

denoising_autoencoder

January 12, 2022

1 18CSC402 - DEEP LEARNING

HANDWRITTEN CHARACTER LANGUAGE : Malayalam

NOISE USED : Salt & Pepper

```
[ ]: import os
import cv2
import random
import numpy as np
from tensorflow import keras
import matplotlib.pyplot as plt
from skimage.util import random_noise
from tensorflow.keras.preprocessing import image
from tensorflow.keras.preprocessing.image import img_to_array
from keras.layers import Conv2D, MaxPooling2D, UpSampling2D, Input
```

```
[ ]: # reading the images from its location

SIZE = 64
X, X_noisy = [], []
img_path = "/content/drive/MyDrive/MalayalamHandwrittenData/"
for folder in os.listdir(img_path):
    path = img_path+folder+"/"
    files = os.listdir(path)
    for i in range(300):
        filename = files[i]
        img = cv2.imread(path+filename,0)
        img = cv2.resize(img,(SIZE,SIZE))
        img_array = img_to_array(img)
        X.append(img_array)
        noisy_img = random_noise(img, mode="s&p", amount=0.25)
        X_noisy.append(noisy_img)

X_noisy = np.reshape(X_noisy, (len(X_noisy), SIZE, SIZE, 1))
X_noisy = X_noisy.astype('float32') / 255.0
X = np.reshape(X, (len(X), SIZE, SIZE, 1))
X = X.astype('float32') / 255.0
```

```
print("Noisy data shape:", X_noisy.shape)
print("Clean data shape:", X.shape)
```

Noisy data shape: (4500, 64, 64, 1)

Clean data shape: (4500, 64, 64, 1)

```
[ ]: # splitting data into train and validation sets
```

```
from sklearn.model_selection import train_test_split
X_train, X_valid, y_train, y_valid = train_test_split(X_noisy, X, test_size=0.2)
print(X_train.shape)
print(X_valid.shape)
print(y_train.shape)
print(y_valid.shape)
```

(3600, 64, 64, 1)

(900, 64, 64, 1)

(3600, 64, 64, 1)

(900, 64, 64, 1)

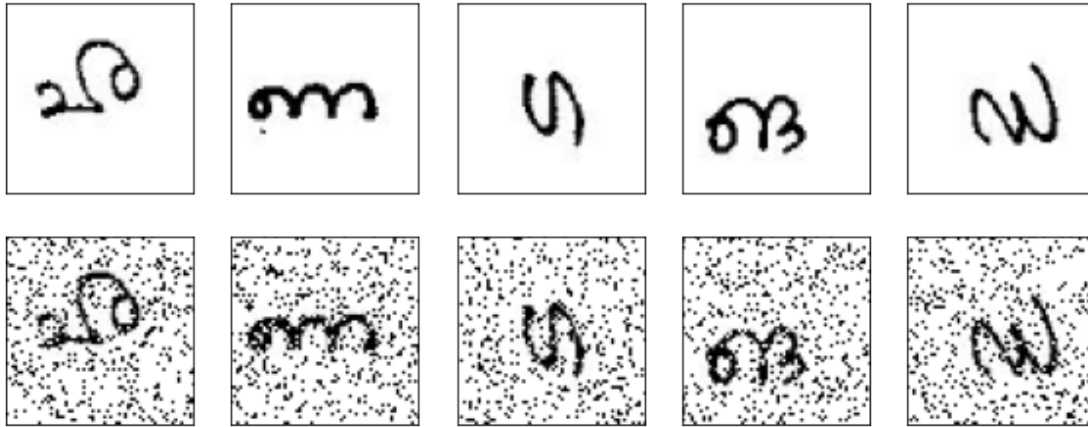
```
[ ]: # plotting some sample clean and corresponding noisy images
```

```
num = [random.randint(0,4500) for i in range(5)]
n = len(num)
plt.figure(figsize=(10, 4))
for i in range(n):

    # Show sample test images
    ax = plt.subplot(2, n, i + 1)
    plt.imshow(X[num[i]].reshape(SIZE, SIZE), cmap="gray")
    ax.get_xaxis().set_visible(False)
    ax.get_yaxis().set_visible(False)

    # Show corresponding output images
    ax = plt.subplot(2, n, i + 1 + n)
    plt.imshow(X_noisy[num[i]].reshape(SIZE, SIZE), cmap="gray")
    ax.get_xaxis().set_visible(False)
    ax.get_yaxis().set_visible(False)

plt.show()
```



```
[ ]: # ARCHITECTURE : convolutional autoencoder
```

```
# Encoder
```

```
input = Input(shape=(64, 64, 1))
```

```
↳ # 64*64*1
```

```
x = Conv2D(64, (3, 3), activation="relu", padding="same")(input)
```

```
↳ # 64*64*64
```

```
x = MaxPooling2D((2, 2), padding="same")(x)
```

```
↳ # 32*32*64
```

```
x = Conv2D(32, (3, 3), activation="relu", padding="same")(x)
```

```
↳ # 32*32*32
```

```
x = MaxPooling2D((2, 2), padding="same")(x)
```

```
↳ # 16*16*32
```

```
x = Conv2D(16, (3, 3), activation="relu", padding="same")(x)
```

```
↳ # 16*16*16
```

```
# encoder model
```

```
encoder = keras.Model(input, x)
```

```
# Decoder
```

```
x = Conv2D(32, (3, 3), activation="relu", padding="same")(x)
```

```
↳ # 16*16*32
```

```
x = UpSampling2D((2, 2))(x)
```

```
↳ # 32*32*32
```

```
x = Conv2D(64, (3, 3), activation="relu", padding="same")(x)
```

```
↳ # 32*32*64
```

```
x = UpSampling2D((2, 2))(x)
```

```
↳ # 64*64*64
```

```
output = Conv2D(1, (3, 3), activation="sigmoid", padding="same")(x)
```

```
↳ # 64*64*1
```

```
# Model
```

```
conv_autoencoder = keras.Model(input, output)
```

```
[ ]: # Compiling the model
```

```
conv_autoencoder.compile(optimizer="adam", loss="mse")
conv_autoencoder.summary()
```

Model: "model_1"

Layer (type)	Output Shape	Param #
input_1 (InputLayer)	[(None, 64, 64, 1)]	0
conv2d (Conv2D)	(None, 64, 64, 64)	640
max_pooling2d (MaxPooling2D)	(None, 32, 32, 64)	0
conv2d_1 (Conv2D)	(None, 32, 32, 32)	18464
max_pooling2d_1 (MaxPooling2D)	(None, 16, 16, 32)	0
conv2d_2 (Conv2D)	(None, 16, 16, 16)	4624
conv2d_3 (Conv2D)	(None, 16, 16, 32)	4640
up_sampling2d (UpSampling2D)	(None, 32, 32, 32)	0
conv2d_4 (Conv2D)	(None, 32, 32, 64)	18496
up_sampling2d_1 (UpSampling2D)	(None, 64, 64, 64)	0
conv2d_5 (Conv2D)	(None, 64, 64, 1)	577

=====
Total params: 47,441
Trainable params: 47,441
Non-trainable params: 0
=====

```
[ ]: # fitting the model
```

```
history = conv_autoencoder.fit(X_train, y_train, validation_data=(X_valid, y_valid), epochs=50, batch_size=128)
```

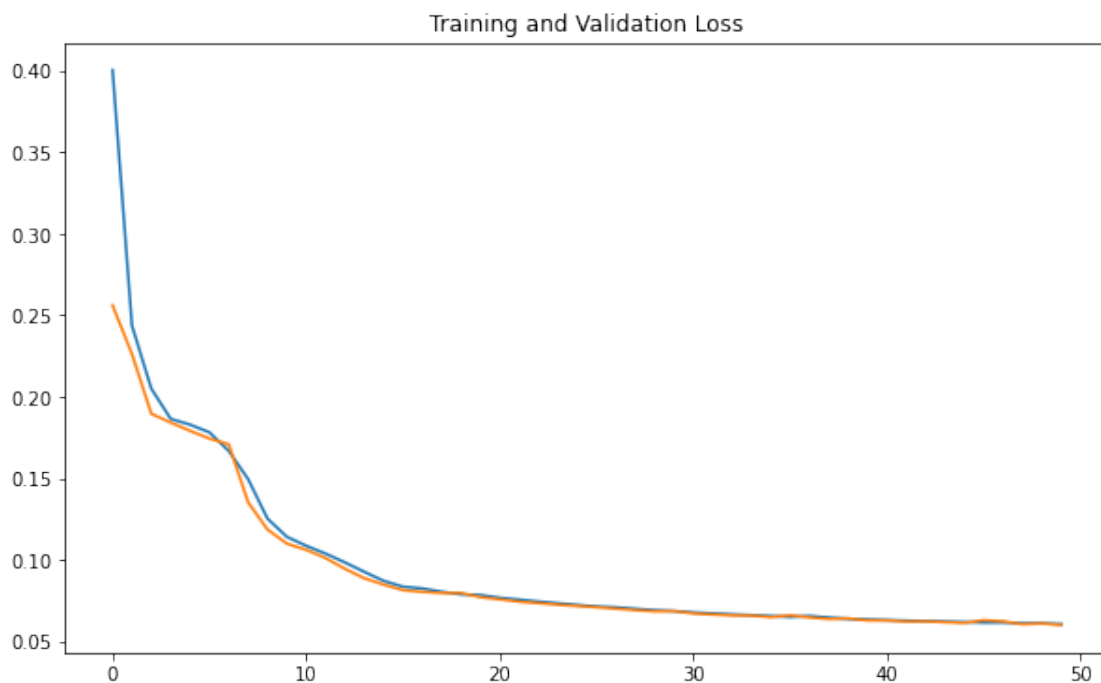
