```
1 %matplotlib inline
 2 import numpy as np
 3 import matplotlib.pyplot as plt
 4 from keras.datasets import fashion mnist
 5 from keras.layers import Dense, Flatten, Reshape
 6 from keras.layers import LeakyReLU
 7 from keras.optimizers import Adam
 8 from keras.models import Sequential
 1 \text{ rows} = 28
 2 cols = 28
 3 \text{ channels} = 1
 4 noise = 100
 1 def generator(noise, rows, cols, channels):
 2
       shape = (rows, cols, channels)
      model = Sequential()
      model.add(Dense(128, input_dim = noise))
 4
 5
      model.add(LeakyReLU(alpha = 0.01))
 6
      model.add(Dense((rows*cols*channels), activation = 'tanh'))
 7
      model.add(Reshape(shape))
 8
       return model
 1 def discriminator(rows, cols, channels):
      shape = (rows, cols, channels)
 2
 3
      model = Sequential()
 4
      model.add(Flatten(input_shape = shape))
      model.add(Dense(128))
      model.add(LeakyReLU(alpha = 0.01))
 6
      model.add(Dense(1, activation = 'sigmoid'))
 8
       return model
 1 def gan(generator, discriminator):
      model = Sequential()
      model.add(generator)
 3
      model.add(discriminator)
 4
       return model
 1 generator = generator(noise, rows, cols, channels)
 2 discriminator = discriminator(rows, cols, channels)
 3 discriminator.compile(loss = 'binary_crossentropy',
 4
                        optimizer = Adam(),
 5
                        metrics = ['accuracy'])
 6 discriminator.trainable = False
 7 gan = gan(generator, discriminator)
 8 gan.compile(loss = 'binary_crossentropy', optimizer = Adam())
 1 check_points = []
 2 loss = []
 3 accuracy = []
 1 def train(iterations, sample interval, batch size):
       (X_{\text{train}}, _), (_, _) = fashion_mnist.load_data()
 3
 4
       X_train = X_train / 127.5 - 1.0
 5
      X_train = np.expand_dims(X_train, axis=3)
 6
 7
       true = np.ones((batch_size, 1))
 8
      fake = np.zeros((batch_size, 1))
 9
       for i in range(iterations):
10
11
12
           idt = np.random.randint(0, X_train.shape[0], batch_size)
           true images = X train[idt]
13
14
15
           idf = np.random.normal(0, 1, (batch_size, 100))
16
           fake images = generator.predict(idf)
17
           disc_loss_true = discriminator.train_on_batch(true_images, true)
18
           disc_loss_fake = discriminator.train_on_batch(fake_images, fake)
           disc_loss, tot_accuracy = 0.5 * np.add(disc_loss_true, disc_loss_fake)
19
```

```
idf = np.random.normal(0, 1, (batch_size, 100))
21
22
           fake_images = generator.predict(idf)
23
24
           gen_loss = gan.train_on_batch(idf, true)
25
           loss=[]
26
           if (i + 1) % sample_interval == 0:
27
               print(type(disc_loss))
28
               print(type(gen_loss))
29
30
31
               loss.append((disc_loss, gen_loss))
32
33
               accuracy.append(100.0 * tot_accuracy)
               check points.append(i + 1)
34
35
36
               print("%d [Discriminator loss: %f, accuracy.: %.2f%%] [Generator loss: %f]" %
37
                     (i + 1, disc_loss, 100.0 * tot_accuracy, gen_loss))
38
39
               display_images(generator, noise)
1 def display_images(generator, noise, grid_rows=4, grid_cols=4):
 2
 3
      n = np.random.normal(0, 1, (grid_rows * grid_cols, noise))
 4
      images = generator.predict(n)
      images = 0.5 * images + 0.5
 5
 6
 7
      fig, axs = plt.subplots(grid_rows,
 8
9
                               figsize=(4, 4),
10
                               sharey=True,
                               sharex=True)
11
12
13
      count = 0
      for i in range(grid_rows):
14
15
           for j in range(grid_cols):
16
               axs[i, j].imshow(images[count, :, :, 0], cmap='gray')
               axs[i, j].axis('off')
17
18
               count += 1
19
1 iterations = 200
2 sample_interval = 100
3 batch_size = 128
4 train(iterations, sample_interval, batch_size)
```