

Project2

February 27, 2020

1 Programming Project #2: Image Quilting

1.1 CS445: Computational Photography - Spring 2020

```
[1]: 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
from PIL import Image
```

1.1.1 Part I: Randomly Sampled Texture (10 pts)

```
[3]: sample_img_dir = 'samples/bricks_small.jpg' # feel free to change
sample_img = None
# feynman = cv2.imread("samples/feynman.tiff")
# sketch = cv2.imread("samples/sketch.tiff")
# text_small = cv2.imread("samples/text_small.jpg")
# textureWeave = cv2.imread("samples/texture.png")
# toast = cv2.imread("samples/toast.jpg")
# white_small = cv2.imread("samples/white_small.jpg")

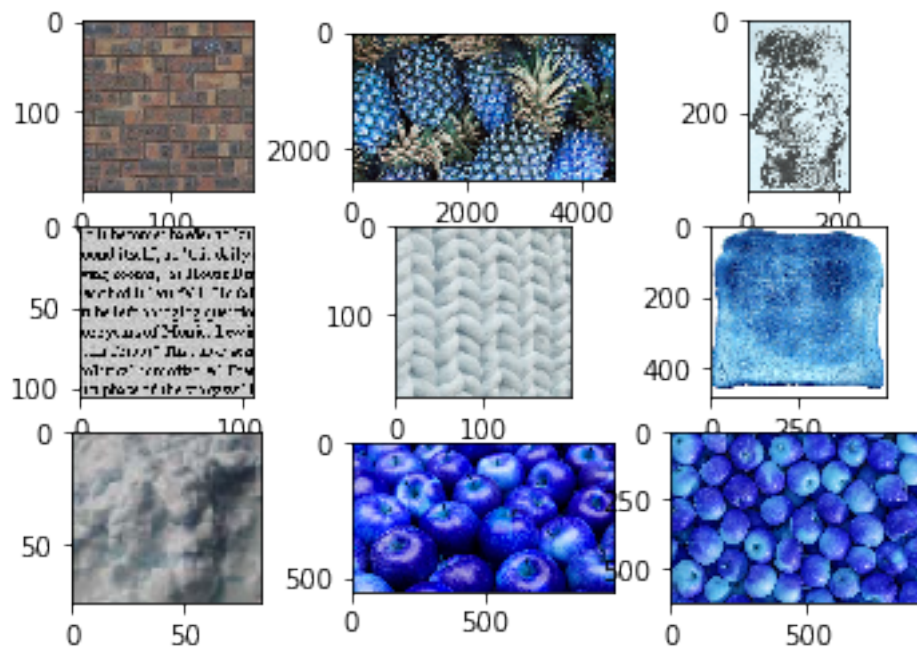
if os.path.exists(sample_img_dir):
    sample_img = cv2.imread(sample_img_dir)
    # plt.imshow(sample_img)

# Visualizing Images on Notebook
# import matplotlib.pyplot as plt
# %matplotlib inline # or `%matplotlib notebook` for interactive plotting
# # from PIL import Image
# # import cv2
# # read image using PIL and OpenCV library
image_1 = Image.open("samples/bricks_small.jpg")
# image_1 = cv2.imread("samples/bricks_small.jpg")
```

```

#     image_2 = cv2.imread("samples/feynman.tiff")
image_2 = cv2.imread("samples/pineapples.jpg")
image_3 = cv2.imread("samples/sketch.tiff")
image_4 = cv2.imread("samples/text_small.jpg")
image_5 = cv2.imread("samples/texture.png")
image_6 = cv2.imread("samples/toast.jpg")
image_7 = cv2.imread("samples/white_small.jpg")
image_8 = cv2.imread("samples/apples.jpg")
image_9 = cv2.imread("samples/fujiapples.jpg")
#For Face on Toast
image_10 = cv2.imread("samples/Aquaman.jpg")
image_11 = cv2.imread("samples/WonderWoman.jpg")
# # Display image on notebook
fig, axes = plt.subplots(3, 3)
axes[0, 0].imshow(image_1)
axes[0, 1].imshow(image_2)
axes[0, 2].imshow(image_3)
axes[1, 0].imshow(image_4)
axes[1, 1].imshow(image_5)
axes[1, 2].imshow(image_6)
axes[2, 0].imshow(image_7)
axes[2, 1].imshow(image_8)
axes[2, 2].imshow(image_9)

```



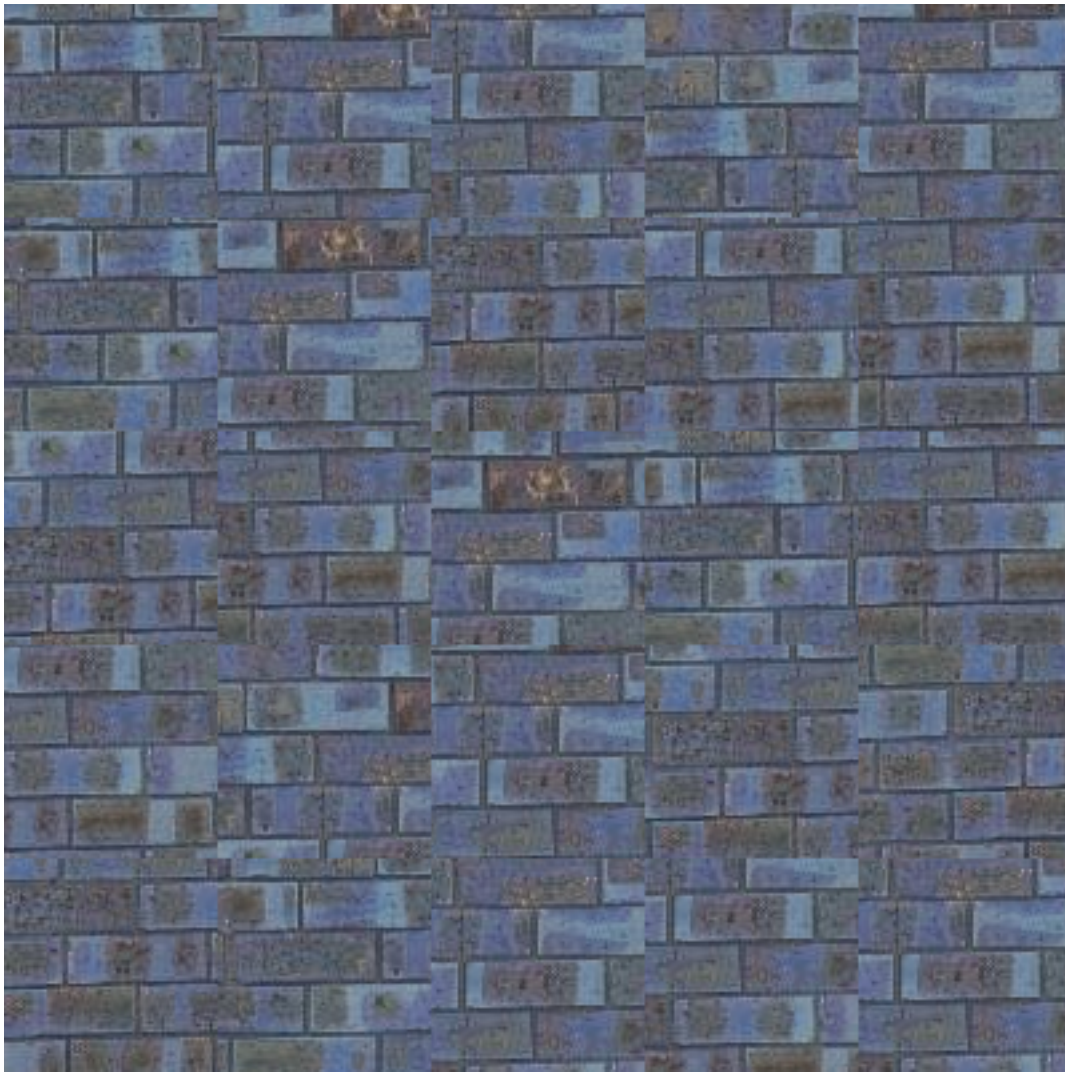
```
[4]: # https://www.geeksforgeeks.org/working-images-python/. use open image library
      ↪ from path, crop a patch, paste the patch
      # https://code-maven.com/create-images-with-python-pil-pillow -> Results array
def random_patch(sample, patch_size):
    """
    Randomly picks a patch on an image
    """
    random_x = np.random.randint(0, sample.size[0] - patch_size)
    random_y = np.random.randint(0, sample.size[0] - patch_size)
    patchCorners = (random_x, random_y, random_x + patch_size, random_y +
    ↪ patch_size) #identify the 4 corners
    patch = sample.crop(patchCorners)
    return patch

def quilt_random(sample, out_size, patch_size):
    """
    Randomly samples square patches of size patchsize from sample in order to
    ↪ create an output image of size outsize.

    :param sample: numpy.ndarray    The image you read from sample directory
    :param out_size: int            The width of the square output image
    :param patch_size: int         The width of the square sample patch
    :return: numpy.ndarray
    """
    # Todo
    sampleImage = Image.fromarray(sample) #convert image matrix parameter to
    ↪ Image
    output = Image.new("RGB", (out_size, out_size), "black") #results array
    # black along the edges ok if doesn't fit evenly, determined by the floor
    ↪ div, like 20/115, or 420/80
    for x in range(0, out_size//patch_size):
        for y in range(0, out_size//patch_size):
            patch = random_patch(sampleImage, patch_size)
            output.paste(patch, (x*patch_size,y*patch_size, (x+1)*patch_size,
    ↪ (y+1)*patch_size) )

    return output
```

```
[5]: out_size = 400 # feel free to change to debug
      patch_size = 80 # feel free to change to debug
      res = quilt_random(sample_img, out_size, patch_size)
      display(res)
      # https://docs.scipy.org/doc/numpy/reference/generated/numpy.ndarray.astype.html
      # https://docs.scipy.org/doc/numpy/reference/arrays.ndarray.html
      # https://www.geeksforgeeks.org/change-data-type-of-given-numpy-array/
      # https://stackoverflow.com/questions/26649716/
      ↪ how-to-show-pil-image-in-ipython-notebook
```



1.1.2 Part II: Overlapping Patches (30 pts)

[6]: *# PseudoCode*

- # a) Generate scan all possible permutations of patches via an AllPatches, ↵*
↪store in obj/dict
- # b) Scan the result with ssd_patch (template matching with the overlapping ↵*
↪region, computing the cost of sampling each patch, based on (SSD) of the ↵
↪overlapping regions)
- # c) Use choose_sample() to run the ssd_patch calculations, build a candidates ↵*
↪array with lowest SSD scores, random selection
- # d) Use quilt_simple() to iterate through the canvas, grab sample patches that ↵*
↪fit the criteria, and place patches in output image

Tip: For efficiency, use filtering to compute SSD.

```

# Note that part of the computation only needs to be done once (not for each
→patch), so that can be cached.
# Suppose I have a template T, a mask M, and an image I: then,  $ssd = ((M*T)**2).$ 
→ $sum() - 2 * cv2.filter2D(I, ddepth=-1, kernel = M*T) + cv2.filter2D(I ** 2,$ 
→ $ddepth=-1, kernel=M)$ 

# def ssd_patch(T, M, I):
#      $ssd = ((M*T)**2).sum() - 2 * cv2.filter2D(I, ddepth=-1, kernel = M*T) +$ 
→ $cv2.filter2D(I ** 2, ddepth=-1, kernel=M)$ 
#     return ssd

# Scan all possible permutations
def allPatches(sample, patch_size):
    allPatchesDict = {}
    countIndex = 0
    #print("Sample", sample.shape)
    for x in range(sample.shape[0]):
        if sample.shape[0] - x < patch_size:
            break
        for y in range(sample.shape[1]):
            #print("Sample[0, 1]", sample.shape[0], sample.shape[1])
            if sample.shape[1] - y < patch_size:
                continue
            allPatchesDict[countIndex] = sample[x : x + patch_size, y : y +
→patch_size, :]
            countIndex += 1
        #print("All Patches", allPatchesDict)
    return allPatchesDict

# patch in the current output image to be filled in (many pixel values will be
→0 because they are not filled in yet)
def ssd_patch(M, T, I):
    template = np.zeros(I.shape)
    template[:, :, 0], template[:, :, 1], template[:, :, 2] = M * I[:, :, 0], M * I[:, :,
→1], M * I[:, :, 2]
    return np.sum((T - template) ** 2)

# run the ssd_patch calculations, build a candidates array with lowest SSD
→scores, random selection
def choose_sample(sample, patch_size, M, T, tol, allPatchesDict):
    tolerance = tol #start with low tolerance (e.g., 0.00001)
    imageCosts = {}
    candidates = []
    randomIndex = (np.random.choice(len(allPatchesDict), int(0.1 *
→len(allPatchesDict))))

```

```

for i in randomIndex:
    imageCosts[i] = ssd_patch(M, T, allPatchesDict[i])
    #minc=max(minc,small_cost_value)
    #grab minimum cost
minc = min(imageCosts.values())
for key in imageCosts.keys():
    # row, col = np.where(cost < minc*(1+tol))
    if imageCosts[key] <= minc * (1 + tolerance):
        candidates.append(allPatchesDict[key])

randomIndexSelection = np.random.choice(len(candidates))
return candidates[randomIndexSelection]

# Note: Should only be used within the quilting functions due to local/global
def addSampleToResultImage(x, y, overlap, patch_size, updatedSample, output):
    overlapPixels = overlap // 2
    endX = x + patch_size
    endY = y + patch_size

    if x == 0 and y == 0:
        output[x : endX, y : endY] = updatedSample
    elif x == 0 and y != 0:
        output[x: endX, y + overlapPixels : endY] = updatedSample[:,
→overlapPixels:, :]
    elif x != 0 and y == 0:
        output[x + overlapPixels : endX, y : endY] =
→updatedSample[overlapPixels:, :, :]
    else:
        output[x + overlapPixels : endX, y + overlapPixels : endY] =
→updatedSample[overlapPixels:, overlapPixels:, :]

def quilt_simple(sample, out_size, patch_size, overlap, tol):
    """
    Randomly samples square patches of size patchsize from sample in order to
→create an 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: float
    :return: numpy.ndarray
    """
    # Todo
    sample = sample.copy() / 255.0
    output = np.zeros([out_size, out_size, 3]) #results array
    patches = allPatches(sample, patch_size)

```



```

for x in range(0, out_size + 1, patch_size - overlap):
    if out_size - x < patch_size:
        break
    for y in range(0, out_size + 1, patch_size - overlap):
        if out_size - y < patch_size:
            continue

        # auto randomize patch for upper-left corner
        if x == 0 and y == 0:
            index = np.random.choice(len(patches))
            addSampleToResultImage(x, y, overlap, patch_size,
→patches[index], output)
            M = (output[x : x + patch_size, y : y + patch_size, 0] != 0).
→astype(int)
            T = (output[x : x + patch_size, y : y + patch_size])
            updatedSample = choose_sample(sample, patch_size, M, T, tol,
→patches)
            addSampleToResultImage(x, y, overlap, patch_size, updatedSample,
→output)
        return output

```

```

[7]: # Use an odd value for patch_size so that its center is well-defined. -> 75
# Extra Image 2 (Used for Testing). Image 1 for this section in next cell.
beardedQuilt = quilt_simple(image_3, 400, 75, 20, 0.00001) #feel free to change
→parameters to get best results
# plt.figure(figsize = (6,6))
# plt.imshow(beardedQuilt[:, :, :])
# plt.imshow(beardedQuilt[:-10, :-10, :])
# plt.imshow(beardedQuilt[:-25, :-25, :])
# plt.imshow(beardedQuilt[:-50, :-50, :])

```

```

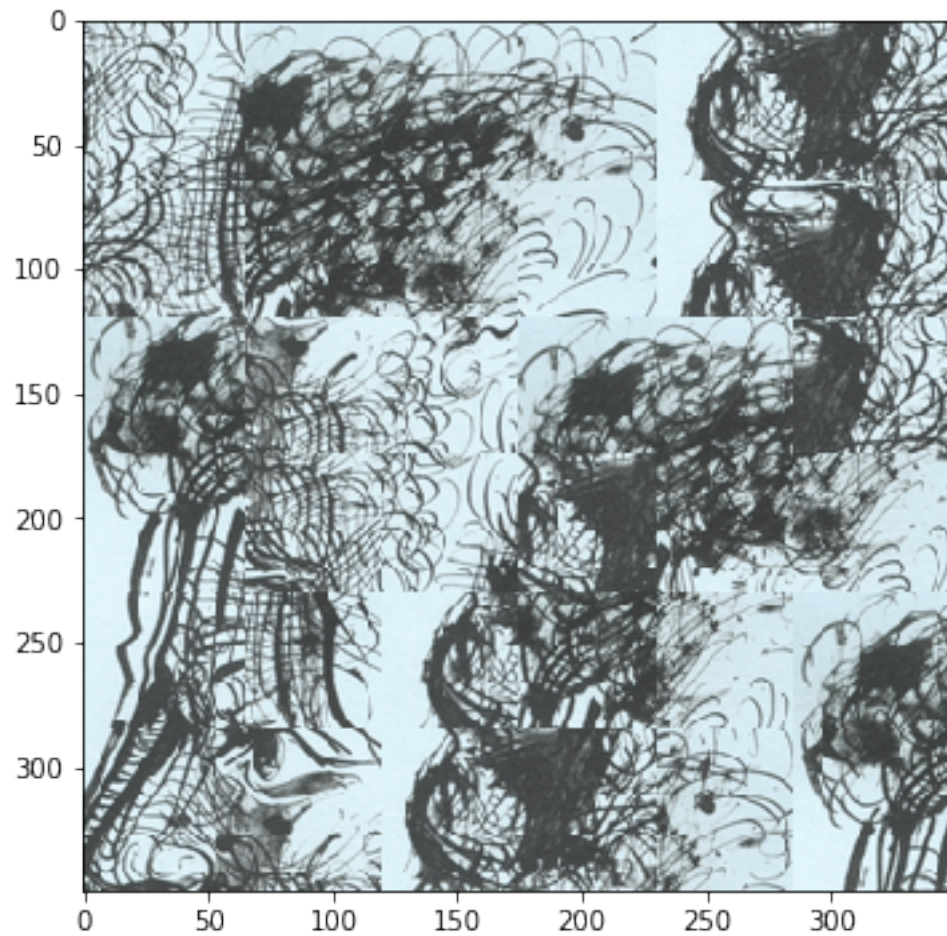
[8]: # Clearer Cell
plt.figure(figsize = (6,6))
plt.imshow(beardedQuilt[:-50, :-50, :])

