

Generic_Endoscopy_Denoising_Framework_DCNN

March 24, 2023

Endoscopy Denoising using DCNN Framework

```
[51]: #Imports
from __future__ import print_function
import os
import numpy as np
import matplotlib.pyplot as plt
import cv2
import tensorflow as tf
from tensorflow.keras.layers import Input, Conv2D, Activation, BatchNormalization, Add, Multiply, Concatenate
from tensorflow.keras.layers import GlobalAveragePooling2D
from tensorflow.keras.utils import plot_model
from tensorflow.keras.models import Model
from tensorflow.keras import metrics
import datetime
import pandas as pd
import time
import warnings
warnings.filterwarnings('ignore')
```

0.0.1 Data Source

```
[52]: H,W,CH=[120,120,3]
BATCH_SIZE=16
NOISE_LEVELS=[15,25,50]

train_files=['Endoscopy/train/images/'+filename for filename in os.
↳listdir('Endoscopy/train/images/')]
test_files=['Endoscopy/test/images/'+filename for filename in os.
↳listdir('Endoscopy/test/images/')]


```

```
[53]: def _parse_function(filename):
    image_string = tf.io.read_file(filename)
    image_decoded = tf.image.decode_jpeg(image_string, channels=3)
    image_decoded=tf.image.resize(image_decoded, [H,W], method='nearest')
    image = tf.cast(image_decoded, tf.float32)/255.
```

```

noise_level=np.random.choice(NOISE_LEVELS)
noisy_image=image+tf.random.normal(shape=(H,W,CH),mean=0,stddev=noise_level/
↪255)
noisy_image=tf.image.resize(noisy_image, [H,W], method='nearest')
noisy_image=tf.clip_by_value(noisy_image, clip_value_min=0.,↪
↪clip_value_max=1.)

return noisy_image,image

```

0.0.2 Creating train dataset

```

[54]: #Creating the Dataset
train_dataset = tf.data.Dataset.from_tensor_slices(np.array(train_files))
train_dataset = train_dataset.map(_parse_function)
train_dataset = train_dataset.batch(BATCH_SIZE)

test_dataset = tf.data.Dataset.from_tensor_slices(np.array(test_files))
test_dataset = test_dataset.map(_parse_function)
test_dataset = test_dataset.batch(BATCH_SIZE)

```

```

[55]: #data dim
iterator = iter(train_dataset)
a, b = iterator.get_next()

print('Shape of single batch of x : ',a.shape)
print('Shape of single batch of y : ',b.shape)

```

Shape of single batch of x : (16, 120, 120, 3)

Shape of single batch of y : (16, 120, 120, 3)

0.0.3 Exploratory data display

```

[56]: # Display samples

fig, axs = plt.subplots(1,10,figsize=(20,4))
for i in range(10):
    axs[i].imshow(a[i])
    axs[i].get_xaxis().set_visible(False)
    axs[i].get_yaxis().set_visible(False)
    fig.suptitle('Noisy Images',fontsize=20)
fig.align_labels()
plt.tight_layout()
plt.show()

fig, axs = plt.subplots(1,10,figsize=(20,4))
for i in range(10):
    axs[i].imshow(b[i])

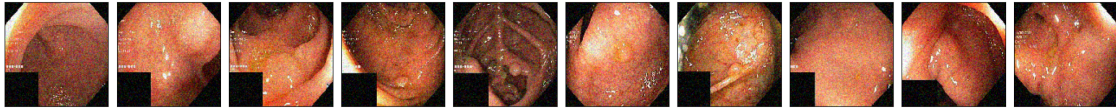
```

```

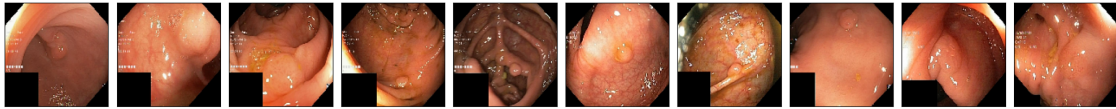
    axs[i].get_xaxis().set_visible(False)
    axs[i].get_yaxis().set_visible(False)
    fig.suptitle('Ground Truth Images',fontsize=20)
fig.align_labels()
plt.tight_layout()
plt.show()

