rowrandom:

⇒rowrandom+patch_size, colrandom:colrandom+patch_size]
```

```
[5]: # while choosing out_size should be a multiple of patch_size, otherwise
# we see black area in the edges.
# Also patch_size should be less than the image sizes;
out_size = 625 # feel free to change to debug
patch_size = 125 # feel free to change to debug
res = quilt_random(sample_img, out_size, patch_size)
if res is not None:
    plt.imshow(res)
```

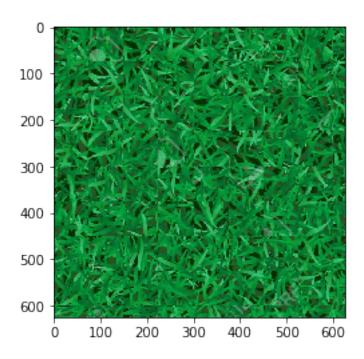


1.1.2 Part II: Overlapping Patches (30 pts)

```
[6]: import pdb
        import random
        import sys
        \# T = Template, M = Mask, I = Image
        \# ssd = ((M*T)**2).sum() - 2 * cv2.filter2D(I, ddepth=-1, kernel = M*T) +
                         cv2.filter2D(I ** 2, ddepth=-1, kernel=M)
        def ssd_patch(T, M, I):
                noOfRowI = I.shape[0]
                noOfColI = I.shape[1]
                channels = I.shape[2] # 0, 1, 2
                ssd_0 = np.zeros((noOfRowI, noOfColI), np.uint8)
                ssd_1 = np.zeros((noOfRowI, noOfColI), np.uint8)
                ssd_2 = np.zeros((noOfRowI, noOfColI), np.uint8)
                ssd_0 = ((M*T[:,:,0])**2).sum() - 2 * cv2.filter2D(I[:,:,0], ddepth=-1,_\textsuperscript{\substack}{\substack}) = ((M*T[:,:,0])**2).sum() - 2 * cv2.filter2D(I[:,:,0], ddepth=-1,_\textsuperscript{\substack}{\substack}{\substack}) = ((M*T[:,:,0])**2).sum() - 2 * cv2.filter2D(I[:,:,0], ddepth=-1,_\textsuperscript{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\substack}{\subs
          →kernel = M*T[:,:,0]) + cv2.filter2D(I[:,:,0] ** 2, ddepth=-1, kernel=M)
                ssd_1 = ((M*T[:,:,1])**2).sum() - 2 * cv2.filter2D(I[:,:,1], ddepth=-1,_u
          →kernel = M*T[:,:,1]) + cv2.filter2D(I[:,:,1] ** 2, ddepth=-1, kernel=M)
                ssd_2 = ((M*T[:,:,2])**2).sum() - 2 * cv2.filter2D(I[:,:,2], ddepth=-1,__
          \rightarrowkernel = M*T[:,:,2]) + cv2.filter2D(I[:,:,2] ** 2, ddepth=-1, kernel=M)
                return ssd_0 + ssd_1 + ssd_2
        # input a cost image (each pixel's value is the cost
        # of selecting the patch centered at that pixel)
        # find the minimum cost minc and then to sample a patch
        # within a % of that value: row, col = np.where(cost < minc*(1+tol))
        # return a randomly sampled patch with low cost
        def choose_sample(cost_image, tol, min_val):
                # If the minimum is approximately zero (which can happen initially),
                # it might make sense to set minc to a larger value
                minc = np.amin(cost_image) # find min value in cost_image
                minc = max(minc, min_val)
                # find co-ordinates of min val. return is a ndarray
                row, col = np.where(cost_image <= minc * (1 + tol))</pre>
                # since it is possible to have multiple values, lets just pick a randon,
          →number form the
                 # list or row values.
                if (len(row) < 1):
                         rand idx = 0
                else:
                         rand_idx = random.randrange(0, len(row)-1)
                return (row[rand_idx], col[rand_idx])
```