```

```

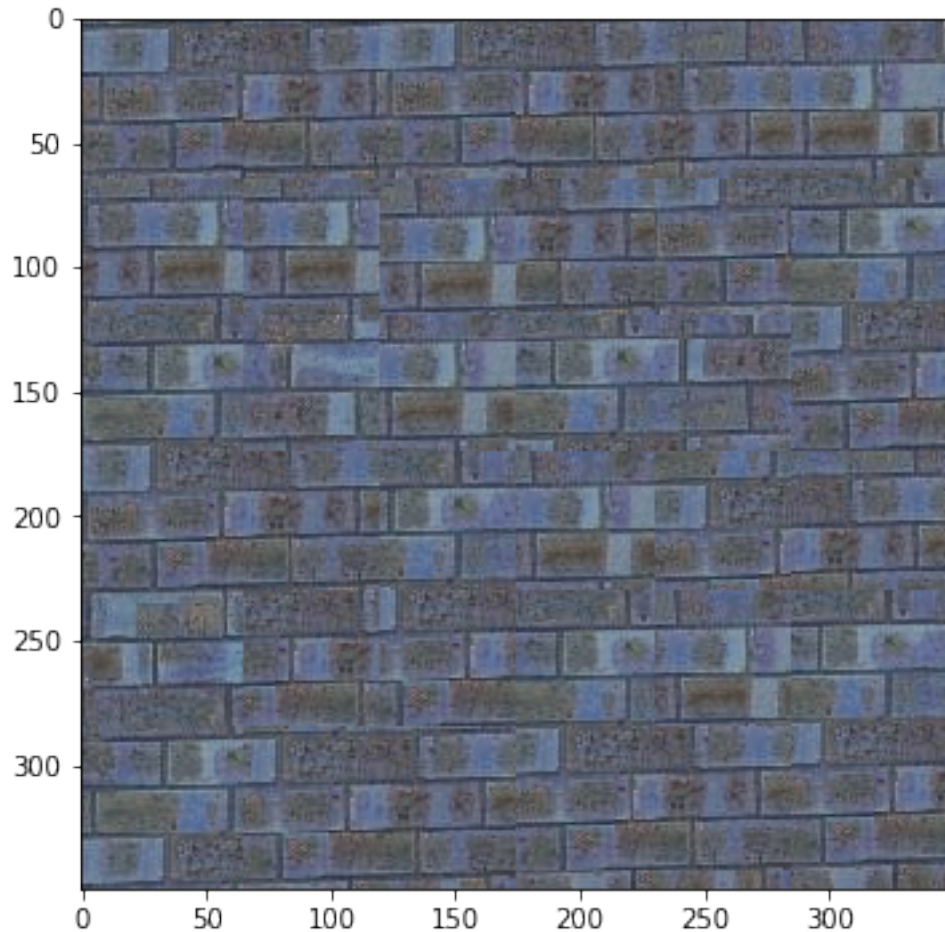
[8]: <matplotlib.image.AxesImage at 0x121315250>

```



```
[9]: # Same Sample from Random
bricksQuilted = quilt_simple(sample_img, 400, 75, 20, 0.1)
plt.figure(figsize = (6,6))
plt.imshow(bricksQuilted[:50,:50,:])
```

```
[9]: <matplotlib.image.AxesImage at 0x1211c5690>
```

1.1.3 Part III: Seam Finding (20 pts)

```
[ ]: # Would this actually have been part of B&W?
      # optional or use cut(err_patch) directly
      def customized_cut(bndcost):
          pass
```

```
[10]: def cut(err_patch):
        """
        Compute the minimum path from the left to right side of the patch

        :param err_patch: numpy.ndarray    cost of cutting through each pixel
        :return: numpy.ndarray             a 0-1 mask that indicates which pixels
        ↪ should be on either side of the cut
        """

        # create padding on top and bottom with very large cost
```

```

padding = np.expand_dims(np.ones(err_patch.shape[1]).T*1e10,0)
err_patch = np.concatenate((padding, err_patch, padding), axis=0)
h, w = err_patch.shape
path = np.zeros([h,w], dtype="int")
cost = np.zeros([h,w])
cost[:,0] = err_patch[:, 0]
cost[0,:] = err_patch[0, :]
cost[cost.shape[0]-1,:] = err_patch[err_patch.shape[0]-1, :]

# for each column, compute the cheapest connected path to the left
# cost of path for each row from left upper/same/lower pixel
for x in range(1,w):
    # cost of path for each row from left upper/same/lower pixel
    tmp = np.vstack((cost[0:h-2,x-1], cost[1:h-1, x-1], cost[2:h, x-1]))
    mi = tmp.argmin(axis=0)
    path[1:h-1, x] = np.arange(1, h-1, 1).T + mi # save the next step of
→ the path
    cost[1:h-1, x] = cost[path[1:h-1, x] - 1, x-1] + err_patch[1:h-1, x]

path = path[1:path.shape[0]-1, :] - 1
cost = cost[1:cost.shape[0]-1, :]

# create the mask based on the best path
mask = np.zeros(path.shape, dtype="int")
best_path = np.zeros(path.shape[1], dtype="int")
best_path[len(best_path)-1] = np.argmin(cost[:, cost.shape[1]-1]) + 1
mask[0:best_path[best_path.shape[0]-1], mask.shape[1]-1] = 1
for x in range(best_path.size-1, 0, -1):
    best_path[x-1] = path[best_path[x]-1, x]
    mask[:best_path[x-1], x-1] = 1
mask ^= 1
return mask

```

```

[11]: # Update helper functions
def ssd_patch2(M, T, I):
    template = np.zeros(I.shape)
    template[:, :, 0], template[:, :, 1], template[:, :, 2] = M * I[:, :, 0], M * I[:, :,
→ 1], M * I[:, :, 2]
    return np.sum((T - template) ** 2, -1)

def choose_sample_with_cost(sample, patch_size, M, T, tol, allPatchesDict):
    tolerance = tol
    imageCosts = {}
    imageCostSSD = {}
    candidates = []
    imageCostsArray = []

```

```

    randomIndex = (np.random.choice(len(allPatchesDict), int(0.01 *
↳len(allPatchesDict))))

    for i in randomIndex:
        cost = ssd_patch2(M, T, allPatchesDict[i])
        imageCosts[i] = np.sum(cost)
        imageCostSSD[i] = cost

    minc = min(imageCosts.values())
    for key in imageCosts.keys():
        if imageCosts[key] <= minc * (1 + tolerance):
            candidates.append(allPatchesDict[key])
            imageCostsArray.append(imageCostSSD[key])

    randomIndexSelection = np.random.choice(len(candidates))
    return candidates[randomIndexSelection],
↳imageCostsArray[randomIndexSelection]

```

```

[12]: def quilt_cut(sample, out_size, patch_size, overlap, tol):
    """
    Samples square patches of size patchsize from sample using seam finding in
↳order to create an 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: float
    :return: numpy.ndarray
    """
    sample = sample.copy() / 255.0
    output = np.zeros([out_size, out_size, 3]) #results array
    patches = allPatches(sample, patch_size) #grab all patches for this image
    #Iterate through x and y of the canvas, generate sample image, keeping in
↳mind the low-cost
    for x in range(0, out_size + 1, patch_size - overlap):
        overlapX = x + overlap
        endX = x + patch_size
        if out_size - x < patch_size:
            break
        for y in range(0, out_size + 1, patch_size - overlap):
            overlapY = y + overlap
            endY = y + patch_size
            if out_size - y < patch_size:
                continue
            # Setup Mask and Template
            M = (output[x : endX, y : endY, 0] != 0)

