

ex7

May 8, 2019

0.0.1 TODO Exercise 7. ex

RENOIR ### You are given a "renoir" set of two images (reference and noisy) from the RENOIR dataset 2. (a) Visualise all color channels of both images. Are the all channels equally affected by the noise?

No green not affected as much as red and blue.

```
[1]: import scipy.ndimage as ndimage
import matplotlib.pyplot as plt
import cv2
import numpy as np

im1 = ndimage.imread("./data/renoir/Reference.bmp")
im1 = np.asarray(im1)
im2 = ndimage.imread("./data/renoir/Noisy.bmp")
im2 = np.asarray(im2)
plt.imshow(im1, cmap=plt.cm.Spectral)
plt.show()
plt.imshow(im2, cmap=plt.cm.Spectral)
plt.show()

#plt.colormaps()

red_images = im2[:, :, 0]
green_images = im2[:, :, 1]
blue_images = im2[:, :, 2]
plt.imshow(red_images, cmap=plt.cm.Red)
plt.show()
plt.imshow(green_images, cmap=plt.cm.Green)
plt.show()
plt.imshow(blue_images, cmap=plt.cm.Blue)
plt.show()
```

```
/usr/local/lib/python3.5/dist-packages/ipykernel_launcher.py:6:
DeprecationWarning: `imread` is deprecated!
`imread` is deprecated in SciPy 1.0.0.
Use ``matplotlib.pyplot.imread`` instead.
```

```
/usr/local/lib/python3.5/dist-packages/ipykernel_launcher.py:8:
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```

<Figure size 640x480 with 1 Axes>

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0.0.2 (b) Try to decrease the noise via image down/up sampling.

b.a) Do it for the RGB image. Measure the PSNR between the reference and de-noised images.

```
[33]: from helpers import mse
import math
# Downsample an image by skipping indicies

def downsample_image(img, skip):
    return img[::skip,::skip]

def psnr(mseries, PIXEL_MAX_SQUARE = 255):
    mse_square = math.sqrt(mseries )
    part1 = math.log10(PIXEL_MAX_SQUARE / mse_square)
    return 20 * part1

# downsampling
psnr_list = [psnr(mse( downsample_image(im1, skip), downsample_image(im2, skip))) for skip in range(1,100)]
plt.plot([skip for skip in range(1,100)], psnr_list)
plt.xlabel("pixel downsample")
plt.ylabel("psnr")
plt.show()

print("no downsampled pair value psnr: " )
print(psnr(mse(im1,im2)))
print("max psnr downsampled value: ")
```

```

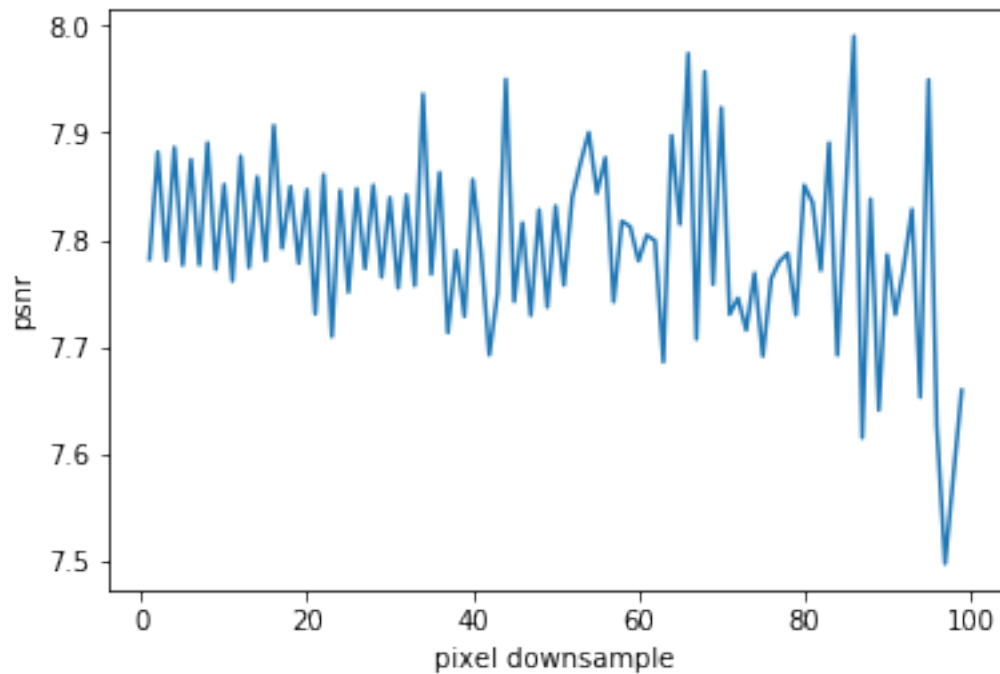
print(max(psnr_list))
print("best downsampling skip value: ")