```

Noisy Images



Ground Truth Images



0.0.4 Patches creation

[57]: *# Image patches creation*

```

def get_patches(file_name,patch_size,crop_sizes):
    image = cv2.imread(file_name)
    image=cv2.cvtColor(image,cv2.COLOR_BGR2RGB)
    height, width , channels= image.shape
    patches = []
    for crop_size in crop_sizes: #We will crop the image to different sizes
        crop_h, crop_w = int(height*crop_size),int(width*crop_size)
        image_scaled = cv2.resize(image, (crop_w,crop_h), interpolation=cv2.
↪INTER_CUBIC)
        for i in range(0, crop_h-patch_size+1, patch_size):
            for j in range(0, crop_w-patch_size+1, patch_size):
                x = image_scaled[i:i+patch_size, j:j+patch_size]
                patches.append(x)
    return patches

def create_image_from_patches(patches,image_shape):
    image=np.zeros(image_shape)
    patch_size=patches.shape[1]
    p=0

```

```

for i in range(0,image.shape[0]-patch_size+1,patch_size):
    for j in range(0,image.shape[1]-patch_size+1,patch_size):
        image[i:i+patch_size,j:j+patch_size]=patches[p]
        p+=1
return np.array(image)

def predict_fun(model,image_path,noise_level=30):
    patches=get_patches(image_path,H,[1])
    test_image=cv2.imread(image_path)
    patches=np.array(patches)
    ground_truth=create_image_from_patches(patches,test_image.shape)

    patches = patches.astype('float32') /255.
    patches_noisy = patches+ tf.random.normal(shape=patches.
↪shape,mean=0,stddev=noise_level/255)
    patches_noisy = tf.clip_by_value(patches_noisy, clip_value_min=0.,↪
↪clip_value_max=1.)
    noisy_image=create_image_from_patches(patches_noisy,test_image.shape)
    denoised_patches=model.predict(patches_noisy)

    denoised_patches=tf.clip_by_value(denoised_patches, clip_value_min=0.,↪
↪clip_value_max=1.)

    denoised_image=create_image_from_patches(denoised_patches,test_image.shape)

    return patches_noisy,denoised_patches,ground_truth/255.
↪,noisy_image,denoised_image

```

0.0.5 Plot image patches

```

[58]: #plot patches
def plot_patches(patches_noisy,denoised_patches):
    fig, axs = plt.subplots(2,10,figsize=(20,4))

    for i in range(10):
        axs[0,i].imshow(patches_noisy[i])
        axs[0,i].title.set_text(' Noisy')
        axs[0,i].get_xaxis().set_visible(False)
        axs[0,i].get_yaxis().set_visible(False)

        axs[1,i].imshow(denoised_patches[i])
        axs[1,i].title.set_text('Denoised')
        axs[1,i].get_xaxis().set_visible(False)
        axs[1,i].get_yaxis().set_visible(False)
    plt.show()

```

```
def plot_predictions(ground_truth, noisy_image, denoised_image):
    fig, axs = plt.subplots(1,3,figsize=(15,15))
    axs[0].imshow(ground_truth)
    axs[0].title.set_text('Ground Truth')
    axs[1].imshow(noisy_image)
    axs[1].title.set_text('Noisy Image')
    axs[2].imshow(denoised_image)
    axs[2].title.set_text('Denoised Image')
    plt.show()
```

0.0.6 PSNR measure

```
[59]: def PSNR(gt, image, max_value=1):
    mse = np.mean((gt - image) ** 2)
    if mse == 0:
        return 100
    return 20 * np.log10(max_value / (np.sqrt(mse)))
```

0.0.7 RPDNET model architecture

```
[60]: #Model
def CNN(input):
    x=Conv2D(64, (3,3), dilation_rate=1,padding='same',activation='relu')(input)
    x=Conv2D(64, (3,3), dilation_rate=2,padding='same',activation='relu')(x)

    y=Conv2D(64, (3,3), dilation_rate=4,padding='same',activation='relu')(input)
    y=Conv2D(64, (3,3), dilation_rate=5,padding='same',activation='relu')(y)

    z=Concatenate(axis=-1)([x,y])
    z=Conv2D(64, (3,3),padding='same',activation='relu')(z)
    add_1=Add()([z, input])

    z=Conv2D(64, (3,3),padding='same',activation='relu')(add_1)
    z=Conv2D(64, (3,3),padding='same')(z)
    add_2=Add()([z,add_1])
    add_2 = Activation('relu')(add_2)

    z=Conv2D(64, (3,3),padding='same',activation='relu')(add_2)
    z=Conv2D(64, (3,3),padding='same',activation='relu')(z)
    z=Conv2D(64, (1,1),padding='same')(z)
    add_3=Add()([z,add_2])
    add_3 = Activation('relu')(add_3)

    z = GlobalAveragePooling2D()(add_3)
    z = tf.expand_dims(z,1)
    z = tf.expand_dims(z,1)
    z=Conv2D(4, (3,3),padding='same',activation='relu')(z)
```

```

z=Conv2D(64, (3,3),padding='same',activation='sigmoid')(z)
mul=Multiply()([z, add_3])

return mul

```

```

[ ]: def RPDNET():
    input = Input((H, W, CH),name='input')
    feat_extraction =Conv2D(64, (3,3),padding='same')(input)
    cnn_1=CNN(feat_extraction)
    cnn_2=CNN(cnn_1)
    cnn_3=CNN(cnn_2)
    cnn_4=CNN(cnn_3)
    x=Conv2D(3, (3,3),padding='same')(cnn_4)
    add_2=Add()([x, input])

    model=Model(input,add_2)

    return model

```

0.0.8 Model summary

```

[62]: #Model train
tf.keras.backend.clear_session()
tf.random.set_seed(6908)
rpdnet = RPDNET()
rpdnet.summary()

```

Model: "model"

Layer (type)	Output Shape	Param #	Connected to
input (InputLayer)	[(None, 120, 120, 3 0)]		[]
conv2d (Conv2D)	(None, 120, 120, 64 1792)		['input[0][0]']
conv2d_1 (Conv2D)	(None, 120, 120, 64 36928 ['conv2d[0][0]'])		
conv2d_3 (Conv2D)	(None, 120, 120, 64 36928 ['conv2d[0][0]'])		

```

conv2d_2 (Conv2D)                (None, 120, 120, 64  36928
['conv2d_1[0][0]']
)

conv2d_4 (Conv2D)                (None, 120, 120, 64  36928
['conv2d_3[0][0]']
)

concatenate (Concatenate)        (None, 120, 120, 12  0
['conv2d_2[0][0]',
8)
'conv2d_4[0][0]']

conv2d_5 (Conv2D)                (None, 120, 120, 64  73792
['concatenate[0][0]']
)

add (Add)                        (None, 120, 120, 64  0
['conv2d_5[0][0]',
)
'conv2d[0][0]']

conv2d_6 (Conv2D)                (None, 120, 120, 64  36928      ['add[0][0]']
)

conv2d_7 (Conv2D)                (None, 120, 120, 64  36928
['conv2d_6[0][0]']
)

add_1 (Add)                      (None, 120, 120, 64  0
['conv2d_7[0][0]',
)
'add[0][0]']

activation (Activation)          (None, 120, 120, 64  0      ['add_1[0][0]']
)

conv2d_8 (Conv2D)                (None, 120, 120, 64  36928
['activation[0][0]']
)

conv2d_9 (Conv2D)                (None, 120, 120, 64  36928
['conv2d_8[0][0]']
)

conv2d_10 (Conv2D)               (None, 120, 120, 64  4160
['conv2d_9[0][0]']
)

```

```

add_2 (Add) (None, 120, 120, 64 0
['conv2d_10[0][0]',
)
'activation[0][0]']

activation_1 (Activation) (None, 120, 120, 64 0 ['add_2[0][0]']
)

global_average_pooling2d (Glob (None, 64) 0
['activation_1[0][0]']
alAveragePooling2D)

tf.expand_dims (TFOpLambda) (None, 1, 64) 0
['global_average_pooling2d[0][0]']

]

tf.expand_dims_1 (TFOpLambda) (None, 1, 1, 64) 0
['tf.expand_dims[0][0]']

conv2d_11 (Conv2D) (None, 1, 1, 4) 2308
['tf.expand_dims_1[0][0]']

conv2d_12 (Conv2D) (None, 1, 1, 64) 2368
['conv2d_11[0][0]']

multiply (Multiply) (None, 120, 120, 64 0
['conv2d_12[0][0]',
)
'activation_1[0][0]']

conv2d_13 (Conv2D) (None, 120, 120, 64 36928
['multiply[0][0]']
)

conv2d_15 (Conv2D) (None, 120, 120, 64 36928
['multiply[0][0]']
)

conv2d_14 (Conv2D) (None, 120, 120, 64 36928
['conv2d_13[0][0]']
)

conv2d_16 (Conv2D) (None, 120, 120, 64 36928
['conv2d_15[0][0]']
)

concatenate_1 (Concatenate) (None, 120, 120, 12 0
['conv2d_14[0][0]',

```



```

8)
'conv2d_16[0][0] '

conv2d_17 (Conv2D)      (None, 120, 120, 64  73792
['concatenate_1[0][0] '
)

add_3 (Add)             (None, 120, 120, 64  0
['conv2d_17[0][0] ',
)
'multiply[0][0] '

conv2d_18 (Conv2D)      (None, 120, 120, 64  36928      ['add_3[0][0] '
)

conv2d_19 (Conv2D)      (None, 120, 120, 64  36928
['conv2d_18[0][0] '
)

add_4 (Add)             (None, 120, 120, 64  0
['conv2d_19[0][0] ',
)
'add_3[0][0] '

activation_2 (Activation) (None, 120, 120, 64  0      ['add_4[0][0] '
)

conv2d_20 (Conv2D)      (None, 120, 120, 64  36928
['activation_2[0][0] '
)

conv2d_21 (Conv2D)      (None, 120, 120, 64  36928
['conv2d_20[0][0] '
)

conv2d_22 (Conv2D)      (None, 120, 120, 64  4160
['conv2d_21[0][0] '
)

add_5 (Add)             (None, 120, 120, 64  0
['conv2d_22[0][0] ',
)
'activation_2[0][0] '

activation_3 (Activation) (None, 120, 120, 64  0      ['add_5[0][0] '
)

global_average_pooling2d_1 (Gl (None, 64)      0
['activation_3[0][0] '

```

```

obalAveragePooling2D)

tf.expand_dims_2 (TFOpLambda) (None, 1, 64) 0
['global_average_pooling2d_1[0][0]

]']

tf.expand_dims_3 (TFOpLambda) (None, 1, 1, 64) 0
['tf.expand_dims_2[0][0]']

conv2d_23 (Conv2D) (None, 1, 1, 4) 2308
['tf.expand_dims_3[0][0]']

conv2d_24 (Conv2D) (None, 1, 1, 64) 2368
['conv2d_23[0][0]']

multiply_1 (Multiply) (None, 120, 120, 64 0
['conv2d_24[0][0]',
)
'activation_3[0][0]']

conv2d_25 (Conv2D) (None, 120, 120, 64 36928
['multiply_1[0][0]']
)

conv2d_27 (Conv2D) (None, 120, 120, 64 36928
['multiply_1[0][0]']
)

conv2d_26 (Conv2D) (None, 120, 120, 64 36928
['conv2d_25[0][0]']
)

conv2d_28 (Conv2D) (None, 120, 120, 64 36928
['conv2d_27[0][0]']
)

concatenate_2 (Concatenate) (None, 120, 120, 12 0
['conv2d_26[0][0]',
8)
'conv2d_28[0][0]']

conv2d_29 (Conv2D) (None, 120, 120, 64 73792
['concatenate_2[0][0]']
)

add_6 (Add) (None, 120, 120, 64 0
['conv2d_29[0][0]',
)

```

```

'multiply_1[0][0]']

conv2d_30 (Conv2D)          (None, 120, 120, 64  36928      ['add_6[0][0]']
)

conv2d_31 (Conv2D)          (None, 120, 120, 64  36928
['conv2d_30[0][0]']
)

add_7 (Add)                 (None, 120, 120, 64  0
['conv2d_31[0][0]',
                                'add_6[0][0]']
)

activation_4 (Activation)    (None, 120, 120, 64  0      ['add_7[0][0]']
)

conv2d_32 (Conv2D)          (None, 120, 120, 64  36928
['activation_4[0][0]']
)

conv2d_33 (Conv2D)          (None, 120, 120, 64  36928
['conv2d_32[0][0]']
)

conv2d_34 (Conv2D)          (None, 120, 120, 64  4160
['conv2d_33[0][0]']
)

add_8 (Add)                 (None, 120, 120, 64  0
['conv2d_34[0][0]',
                                'activation_4[0][0]']
)

activation_5 (Activation)    (None, 120, 120, 64  0      ['add_8[0][0]']
)

global_average_pooling2d_2 (GlobalAveragePooling2D) (None, 64) 0
['activation_5[0][0]']

tf.expand_dims_4 (TFOpLambda) (None, 1, 64) 0
['global_average_pooling2d_2[0][0]']

tf.expand_dims_5 (TFOpLambda) (None, 1, 1, 64) 0
['tf.expand_dims_4[0][0]']

conv2d_35 (Conv2D)          (None, 1, 1, 4) 2308

```

```

['tf.expand_dims_5[0][0]']

conv2d_36 (Conv2D)          (None, 1, 1, 64)      2368
['conv2d_35[0][0]']

multiply_2 (Multiply)      (None, 120, 120, 64   0
['conv2d_36[0][0]',
)
'activation_5[0][0]']

conv2d_37 (Conv2D)          (None, 120, 120, 64   36928
['multiply_2[0][0]']
)

conv2d_39 (Conv2D)          (None, 120, 120, 64   36928
['multiply_2[0][0]']
)

conv2d_38 (Conv2D)          (None, 120, 120, 64   36928
['conv2d_37[0][0]']
)

conv2d_40 (Conv2D)          (None, 120, 120, 64   36928
['conv2d_39[0][0]']
)

concatenate_3 (Concatenate) (None, 120, 120, 12   0
['conv2d_38[0][0]',
8)
'conv2d_40[0][0]']

conv2d_41 (Conv2D)          (None, 120, 120, 64   73792
['concatenate_3[0][0]']
)

add_9 (Add)                 (None, 120, 120, 64   0
['conv2d_41[0][0]',
)
'multiply_2[0][0]']

conv2d_42 (Conv2D)          (None, 120, 120, 64   36928      ['add_9[0][0]']
)

conv2d_43 (Conv2D)          (None, 120, 120, 64   36928
['conv2d_42[0][0]']
)

add_10 (Add)                (None, 120, 120, 64   0

```

```

['conv2d_43[0][0]',
                                )                                'add_9[0][0]']

    activation_6 (Activation)    (None, 120, 120, 64  0
['add_10[0][0]']
                                )

    conv2d_44 (Conv2D)          (None, 120, 120, 64  36928
['activation_6[0][0]']
                                )

    conv2d_45 (Conv2D)          (None, 120, 120, 64  36928
['conv2d_44[0][0]']
                                )

    conv2d_46 (Conv2D)          (None, 120, 120, 64  4160
['conv2d_45[0][0]']
                                )

    add_11 (Add)                (None, 120, 120, 64  0
['conv2d_46[0][0]',
                                )
'activation_6[0][0]']

    activation_7 (Activation)    (None, 120, 120, 64  0
['add_11[0][0]']
                                )

    global_average_pooling2d_3 (GlobalAveragePooling2D) (None, 64) 0
['activation_7[0][0]']

    tf.expand_dims_6 (TFExpandDims) (None, 1, 64) 0
['global_average_pooling2d_3[0][0]']

    tf.expand_dims_7 (TFExpandDims) (None, 1, 1, 64) 0
['tf.expand_dims_6[0][0]']

    conv2d_47 (Conv2D)          (None, 1, 1, 4) 2308
['tf.expand_dims_7[0][0]']

    conv2d_48 (Conv2D)          (None, 1, 1, 64) 2368
['conv2d_47[0][0]']

    multiply_3 (Multiply)       (None, 120, 120, 64  0
['conv2d_48[0][0]',
                                )

```

```

'activation_7[0][0]']

conv2d_49 (Conv2D)          (None, 120, 120, 3) 1731
['multiply_3[0][0]']

add_12 (Add)                (None, 120, 120, 3) 0
['conv2d_49[0][0]',

'input[0][0]']

```

```

=====
=====
Total params: 1,515,731
Trainable params: 1,515,731
Non-trainable params: 0
-----
-----

```

```

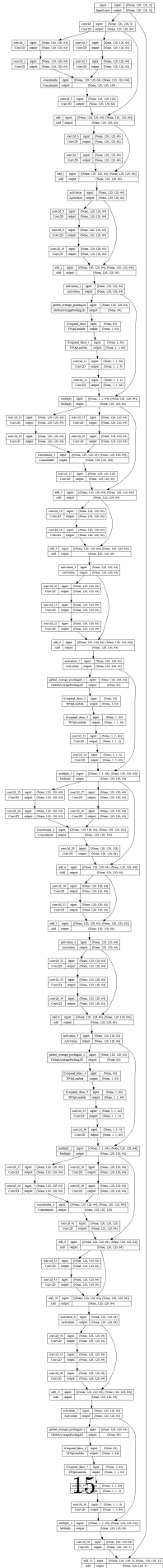
[63]: dot_img_file = 'rpdnet.png'
      tf.keras.utils.plot_model(rpdnet, to_file=dot_img_file, show_shapes=True)

```

```

[63]:

```



```
[64]: #Compile model
rpddnet.compile(optimizer=tf.keras.optimizers.Adam(1e-03), loss=tf.keras.losses.
↳MeanAbsoluteError(),
           metrics=['mean_absolute_error'])
```

0.0.9 Set callback

```
[65]: #Callback
def scheduler(epoch,lr):
    return lr*0.9

checkpoint_path = "rpddnet.h5"
checkpoint_dir = os.path.dirname(checkpoint_path)
cp_callback = tf.keras.callbacks.ModelCheckpoint(filepath=checkpoint_path,
↳save_weights_only=False,verbose=0,save_best_only=False)

lrScheduler = tf.keras.callbacks.LearningRateScheduler(scheduler)
callbacks = [cp_callback,lrScheduler]
```

0.0.10 Train model

```
[ ]: #Train model
nepochs=25
history=rpddnet.fit( train_dataset,shuffle=True,epochs=nepochs,
                    validation_data= test_dataset,callbacks=callbacks,
                    max_queue_size=10,workers=1,verbose=1,
                    use_multiprocessing=True
                    )
```

```
Epoch 1/25
13/13 [=====] - 519s 40s/step - loss: 0.0858 -
mean_absolute_error: 0.0858 - val_loss: 0.1317 - val_mean_absolute_error: 0.1317
- lr: 9.0000e-04
Epoch 2/25
13/13 [=====] - 531s 41s/step - loss: 0.0692 -
mean_absolute_error: 0.0692 - val_loss: 0.1294 - val_mean_absolute_error: 0.1294
- lr: 8.1000e-04
Epoch 3/25
13/13 [=====] - 492s 38s/step - loss: 0.0673 -
mean_absolute_error: 0.0673 - val_loss: 0.1225 - val_mean_absolute_error: 0.1225
- lr: 7.2900e-04
Epoch 4/25
11/13 [=====>...] - ETA: 1:14 - loss: 0.0629 -
```


mean_absolute_error: 0.0629

```
[ ]: #history info.  
print(history.history.keys())
```

0.0.11 Plot training performance

```
[ ]: def plot_history(model_history,keys):  
    m,val_m = keys  
    plt.plot(model_history.history[m])  
    plt.plot(model_history.history[val_m])  
    plt.ylabel(m)  
    plt.xlabel('epoch')  
    plt.legend(['train', 'validation'], loc='upper left')  
    plt.show()  
  
plot_history(history,['loss','val_loss'])
```

0.0.12 Model evaluation

```
[ ]: path='Endoscopy/data/e2.jpg'  
patches_noisy,denoised_patches,ground_truth,noisy_image,denoised_image=predict_fun(rpdnet,path)  
print('PSNR of Noisy Image : ',PSNR(ground_truth,noisy_image))  
print('PSNR of Denoised Image : ',PSNR(ground_truth,denoised_image))  
plot_patches(patches_noisy,denoised_patches)  
  
#result  
plot_predictions(ground_truth,noisy_image,denoised_image)
```

```
[ ]: path='Endoscopy/data/e3.jpg'  
patches_noisy,denoised_patches,ground_truth,noisy_image,denoised_image=predict_fun(rpdnet,path)  
print('PSNR of Noisy Image : ',PSNR(ground_truth,noisy_image))  
print('PSNR of Denoised Image : ',PSNR(ground_truth,denoised_image))  
plot_patches(patches_noisy,denoised_patches)  
  
#result  
plot_predictions(ground_truth,noisy_image,denoised_image)
```

```
[ ]: path='Endoscopy/data/e4.jpg'  
patches_noisy,denoised_patches,ground_truth,noisy_image,denoised_image=predict_fun(rpdnet,path)  
print('PSNR of Noisy Image : ',PSNR(ground_truth,noisy_image))  
print('PSNR of Denoised Image : ',PSNR(ground_truth,denoised_image))  
plot_patches(patches_noisy,denoised_patches)  
  
#result  
plot_predictions(ground_truth,noisy_image,denoised_image)
```

0.0.13 Residual learning

```
[ ]: # Residual Learning
model_res=Model(rpdnet.input,rpdnet.get_layer('conv2d_49').output)
path='Endoscopy/data/e4.jpg'
patches_noisy,denoised_patches,ground_truth,noisy_image,residual_image=predict_fun(model_res,p
```

0.0.14 Residual data visualisation

```
[ ]: #Display predicted result
fig, axs = plt.subplots(1,3,figsize=(15,15))
axs[0].imshow(ground_truth)
axs[0].title.set_text('Ground Truth')
axs[1].imshow(noisy_image)
axs[1].title.set_text('Noisy Image')
axs[2].imshow(residual_image)
axs[2].title.set_text('Residual Image')

plt.show()
```

```
[ ]: import keract
from tensorflow.keras.applications.mobilenet import decode_predictions, \
    preprocess_input
from tensorflow.keras.preprocessing.image import load_img, img_to_array
```

0.0.15 Activation silency map

```
[ ]: #Activation silency map
image = load_img(path, target_size= (H, W))
image = img_to_array(image)
image = image.reshape((1, image.shape[0], image.shape[1], image.shape[2]))
image = preprocess_input(image)
y_hat = rpdnet.predict(image)
```

```
[ ]: #layers
layers=['conv2d_1','conv2d_3','conv2d_2','conv2d_4']

activations= keract.get_activations(rpdnet, image, layer_names= layers,
                                   nodes_to_evaluate= None, output_format= \
    'simple', auto_compile= True)
keract.display_activations(activations, cmap='viridis', save= False, directory= \
    'activations')
```

```
[ ]:
```