```

```

        T = (output[x : endX, y : endY, :])
        newSampleImage, cost = choose_sample_with_cost(sample, patch_size,
↪M, T, tol, patches)

        # Upper Corner
        if x == 0 and y == 0:
            addSampleToResultImage(x, y, overlap, patch_size, patches[np.
↪random.choice(len(patches))], output)
            # Account for top and/or left over flaps. Using cut to find the
↪min-cost contiguous path and generate masks
            elif x == 0 and y != 0:
                template = cost[:, 0 : overlap]
                mask1 = cut(template.T).T
                #print(mask1)
                overlappedRegion = output[x : endX, y: overlapY, :]
                for z in range(3):
                    overlappedRegion[:, :, z] = newSampleImage[:, :, overlap, z] *
↪mask1 + (1 - mask1) * overlappedRegion[:, :, z]
                    output[x: endX, overlapY: endY, :] = newSampleImage[:, overlap :, :]
            elif x != 0 and y == 0:
                mask2 = cut(cost[0 : overlap, :])
                overlappedRegion = output[x : overlapX, y: endY, :]
                for z in range(3):
                    overlappedRegion[:, :, z] = newSampleImage[:, overlap :, z] *
↪mask2 + (1 - mask2) * overlappedRegion[:, :, z]
                    output[overlapX: endX, y: endY, :] = newSampleImage[overlap :, :, :]
            else:
                template = cost[:, 0 : overlap]
                mask3 = cut(template.T).T
                overlappedRegion = output[x : endX, y: overlapY, :]
                for z in range(3):
                    overlappedRegion[:, :, z] = newSampleImage[:, :, overlap, z] *
↪mask3 + (1 - mask3) * overlappedRegion[:, :, z]
                    mask4 = cut(cost[0 : overlap, :])
                    overlappedRegion = output[x : overlapX, y: endY, :]
                    for z in range(3):
                        overlappedRegion[:, :, z] = newSampleImage[:, overlap :, z] *
↪mask4 + (1 - mask4) * overlappedRegion[:, :, z]
                        output[overlapX: endX, overlapY: endY, :] = newSampleImage[overlap :
↪, overlap :, :]
        return output

```

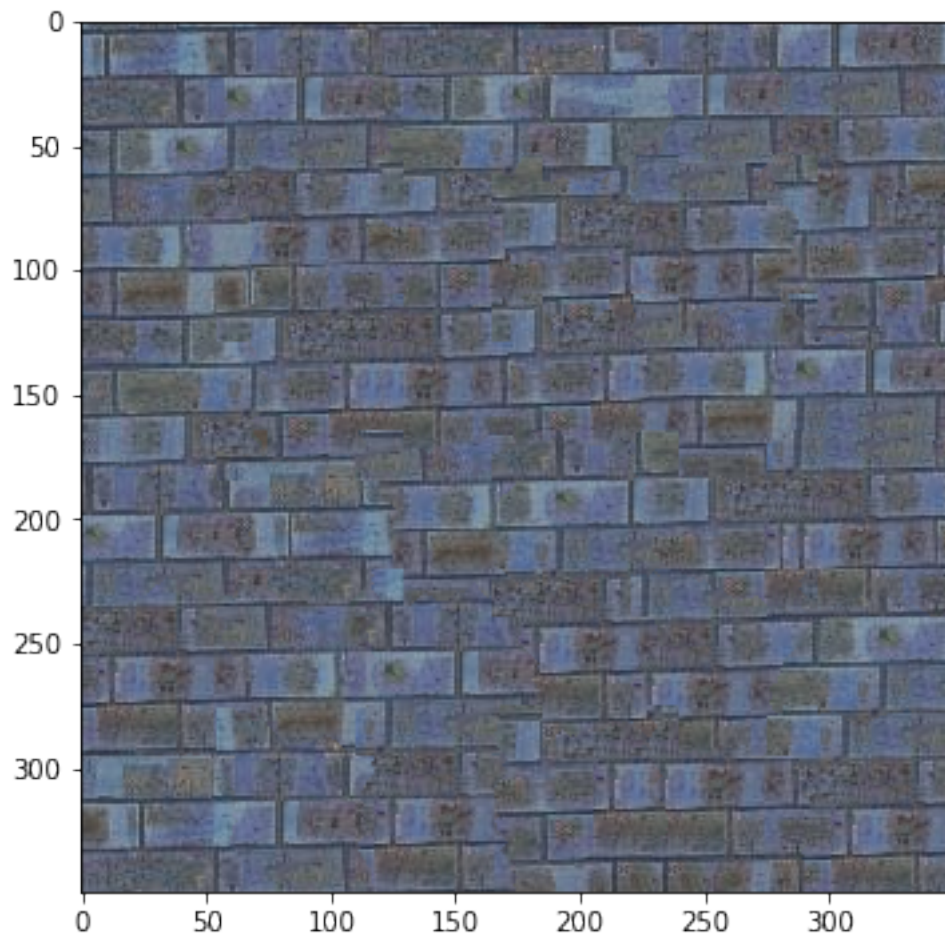
```

[13]: # Image 1. Default to compare against random and overlapping texture samples
bricksCut = quilt_cut(sample_img, 400, 75, 20, 0.001)
plt.figure(figsize = (6,6))
# plt.imshow(bricksCut[:, :, :])

```

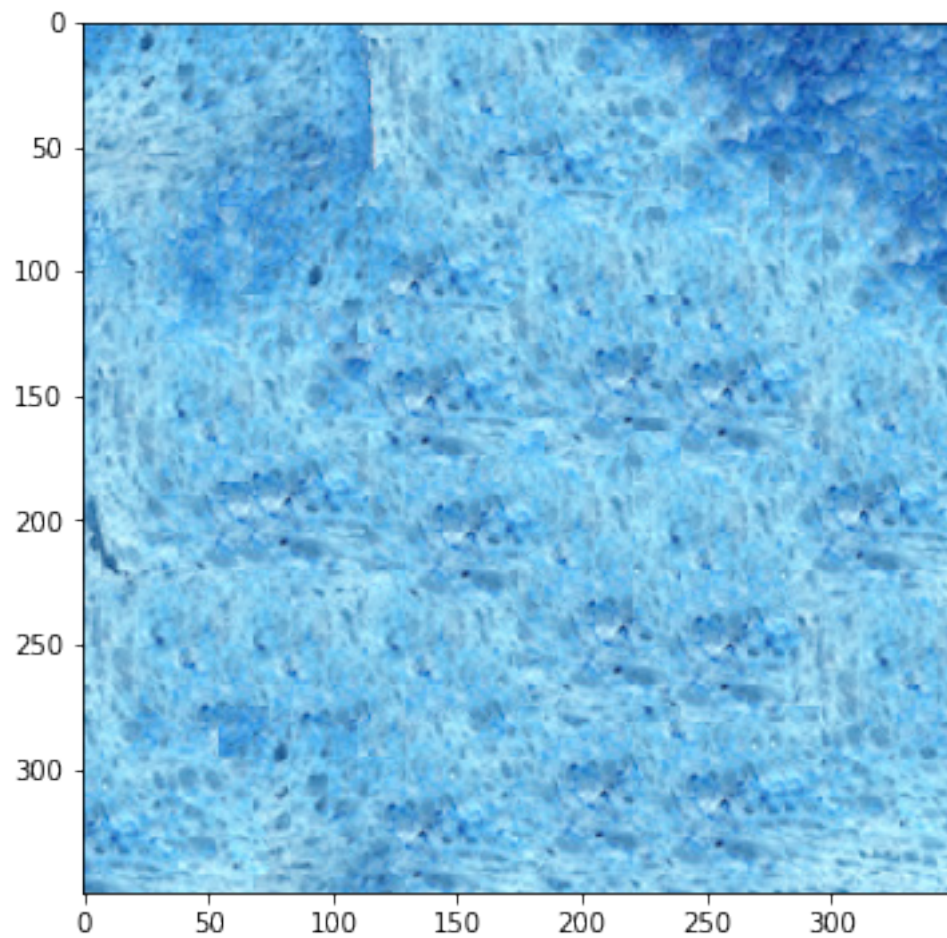
```
plt.imshow(bricksCut[:50,:50,:])
```

[13]: <matplotlib.image.AxesImage at 0x10418ea10>



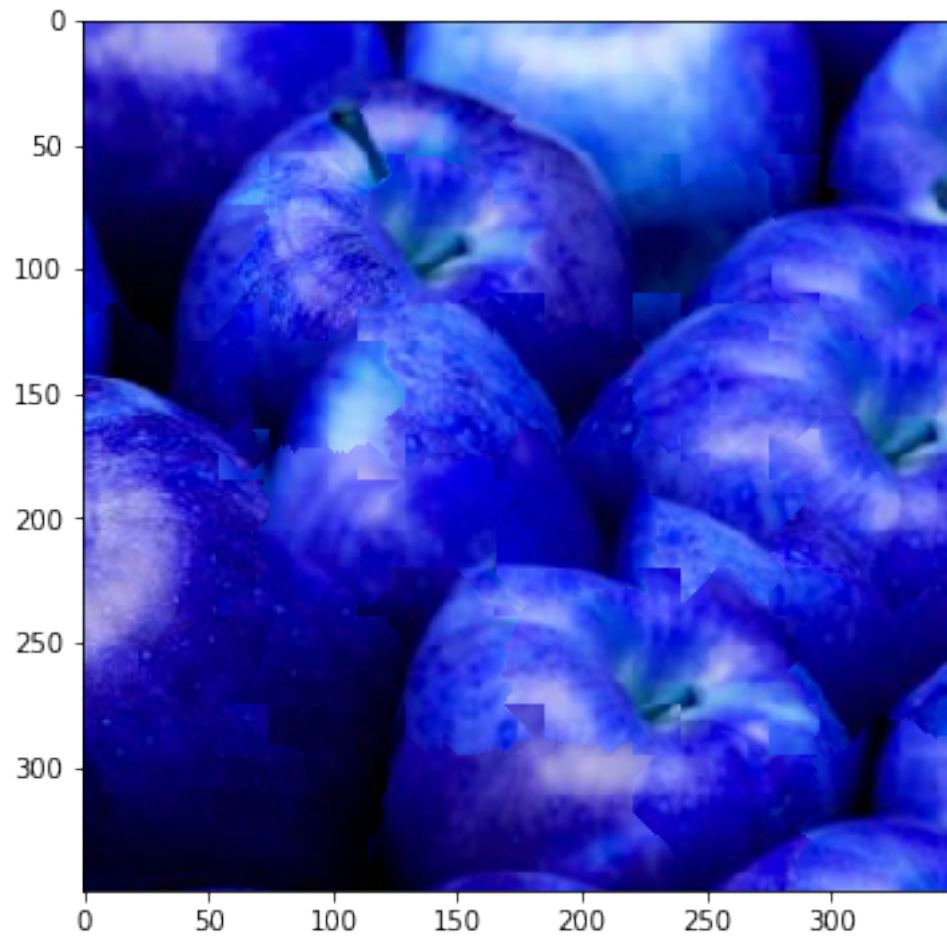
```
[14]: # Image 2
toast = quilt_cut(image_6, 400, 75, 20, 0.002)
plt.figure(figsize = (6,6))
plt.imshow(toast[:50,:50,:])
```

[14]: <matplotlib.image.AxesImage at 0x121446790>



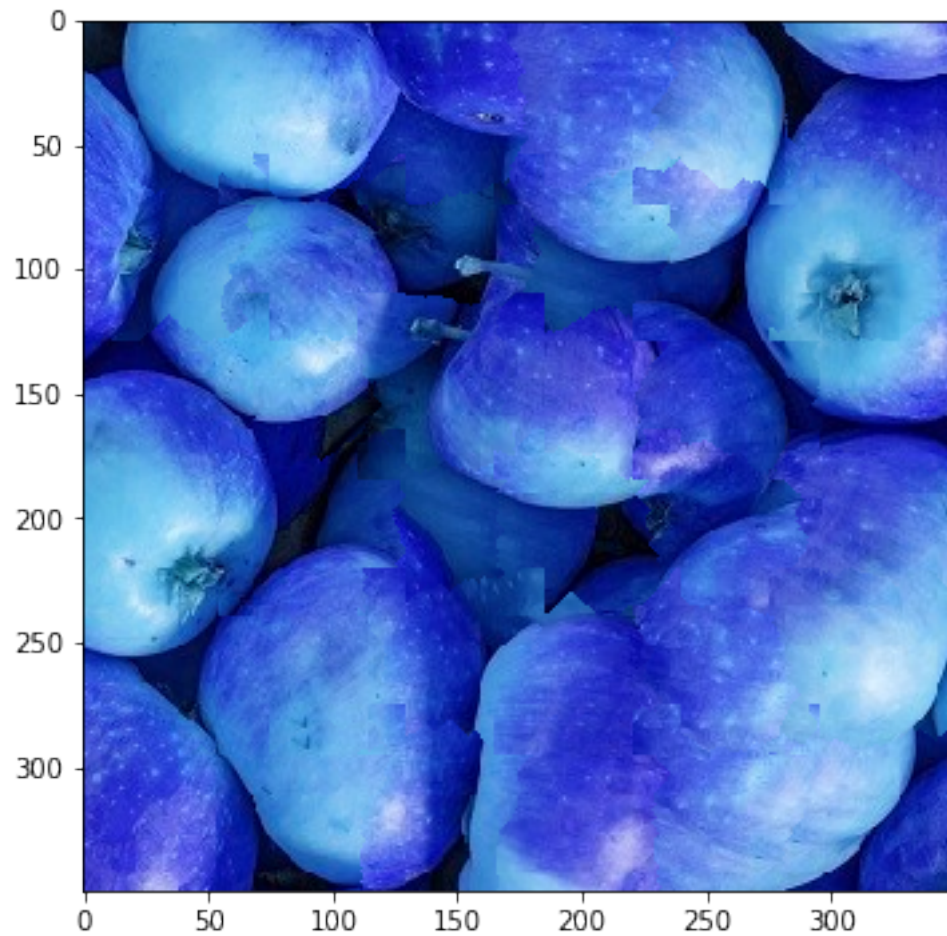
```
[15]: # Image 3. Personal Image
appleQuilted = quilt_cut(image_8, 400, 75, 20, 0.002)
plt.figure(figsize = (6,6))
plt.imshow(appleQuilted[:50,:50,:])
```

```
[15]: <matplotlib.image.AxesImage at 0x12127f6d0>
```



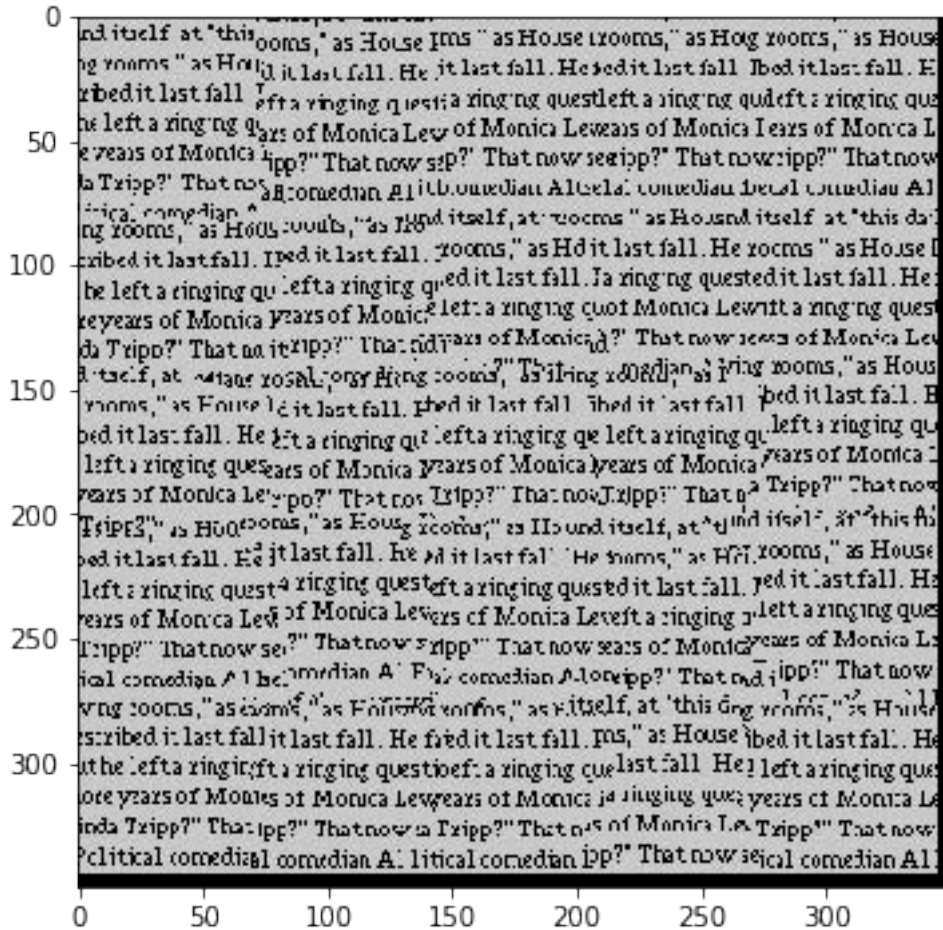
```
[18]: # Image 4. Personal Image
pineappleQuilted = quilt_cut(image_9, 400, 75, 20, 0.002)
plt.figure(figsize = (6,6))
plt.imshow(pineappleQuilted[:50,:50,:])
```

```
[18]: <matplotlib.image.AxesImage at 0x26e931d90>
```



```
[19]: # Image 5. Sample Image
textQuilted = quilt_cut(image_4, 400, 85, 20, 0.1)
plt.figure(figsize = (6,6))
plt.imshow(textQuilted[:50,:50,:])
```

```
[19]: <matplotlib.image.AxesImage at 0x27c61fb10>
```



1.1.4 part III-B: Seam Finding Illustration

```
[20]: def cut2(err_patch):
    """
    Compute the minimum path from the left to right side of the patch

    :param err_patch: numpy.ndarray    cost of cutting through each pixel
    :return: numpy.ndarray             a 0-1 mask that indicates which pixels
    ↪ should be on either side of the cut
    """

    # create padding on top and bottom with very large cost
    padding = np.expand_dims(np.ones(err_patch.shape[1]).T*1e10,0)
    err_patch = np.concatenate((padding, err_patch, padding), axis=0)
    h, w = err_patch.shape
    path = np.zeros([h,w], dtype="int")
    cost = np.zeros([h,w])
```



```

cost[:,0] = err_patch[:, 0]
cost[0,:] = err_patch[0, :]
cost[cost.shape[0]-1,:] = err_patch[err_patch.shape[0]-1, :]

# for each column, compute the cheapest connected path to the left
# cost of path for each row from left upper/same/lower pixel
for x in range(1,w):
    # cost of path for each row from left upper/same/lower pixel
    tmp = np.vstack((cost[0:h-2,x-1], cost[1:h-1, x-1], cost[2:h, x-1]))
    mi = tmp.argmin(axis=0)
    path[1:h-1, x] = np.arange(1, h-1, 1).T + mi # save the next step of
→the path
    cost[1:h-1, x] = cost[path[1:h-1, x] - 1, x-1] + err_patch[1:h-1, x]

path = path[1:path.shape[0]-1, :] - 1
cost = cost[1:cost.shape[0]-1, :]

# create the mask based on the best path
mask = np.zeros(path.shape, dtype="int")
best_path = np.zeros(path.shape[1], dtype="int")
best_path[len(best_path)-1] = np.argmin(cost[:, cost.shape[1]-1]) + 1
mask[0:best_path[best_path.shape[0]-1], mask.shape[1]-1] = 1
for x in range(best_path.size-1, 0, -1):
    best_path[x-1] = path[best_path[x]-1, x]
    mask[:,best_path[x-1], x-1] = 1
mask ^= 1
return mask, best_path

```

```

[21]: def quilt_cut2(sample, out_size, patch_size, overlap, tol):
    """
    Samples square patches of size patchsize from sample using seam finding in
    →order to create an 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: float
    :return: numpy.ndarray
    """
    #pass
    sample = sample.copy() / 255.0 #
    output = np.zeros([out_size, out_size, 3]) #results array
    patches = allPatches(sample, patch_size)
    patchArray = []
    for x in range(0, out_size + 1, patch_size - overlap):
        overlapX = x + overlap

```



```

endX = x + patch_size

if out_size - x < patch_size:
    break
for y in range(0, out_size + 1, patch_size - overlap):
    overlapY = y + overlap
    endY = y + patch_size
    if out_size - y < patch_size:
        continue

    M = (output[x : endX, y : endY, 0] != 0)
    T = (output[x : endX, y : endY, :])
    newSampleImage, cost = choose_sample_with_cost(sample, patch_size,
    ↪M, T, tol, patches)
    patchArray.append(newSampleImage)
    if x == 0 and y == 0:
        addSampleToResultImage(x, y, overlap, patch_size, patches[np.
    ↪random.choice(len(patches))], output)

#         elif x == 0 and y != 0:
#             template = cost[:, 0 : overlap]
#             mask1, bestPath1 = cut2(template.T)
#             overlappedRegion = output[x : endX, y : overlapY, :]

#             for z in range(3):
#                 overlappedRegion[:, :, z] = newSampleImage[:, :overlap, z] *
    ↪mask1.T + (1 - mask1.T) * overlappedRegion[:, :, z]
#                 output[x:endX, overlapY:endY, :] = newSampleImage[:, overlap:,:
    ↪]

#             patch1 = patchArray[0]
#             patch2 = patchArray[1]
#             costImage = cost
#             mask1 = mask1.T
#             #bestPath1
#             templateT = template.T
#             break;
    elif x != 0 and y == 0:
        mask2, bestPath2 = cut2(cost[0 : overlap, :])
        overlappedRegion = output[x : overlapX, y : endY, :]
        for z in range(3):
            overlappedRegion[:, :, z] = newSampleImage[:, :overlap, :, z] *
    ↪mask2 + (1 - mask2) * overlappedRegion[:, :, z]
        output[overlapX:endX, y:endY, :] = newSampleImage[overlap:,:,:]
        # Setting up the exports
        patch1 = patchArray[0]
        patch2 = patchArray[1]

```

```

        costImage = cost
        overlapping = overlappedRegion
        #bestPath
        break;
#     else:
#         template = cost[:, 0 : overlap]
#         mask3, bestPath3 = cut2(template.T).T
#         mask3 = mask3.T
#         overlappedRegion = output[x : endX, y: overlapY, :]
#         for z in range(3):
#             overlappedRegion[:, :, z] = newSampleImage[:, :overlap, z] *
↪mask3 + (1 - mask3) * overlappedRegion[:, :, z]
#             mask4, bestPath4 = cut2(cost[0 : overlap, :])
#             overlappedRegion = output[x : overlapX, y: endY, :]
#             for z in range(3):
#                 overlappedRegion[:, :, z] = newSampleImage[:, :overlap, :, z] *
↪mask4 + (1 - mask4) * overlappedRegion[:, :, z]
#             output[overlapX:endX, overlapY:endY, :] =
↪newSampleImage[overlap:, overlap:, :]
        return patch1, patch2, costImage, bestPath2, mask2

```

```

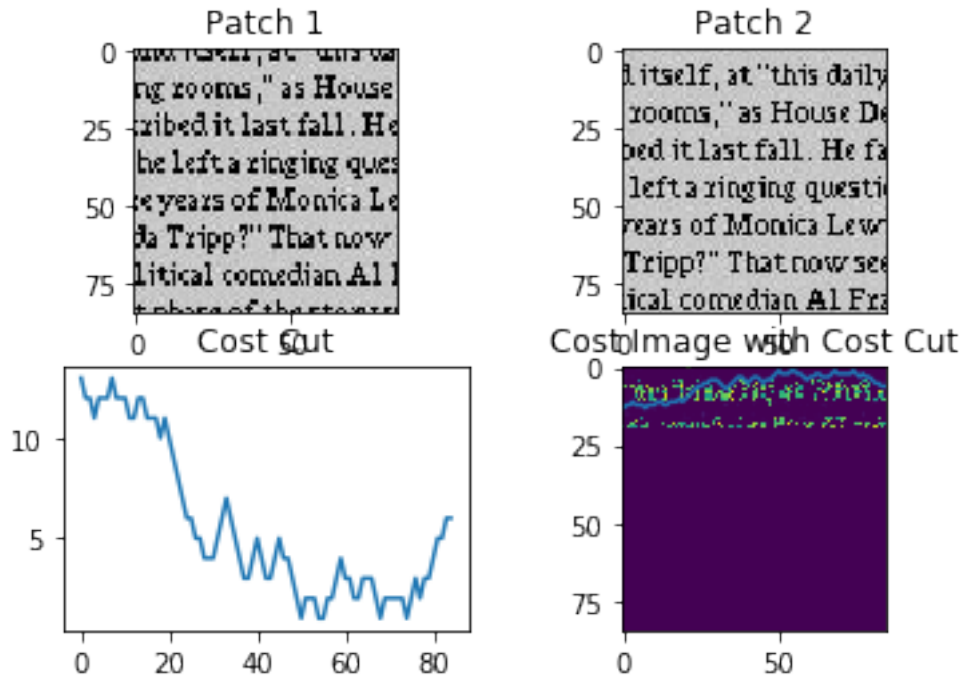
[22]: #Seam finding illustration
patch1, patch2, costImage, bestPath, mask2 = quilt_cut2(image_4, 400, 85, 20, 0.
↪1)
fig, axes = plt.subplots(2, 2)
axes[0, 0].imshow(patch1)
axes[0, 0].set_title('Patch 1')
axes[0, 1].imshow(patch2)
axes[0, 1].set_title('Patch 2')
axes[1, 0].plot(bestPath)
axes[1, 0].set_title('Cost Cut')
axes[1, 1].imshow(costImage)
axes[1, 1].plot(bestPath)
axes[1, 1].set_title('Cost Image with Cost Cut')

```

```

[22]: Text(0.5, 1.0, 'Cost Image with Cost Cut')

```



1.1.5 part IV: Texture Transfer (30 pts)

```
[23]: # Update helper functions
# The main difference between this function and quilt_cut is that there is an
# additional cost term
# based on the difference between the sampled source patch and the target patch
# at the location to be filled.
def choose_sample_with_extra_cost(targetImage, patch_size, M, T, tol,
    allPatchesDict):
    tolerance = tol
    imageCosts = {}
    imageCostSSD = {}
    candidates = []
    imageCostsArray = []
    randomIndex = (np.random.choice(len(allPatchesDict), int(0.01 *
    len(allPatchesDict))))

    for i in randomIndex:
        # calculating the additional cost. as noted from the paper
        alpha = np.sum(M) / (M.shape[0] * M.shape[1])
        cost = ssd_patch2(M, T, allPatchesDict[i])
        cost2 = ssd_patch2(np.ones([patch_size, patch_size]), targetImage,
    allPatchesDict[i])
        imageCosts[i] = alpha * np.sum(cost) + (1 - alpha) * np.sum(cost2)
```

```

        imageCostSSD[i] = cost

    minc = min(imageCosts.values())
    for key in imageCosts.keys():
        if imageCosts[key] <= minc * (1 + tolerance):
            candidates.append(allPatchesDict[key])
            imageCostsArray.append(imageCostSSD[key])

    randomIndexSelection = np.random.choice(len(candidates))
    return candidates[randomIndexSelection], ␣
↪imageCostsArray[randomIndexSelection]

```

```

[24]: def texture_transfer(texture, target, patch_size, overlap, tol):
    texture = texture.copy() / 255.0
    output = target.copy() / 255.0
    patches = allPatches(texture, patch_size)

    for x in range(0, target.shape[0], patch_size - overlap):
        overlapX = x + overlap
        endX = x + patch_size
        if target.shape[0] - x < patch_size:
            break
        for y in range(0, target.shape[1], patch_size - overlap):
            overlapY = y + overlap
            endY = y + patch_size
            if target.shape[1] - y < patch_size:
                continue
            M = (output[x : endX, y : endY, 0] != 0).astype(int)
            T = (output[x : endX, y : endY, :])
            newTexturedImage, cost = choose_sample_with_extra_cost(output[x:
↪endX, y:endY], patch_size, M, T, tol, patches)

            if x == 0 and y == 0:
                addSampleToResultImage(x, y, overlap, patch_size, ␣
↪newTexturedImage, output)
            elif x == 0 and y != 0:
                template = cost[:, 0 : overlap]
                mask1 = cut(template.T).T
                overlappedRegion = output[x : endX, y: overlapY, :]
                for z in range(3):
                    overlappedRegion[:, :, z] = newTexturedImage[:, :, overlap, z] ␣
↪mask1 + (1 - mask1) * overlappedRegion[:, :, z]
                    output[x:endX, overlapY:endY, :] = newTexturedImage[:, overlap:, :
↪]

            elif x != 0 and y == 0:
                mask2 = cut(cost[0 : overlap, :])
                overlappedRegion = output[x : overlapX, y: endY, :]

```

```

        for z in range(3):
            overlappedRegion[:, :, z] = newTexturedImage[:, :, z] *
↪mask2 + (1 - mask2) * overlappedRegion[:, :, z]
            output[overlapX:endX, y:endY, :] = newTexturedImage[overlap[:, :, :]]
        else:
            template = cost[:, 0 : overlap]
            mask3 = cut(template.T).T
            overlappedRegion = output[x : endX, y : overlapY, :]
            for z in range(3):
                overlappedRegion[:, :, z] = newTexturedImage[:, :, z] *
↪mask3 + (1 - mask3) * overlappedRegion[:, :, z]
                mask4 = cut(cost[0 : overlap, :])
                overlappedRegion = output[x : overlapX, y : endY, :]
                for z in range(3):
                    overlappedRegion[:, :, z] = newTexturedImage[:, :, z] *
↪mask4 + (1 - mask4) * overlappedRegion[:, :, z]
                    output[overlapX:endX, overlapY:endY, :] =
↪newTexturedImage[overlap[:, :, :]]
            return output

```

```

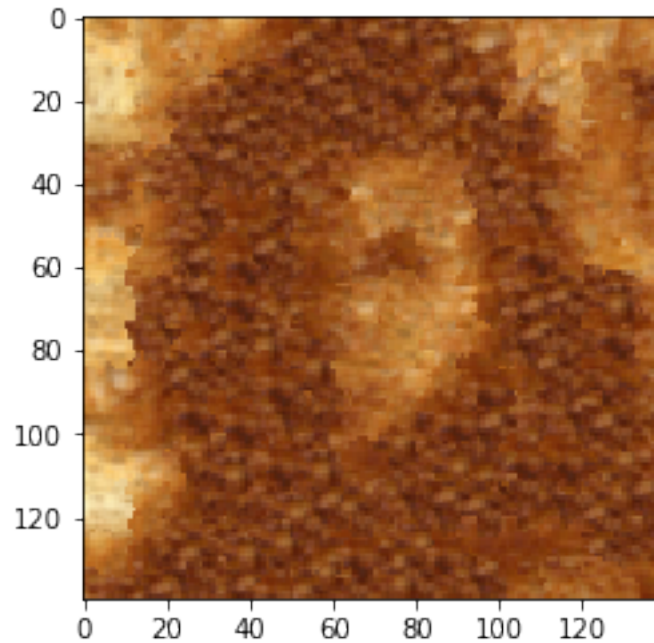
[25]: # Image Result #1. One is mine as the target and the other is the source texture
sizeDown = cv2.pyrDown(image_6)
sizeDownPerson = cv2.pyrDown(image_11)
coloredToast = cv2.cvtColor(sizeDown, cv2.COLOR_BGR2RGB)
coloredPerson = cv2.cvtColor(sizeDownPerson, cv2.COLOR_BGR2RGB)[: , 75:-75, :]
# print(coloredToast.shape)
# print(coloredPerson.shape)
# plt.imshow(coloredToast)
# plt.imshow(coloredPerson)
# plt.show()
wonderToast = texture_transfer(coloredToast, coloredPerson, 15, 5, 0.1)
# plt.imshow(wonderToast[:, :, :])
plt.imshow(wonderToast[: -10, : -10, :])

```

```

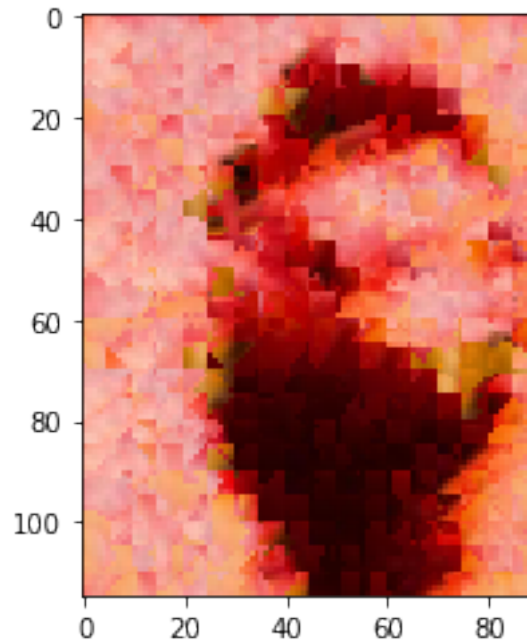
[25]: <matplotlib.image.AxesImage at 0x26f16db50>

```

```
[26]: # Image Result #1. One is mine as the target and the other is the source texture
sizeDown = cv2.pyrDown(image_8)
sizeDownPerson = cv2.pyrDown(image_10)
coloredFruit = cv2.cvtColor(sizeDown, cv2.COLOR_BGR2RGB)[125:225,250:350,:]
coloredPerson = cv2.cvtColor(sizeDownPerson, cv2.COLOR_BGR2RGB)[25:150,200:300,:
↪]
# plt.imshow(coloredFruit)
# plt.imshow(coloredPerson)
# plt.show()
aquaApple = texture_transfer(coloredFruit, coloredPerson, 10, 5, 0.1)
# plt.imshow(aquaApple[:, :, :])
plt.imshow(aquaApple[:-10, :-10, :])
```

```
[26]: <matplotlib.image.AxesImage at 0x1a330eb90>
```



1.1.6 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.

[]:

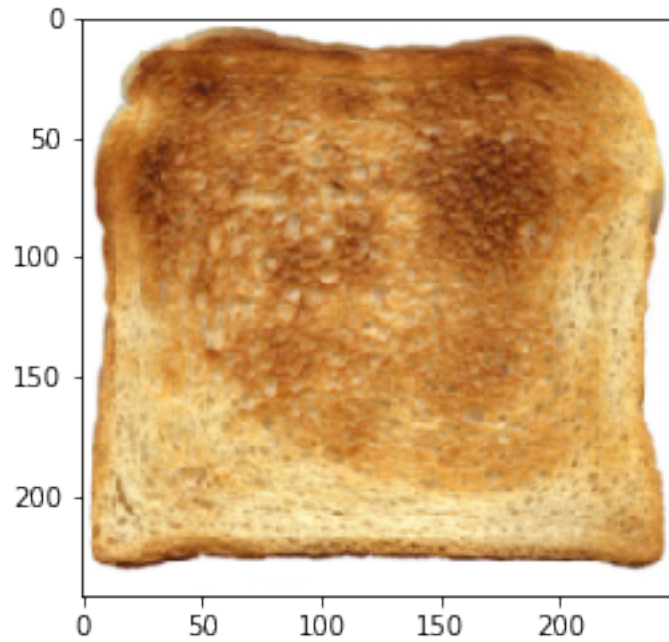
(up to 20 pts) Use a combination of texture transfer and blending to create a face-in-toast image like the one on top. To get full points, you must use some type of blending, such as feathering or Laplacian pyramid blending.

```
[27]: # Texture Transfer
sizeDown2 = cv2.pyrDown(image_6)
sizeDownPerson2 = cv2.pyrDown(image_11)
coloredToast2 = cv2.cvtColor(sizeDown2, cv2.COLOR_BGR2RGB)
coloredPerson2 = cv2.cvtColor(sizeDownPerson2, cv2.COLOR_BGR2RGB)[:,-50,:]
coloredPerson3 = cv2.resize(coloredPerson2, (250, 242), 1, 1)
# print(coloredToast2.shape)
```

```

# print(coloredPerson2.shape)
# print(coloredPerson3.shape)
plt.figure(figsize = (4,4))
plt.imshow(coloredToast)
# plt.imshow(coloredPerson3)
plt.show()
# wonderToast = texture_transfer(coloredToast, coloredPerson3, 15, 5, 0.1)
→ #toast is target

```

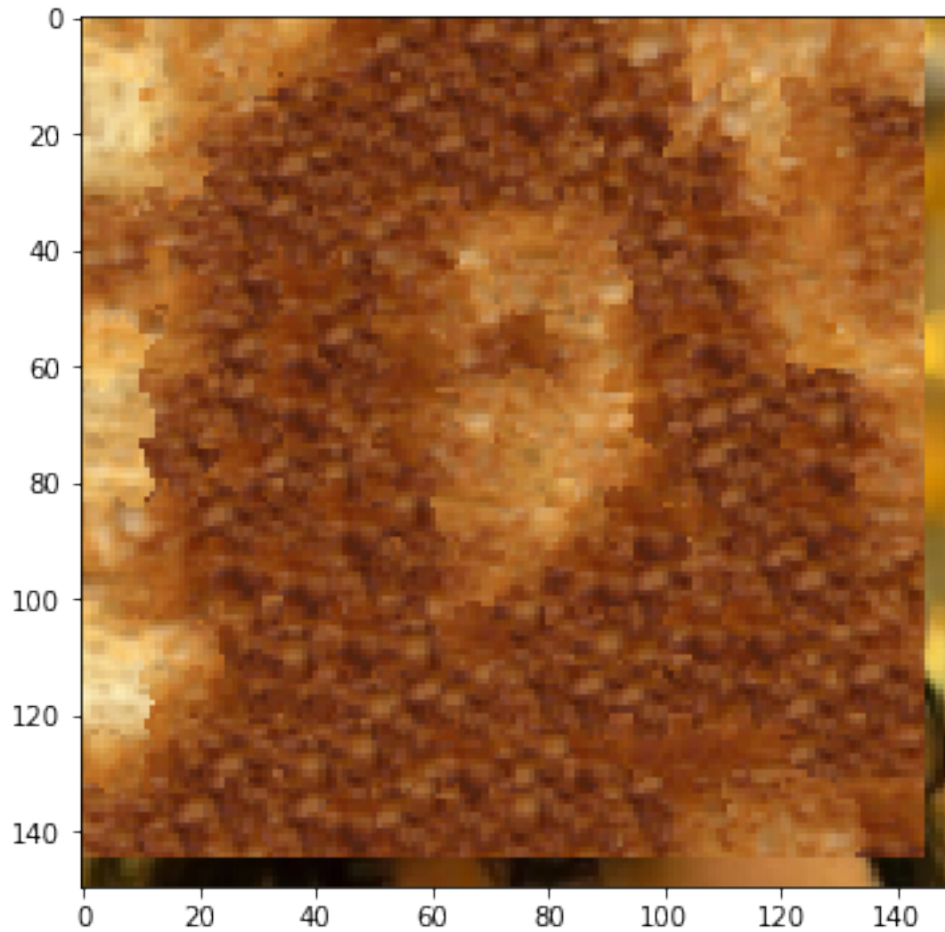


```

[28]: # Check Before Combination
plt.figure(figsize = (6,6))
plt.imshow(wonderToast[:, :, :])

```

[28]: <matplotlib.image.AxesImage at 0x26f557390>

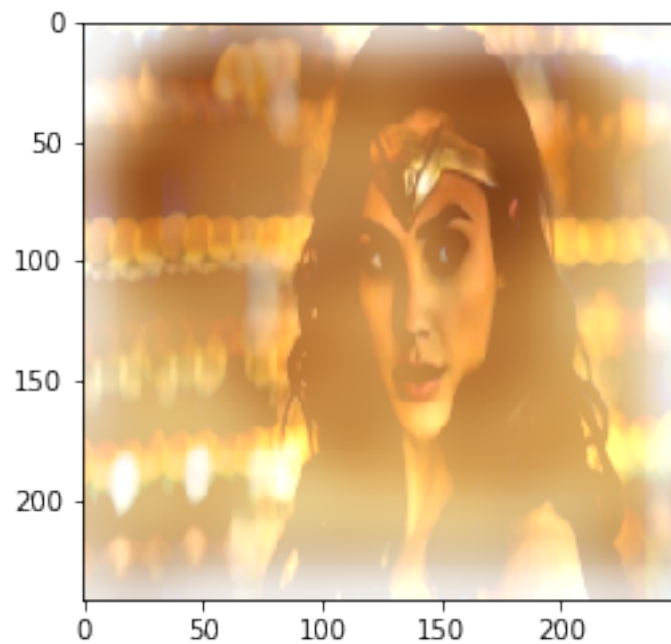


```
[29]: #Blending: Gaussian Blur
def hybridImage(im1, im2, cutoff_low, cutoff_high):
    lowPassImage = cv2.GaussianBlur(im1, (45,45), cutoff_low)
    lowPassImage2 = cv2.GaussianBlur(im2, (45,45), cutoff_high)
    highPassImage = cv2.subtract(im2, lowPassImage2)
    hybridImage = cv2.add(lowPassImage, highPassImage)
    return hybridImage

im1 = cv2.cvtColor(image_6, cv2.COLOR_BGR2RGB) / 255.0
im2 = cv2.cvtColor(image_11, cv2.COLOR_BGR2RGB) / 255.0

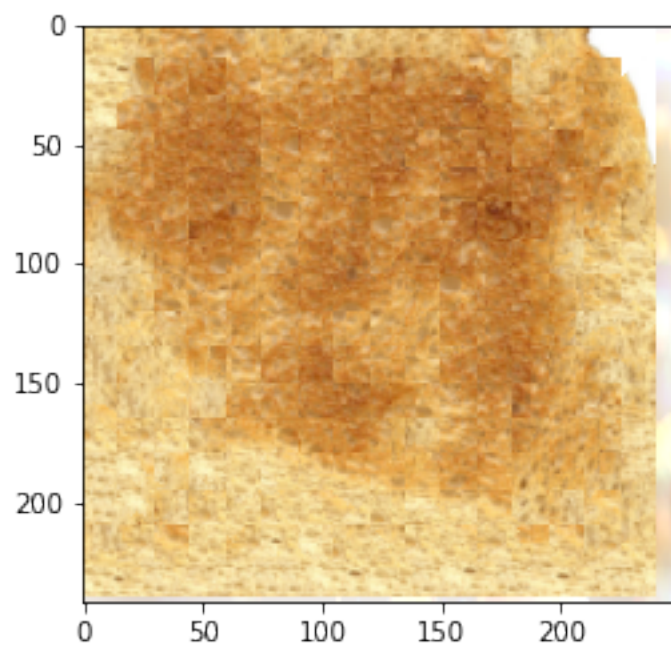
im_hybrid = hybridImage(coloredToast2, coloredPerson3, 10, 30)
plt.figure(figsize = (4,4))
plt.imshow(im_hybrid)
```

```
[29]: <matplotlib.image.AxesImage at 0x26f603bd0>
```



```
[30]: wonderToast2 = texture_transfer(coloredToast, im_hybrid, 30, 15, 0.01)
plt.figure(figsize = (4,4))
plt.imshow(wonderToast2[:, :, :])
```

[30]: <matplotlib.image.AxesImage at 0x2704bf310>




```
[31]: # If you look closely, you can see the outline of her hair from im_hybrid and  
      ↪ the subsequent face.
```

(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.).

```
[ ]:
```