print(psnr_list.index(max(psnr_list)))
best_ratio = psnr_list.index(max(psnr_list))

plt.imshow(downsample_image(im1, best_ratio))
plt.show()
plt.imshow(downsample_image(im2, best_ratio))
plt.show()

# best_ratio, im1.shape, downsample_image(im1, best_ratio).shape

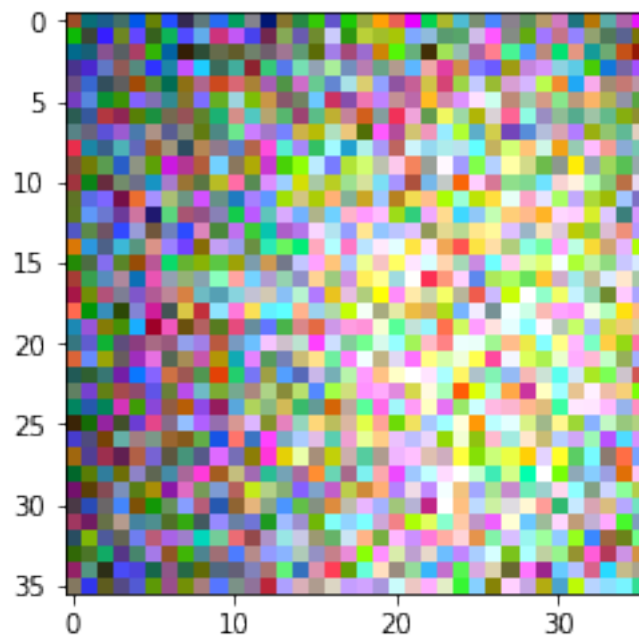
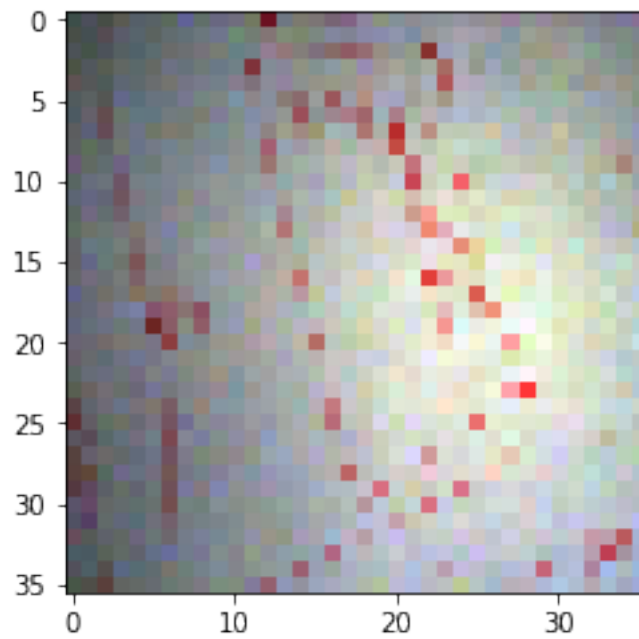
```



```

no downsampled pair value psnr:
7.781946412448223
max psnr downsampled value:
7.989865945271479
best downsampling skip value:
85

```



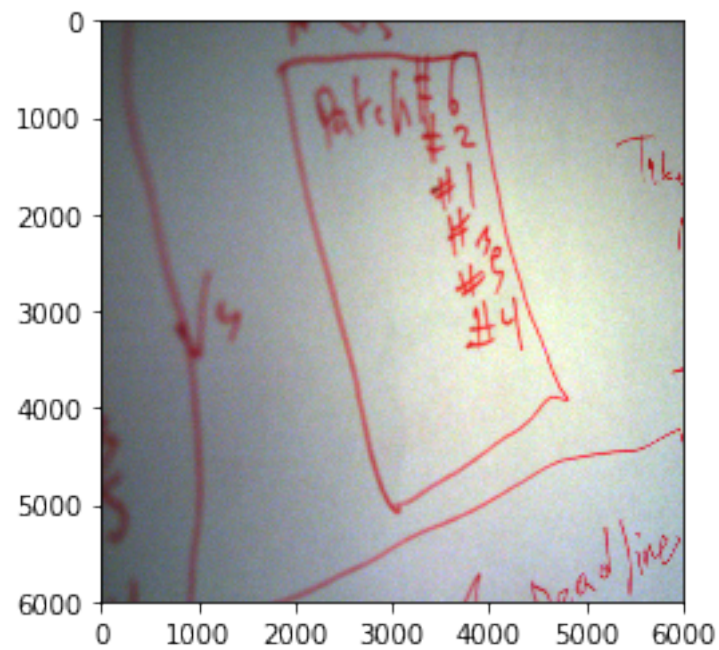
```
[5]: #upsampling using scipy.ndimage
import skimage.transform
def upsample_skimage(factor, input_img):
    # Order=1 is bilinear upsampling
```

```

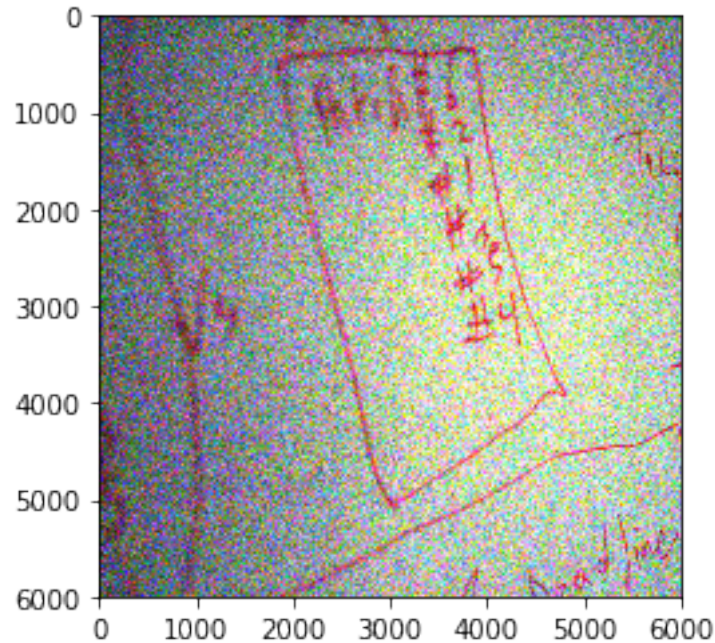
return skimage.transform.rescale(input_img,
                                factor,
                                mode='constant',
                                cval=0,
                                order=1)

upsampled_im1 = upsample_skimage(factor=2, input_img=im1)
upsampled_im2 = upsample_skimage(factor=2, input_img=im2)
plt.imshow(upsampled_im1)
plt.show()
plt.imshow(upsampled_im2)
psnr(mse(upsampled_im1,upsampled_im2))

```



[5]: 57.29515427642124



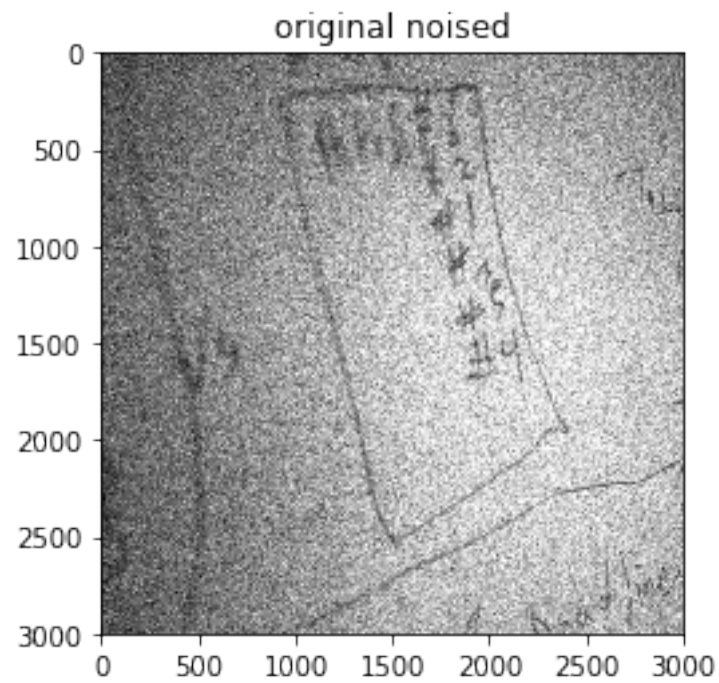
b.b) Do it for the grayscale image. Measure the PSNR between the grayscale reference and de-noised images.

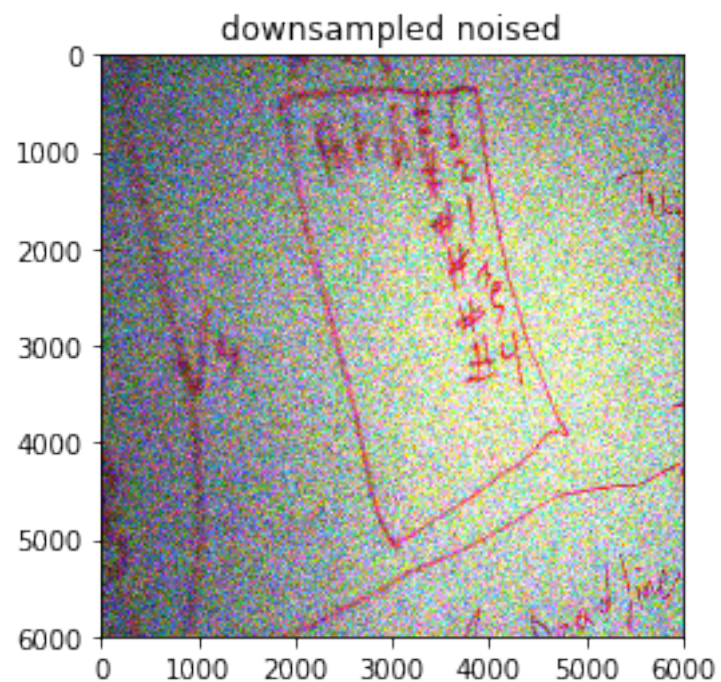
```
[38]: im1_gr = cv2.cvtColor(
        im1,
        cv2.COLOR_RGB2GRAY
    )
    im2_gr = cv2.cvtColor(
        im2,
        cv2.COLOR_RGB2GRAY
    )
    plt.title("original noised")
    plt.imshow(im2_gr, cmap = 'gray' )
    plt.show()

    d1=downsample_image(im1_gr, 3)
    d2=downsample_image(im2_gr, 3)
    plt.imshow(upsampled_im2, cmap = 'gray')
    plt.title("downsampled noised")
    plt.show()
    print("downsampled psnr")
    print(psnr(mse( d1, d2)))

    #upscale
    upsampled_im1 = upsample_skimage(factor=3, input_img=im1_gr)
```

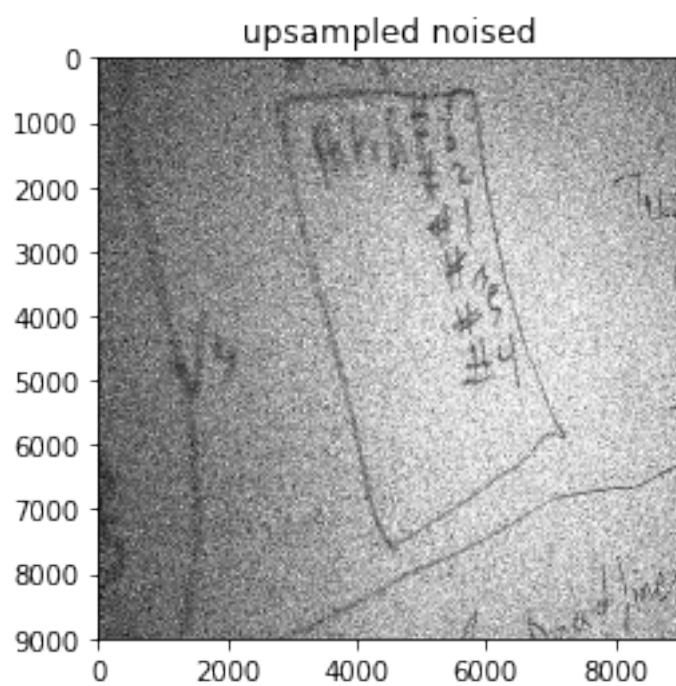
```
upsampled_im2 = upsample_skimage(factor=3, input_img=im2_gr)
plt.title("upsampled noised")
plt.imshow(upsampled_im2, cmap = 'gray')
print("upsampled psnr")
psnr(mse(upsampled_im1, upsampled_im2))
```





downsampled psnr
19.06967847453363
upsampled psnr

[38]: 69.10304970013075

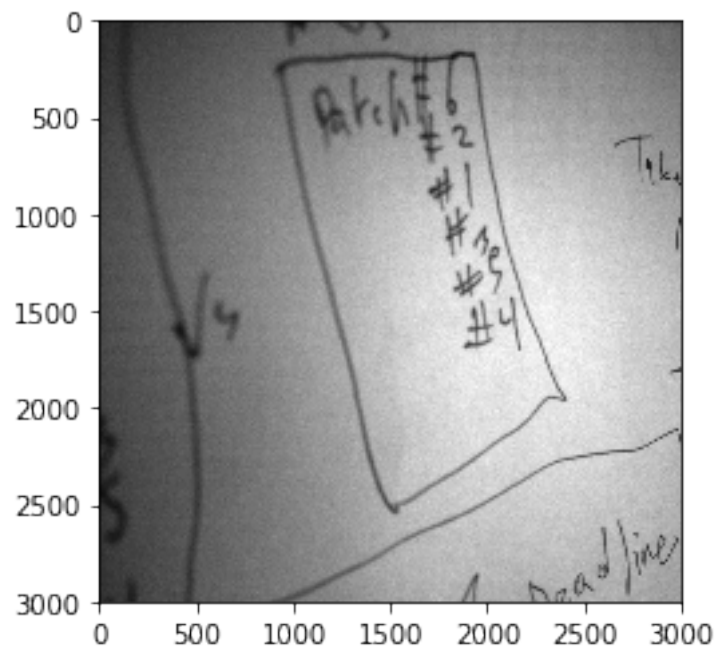


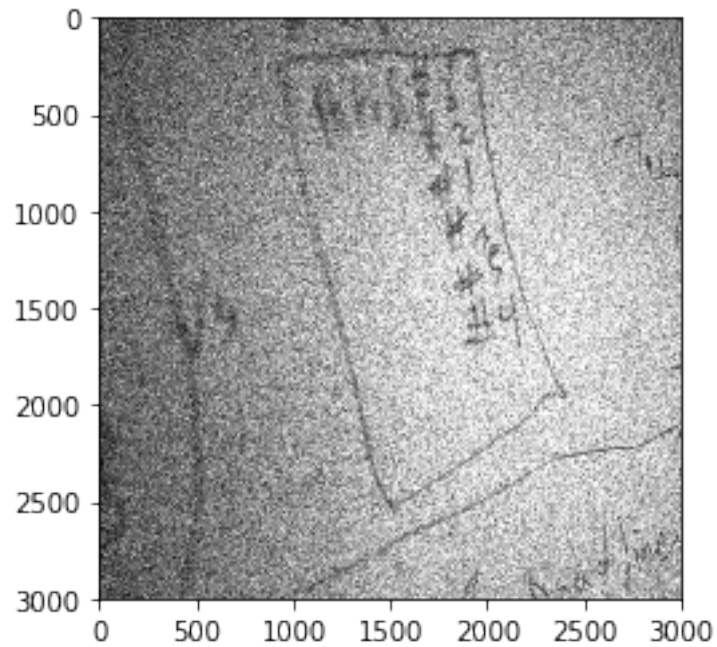
b.c) Do the denoising for the RGB image and after convert it to the grayscale. Measure the PSNR. Does the obtained result is different from the (b)? Explain the result. I used the best from the two techniques the psnr dropped which implies quality drop. Upsampling is sensitive to color mapping since the transform from rgb to grayscale is not linear.

```
[39]: upsampled_im1 = upsample_skiimage(factor=3, input_img=im1)
      upsampled_im2 = upsample_skiimage(factor=3, input_img=im2)

      upsampled_im1_gr = cv2.cvtColor(
          im1,
          cv2.COLOR_RGB2GRAY
      )
      upsampled_im2_gr = cv2.cvtColor(
          im2,
          cv2.COLOR_RGB2GRAY
      )
      plt.imshow(upsampled_im1_gr, cmap='gray')
      plt.show()
      plt.imshow(upsampled_im2_gr, cmap='gray')
      plt.show()

      print(psnr(mse(upsampled_im1_gr, upsampled_im2_gr)))
```





19.076533001482233

0.0.3 (c) What other methods can you suggest to improve the noisy image quality?

Deep neural network autoencoder eg. <https://papers.nips.cc/paper/4686-image-denoising-and-inpainting-with-deep-neural-networks.pdf>