```
[ ]: # plotting the train and validation loss curves
```

```
loss = history.history["loss"]
val_loss = history.history["val_loss"]

plt.figure(figsize=(10,6))
epochs = 50

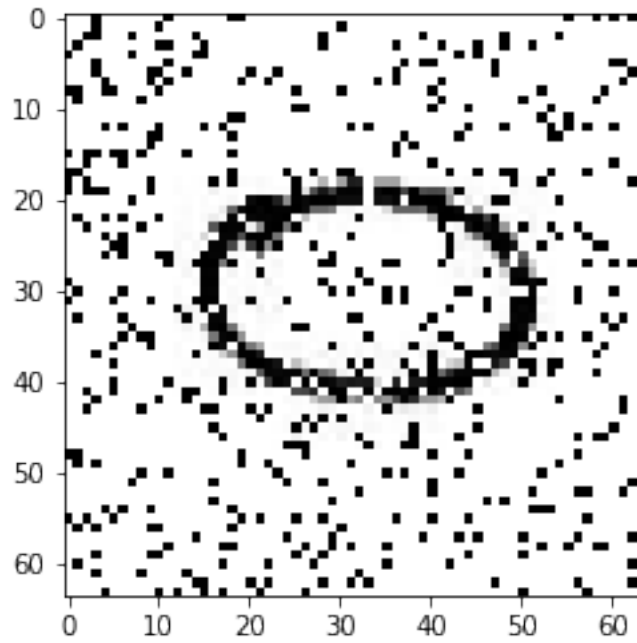
plt.plot(loss, label='Training Loss')
plt.plot(val_loss, label='Validation Loss')
plt.title('Training and Validation Loss')

plt.show()
```



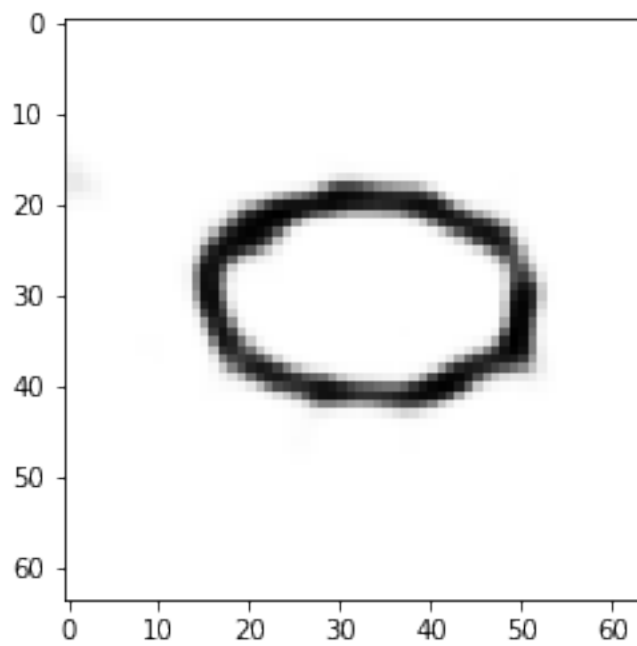
```
[ ]: # demonstrating the prediction on an image in the validation set
plt.imshow(X_valid[1].reshape(64,64), cmap="gray")
```

```
[ ]: <matplotlib.image.AxesImage at 0x7fe0f6b5f790>
```



```
[ ]: # denoised output of the image
output = conv_autoencoder.predict([X_valid[1].reshape(-1,64,64,1))][0]
plt.imshow(output.reshape(64,64), cmap="gray")
```

```
[ ]: <matplotlib.image.AxesImage at 0x7fe0f6aff110>
```

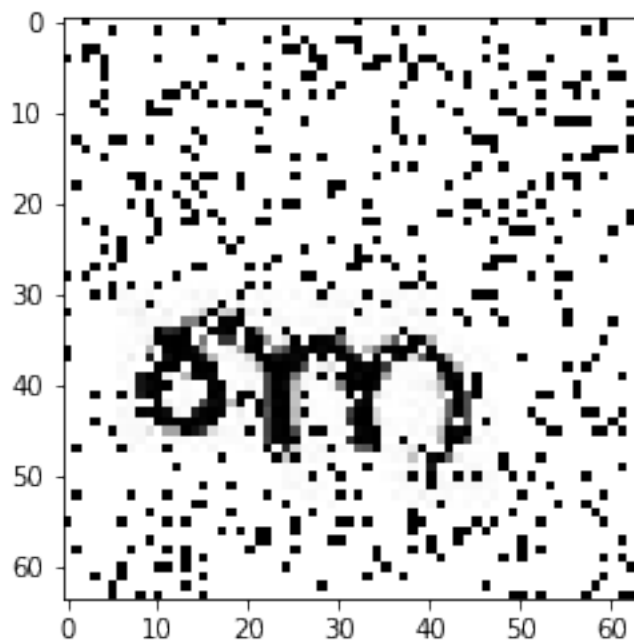


```
[ ]: # checking the reconstruction power on unseen data

test_img = cv2.imread("/content/drive/MyDrive/MalayalamHandwrittenSubset/
↳Malayalam Handwritten Subset/15.jpg",0)
test_img = cv2.resize(test_img,(SIZE,SIZE))
test_noisy_img = random_noise(test_img, mode="s&p", amount=0.25)
test_noisy_img_reshaped = np.reshape(test_noisy_img, (SIZE, SIZE,1))
test_noisy_img_reshaped = test_noisy_img_reshaped.astype('float32') / 255.0
denoised_img = conv_autoencoder.predict([test_noisy_img_reshaped.
↳reshape(-1,64,64,1)])[0]
print(denoised_img.shape)
plt.imshow(test_noisy_img, cmap="gray")
```

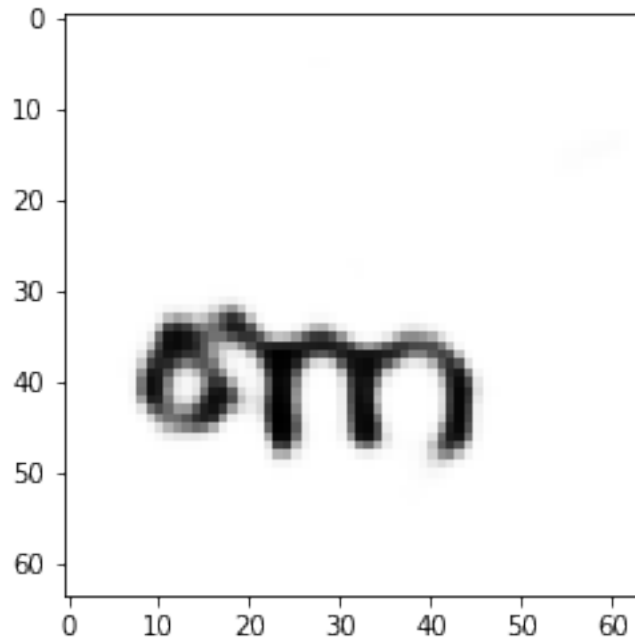
(64, 64, 1)

```
[ ]: <matplotlib.image.AxesImage at 0x7fe16d9d6350>
```



```
[ ]: plt.imshow(denoised_img.reshape(64,64), cmap="gray")
```

```
[ ]: <matplotlib.image.AxesImage at 0x7fe16daf9510>
```



PREDICTING ONE SAMPLE FROM EACH OF THE 15 CHARACTERS

```
[ ]: dir_path = "/content/drive/MyDrive/MalayalamHandwrittenSubset/Malayalam_
      ↳Handwritten Subset/"
SIZE = 64
test_clean, test_noisy = [], []
for filename in os.listdir(dir_path):
    img = cv2.imread(dir_path+filename,0)
    img = cv2.resize(img,(SIZE,SIZE))
    img_array = img_to_array(img)
    test_clean.append(img_array)
    noisy_img = random_noise(img, mode="s&p", amount=0.25)
    test_noisy.append(noisy_img)

test_noisy = np.reshape(test_noisy, (len(test_noisy), SIZE, SIZE, 1))
test_noisy = test_noisy.astype('float32') / 255.0
test_clean = np.reshape(test_clean, (len(test_clean), SIZE, SIZE, 1))
test_clean = test_clean.astype('float32') / 255.0
print("Noisy test data shape:", test_noisy.shape)
print("Clean test data shape:", test_clean.shape)
```

Noisy test data shape: (15, 64, 64, 1)

Clean test data shape: (15, 64, 64, 1)

```
[ ]: # decoding the noisy images
test_decoded = conv_autoencoder.predict([test_noisy.reshape(-1,64,64,1)])
```

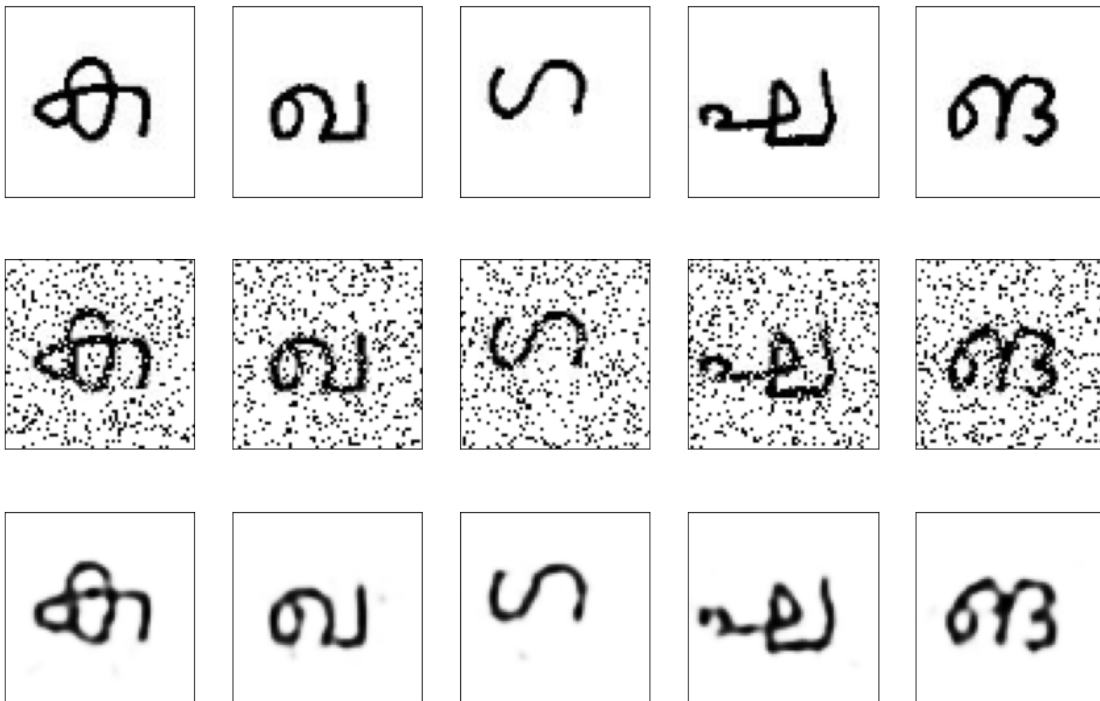


```
[ ]: # visualisation of adding noise to data and predicting (first 5 characters)
```

```
plt.figure(figsize=(15, 10))
for i in range(5):
    # original clean images
    fig = plt.subplot(3,5,i+1)
    plt.imshow(test_clean[i].reshape(64,64), cmap="gray")
    fig.get_xaxis().set_visible(False)
    fig.get_yaxis().set_visible(False)

    # noisy images
    fig = plt.subplot(3,5,i+1+5)
    plt.imshow(test_noisy[i].reshape(64,64), cmap="gray")
    fig.get_xaxis().set_visible(False)
    fig.get_yaxis().set_visible(False)

    # decoded images
    fig = plt.subplot(3,5,i+1+(2*5))
    plt.imshow(test_decoded[i].reshape(64,64), cmap="gray")
    fig.get_xaxis().set_visible(False)
    fig.get_yaxis().set_visible(False)
```

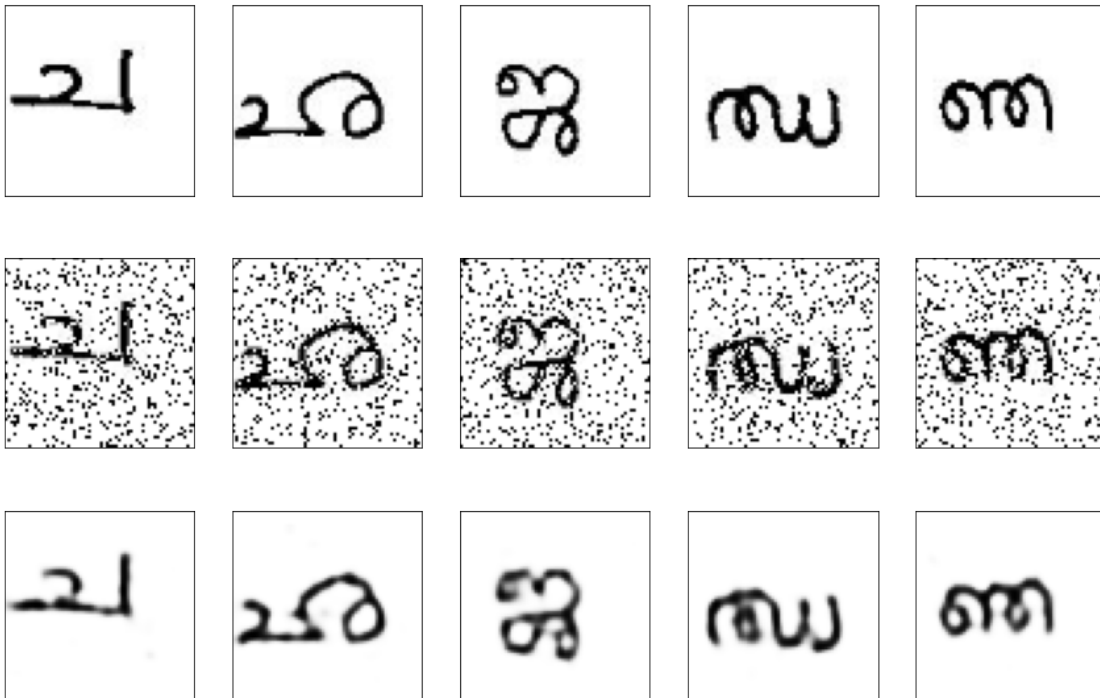


```
[ ]: # visualisation of adding noise to data and predicting (next 5 characters)
```

```
plt.figure(figsize=(15, 10))
for i in range(5):
    # original clean images
    fig = plt.subplot(3,5,i+1)
    plt.imshow(test_clean[i+5].reshape(64,64), cmap="gray")
    fig.get_xaxis().set_visible(False)
    fig.get_yaxis().set_visible(False)

    # noisy images
    fig = plt.subplot(3,5,i+1+5)
    plt.imshow(test_noisy[i+5].reshape(64,64), cmap="gray")
    fig.get_xaxis().set_visible(False)
    fig.get_yaxis().set_visible(False)

    # decoded images
    fig = plt.subplot(3,5,i+1+(2*5))
    plt.imshow(test_decoded[i+5].reshape(64,64), cmap="gray")
    fig.get_xaxis().set_visible(False)
    fig.get_yaxis().set_visible(False)
```



```
[ ]: # visualisation of adding noise to data and predicting (last 5 characters)

plt.figure(figsize=(15, 10))
for i in range(5):
```

```

# original clean images
fig = plt.subplot(3,5,i+1)
plt.imshow(test_clean[i+10].reshape(64,64), cmap="gray")
fig.get_xaxis().set_visible(False)
fig.get_yaxis().set_visible(False)

# noisy images
fig = plt.subplot(3,5,i+1+5)
plt.imshow(test_noisy[i+10].reshape(64,64), cmap="gray")
fig.get_xaxis().set_visible(False)
fig.get_yaxis().set_visible(False)

# decoded images
fig = plt.subplot(3,5,i+1+(2*5))
plt.imshow(test_decoded[i+10].reshape(64,64), cmap="gray")
fig.get_xaxis().set_visible(False)
fig.get_yaxis().set_visible(False)

```

