An Introduction to Deep Learning With Python

[8.4] Generating images with variational autoencoders

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pgs: 300 - 303

VAE encoder network

In [1]:
```
import keras
from keras import layers
from keras import backend as K
from keras.models import Model
import numpy as np
```

```
Using TensorFlow backend.
```

In [2]:
```
img_shape = (28, 28, 1)
batch_size = 16
latent_dim = 2  # Dimensionality of the latent space: a plane

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)
```

```
WARNING:tensorflow:From C:\Users\pablo\AppData\Roaming\Python\Python36\site-packages\tensorflow\python\framework\op_def_library.py:263: colocate_with (from tensorflow.python.framework.ops) is deprecated and will be removed in a future version.
Instructions for updating:
Colocations handled automatically by placer.
```

Latent-space-sampling function

In [3]:
```
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
```

In [4]:
```
z = layers.Lambda(sampling)([z_mean, z_log_var])
```

VAE decoder network, mapping latent space points to images

In [5]:
```
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)
```

Custom layer used to compute the VAE loss

In [6]:
```
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
```

In [7]:
```
y = CustomVariationalLayer()([input_img, z_decoded])
```

Training the VAE

In [8]:
```
from keras.datasets import mnist

vae = Model(input_img, y)
vae.compile(optimizer='rmsprop', loss=None)
vae.summary()
```

```
_________________________________________________________________________________________________
Layer (type)                 Output Shape              Param #     Connected to
=================================================================================================
input_1 (InputLayer)         (None, 28, 28, 1)         0
_________________________________________________________________________________________________
conv2d_1 (Conv2D)            (None, 28, 28, 32)        320         input_1[0][0]
_________________________________________________________________________________________________
conv2d_2 (Conv2D)            (None, 14, 14, 64)        18496       conv2d_1[0][0]
_________________________________________________________________________________________________
conv2d_3 (Conv2D)            (None, 14, 14, 64)        36928       conv2d_2[0][0]
_________________________________________________________________________________________________
conv2d_4 (Conv2D)            (None, 14, 14, 64)        36928       conv2d_3[0][0]
_________________________________________________________________________________________________
flatten_1 (Flatten)          (None, 12544)             0           conv2d_4[0][0]
_________________________________________________________________________________________________
dense_1 (Dense)              (None, 32)                401440      flatten_1[0][0]
_________________________________________________________________________________________________
dense_2 (Dense)              (None, 2)                 66          dense_1[0][0]
_________________________________________________________________________________________________
dense_3 (Dense)              (None, 2)                 66          dense_1[0][0]
_________________________________________________________________________________________________
lambda_1 (Lambda)            (None, 2)                 0           dense_2[0][0]
                                                                   dense_3[0][0]
_________________________________________________________________________________________________
model_1 (Model)              (None, 28, 28, 1)         56385       lambda_1[0][0]
_________________________________________________________________________________________________
custom_variational_layer_1 (Cus [(None, 28, 28, 1),  0           input_1[0][0]
                                                                   model_1[1][0]
=================================================================================================
Total params: 550,629
Trainable params: 550,629
Non-trainable params: 0
_________________________________________________________________________________________________
```

In [9]:
```
(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,))
x_test = x_test.astype('float32') / 255.
x_test = x_test.reshape(x_test.shape + (1,))

vae.fit(x=x_train, y=None,
        shuffle=True,
        epochs=10,
        batch_size=batch_size,
        validation_data=(x_test, None))
```

```
WARNING:tensorflow:From C:\Users\pablo\AppData\Roaming\Python\Python36\site-packages\tensorflow\python\ops\math_op
s.py:3066: to_int32 (from tensorflow.python.ops.math_ops) is deprecated and will be removed in a future version.
Instructions for updating:
Use tf.cast instead.
Train on 60000 samples, validate on 10000 samples
Epoch 1/10
60000/60000 [==============================] - 314s 5ms/step - loss: 2.2031 - val_loss: 0.1994
Epoch 2/10
60000/60000 [==============================] - 455s 8ms/step - loss: 0.1955 - val_loss: 0.1920
Epoch 3/10
60000/60000 [==============================] - 447s 7ms/step - loss: 0.1911 - val_loss: 0.1893
Epoch 4/10
60000/60000 [==============================] - 428s 7ms/step - loss: 0.1887 - val_loss: 0.1875
Epoch 5/10
60000/60000 [==============================] - 426s 7ms/step - loss: 0.1870 - val_loss: 0.1864
Epoch 6/10
60000/60000 [==============================] - 428s 7ms/step - loss: 0.1857 - val_loss: 0.1856
Epoch 7/10
60000/60000 [==============================] - 430s 7ms/step - loss: 0.1847 - val_loss: 0.1847
Epoch 8/10
60000/60000 [==============================] - 402s 7ms/step - loss: 0.1839 - val_loss: 0.1834
Epoch 9/10
60000/60000 [==============================] - 402s 7ms/step - loss: 0.1832 - val_loss: 0.1823
Epoch 10/10
60000/60000 [==============================] - 401s 7ms/step - loss: 0.1828 - val_loss: 0.1833
```

Out[9]:
```
<keras.callbacks.History at 0x1f921503c18>
```

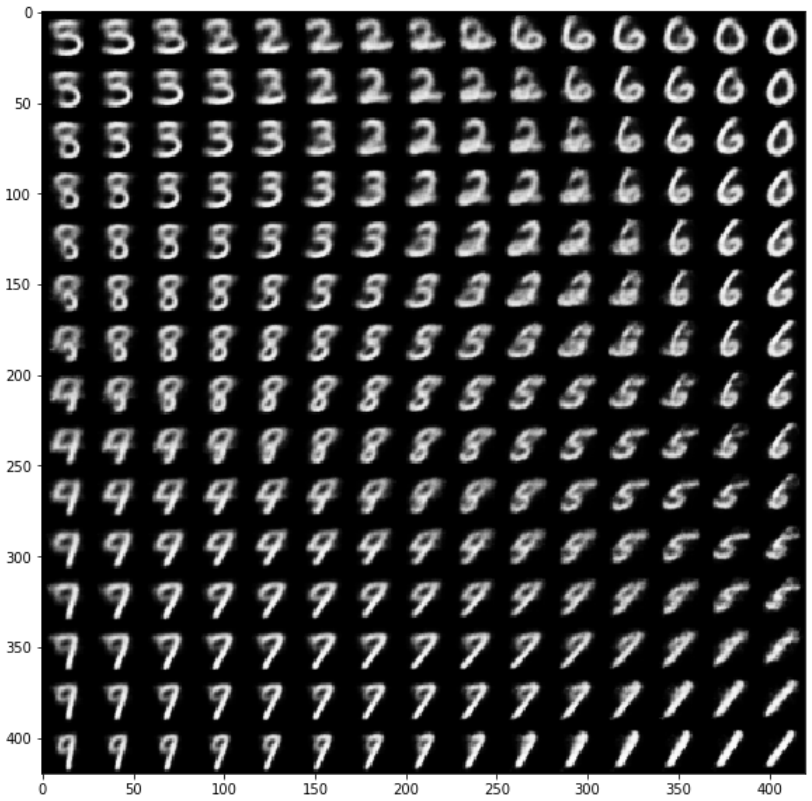
Sampling a grid of points from the 2D latent space and decoding them to images

In [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')
plt.show()
```



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