```
def quilt_simple(sample, out_size, patch_size, overlap, tol):
    Quilts square patches of size patchsize from sample in order to create anu
 \rightarrow output image of size outsize.
    Feel free to add function parameters
    :param sample: numpy.ndarray
    :param out size: int
    :param patch_size: int
    :param overlap: int
    :param tol: int
    :return: numpy.ndarray
    min_val = 0.1
    res = np.zeros((out_size, out_size, 3), np.uint8)
    for i in range(0, out_size, patch_size - overlap):
        for j in range(0, out_size, patch_size - overlap):
            # shrink the patch to accommodate the smaller regions in the borders.
            patch_row = min(i + patch_size, out_size) - i
            patch col = min(j + patch size, out size) - j
            # The mask has the same size as the patch template
            # and has values of 1 in the overlapping region
            # and values of 0 elsewhere.
            mask = np.zeros((patch_row, patch_col))
            if (i > 0):
                mask[:overlap, :] = 1
            if (j > 0):
                mask[:, :overlap] = 1
            # get the area tha needs to patched from the final result image
            patch = res[i:i+patch_row, j:j+patch_col]
            # calculate the ssd value of the image patch, sample image
            # and the mask. mask has the overlap area covered.
            # 192,192,3:5x5:5,5,3
            cost_image = np.zeros((patch_row, patch_col), np.uint8)
            cost_image = ssd_patch(patch, mask, sample)
            # since we are looking for the minc from the patch, need to make
            # sure that other values are not selected. So marking all the other \square
 \rightarrow values
            # to max size. Since we are looking for only min value.
            patch_row_mid = int(patch_row/2) + 1
            patch_col_mid = int(patch_col/2) + 1
            cost_image[:patch_row_mid, :] = sys.maxsize
```

```
cost_image[-patch_row_mid:, :] = sys.maxsize
                cost_image[:, :patch_col_mid] = sys.maxsize
                cost_image[:, -patch_col_mid:] = sys.maxsize
                # generate the quilt patch/
                row, col = choose_sample(cost_image, tol, min_val)
                quilt_patch = sample[row - int(patch_row/2):row + int((patch_row+1)/
     \rightarrow2), col - int(patch_col/2):col + int((patch_col+1)/2)]
                # update the final result image with the quilt patch
                res[i:i+patch_row, j:j+patch_col] = quilt_patch
        return res
[7]: out_size = 625
    patch_size = 125
    overlap = 50
    tol = 0.00006
    res = quilt_simple(sample_img, out_size, patch_size, overlap, tol) #feel free_u
    →to change parameters to get best results
    if res is not None:
        plt.imshow(res)
```



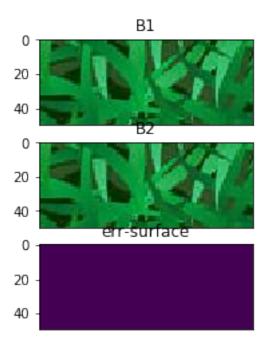
1.1.3 Part III: Seam Finding (20 pts)

```
[8]: # optional or use cut(err_patch) directly
   def customized cut(bndcost):
       pass
[9]: def quilt_cut(sample, out_size, patch_size, overlap, tol):
        Quilts square patches of size patchsize from sample in order to create anu
     →output image of size outsize.
       Feel free to add function parameters
        :param sample: numpy.ndarray
        :param out_size: int
        :param patch_size: int
        :param overlap: int
        :param tol: int
        :return: numpy.ndarray
        11 11 11
       min_val = 0.1
       flag = 1
       res = np.zeros((out size, out size, 3), np.uint8)
       for i in range(0, out_size, patch_size - overlap):
            for j in range(0, out_size, patch_size - overlap):
                # shrink the patch to accommodate the smaller regions in the borders.
                patch_row = min(i + patch_size, out_size) - i
                patch_col = min(j + patch_size, out_size) - j
                # The mask has the same size as the patch template
                # and has values of 1 in the overlapping region
                # and values of O elsewhere.
                mask = np.zeros((patch_row, patch_col))
                if (i > 0):
                    mask[:overlap, :] = 1
                if (j > 0):
                    mask[:, :overlap] = 1
                # get the area tha needs to patched from the final result image
                patch = res[i:i+patch_row, j:j+patch_col]
                # calculate the ssd value of the image patch, sample image
                # and the mask. mask has the overlap area covered.
                # 192,192,3:5x5:5,5,3
                cost_image = np.zeros((patch_row, patch_col), np.uint8)
                cost_image = ssd_patch(patch, mask, sample)
                # since we are looking for the minc from the patch, need to make
```

```
# sure that other values are not selected. So marking all the other
\rightarrow values
           # to max size. Since we are looking for only min value.
           patch row mid = int(patch row/2) + 1
           patch_col_mid = int(patch_col/2) + 1
           cost image[:patch row mid, :] = sys.maxsize
           cost_image[-patch_row_mid:, :] = sys.maxsize
           cost_image[:, :patch_col_mid] = sys.maxsize
           cost_image[:, -patch_col_mid:] = sys.maxsize
           # generate the quilt patch
           row, col = choose_sample(cost_image, tol, min_val)
           quilt_patch = sample[row - int(patch_row/2):row + int((patch_row+1)/
\rightarrow2), col - int(patch_col/2):col + int((patch_col+1)/2)]
           # update the final result image with the quilt patch
           res[i:i+patch_row, j:j+patch_col] = quilt_patch
           cut_mask = np.ones((patch_size, patch_size))
           if (i > 0):
               B1 = res[i:i+overlap, j:j+patch_col]
               B2 = sample[row-int(patch_row/2):row-int(patch_row/
→2)+overlap,col-int(patch_col/2):col+int((patch_col+1)/2)]
               err_surface = (B1 - B2) ** 2
               cut_mask[:overlap, :patch_col] = cut(err_surface.sum(axis=2))
               # seam display
               if (flag == 1):
                   fig, axes = plt.subplots(3)
                   axes[0].set_title('B1'), axes[0].set_xticks([])
                   axes[0].imshow(B1), axes[0].set_xticks([])
                   axes[1].set_title('B2'), axes[1].set_xticks([])
                   axes[1].imshow(B2), axes[1].set_xticks([])
                   axes[2].set_title('err-surface'), axes[2].set_xticks([])
                   axes[2].imshow(err_surface.sum(axis=2))
                   flag = 0
             if(j > 0):
                 B1 = res[i:i+patch row, j:j+overlap]
                 B2 = sample[row-int(patch_row/2):row+int((patch_row+1)/
\rightarrow2), col-int(patch_col/2):col-int(patch_col/2)+overlap]
                 err_surface = (B1 - B2) ** 2
                 cut_mask[:patch_row, :overlap] = np.transpose(cut(np.
→ transpose(err_surface.sum(axis=2))))
           for idx in range(3):
```

```
sam = sample[row-int(patch_row/2):row+int((patch_row+1)/2),__
      →col-int(patch_col/2):col+int((patch_col+1)/2), idx]
                     cut_img = sam * cut_mask[:patch_row,:patch_col]
                     for x in range(patch_row):
                         for y in range(patch_col):
                             res[i+x, j+y, idx] = cut_img[x, y]
         return res
[10]: # out_size = 50
     # patch_size = 5
     # overlap = 2
     # tol = 0.00001
     out_size = 625
     patch_size = 125
     overlap = 50
     tol = 0.00006
     res = quilt_cut(sample_img, out_size, patch_size, overlap, tol) #feel free tou
     → change parameters to get best results
     if res is not None:
         plt.imshow(res)
            ValueError
                                                       Traceback (most recent call⊔
     →last)
            <ipython-input-10-54c91d362350> in <module>
              9 \text{ tol} = 0.00006
        ---> 11 res = quilt_cut(sample_img, out_size, patch_size, overlap, tol)_
     →#feel free to change parameters to get best results
             12 if res is not None:
             13
                    plt.imshow(res)
            <ipython-input-9-2b82e203e49c> in quilt_cut(sample, out_size,_
     →patch_size, overlap, tol)
             59
                                 B1 = res[i:i+overlap, j:j+patch_col]
             60
                                 B2 = sample[row-int(patch_row/2):row-int(patch_row/
     →2)+overlap,col-int(patch_col/2):col+int((patch_col+1)/2)]
        ---> 61
                                err surface = (B1 - B2) ** 2
             62
                                 cut_mask[:overlap, :patch_col] = cut(err_surface.
     →sum(axis=2))
```

ValueError: operands could not be broadcast together with shapes \hookrightarrow (25,125,3) (50,125,3)



1.1.4 part IV: Texture Transfer (30 pts)

```
[]: def texture_transfer(sample, target):
    """
    Feel free to add function parameters
    """
    pass
```

1.1.5 Bells & Whistles

(10 pts) Create and use your own version of cut.m. To get these points, you should create your own implementation without basing it directly on the provided function (you're on the honor code for this one).

You can simply copy your customized_cut(bndcost) into the box below so that it is easier for us to grade

	(15 pts) Implement the iterative texture transfer method described in the paper. Compare to the non-iterative method for two examples.
[]:	
	(up to 20 pts) Use a combination of texture transfer and blending to create a face-in-toast image like the one on top. To get full points, you must use some type of blending, such as feathering or
	Laplacian pyramid blending.
[]:	
	(up to 40 pts) Extend your method to fill holes of arbitrary shape for image completion. In this case, patches are drawn from other parts of the target image. For the full 40 pts, you should implement a smart priority function (e.g., similar to Criminisi et al.).
[]:	
[]:	
[]:	
[]:	