

keras2.ipynb ☆

파일 수정 보기 삽입 런타임 도구 도움말 모드 변경사항이 저장됨

RAM
디스크

수정 가능

+ 코드 + 텍스트

✓

1 from tensorflow.keras.datasets import mnist
2 import time

1 (x_train, y_train), (x_test, y_test) = mnist.load_data()

Downloading data from https://storage.googleapis.com/tensorflow/tf-keras-datasets/mnist_noz11493376/11490434 [=====] - 0s 0us/step

1 x_train = x_train / 127.5 - 1
2 x_test = x_test / 127.5 - 1

1 x_train.min(), x_train.max()

(-1.0, 1.0)

1 x_train = x_train.reshape(-1, 784)
2 x_train.shape

(60000, 784)

1 from tensorflow.keras.layers import Dense, LeakyReLU, Dropout, Input
2 from tensorflow.keras.models import Sequential, Model
3 from tensorflow.keras.optimizers import Adam
4 from tensorflow.keras.initializers import RandomNormal
5 import numpy as np
6 import matplotlib.pyplot as plt

1 # Gan에 입력되는 noise에 대한 dimension
2 NOISE_DIM = 10
3
4 # adam optimizer 정의, learning_rate = 0.0002, beta_1로 줍니다.
5 # Vanilla Gan과 DCGAN에서 이렇게 셋팅을 해주는데
6 # 이렇게 해줘야 좋은 학습을 할합니다.
7 adam = Adam(lr=0.0002, beta_1=0.5)

/usr/local/lib/python3.7/dist-packages/tensorflow/python/keras/optimizer_v2/optimizer_v2.py:375: UserWarning: The "lr" argument is deprecated, use "learning_rate" instead.
"The "lr" argument is deprecated, use "learning_rate" instead."

1 generator = Sequential([
2 Dense(256, input_dim=NOISE_DIM),
3 LeakyReLU(0.2),
4 Dense(512),
5 LeakyReLU(0.2),
6 Dense(1024),
7 LeakyReLU(0.2),
8 Dense(28*28, activation='tanh'),
9])

1 generator.summary()

Model: "sequential"

Layer (type) Output Shape Param #

dense_1 (Dense) (None, 256) 2616

leaky_re_lu_1 (LeakyReLU) (None, 256) 0

dense_2 (Dense) (None, 512) 131584

leaky_re_lu_2 (LeakyReLU) (None, 512) 0

dense_3 (Dense) (None, 1024) 525312

leaky_re_lu_3 (LeakyReLU) (None, 1024) 0

dense_4 (Dense) (None, 784) 803600

Total params: 1,463,312
Trainable params: 1,463,312
Non-trainable params: 0

1 discriminator = Sequential([
2 Dense(1024, input_shape=(784,)), kernel_initializer=RandomNormal(stddev=0.02)),
3 LeakyReLU(0.2),
4 Dropout(0.3),
5 Dense(512),
6 LeakyReLU(0.2),
7 Dropout(0.3),
8 Dense(256),
9 LeakyReLU(0.2),
10 Dropout(0.3),
11 Dense(1, activation='sigmoid')
12])

1 discriminator.summary()

Model: "sequential_1"

Layer (type) Output Shape Param #

dense_4 (Dense) (None, 1024) 803840

leaky_re_lu_3 (LeakyReLU) (None, 1024) 0

dropout_1 (Dropout) (None, 1024) 0

dense_5 (Dense) (None, 512) 524800

leaky_re_lu_4 (LeakyReLU) (None, 512) 0

dropout_2 (Dropout) (None, 512) 0

dense_6 (Dense) (None, 256) 131328

leaky_re_lu_5 (LeakyReLU) (None, 256) 0

dropout_3 (Dropout) (None, 256) 0

dense_7 (Dense) (None, 1) 257

Total params: 1,460,225
Trainable params: 1,460,225
Non-trainable params: 0

1 discriminator.compile(loss='binary_crossentropy', optimizer=adam)

1 # discriminator는 학습을 하지 않도록 하며, Gan 모델에서는 generator만 학습하도록 합니다.
2 discriminator.trainable = False
3 gan_input = Input(shape=(NOISE_DIM,))

```
4 x = generator(inputs=gan_input)
5 output = discriminator(x)
```

```
[14] 1 gan = Model(gan_input, output)
```

```
[15] 1 gan.summary()
```

Model: "model"

Layer (type)	Output Shape	Param #
input_1 (InputLayer)	[(None, 10)]	0
sequential (Sequential)	(None, 784)	1463312
sequential_1 (Sequential)	(None, 1)	1460225

Total params: 2,923,537
Trainable params: 1,463,312
Non-trainable params: 1,460,225

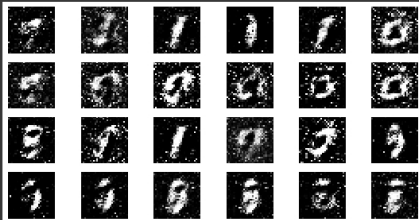
```
[16] 1 gan.compile(loss='binary_crossentropy', optimizer=adam)
```

```
[17] 1 def get_batches(data, batch_size):
2     batches = []
3     for i in range(int(data.shape[0] // batch_size)):
4         batch = data[i * batch_size: (i + 1) * batch_size]
5         batches.append(batch)
6     return np.asarray(batches)
```

```
[18] 1 def visualize_training(epoch, d_losses, g_losses):
2     ...
3     # 오차에 대한 시각화
4     plt.figure(figsize=(8, 4))
5     plt.plot(d_losses, label='Discriminator Loss')
6     plt.plot(g_losses, label='Generator Loss')
7     plt.xlabel('Epoch')
8     plt.ylabel('Loss')
9     plt.legend()
10    plt.show()
11
12    print('epoch: {}, Discriminator Loss: {}, Generator Loss: {}'.format(epoch, np.asarray(d_losses).mean(), np.asarray(g_losses).mean()))
13    ...
14    #샘플 데이터 생성 후 시각화
15    noise = np.random.normal(0, 1, size=(24, NOISE_DIM))
16    generated_images = generator.predict(noise)
17    generated_images = generated_images.reshape(-1, 28, 28)
18
19    plt.figure(figsize=(8, 4))
20    for i in range(generated_images.shape[0]):
21        plt.subplot(4, 6, i+1)
22        plt.imshow(generated_images[i], interpolation='nearest', cmap='gray')
23        plt.axis('off')
24    plt.tight_layout()
25    plt.show()
```

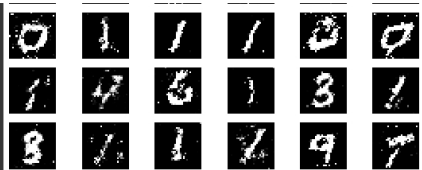
```
[19] 1 BATCH_SIZE = 128
2     EPOCHS= 10
```

```
[20] 1 # discriminator와 gan 모델의 loss 측정을 위한 list 인italize.
2     d_losses = []
3     g_losses = []
4
5     for epoch in range(1, EPOCHS + 1):
6         start = time.time()
7
8         # 각 배치별 학습
9         for real_images in get_batches(x_train, BATCH_SIZE):
10            # 랜덤 노이즈 생성
11            input_noise = np.random.uniform(-1, 1, size=(BATCH_SIZE, NOISE_DIM))
12
13            # 가짜 이미지 데이터 생성
14            generated_images = generator.predict(input_noise)
15
16            # Gan에 학습할 X 데이터 정의
17            x_dis = np.concatenate([real_images, generated_images])
18
19            # Gan에 학습할 Y 데이터 정의
20            y_dis = np.zeros(2 * BATCH_SIZE)
21            y_dis[:BATCH_SIZE] = 0.9
22
23            # Discriminator 훈련
24            discriminator.trainable = True
25            d_loss = discriminator.train_on_batch(x_dis, y_dis)
26
27            # Gan 훈련
28            noise = np.random.uniform(-1, 1, size=(BATCH_SIZE, NOISE_DIM))
29            y_gan = np.ones(BATCH_SIZE)
30
31            # Discriminator의 판별 학습을 방지합니다.
32            discriminator.trainable = False
33            g_loss = gan.train_on_batch(noise, y_gan)
34
35            d_losses.append(d_loss)
36            g_losses.append(g_loss)
37
38            if epoch == 1 or epoch % 10 == 0:
39                visualize_training(epoch, d_losses, g_losses)
40
41            # print (' 에포크 {}에서 걸린 시간은 {} 초 입니다'.format(epoch + 1, time.time() - start))
42            print ('Time for epoch {} is {} sec'.format(epoch, time.time() - start))
```



Time for epoch 1 is 35.962157735824565 sec
Time for epoch 2 is 21.332537889481059 sec
Time for epoch 3 is 21.5112485856201 sec
Time for epoch 4 is 21.60704803466797 sec
Time for epoch 5 is 21.69836682739258 sec
Time for epoch 6 is 21.824479818344116 sec
Time for epoch 7 is 21.75549563217163 sec
Time for epoch 8 is 21.749380350112915 sec
Time for epoch 9 is 21.793075941933837 sec





Time for epoch 10 is 22.54566131362915 sec