```
In [1]:
```

```
%matplotlib inline
from matplotlib import pyplot as plt
import numpy as np
import tensorflow as tf
from tensorflow import keras
from tensorflow.keras import Sequential, layers
from tensorflow.keras.layers import Conv2D, MaxPooling2D, Flatten, Dense
from tensorflow.keras.utils import to_categorical
```

#### In [2]:

```
import glob
import os
import PIL
import time

from IPython import display
```

### In [3]:

```
fashion_mnist = keras.datasets.fashion_mnist

(train_images, train_labels), (test_images, test_labels) = fashion_mnist.load_data()

print(train_images.shape, train_labels.shape, '\n')

# set random seed

np.random.seed(42)

trainsize = 10000 # number of training images we use, selected randomly
randtrain = np.random.choice(train_images.shape[0], trainsize, replace=False)
train_images, train_labels = train_images[randtrain], train_labels[randtrain]

print(train_images.shape, train_labels.shape)
```

### In [4]:

```
# demonstrate an image
plt.figure()
plt.imshow(train_images[10])
plt.imshow(train_images[10], cmap='gray')
plt.show()
```



```
25 -
```

#### In [5]:

```
train_images = train_images.reshape(train_images.shape[0], 28, 28, 1).astype('float32')
train_images = (train_images - 127.5) / 127.5 # Normalize the images to [-1, 1]

BUFFER_SIZE = trainsize
BATCH_SIZE = 100
g_loss = []
d_loss = []
# Batch and shuffle the data
train_dataset = tf.data.Dataset.from_tensor_slices(train_images).shuffle(BUFFER_SIZE).ba
tch(BATCH_SIZE)
```

#### In [6]:

```
def make generator model():
   model = tf.keras.Sequential()
   model.add(layers.Dense(7*7*256, use bias=False, input shape=(100,)))
   model.add(layers.BatchNormalization())
   model.add(layers.LeakyReLU())
   model.add(layers.Reshape((7, 7, 256)))
   assert model.output shape == (None, 7, 7, 256) # Note: None is the batch size
   model.add(layers.Conv2DTranspose(128, (5, 5), strides=(1, 1), padding='same', use bi
as=False))
   assert model.output shape == (None, 7, 7, 128)
   model.add(layers.BatchNormalization())
   model.add(layers.LeakyReLU())
   model.add(layers.Conv2DTranspose(64, (5, 5), strides=(2, 2), padding='same', use bia
s=False))
   assert model.output shape == (None, 14, 14, 64)
   model.add(layers.BatchNormalization())
   model.add(layers.LeakyReLU())
   model.add(layers.Conv2DTranspose(1, (5, 5), strides=(2, 2), padding='same', use bias
=False, activation='tanh'))
   assert model.output shape == (None, 28, 28, 1)
   return model
```

### In [7]:

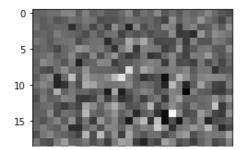
```
generator = make_generator_model()

noise = tf.random.normal([1, 100])
generated_image = generator(noise, training=False)

plt.imshow(generated_image[0, :, :, 0], cmap='gray')
```

# Out[7]:

<matplotlib.image.AxesImage at 0x7ff5603ff550>



```
20 -
25 -
0 5 10 15 20 25
```

#### In [8]:

### In [9]:

```
discriminator = make_discriminator_model()
decision = discriminator(generated_image)
print (decision)
```

tf.Tensor([[-0.00055553]], shape=(1, 1), dtype=float32)

#### In [10]:

```
# This method returns a helper function to compute cross entropy loss cross_entropy = tf.keras.losses.BinaryCrossentropy(from_logits=True)
```

#### In [11]:

```
def discriminator_loss(real_output, fake_output, d_loss):
    real_loss = cross_entropy(tf.ones_like(real_output), real_output)
    fake_loss = cross_entropy(tf.zeros_like(fake_output), fake_output)
    total_loss = real_loss + fake_loss
    d_loss.append(total_loss)
    return total_loss
```

### In [12]:

```
def generator_loss(fake_output, g_loss):
    fake_loss = cross_entropy(tf.ones_like(fake_output), fake_output)
    g_loss.append(fake_loss)
    return fake_loss
```

### In [13]:

```
generator_optimizer = tf.keras.optimizers.Adam(1e-4)
discriminator_optimizer = tf.keras.optimizers.Adam(1e-4)
```

#### In [14]:

#### In [28]:

```
EPOCHS = 100
```

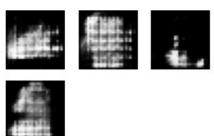
```
noise_dim = 100
num_examples_to_generate = 4

# We will reuse this seed overtime (so it's easier)
# to visualize progress in the animated GIF)
seed = tf.random.normal([num_examples_to_generate, noise_dim])
```

### In [29]:

```
# Notice the use of `tf.function`
# This annotation causes the function to be "compiled".
# @tf.function
def train step(images):
   noise = tf.random.normal([BATCH SIZE, noise dim])
   with tf.GradientTape() as gen tape, tf.GradientTape() as disc tape:
     generated images = generator(noise, training=True)
      real output = discriminator(images, training=True)
      fake output = discriminator(generated images, training=True)
      gen loss = generator loss(fake output, g loss)
     disc_loss = discriminator_loss(real_output, fake_output, d_loss)
    gradients of generator = gen tape.gradient(gen loss, generator.trainable variables)
    gradients of discriminator = disc tape.gradient(disc loss, discriminator.trainable va
riables)
    generator optimizer.apply gradients(zip(gradients of generator, generator.trainable v
ariables))
   discriminator optimizer.apply gradients(zip(gradients of discriminator, discriminator
.trainable variables))
def train(dataset, epochs):
  for epoch in range(epochs):
   start = time.time()
    for image batch in dataset:
      train step(image batch)
    # Produce images for the GIF as we go
    display.clear output (wait=True)
    generate and save images (generator,
                             epoch + 1,
    # Save the model every 15 epochs
    if (epoch + 1) % 15 == 0:
      checkpoint.save(file prefix = checkpoint prefix)
   print ('Time for epoch {} is {} sec'.format(epoch + 1, time.time()-start))
  # Generate after the final epoch
  display.clear output(wait=True)
  generate and save images (generator,
                           epochs,
                           seed)
def generate and save images (model, epoch, test input):
  # Notice `training` is set to False.
  # This is so all layers run in inference mode (batchnorm).
 predictions = model(test input, training=False)
 fig = plt.figure(figsize=(4,4))
  for i in range(predictions.shape[0]):
     plt.subplot(3, 3, i+1)
      plt.imshow(predictions[i, :, :, 0] * 127.5 + 127.5, cmap='gray')
      plt.axis('off')
```

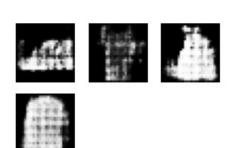
```
plt.savefig('image_at_epoch_{:04d}.png'.format(epoch))
  plt.show()
In [30]:
train(train dataset, EPOCHS)
In [31]:
checkpoint.restore(tf.train.latest checkpoint(checkpoint dir))
Out[31]:
<tensorflow.python.training.tracking.util.CheckpointLoadStatus at 0x7ff4d0377a90>
In [32]:
# Display a single image using the epoch number
def display image(epoch no):
  return PIL.Image.open('image_at_epoch_{:04d}.png'.format(epoch_no))
In [40]:
display_image(1)
Out[40]:
```



## In [38]:

display\_image(20)

### Out[38]:



# In [34]:

display\_image(40)

# Out[34]:









# In [35]:

display\_image(60)

# Out[35]:









# In [36]:

display\_image(80)

# Out[36]:





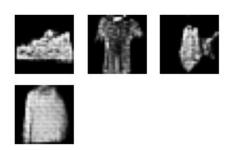




### In [37]:

```
display_image(100)
```

#### Out[37]:

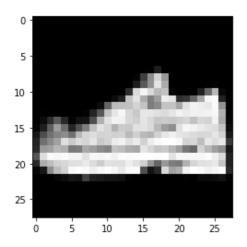


# In [124]:

```
# generate a new image from the trained generator
noi = tf.random.normal([1, 100])
sample = checkpoint.generator(noi, training=False)
plt.imshow(sample[0, :, :, 0], cmap='gray')
```

# Out[124]:

<matplotlib.image.AxesImage at 0x7ff4451507d0>



### In [125]:

```
# detect the image using the trained discriminator
val = checkpoint.discriminator(sample)
print (val)
```

tf.Tensor([[-0.3878896]], shape=(1, 1), dtype=float32)