## Assignment12

June 2, 2021

[4]: from keras.datasets import mnist

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from keras.models import Model
     import keras
     from keras import layers
     from keras import backend as K
     from keras.models import Model
     #from tensorflow.keras.models import Model
     import tensorflow.compat.v1.keras.backend as K
     import tensorflow as tf
     tf.compat.v1.disable_eager_execution()
     from pathlib import Path
[5]: results dir = Path('results').joinpath('vae')
     results_dir.mkdir(parents=True, exist_ok=True)
[6]: import numpy as np
     img_shape = (28, 28, 1)
     batch size = 16
     latent_dim = 2
     input_img = keras.Input(shape=img_shape)
     x = layers.Conv2D(32, 3,
                       padding='same', activation='relu')(input_img)
     x = layers.Conv2D(64, 3,
                       padding='same', activation='relu',
                       strides=(2, 2))(x)
     x = layers.Conv2D(64, 3,padding='same', activation='relu')(x)
     x = layers.Conv2D(64, 3,padding='same', activation='relu')(x)
     shape_before_flattening = K.int_shape(x)
     x = layers.Flatten()(x)
     x = layers.Dense(32, activation='relu')(x)
     z_mean = layers.Dense(latent_dim)(x)
     z_log_var = layers.Dense(latent_dim)(x)
```

```
[7]: def sampling(args):
          z_mean, z_log_var = args
          epsilon = K.random_normal(shape=(K.shape(z_mean)[0], latent_dim),
                                    mean=0., stddev=1.)
          return z_mean + K.exp(z_log_var) * epsilon
      z = layers.Lambda(sampling)([z_mean, z_log_var])
      decoder_input = layers.Input(K.int_shape(z)[1:])
      x = layers.Dense(np.prod(shape_before_flattening[1:]),
                       activation='relu')(decoder input)
      x = layers.Reshape(shape_before_flattening[1:])(x)
      x = layers.Conv2DTranspose(32, 3,
                                      padding='same',
                                      activation='relu',
                                      strides=(2, 2))(x)
      x = layers.Conv2D(1, 3,
                             padding='same',
                            activation='sigmoid')(x)
      decoder = Model(decoder_input, x)
      z_decoded = decoder(z)
[8]: class CustomVariationalLayer(keras.layers.Layer):
          def vae_loss(self, x, z_decoded):
              x = K.flatten(x)
              z_decoded = K.flatten(z_decoded)
              xent_loss = keras.metrics.binary_crossentropy(x, z_decoded)
              kl_loss = -5e-4 * K.mean(
              1 + z_log_var - K.square(z_mean) - K.exp(z_log_var), axis=-1)
              return K.mean(xent_loss + kl_loss)
          def call(self, inputs):
              x = inputs[0]
              z_decoded = inputs[1]
              loss = self.vae_loss(x, z_decoded)
              self.add_loss(loss, inputs=inputs)
              return x
[9]: y = CustomVariationalLayer()([input_img, z_decoded])
[10]: vae = Model(input_img, y)
      vae.compile(optimizer='rmsprop', loss=None)
      vae.summary()
      (x_train, _), (x_test, y_test) = mnist.load_data()
      x_train = x_train.astype('float32') / 255.
      x_train = x_train.reshape(x_train.shape + (1,))
```

WARNING:tensorflow:Output custom\_variational\_layer missing from loss dictionary. We assume this was done on purpose. The fit and evaluate APIs will not be expecting any data to be passed to custom\_variational\_layer.

Model: "model\_1"

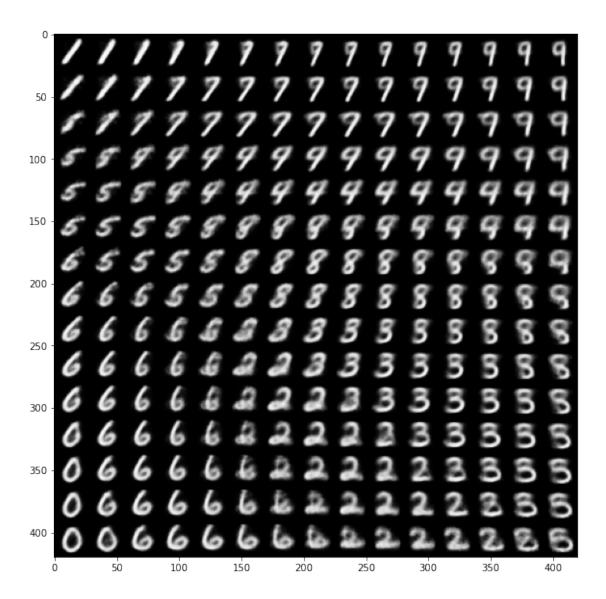
Layer (type)	Output Shape		Connected to
input_1 (InputLayer)			
conv2d (Conv2D)	(None, 28, 28, 32)	320	input_1[0][0]
conv2d_1 (Conv2D)	(None, 14, 14, 64)	18496	conv2d[0][0]
conv2d_2 (Conv2D)	(None, 14, 14, 64)	36928	conv2d_1[0][0]
conv2d_3 (Conv2D)	(None, 14, 14, 64)	36928	conv2d_2[0][0]
flatten (Flatten)	(None, 12544)	0	conv2d_3[0][0]
dense (Dense)	(None, 32)	401440	flatten[0][0]
dense_1 (Dense)	(None, 2)	66	dense[0][0]
dense_2 (Dense)	(None, 2)	66	dense[0][0]
lambda (Lambda)	(None, 2)	0	dense_1[0][0] dense_2[0][0]

```
(None, 28, 28, 1) 56385
                                                     lambda[0][0]
model (Functional)
custom_variational_layer (Custo (None, 28, 28, 1) 0
                                                     input_1[0][0]
                                                      model[0][0]
______
Total params: 550,629
Trainable params: 550,629
Non-trainable params: 0
Train on 60000 samples, validate on 10000 samples
Epoch 1/10
60000/60000 [============= ] - ETA: Os - loss: 3590.0738
/opt/conda/lib/python3.8/site-
packages/tensorflow/python/keras/engine/training.py:2325: UserWarning:
`Model.state_updates` will be removed in a future version. This property should
not be used in TensorFlow 2.0, as `updates` are applied automatically.
 warnings.warn('`Model.state_updates` will be removed in a future version. '
60000/60000 [============= ] - 88s 1ms/sample - loss: 3590.0738
- val_loss: 0.1970
Epoch 2/10
60000/60000 [============ ] - 85s 1ms/sample - loss: 0.1932 -
val_loss: 0.1905
Epoch 3/10
60000/60000 [============= ] - 84s 1ms/sample - loss: 0.1887 -
val_loss: 0.1889
Epoch 4/10
60000/60000 [============= ] - 85s 1ms/sample - loss: 0.1863 -
val_loss: 0.1874
Epoch 5/10
60000/60000 [============ ] - 87s 1ms/sample - loss: 0.1846 -
val loss: 0.1838
Epoch 6/10
60000/60000 [============= ] - 85s 1ms/sample - loss: 0.1835 -
val_loss: 0.1826
Epoch 7/10
60000/60000 [============= ] - 86s 1ms/sample - loss: 0.1825 -
val_loss: 0.1832
Epoch 8/10
60000/60000 [============ ] - 87s 1ms/sample - loss: 0.1819 -
val_loss: 0.1828
Epoch 9/10
60000/60000 [============ ] - 88s 1ms/sample - loss: 0.1812 -
val_loss: 0.1817
```

[10]: <tensorflow.python.keras.callbacks.History at 0x7f60d13758b0>

```
[11]: import matplotlib.pyplot as plt
      from scipy.stats import norm
      n = 15
      digit_size = 28
      figure = np.zeros((digit_size * n, digit_size * n))
      grid_x = norm.ppf(np.linspace(0.05, 0.95, n))
      grid_y = norm.ppf(np.linspace(0.05, 0.95, n))
      for i, yi in enumerate(grid_x):
          for j, xi in enumerate(grid_y):
              z_sample = np.array([[xi, yi]])
              z_sample = np.tile(z_sample, batch_size).reshape(batch_size, 2)
              x_decoded = decoder.predict(z_sample, batch_size=batch_size)
              digit = x_decoded[0].reshape(digit_size, digit_size)
              figure[i * digit_size: (i + 1) * digit_size,
                             j * digit_size: (j + 1) * digit_size] = digit
      plt.figure(figsize=(10, 10))
      plt.imshow(figure, cmap='Greys_r')
      img_file = results_dir.joinpath('Assignment_12_15x15_Grid.png')
      plt.savefig(img_file)
     plt.show()
```

/opt/conda/lib/python3.8/site-



[]: