```
from keras.datasets import mnist
from keras.layers import Input, Dense, Reshape, Flatten
from keras.layers import BatchNormalization
from keras.layers import LeakyReLU
from keras.models import Sequential, Model
from keras.optimizers import Adam
import matplotlib.pyplot as plt
import numpy as np


img_rows = 28
img_cols = 28
channels = 1
img_shape = (img_rows, img_cols, channels)


####### Sequential API #########

# def build_generator():
#     noise_shape = (100,)

#     #Generator model definition
#     model = Sequential()
#     model.add(Dense(256, input_shape=noise_shape))
#     model.add(LeakyReLU(alpha=0.2))
#     model.add(BatchNormalization(momentum=0.8))
#     model.add(Dense(512))
#     model.add(LeakyReLU(alpha=0.2))
#     model.add(BatchNormalization(momentum=0.8))
#     model.add(Dense(1024))
#     model.add(LeakyReLU(alpha=0.2))
#     model.add(BatchNormalization(momentum=0.8))

#     model.add(Dense(np.prod(img_shape), activation='tanh'))
#     model.add(Reshape(img_shape))

#     model.summary()

#     noise = Input(noise_shape)
#     img = model(noise)

#     return Model(noise, img)


##### Functional API ######

def build_generator():
    noise_shape = (100,)

    # Define the input for the generator
    noise_input = Input(shape=noise_shape)

    # Define the hidden layers of the generator
    generator = Dense(256, input_shape=noise_shape)(noise_input)
    generator = LeakyReLU(alpha=0.2)(generator)
    generator = BatchNormalization(momentum=0.8)(generator)
    generator = Dense(512)(generator)
    generator = LeakyReLU(alpha=0.2)(generator)
    generator = BatchNormalization(momentum=0.8)(generator)
    generator = Dense(1024)(generator)
    generator = LeakyReLU(alpha=0.2)(generator)
    generator = BatchNormalization(momentum=0.8)(generator)

    # Define the output of the generator
    generator = Dense(np.prod(img_shape), activation='tanh')(generator)
    generator = Reshape(img_shape)(generator)

    # Define the generator model
    generator_model = Model(inputs=noise_input, outputs=generator)
    generator_model.summary()

    return generator_model


##### Sequential API ######

# def build_discriminator():

#     #Discriminator model definition
#     model = Sequential()
#     model.add(Flatten(input_shape=img_shape))
#     model.add(Dense(512))
```

```
#     model.add(LeakyReLU(alpha=0.2))
#     model.add(Dense(256))
#     model.add(LeakyReLU(alpha=0.2))
#     model.add(Dense(1, activation='sigmoid'))

#     model.summary()

#     img = Input(shape=img_shape)
#     validity = model(img)

#     return Model(img, validity)


##### Functional API #####

def build_discriminator():

    # Input layer
    img = Input(shape=img_shape)

    # Hidden layers
    x = Flatten()(img)
    x = Dense(512)(x)
    x = LeakyReLU(alpha=0.2)(x)
    x = Dense(256)(x)
    x = LeakyReLU(alpha=0.2)(x)

    # Output layer
    validity = Dense(1, activation='sigmoid')(x)

    # Model
    model = Model(img, validity)
    model.summary()

    return model


def train(epochs, batch_size=128, save_interval=10):

    #Training Discriminator
    (X_train, _), (_, _) = mnist.load_data()
    X_train = (X_train.astype(np.float32) - 127.5) / 127.5
    X_train = np.expand_dims(X_train, axis=3)
    half_batch = int(batch_size / 2)

    for epoch in range(epochs):
        idx = np.random.randint(0, X_train.shape[0], half_batch)
        imgs = X_train[idx]
        noise = np.random.normal(0, 1, (half_batch, 100))

        gen_imgs = generator.predict(noise)

        d_loss_real = discriminator.train_on_batch(imgs, np.ones((half_batch, 1)))
        d_loss_fake = discriminator.train_on_batch(imgs, np.zeros((half_batch, 1)))

        d_loss = 0.5 * np.add(d_loss_real, d_loss_fake)

        #Training Generator
        noise = np.random.normal(0, 1, (batch_size, 100))
        valid_y = np.array([1] * batch_size)

        g_loss = combined.train_on_batch(noise, valid_y)

        print("%d [D loss: %f, acc. : %.2f%%] [G loss: %f]" %(epoch, d_loss[0], 100*d_loss[1], g_loss))

        if epoch % save_interval == 0:
            save_imgs(epoch)


def save_imgs(epoch):
    r, c = 5, 5
    noise = np.random.normal(0, 1, (r*c, 100))
    gen_imgs = generator.predict(noise)
    gen_imgs = 0.5 * gen_imgs + 0.5

    fig, axs = plt.subplots(r, c)
    cnt = 0

    for i in range(r):
        for j in range(c):
            axs[i, j].imshow(gen_imgs[cnt, :, :, 0], cmap='gray')
            axs[i, j].axis('off')
            cnt += 1
```

```
        fig.savefig("images/mnist_%d.png" %epoch)
        plt.close()


    # Training combined model
    optimizer = Adam(0.0002, 0.5)

    discriminator = build_discriminator()
    discriminator.compile(loss='binary_crossentropy', optimizer=optimizer, metrics=['accuracy'])

    generator = build_generator()
    generator.compile(loss='binary_crossentropy', optimizer=optimizer)

    z = Input(shape=(100,))
    img = generator(z)

    discriminator.trainable = False
    valid = discriminator(img)

    combined = Model(z, valid)
    combined.compile(loss='binary_crossentropy', optimizer=optimizer)

    train(epochs=100, batch_size=32, save_interval=5)
    generator.save('generator_model_test.h5')
```

```
Model: "model"
_________________________________________________________________
 Layer (type)                Output Shape              Param #
=================================================================
 input_1 (InputLayer)        [(None, 28, 28, 1)]       0

 flatten (Flatten)           (None, 784)               0

 dense (Dense)               (None, 512)               401920

 leaky_re_lu (LeakyReLU)     (None, 512)               0

 dense_1 (Dense)             (None, 256)               131328

 leaky_re_lu_1 (LeakyReLU)   (None, 256)               0

 dense_2 (Dense)             (None, 1)                 257

=================================================================
Total params: 533,505
Trainable params: 533,505
Non-trainable params: 0
_________________________________________________________________
Model: "model_1"
_________________________________________________________________
 Layer (type)                Output Shape              Param #
=================================================================
 input_2 (InputLayer)        [(None, 100)]             0

 dense_3 (Dense)             (None, 256)               25856

 leaky_re_lu_2 (LeakyReLU)   (None, 256)               0

 batch_normalization (BatchN  (None, 256)              1024
 ormalization)

 dense_4 (Dense)             (None, 512)               131584

 leaky_re_lu_3 (LeakyReLU)   (None, 512)               0

 batch_normalization_1 (Batc  (None, 512)              2048
 hNormalization)

 dense_5 (Dense)             (None, 1024)              525312

 leaky_re_lu_4 (LeakyReLU)   (None, 1024)              0

 batch_normalization_2 (Batc  (None, 1024)             4096
 hNormalization)

 dense_6 (Dense)             (None, 784)               803600

 reshape (Reshape)           (None, 28, 28, 1)         0

=================================================================
Total params: 1,493,520
Trainable params: 1,489,936
Non-trainable params: 3,584
```

```
    from keras.models import load_model
```

```
# Load the trained generator model
generator = load_model('generator_model_test.h5')

# Generate new images
noise = np.random.normal(0, 1, size=(10, 100))
generated_images = generator.predict(noise)

# Rescale images from [-1, 1] to [0, 1]
generated_images = (generated_images + 1) / 2.0

# Plot the generated images
for i in range(10):
    plt.subplot(2, 5, i+1)
    plt.imshow(generated_images[i].reshape(28,28), cmap='gray')
    plt.axis('off')
plt.show()
```

```
WARNING:tensorflow:Error in loading the saved optimizer state. As a result, your m
1/1 [==============================] - 0s 255ms/step
```