

Texture Quilting (Due Saturday 2/23/2019)

In this assignment, you will develop code to stitch together image patches sampled from an input texture in order to synthesize new texture images. You can download the test image used to generate the example above from assignment folder Canvas.

You should start by reading through the whole assignment, looking at the provided code in detail to make sure you understand what it does. The main function ***quilt_demo*** appears at the end. You will need to write several subroutines in order for it to function properly.

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1. Shortest Path [25 pts]

Write a function ***shortest_path*** that takes an 2D array of ***costs***, of shape HxW, as input and finds the *shortest vertical path* from top to bottom through the array. A vertical path is specified by a single horizontal location for each row of the H rows. Locations in successive rows should not differ by more than 1 so that at each step the path either goes straight or moves at most one pixel to the left or right. The cost is the sum of the costs of each entry the path traverses. Your function should return an length H vector that contains the index of the path location (values in the range 0..W-1) for each of the H rows.

You should solve the problem by implementing the dynamic programming algorithm described in class. You will have a for-loop over the rows of the "cost-to-go" array (M in the slides), computing the cost of the shortest path up to that row using the recursive formula that depends on the costs-to-go for the previous row. Once you have get to the last row, you can then find the smallest total cost. To find the path which actually has this smallest cost, you will need to do backtracking. The easiest way to do this is to also store the index of whichever minimum was selected at each location. These indices will also be an HxW array. You can then backtrack through these indices, reading out the path.

Finally, you should create at least three test cases by hand where you know the shortest path and see that the code gives the correct answer.

```
In [131]: #modules used in your code
import numpy as np
import matplotlib.pyplot as plt
```

```
In [132]: def shortest_path(costs):
```

```
    """
```

This function takes an array of costs and finds a shortest path from the top to the bottom of the array which minimizes the total costs along the path. The path should not shift more than 1 location left or right between successive rows.

In other words, the returned path is one that minimizes the total cost:

```
total_cost = costs[0,path[0]] + costs[1,path[1]] + costs[2,path[2]] +
```

```
...
```

subject to the constraint that:

```
abs(path[i]-path[i+1])<=1
```

Parameters

*costs : 2D float array of shape HxW
An array of cost values*

Returns

*path : 1D array of length H
indices of a vertical path. path[i] contains the column index of
the path for each row i.*

```
"""
```

```
M = costs.copy();
minCostArrayIndex = np.zeros(shape = M.shape);
minima = 0;
for i in reversed(range(M.shape[0])):
    for j in range(0, M.shape[1]):
        if(i == M.shape[0]-1):
            temp = j;
            minCostArrayIndex[i,j] = j;
            minima = 0;
        elif (j == 0):
            temp = M[i+1, j:j+2].argmin();
            minCostArrayIndex[i,j] = temp+j;
            minima = M[i+1, temp+j];
        else:
            temp = M[i+1, j-1:j+2].argmin();
            minCostArrayIndex[i,j] = temp+j-1;
            minima = M[i+1, temp+j-1];
        M[i,j] += minima;
path = np.zeros(shape = M.shape[0], dtype = int)
path[0] = np.argmin(M[0])
for i in range(1,M.shape[0]):
    path[i] = minCostArrayIndex[(i-1), path[i-1]]
return path
```

In [133]:

```
# your test code goes here. come up with at least 3 test cases
#
costs1 = np.array([[2, 1, 3, 4],[6, 11, 2, 12],[4, 5, 13, 5]]);
path1 = shortest_path(costs1)
print(costs1)
print(path1)

costs2 = np.array([[9, 9, 9, 0, 9],[2, 4, 0, 2, 0],[2, 4, 2, 0, 2],[2, 4, 2, 0, 2],[2, 4, 2, 0, 2],[0, 2, 3, 4, 1]]);
path2 = shortest_path(costs2)
print(costs2)
print(path2)

costs3 = np.array([[25, 0, 4, 2, 0],[4, 35, 7, 1, 1],[1, 5, 1, 8, 3],[1, 40, 0, 55, 3],[7, 2, 1, 65, 0],[0, 1, 10, 50, 100]]);
path3 = shortest_path(costs3)
print(costs3)
print(path3)
```

```
[[ 2  1  3  4]
 [ 6 11  2 12]
 [ 4  5 13  5]]
[1 2 1]
[[9 9 9 0 9]
 [2 4 0 2 0]
 [2 4 2 0 2]
 [2 4 2 0 2]
 [2 4 2 0 2]
 [0 2 3 4 1]]
[3 2 3 3 3 4]
[[ 25    0    4    2    0]
 [ 4   35    7    1    1]
 [ 1    5    1    8    3]
 [ 1   40    0   55    3]
 [ 7    2    1   65    0]
 [ 0    1   10   50 100]]
[4 3 2 2 1 0]
```

2. Image Stitching: [25 pts]

Write a function ***stitch*** that takes two gray-scale images, ***left_image*** and ***right_image*** and a specified ***overlap*** and returns a new output image by stitching them together along a seam where the two images have very similar brightness values. If the input images are of widths ***w1*** and ***w2*** then your stitched result image returned by the function should be of width ***w1+w2-overlap*** and have the same height as the two input images.

You will want to first extract the overlapping strips from the two input images and then compute a cost array given by the absolute value of their difference. You can then use your ***shortest_path*** function to find the seam along which to stitch the images where they differ the least in brightness. Finally you need to generate the output image by using pixels from the left image on the left side of the seam and from the right image on the right side of the seam. You may find it easiest to code this by first turning the path into an alpha mask for each image and then using the standard equation for compositing.

```
In [134]: def stitch(left_image, right_image, overlap):
    """
        This function takes a pair of images with a specified overlap and stitches
        them
        togther by finding a minimal cost seam in the overlap region.

    Parameters
    -----
    left_image : 2D float array of shape HxW1
        Left image to stitch

    right_image : 2D float array of shape HxW2
        Right image to stitch

    overlap : int
        Width of the overlap zone between left and right image

    Returns
    -----
    stitched : 2D float array of shape Hx(W1+W2-overlap)
        The resulting stitched image
    """

# inputs should be the same height
assert(left_image.shape[0]==right_image.shape[0])

left_overlap = left_image.shape[1] - overlap;
overlap_image = abs(left_image[:,left_overlap:] - right_image[:,overlap]);
shortest = shortest_path(overlap_image)
stitched = np.zeros(shape = (left_image.shape[0],left_image.shape[1]+right_image.shape[1]-overlap))
for i in range(0,left_image.shape[0]):
    stitched[i,:(left_image.shape[1] - overlap + shortest[i]+1)] = left_image[i,: (left_image.shape[1] - overlap + shortest[i]+1)]
    stitched[i,(left_image.shape[1] - overlap + shortest[i]+1):] = right_image[i,(shortest[i]+1):]

assert(stitched.shape[0]==left_image.shape[0])
assert(stitched.shape[1]==(left_image.shape[1]+right_image.shape[1]-overlap))

return stitched
```

In []:

3. Texture Quilting: [25 pts]

Write a function ***synth_quilt*** that takes as input an array indicating the set of texture tiles to use, an array containing the set of available texture tiles, the ***tilesize*** and ***overlap*** parameters and synthesizes the output texture by stitching together the tiles. ***synth_quilt*** should utilize your ***stitch*** function repeatedly. First, for each horizontal row of tiles, construct the stitched row by repeatedly stitching the next tile in the row on to the right side of your row image. Once you have row images for all the rows, you can stitch them together to get the final image. Since your ***stitch*** function only works for vertical seams, you will want to transpose the rows, stitch them together, and then transpose the result. You may find it useful to look at the provided code below which simply puts down the tiles with the specified overlap but doesn't do stitching. Your quilting function will return a similar result but with much smoother transitions between the tiles.

```
In [135]: def synth_quilt(tile_map,tiledb,tilesize,overlap):
    """
    This function takes the name of an image and quilting parameters and synthesizes a new texture image by stitching together sampled tiles from the source image.
    """

    Parameters
    -----
    tile_map : 2D array of int
        Array storing the indices of which tiles to paste down at each output location

    tiledb : list of int
        Dimensions of output in tiles, e.g. (3,4)

    tilesize : (int,int)
        Size of a tile in pixels

    overlap : int
        Amount of overlap between tiles

    Returns
    -----
    output : 2D float array
        The resulting synthesized texture of size
    """

    # determine output size based on overlap and tile size
    outh = (tilesize[0]-overlap)*tile_map.shape[0] + overlap
    outw = (tilesize[1]-overlap)*tile_map.shape[1] + overlap
    output = np.zeros((outh,outw))

    # The code below is a dummy implementation that pastes down each
    # tile in the correct position in the output image. You need to
    # replace this with your own version that stitches each row and then
    # stitches together the columns
    rowImages = np.zeros(shape = (tilesize[0],output.shape[1],output.shape[0]))
    for i in range(tile_map.shape[0]):
        row = tiledb[tile_map[i,0],:];
        row_image = np.reshape(row,tilesize);
        for j in range(1,tile_map.shape[1]):
            tile = tiledb[tile_map[i,j],:];
            tile_image = np.reshape(tile, tilesize)
            row_image = stitch(row_image, tile_image, overlap)
            rowImages[:, :, i] = row_image
        output = rowImages[:, :, 0].T
    for i in range(1,tile_map.shape[0]):
        output = stitch(output, rowImages[:, :, i].T, overlap);

    output = output.T
    return output
```

In []:

4. Texture Synthesis Demo [25pts]

The function provided below ***quilt_demo*** puts together the pieces. It takes a sample texture image and a specified output size and uses the functions you've implemented previously to synthesize a new texture sample.

You should write some additional code in the cells that follow to in order demonstrate the final result and experiment with the algorithm parameters in order to produce a compelling visual result and write explanations of what you discovered.

Test your code on the provided image *rock_wall.jpg*. There are three parameters of the algorithm. The *tilesize*, *overlap* and *K*. In the provided ***texture_demo*** code below, these have been set at some default values. Include in your demo below images of three example texture outputs when you: (1) increase the tile size, (2) decrease the overlap, and (3) decrease the value for *K*. For each result explain how it differs from the default setting of the parameters and why.

Test your code on two other texture source images of your choice. You can use images from the web or take a picture of a texture yourself. You may need to resize or crop your input image to make sure that the ***tiledb*** is not overly large. You will also likely need to modify the ***tilesize*** and ***overlap*** parameters depending on your choice of texture. Once you have found good settings for these parameters, synthesize a nice output texture. Make sure you display both the image of the input sample and the output synthesis for your two other example textures in your submitted pdf.

```
In [138]: #skimage is only needed for sample_tiles code provided below
#you should not use it in your own code
import skimage as ski

def sample_tiles(image,tilesize,randomize=True):
    """
        This function generates a library of tiles of a specified size from a given source image

    Parameters
    -----
    image : float array of shape HxW
        Input image

    tilesize : (int,int)
        Dimensions of the tiles in pixels

    Returns
    -----
    tiles : float array of shape npixels x numtiles
        The library of tiles stored as vectors where npixels is the product of the tile height and width
    """

    tiles = ski.util.view_as_windows(image,tilesize)
    ntiles = tiles.shape[0]*tiles.shape[1]
    npix = tiles.shape[2]*tiles.shape[3]
    assert(npix==tilesize[0]*tilesize[1])

    print("library has ntiles = ",ntiles,"each with npix = ",npix)

    tiles = tiles.reshape((ntiles,npix))

    # randomize tile order
    if randomize:
        tiles = tiles[np.random.permutation(ntiles),:]
    return tiles

def topkmatch(tilestrip,dbstrips,k):
    """
        This function finds the top k candidate matches in dbstrips that are most similar to the provided tile strip.

    Parameters
    -----
    tilestrip : 1D float array of Length npixels
        Grayscale values of the query strip

    dbstrips : 2D float array of size npixels x numtiles
        Array containing brightness values of numtiles strips in the database to match to the npixels brightness values in tilestrip

    k : int
        Number of top candidate matches to sample from
```

```

>Returns
-----
matches : List of ints of length k
    The indices of the k top matching tiles
"""

assert(k>0)
assert(dbstrips.shape[0]>k)
error = (dbstrips-tilestrip)
ssd = np.sum(error*error, axis=1)
ind = np.argsort(ssd)
matches = ind[0:k]
return matches

def quilt_demo(sample_image, ntilesout=(10,20), tilesize=(30,30), overlap=5, k=5):
    """
    This function takes the name of an image

    Parameters
    -----
    sample_image : 2D float array
        Grayscale image containing sample texture

    ntilesout : List of int
        Dimensions of output in tiles, e.g. (3,4)

    tilesize : int
        Size of the square tile in pixels

    overlap : int
        Amount of overlap between tiles

    k : int
        Number of top candidate matches to sample from

>Returns
-----
img : List of int of length K
    The resulting synthesized texture of size
"""

# generate database of tiles from sample
tiledb = sample_tiles(sample_image,tilesize)
# number of tiles in the database
nsampletiles = tiledb.shape[1]

if (nsampletiles<k):
    print("Error: tile database is not big enough!")

# generate indices of the different tile strips
i,j = np.mgrid[0:tilesize[0],0:tilesize[1]]
top_ind = np.ravel_multi_index(np.where(i<overlap),tilesize)
bottom_ind = np.ravel_multi_index(np.where(i>=tilesize[0]-overlap),tilesize)
left_ind = np.ravel_multi_index(np.where(j<overlap),tilesize)

```

```

    right_ind = np.ravel_multi_index(np.where(j>=tilesize[1]-overlap),tilesize
)

# initialize an array to store which tile will be placed
# in each location in the output image
tile_map = np.zeros(ntilesout,'int')

#print('row:')
for i in range(ntilesout[0]):
    #print(i)
    for j in range(ntilesout[1]):

        if (i==0)&(j==0):                      # first row first tile
            matches = np.zeros(k) #range(nsamplletiles)

        elif (i==0):                           # first row (but not first tile)
            left_tile = tile_map[i,j-1]
            tilestrip = tiledb[left_tile,right_ind]
            dbstrips = tiledb[:,left_ind]
            matches = topkmatch(tilestrip,dbstrips,k)

        elif (j==0):                           # first column (but not first row)
            above_tile = tile_map[i-1,j]
            tilestrip = tiledb[above_tile,bottom_ind]
            dbstrips = tiledb[:,top_ind]
            matches = topkmatch(tilestrip,dbstrips,k)

        else:                                 # neighbors above and to the left
            left_tile = tile_map[i,j-1]
            tilestrip_1 = tiledb[left_tile,right_ind]
            dbstrips_1 = tiledb[:,left_ind]
            above_tile = tile_map[i-1,j]
            tilestrip_2 = tiledb[above_tile,bottom_ind]
            dbstrips_2 = tiledb[:,top_ind]
            # concatenate the two strips
            tilestrip = np.concatenate((tilestrip_1,tilestrip_2))
            dbstrips = np.concatenate((dbstrips_1,dbstrips_2),axis=1)
            matches = topkmatch(tilestrip,dbstrips,k)

        #choose one of the k matches at random
        tile_map[i,j] = matches[np.random.randint(0,k)]

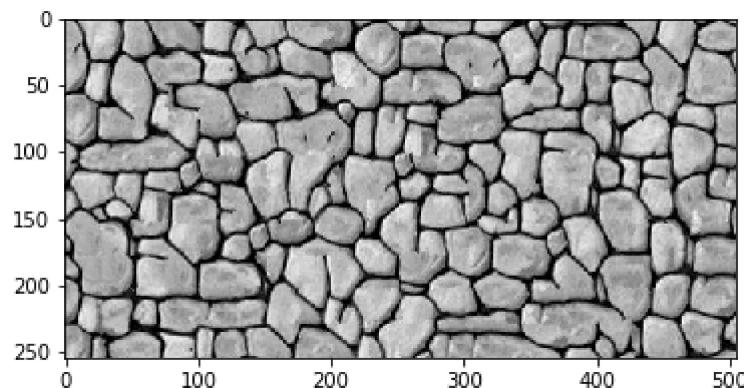

output = synth_quilt(tile_map,tiledb,tilesize,overlap)

return output

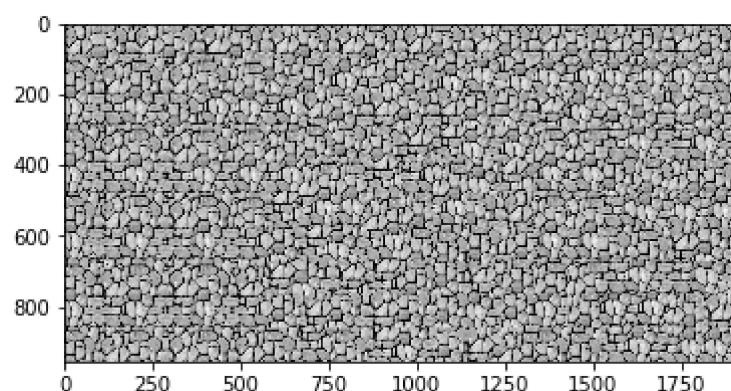
```

```
In [139]: # Load in rock_wall.jpg
# run and display results for quilt_demo with
#
# (0) default parameters
# (1) increased tile size
# (2) decrease the overlap
# (3) increase the value for K.
np.random.seed(0)
img = plt.imread('rock_wall.jpg')
if(img.dtype == np.uint8):
    img = img.astype(float)/256
img_g = np.zeros((img.shape[0], img.shape[1]))
img_g[:, :] = (img[:, :, 0] + img[:, :, 1] + img[:, :, 2]) / 3
#DEFAULT
o = quilt_demo(img_g)
plt.imshow(o, cmap = 'gray')
plt.show()
#INCREASED TILE SIZE
o = quilt_demo(img_g, tilesize = (100,100))
plt.imshow(o, cmap = 'gray')
plt.show()
#DECREASED OVERLAP
o = quilt_demo(img_g, overlap = 2)
plt.imshow(o, cmap = 'gray')
plt.show()
#INCREASED K
o = quilt_demo(img_g, k = 30)
plt.imshow(o, cmap = 'gray')
plt.show()
```

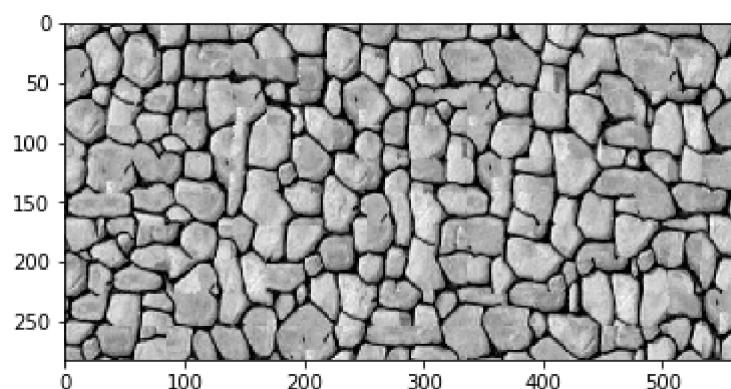
library has ntiles = 29241 each with npix = 900



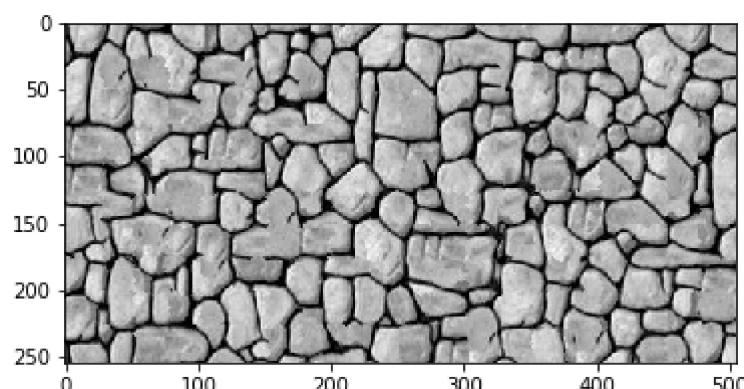
library has ntiles = 10201 each with npix = 10000



library has ntiles = 29241 each with npix = 900



library has ntiles = 29241 each with npix = 900



For each result shown, explain here how it differs visually from the default setting of the parameters and explain why:

DEFAULT Default texture has similar looks to the original picture, which makes sense since we are sampling a small portion of it as a tile with a decent overlap so the texture looks as we expected. However we can still see some imperfections .

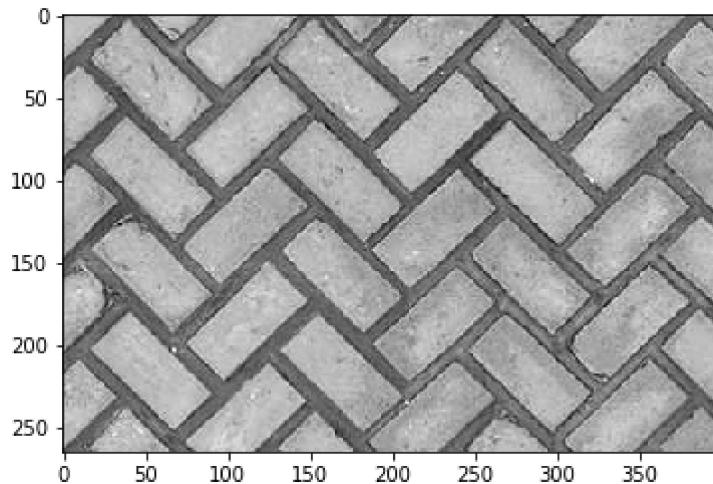
Increased tile This increased tile size again is very apparent, we can see how a much larger portion of the image was sampled as a tile so the whole texture looks more small and concentrated

decreased overlap Here we have decreased overlap of the texture, which means fewer points where the shortest path function can use to find a good place to stitch, while this may reduce the overall quality of the seam, it can also help make things faster.

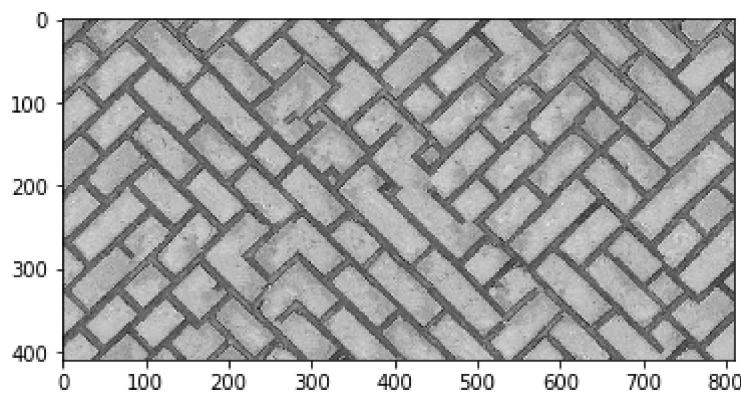
increased K in here we have a k value much higher, which means our subsampling is more than it used to be in default settings, this means stitching will be a lot smoother and as we can see the imperfections from previous part have been more or less eliminated.

In [146]:

```
# Load in yourimage1.jpg
#
# call quilt_demo, experiment with parameters if needed to get a good result
#
# display your source image and the resulting synthesized texture
#
img = plt.imread('brick_floor.jpg')
if(img.dtype == np.uint8):
    img = img.astype(float)/256
img_g = np.zeros((img.shape[0], img.shape[1]))
img_g[:, :] = (img[:, :, 0] + img[:, :, 1] + img[:, :, 2]) / 3
plt.imshow(img_g, cmap = 'gray')
plt.show()
o = quilt_demo(img_g, (10,20), (50,50), 10, 10)
plt.imshow(o, cmap = 'gray')
plt.show()
```

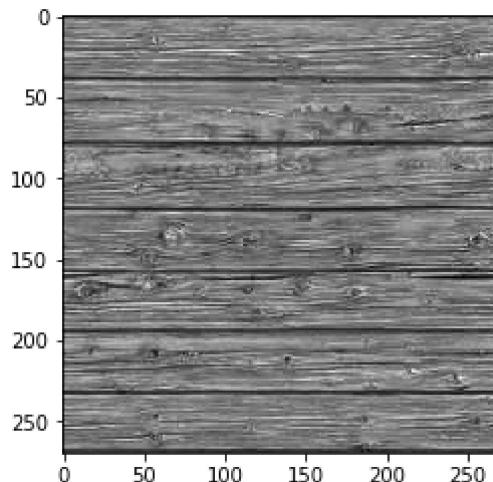


library has ntiles = 75384 each with npix = 2500

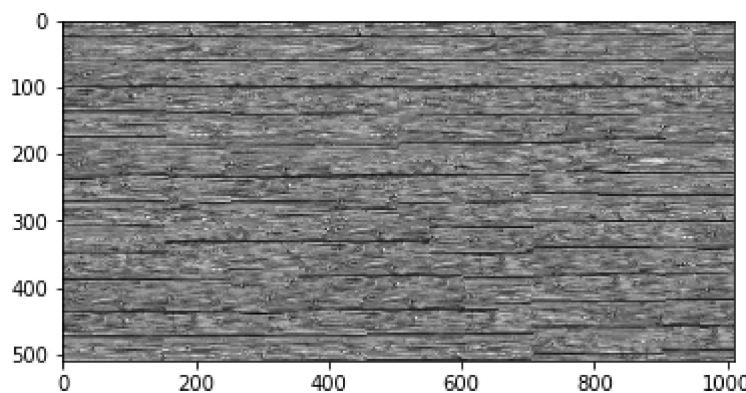


In [144]:

```
# Load in yourimage2.jpg
#
# call quilt_demo, experiment with parameters if needed to get a good result
#
# display your source image and the resulting synthesized texture
#
img = plt.imread('wood.jpg')
if(img.dtype == np.uint8):
    img = img.astype(float)/256
img_g = np.zeros((img.shape[0], img.shape[1]))
img_g[:, :] = (img[:, :, 0] + img[:, :, 1] + img[:, :, 2]) / 3
plt.imshow(img_g, cmap = 'gray')
plt.show()
o = quilt_demo(img_g, (10,20), (60,60), 10, 10)
plt.imshow(o, cmap = 'gray')
plt.show()
```



library has ntiles = 44521 each with npix = 3600



In []:

In []: