


Minggu Ke 2

 Praktikum Training GAN minggu ke 2.ipynb ☆

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1) Importing Python Packages for GAN

```
[1] from keras.datasets import mnist

    from keras.models import Sequential
    from keras.layers import BatchNormalization
    from keras.layers import Dense, Reshape, Flatten
    from keras.layers.advanced_activations import LeakyReLU
    from tensorflow.keras.optimizers import Adam

    import numpy as np
    !mkdir generated_images
```

2) Variables for Neural Networks & Data

```
[2] img_width = 28
    img_height = 28
    channels = 1
    img_shape = (img_width, img_height, channels)
    latent_dim = 100
    adam = Adam(lr=0.0001)
```

/usr/local/lib/python3.7/dist-packages/keras/optimizer_v2/adam.py:105: UserWarning: The `lr` argument is deprecated, use `learning_rate` instead.
super(Adam, self).__init__(name, **kwargs)

3) Building Generator

```
[3] def build_generator():
    model = Sequential()

    model.add(Dense(256, input_dim=latent_dim))
    model.add(LeakyReLU(alpha=0.2))
    model.add(BatchNormalization(momentum=0.8))

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    model.add(BatchNormalization(momentum=0.8))

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

    model.summary()
    return model

generator = build_generator()
```

Model: "sequential"

Layer (type)	Output Shape	Param #
dense (Dense)	(None, 256)	25856
leaky_re_lu (LeakyReLU)	(None, 256)	0
batch_normalization (Batch Normalization)	(None, 256)	1024
dense_1 (Dense)	(None, 256)	65792
leaky_re_lu_1 (LeakyReLU)	(None, 256)	0
batch_normalization_1 (Batch Normalization)	(None, 256)	1024
dense_2 (Dense)	(None, 256)	65792
leaky_re_lu_2 (LeakyReLU)	(None, 256)	0
batch_normalization_2 (Batch Normalization)	(None, 256)	1024
dense_3 (Dense)	(None, 784)	201488
reshape (Reshape)	(None, 28, 28, 1)	0

=====
Total params: 362,000
Trainable params: 360,464
Non-trainable params: 1,536

4) Building Discriminator

```
[4] def build_discriminator():  
    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(Dense(1, activation='sigmoid'))  
  
    model.summary()  
    return model  
  
discriminator = build_discriminator()  
discriminator.compile(loss='binary_crossentropy', optimizer=adam, metrics=['accuracy'])
```

Model: "sequential_1"

Layer (type)	Output Shape	Param #
flatten (Flatten)	(None, 784)	0
dense_4 (Dense)	(None, 512)	401920
leaky_re_lu_3 (LeakyReLU)	(None, 512)	0
dense_5 (Dense)	(None, 256)	131328
dense_6 (Dense)	(None, 1)	257

=====
Total params: 533,505
Trainable params: 533,505
Non-trainable params: 0

5) Connecting Neural Networks to build GAN

```
[5] GAN = Sequential()
    discriminator.trainable = False
    GAN.add(generator)
    GAN.add(discriminator)

    GAN.compile(loss='binary_crossentropy', optimizer=adam)
    GAN.summary()
```

Model: "sequential_2"

Layer (type)	Output Shape	Param #
sequential (Sequential)	(None, 28, 28, 1)	362000
sequential_1 (Sequential)	(None, 1)	533505

=====
Total params: 895,505
Trainable params: 360,464
Non-trainable params: 535,041
=====

▼ 6) Outputting Images

```
✓ [6] #@title
      ## **7) Outputting Images**
      import matplotlib.pyplot as plt
      import glob
      import imageio
      import PIL

      save_name = 0.00000000

      def save_imgs(epoch):
          r, c = 5, 5
          noise = np.random.normal(0, 1, (r * c, latent_dim))
          gen_imgs = generator.predict(noise)
          global save_name
          save_name += 0.00000001
          print("%.8f" % save_name)

          # Rescale images 0 - 1
          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].imshow(gen_imgs[cnt])
                  axs[i,j].axis('off')
                  cnt += 1
          fig.savefig("generated_images/%.8f.png" % save_name)
          print('saved')
          plt.close()
```

7) Training GAN

```
def train(epochs, batch_size=64, save_interval=200):
    (X_train, _), (_, _) = mnist.load_data()

    # print(X_train.shape)
    # Rescale data between -1 and 1
    X_train = X_train / 127.5 - 1.
    # X_train = np.expand_dims(X_train, axis=3)
    # print(X_train.shape)

    # Create our Y for our Neural Networks
    valid = np.ones((batch_size, 1))
    fakes = np.zeros((batch_size, 1))

    for epoch in range(epochs):
        # Get Random Batch
        idx = np.random.randint(0, X_train.shape[0], batch_size)
        imgs = X_train[idx]

        # Generate Fake Images
        noise = np.random.normal(0, 1, (batch_size, latent_dim))
        gen_imgs = generator.predict(noise)

        # Train discriminator
        d_loss_real = discriminator.train_on_batch(imgs, valid)
        d_loss_fake = discriminator.train_on_batch(gen_imgs, fakes)
        d_loss = 0.5 * np.add(d_loss_real, d_loss_fake)

        noise = np.random.normal(0, 1, (batch_size, latent_dim))

        # Inverse y label
        g_loss = GAN.train_on_batch(noise, valid)

        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)

    # print(valid)

train(30000, batch_size=64, save_interval=200)
```

*****	29951	[D loss: 0.589171, acc: 67.97%]	[G loss: 1.395642]
*****	29952	[D loss: 0.558636, acc: 71.09%]	[G loss: 1.352116]
*****	29953	[D loss: 0.509654, acc: 68.75%]	[G loss: 1.188756]
*****	29954	[D loss: 0.520957, acc: 71.88%]	[G loss: 1.620432]
*****	29955	[D loss: 0.602423, acc: 71.09%]	[G loss: 1.591233]
*****	29956	[D loss: 0.588075, acc: 70.31%]	[G loss: 1.427646]
*****	29957	[D loss: 0.600280, acc: 66.41%]	[G loss: 1.279619]
*****	29958	[D loss: 0.543135, acc: 70.31%]	[G loss: 1.429614]
*****	29959	[D loss: 0.517202, acc: 76.56%]	[G loss: 1.734145]
*****	29960	[D loss: 0.631524, acc: 64.84%]	[G loss: 1.572028]
*****	29961	[D loss: 0.664978, acc: 65.62%]	[G loss: 1.397924]
*****	29962	[D loss: 0.627157, acc: 68.75%]	[G loss: 1.363812]
*****	29963	[D loss: 0.566750, acc: 69.53%]	[G loss: 1.481601]
*****	29964	[D loss: 0.547315, acc: 77.34%]	[G loss: 1.559325]
*****	29965	[D loss: 0.536611, acc: 78.12%]	[G loss: 1.421046]
*****	29966	[D loss: 0.550907, acc: 72.66%]	[G loss: 1.263606]
*****	29967	[D loss: 0.660345, acc: 61.72%]	[G loss: 1.153933]
*****	29968	[D loss: 0.538504, acc: 75.00%]	[G loss: 1.307243]
*****	29969	[D loss: 0.532854, acc: 78.12%]	[G loss: 1.593028]
*****	29970	[D loss: 0.462812, acc: 75.00%]	[G loss: 1.746032]
*****	29971	[D loss: 0.456108, acc: 80.47%]	[G loss: 1.841790]
*****	29972	[D loss: 0.583858, acc: 66.41%]	[G loss: 1.461316]
*****	29973	[D loss: 0.619556, acc: 73.44%]	[G loss: 1.292511]
*****	29974	[D loss: 0.539022, acc: 70.31%]	[G loss: 1.429295]
*****	29975	[D loss: 0.501318, acc: 72.66%]	[G loss: 1.671717]
*****	29976	[D loss: 0.398879, acc: 85.16%]	[G loss: 1.906677]
*****	29977	[D loss: 0.647697, acc: 65.62%]	[G loss: 1.524808]
*****	29978	[D loss: 0.634908, acc: 71.09%]	[G loss: 1.205254]
*****	29979	[D loss: 0.640701, acc: 63.28%]	[G loss: 1.311170]
*****	29980	[D loss: 0.584872, acc: 71.09%]	[G loss: 1.338706]
*****	29981	[D loss: 0.582139, acc: 71.88%]	[G loss: 1.609013]
*****	29982	[D loss: 0.496531, acc: 75.78%]	[G loss: 1.774616]
*****	29983	[D loss: 0.573115, acc: 70.31%]	[G loss: 1.567858]
*****	29984	[D loss: 0.488099, acc: 72.66%]	[G loss: 1.230682]
*****	29985	[D loss: 0.571470, acc: 72.66%]	[G loss: 1.097119]
*****	29986	[D loss: 0.610216, acc: 63.28%]	[G loss: 1.478118]
*****	29987	[D loss: 0.513266, acc: 75.78%]	[G loss: 1.715492]
*****	29988	[D loss: 0.448614, acc: 79.69%]	[G loss: 1.665279]
*****	29989	[D loss: 0.480862, acc: 75.00%]	[G loss: 1.635394]
*****	29990	[D loss: 0.593129, acc: 66.41%]	[G loss: 1.359214]
*****	29991	[D loss: 0.499047, acc: 71.88%]	[G loss: 1.414220]
*****	29992	[D loss: 0.526219, acc: 71.88%]	[G loss: 1.494169]
*****	29993	[D loss: 0.515515, acc: 69.53%]	[G loss: 1.756819]
*****	29994	[D loss: 0.524061, acc: 73.44%]	[G loss: 1.669944]
*****	29995	[D loss: 0.545207, acc: 68.75%]	[G loss: 1.393118]
*****	29996	[D loss: 0.628788, acc: 64.06%]	[G loss: 1.135735]
*****	29997	[D loss: 0.590283, acc: 69.53%]	[G loss: 1.451259]
*****	29998	[D loss: 0.558456, acc: 67.97%]	[G loss: 1.557040]
*****	29999	[D loss: 0.514513, acc: 74.22%]	[G loss: 1.758997]

8) Making GIF

```
✓ [12] # Display a single image using the epoch number
3a # def display_image(epoch_no):
    # return PIL.Image.open('generated_images/{}.png'.format(epoch_no))

anim_file = 'dcgan.gif'

with imageio.get_writer(anim_file, mode='I') as writer:
    filenames = glob.glob('generated_images/*.png')
    filenames = sorted(filenames)
    for filename in filenames:
        image = imageio.imread(filename)
        writer.append_data(image)
    image = imageio.imread(filename)
    writer.append_data(image)
```