

Thasina Tabashum

```
In [1]: # to generate gifs
!pip install -q imageio
```

```
In [2]: import tensorflow as tf

import os
import time
import numpy as np
import glob
import matplotlib.pyplot as plt
import PIL
import imageio

from IPython import display
```

```
In [3]: (train_images, _), (test_images, _) = tf.keras.datasets.mnist.load_data()
```

Downloading data from <https://storage.googleapis.com/tensorflow/tf-keras-datasets/mnist.npz> (<https://storage.googleapis.com/tensorflow/tf-keras-datasets/mnist.npz>)
 11493376/11490434 [=====] - 0s 0us/step

```
In [4]: train_images = train_images.reshape(train_images.shape[0], 28, 28, 1).astype('float32')
test_images = test_images.reshape(test_images.shape[0], 28, 28, 1).astype('float32')

# Normalizing the images to the range of [0., 1.]
train_images /= 255.
test_images /= 255.

# Binarization
train_images[train_images >= .5] = 1.
train_images[train_images < .5] = 0.
test_images[test_images >= .5] = 1.
test_images[test_images < .5] = 0.
```

```
In [6]: TRAIN_BUF = 60000
BATCH_SIZE = 100

TEST_BUF = 10000
```

```
In [7]: train_dataset = tf.data.Dataset.from_tensor_slices(train_images).shuffle(TRAIN_BUF)
test_dataset = tf.data.Dataset.from_tensor_slices(test_images).shuffle(TEST_BUF)
```

```

In [8]: class CVAE(tf.keras.Model):
    def __init__(self, latent_dim):
        super(CVAE, self).__init__()
        self.latent_dim = latent_dim
        self.inference_net = tf.keras.Sequential(
            [
                tf.keras.layers.InputLayer(input_shape=(28, 28, 1)),
                tf.keras.layers.Conv2D(
                    filters=32, kernel_size=3, strides=(2, 2), activation='relu'),
                tf.keras.layers.Conv2D(
                    filters=64, kernel_size=3, strides=(2, 2), activation='relu'),
                tf.keras.layers.Flatten(),
                # No activation
                tf.keras.layers.Dense(latent_dim + latent_dim),
            ]
        )

        self.generative_net = tf.keras.Sequential(
            [
                tf.keras.layers.InputLayer(input_shape=(latent_dim,)),
                tf.keras.layers.Dense(units=7*7*32, activation=tf.nn.relu),
                tf.keras.layers.Reshape(target_shape=(7, 7, 32)),
                tf.keras.layers.Conv2DTranspose(
                    filters=64,
                    kernel_size=3,
                    strides=(2, 2),
                    padding="SAME",
                    activation='relu'),
                tf.keras.layers.Conv2DTranspose(
                    filters=32,
                    kernel_size=3,
                    strides=(2, 2),
                    padding="SAME",
                    activation='relu'),
                # No activation
                tf.keras.layers.Conv2DTranspose(
                    filters=1, kernel_size=3, strides=(1, 1), padding="SAME"),
            ]
        )

    @tf.function
    def sample(self, eps=None):
        if eps is None:
            eps = tf.random.normal(shape=(100, self.latent_dim))
        return self.decode(eps, apply_sigmoid=True)

    def encode(self, x):
        mean, logvar = tf.split(self.inference_net(x), num_or_size_splits=2, axis=1)
        return mean, logvar

    def reparameterize(self, mean, logvar):
        eps = tf.random.normal(shape=mean.shape)
        return eps * tf.exp(logvar * .5) + mean

    def decode(self, z, apply_sigmoid=False):
        logits = self.generative_net(z)

```

```

    if apply_sigmoid:
        probs = tf.sigmoid(logits)
        return probs

    return logits

```

```

In [9]: optimizer = tf.keras.optimizers.Adam(1e-4)

def log_normal_pdf(sample, mean, logvar, raxis=1):
    log2pi = tf.math.log(2. * np.pi)
    return tf.reduce_sum(
        -.5 * ((sample - mean) ** 2. * tf.exp(-logvar) + logvar + log2pi),
        axis=raxis)

@tf.function
def compute_loss(model, x):
    mean, logvar = model.encode(x)
    z = model.reparameterize(mean, logvar)
    x_logit = model.decode(z)

    cross_ent = tf.nn.sigmoid_cross_entropy_with_logits(logits=x_logit, labels=x)
    logpx_z = -tf.reduce_sum(cross_ent, axis=[1, 2, 3])
    logpz = log_normal_pdf(z, 0., 0.)
    logqz_x = log_normal_pdf(z, mean, logvar)
    return -tf.reduce_mean(logpx_z + logpz - logqz_x)

@tf.function
def compute_apply_gradients(model, x, optimizer):
    with tf.GradientTape() as tape:
        loss = compute_loss(model, x)
    gradients = tape.gradient(loss, model.trainable_variables)
    optimizer.apply_gradients(zip(gradients, model.trainable_variables))

```

```

In [10]: epochs = 100
latent_dim = 50
num_examples_to_generate = 16

# keeping the random vector constant for generation (prediction) so
# it will be easier to see the improvement.
random_vector_for_generation = tf.random.normal(
    shape=[num_examples_to_generate, latent_dim])
model = CVAE(latent_dim)

```

```
In [11]: def generate_and_save_images(model, epoch, test_input):
    predictions = model.sample(test_input)
    fig = plt.figure(figsize=(4,4))

    for i in range(predictions.shape[0]):
        plt.subplot(4, 4, i+1)
        plt.imshow(predictions[i, :, :, 0], cmap='gray')
        plt.axis('off')

    # tight_layout minimizes the overlap between 2 sub-plots
    plt.savefig('image_at_epoch_{:04d}.png'.format(epoch))
    plt.show()
```

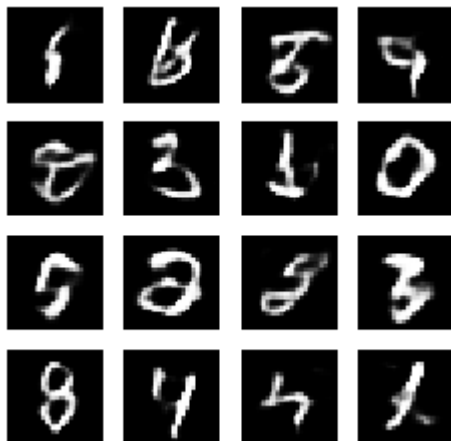
```
In [12]: generate_and_save_images(model, 0, random_vector_for_generation)

for epoch in range(1, epochs + 1):
    start_time = time.time()
    for train_x in train_dataset:
        compute_apply_gradients(model, train_x, optimizer)
    end_time = time.time()

    if epoch % 1 == 0:
        loss = tf.keras.metrics.Mean()
        for test_x in test_dataset:
            loss.compute_loss(model, test_x)
        elbo = -loss.result()
        display.clear_output(wait=False)
        print('Epoch: {}, Test set ELBO: {}, '
              'time elapse for current epoch {}'.format(epoch,
                                                            elbo,
                                                            end_time - start_time))

        generate_and_save_images(
            model, epoch, random_vector_for_generation)
```

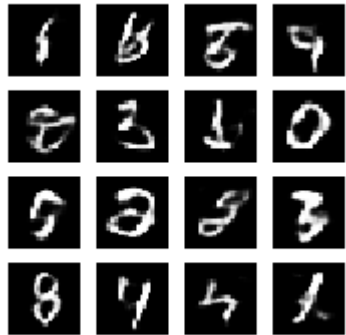
Epoch: 100, Test set ELBO: -78.22090148925781, time elapse for current epoch 28.378284692764282



```
In [13]: def display_image(epoch_no):
    return PIL.Image.open('image_at_epoch_{:04d}.png'.format(epoch_no))
```

```
In [14]: plt.imshow(display_image(epochs))
plt.axis('off')# Display images
```

Out[14]: (-0.5, 287.5, 287.5, -0.5)



```
In [15]: anim_file = 'cvae.gif'

with imageio.get_writer(anim_file, mode='I') as writer:
    filenames = glob.glob('image*.png')
    filenames = sorted(filenames)
    last = -1
    for i,filename in enumerate(filenames):
        frame = 2*(i**0.5)
        if round(frame) > round(last):
            last = frame
        else:
            continue
        image = imageio.imread(filename)
        writer.append_data(image)
    image = imageio.imread(filename)
    writer.append_data(image)

import IPython
if IPython.version_info >= (6,2,0,''):
    display.Image(filename=anim_file)
```

Fashion Minst Dataset

```
In [1]: from keras.datasets import fashion_mnist
```

Using TensorFlow backend.

```
In [2]: import tensorflow as tf

import os
import time
import numpy as np
import glob
import matplotlib.pyplot as plt
import PIL
import imageio

from IPython import display
```

```
In [3]: (train_images, _), (test_images, _) = tf.keras.datasets.fashion_mnist.load_data()
```

```
In [13]: train_images = train_images.reshape(train_images.shape[0], 28, 28, 1).astype('float32')
test_images = test_images.reshape(test_images.shape[0], 28, 28, 1).astype('float32')

# Normalizing the images to the range of [0., 1.]
train_images /= 255.
test_images /= 255.

# Binarization
train_images[train_images >= .5] = 1.
train_images[train_images < .5] = 0.
test_images[test_images >= .5] = 1.
test_images[test_images < .5] = 0.
```

```
In [14]: TRAIN_BUF = 60000
BATCH_SIZE = 100

TEST_BUF = 10000
```

```
In [15]: train_dataset = tf.data.Dataset.from_tensor_slices(train_images).shuffle(TRAIN_BUF)
test_dataset = tf.data.Dataset.from_tensor_slices(test_images).shuffle(TEST_BUF)
```

```

In [16]: class CVAE(tf.keras.Model):
    def __init__(self, latent_dim):
        super(CVAE, self).__init__()
        self.latent_dim = latent_dim
        self.inference_net = tf.keras.Sequential(
            [
                tf.keras.layers.InputLayer(input_shape=(28, 28, 1)),
                tf.keras.layers.Conv2D(
                    filters=32, kernel_size=3, strides=(2, 2), activation='relu'),
                tf.keras.layers.Conv2D(
                    filters=64, kernel_size=3, strides=(2, 2), activation='relu'),
                tf.keras.layers.Flatten(),
                # No activation
                tf.keras.layers.Dense(latent_dim + latent_dim),
            ]
        )

        self.generative_net = tf.keras.Sequential(
            [
                tf.keras.layers.InputLayer(input_shape=(latent_dim,)),
                tf.keras.layers.Dense(units=7*7*32, activation=tf.nn.relu),
                tf.keras.layers.Reshape(target_shape=(7, 7, 32)),
                tf.keras.layers.Conv2DTranspose(
                    filters=64,
                    kernel_size=3,
                    strides=(2, 2),
                    padding="SAME",
                    activation='relu'),
                tf.keras.layers.Conv2DTranspose(
                    filters=32,
                    kernel_size=3,
                    strides=(2, 2),
                    padding="SAME",
                    activation='relu'),
                # No activation
                tf.keras.layers.Conv2DTranspose(
                    filters=1, kernel_size=3, strides=(1, 1), padding="SAME"),
            ]
        )

    @tf.function
    def sample(self, eps=None):
        if eps is None:
            eps = tf.random.normal(shape=(100, self.latent_dim))
        return self.decode(eps, apply_sigmoid=True)

    def encode(self, x):
        mean, logvar = tf.split(self.inference_net(x), num_or_size_splits=2, axis=1)
        return mean, logvar

    def reparameterize(self, mean, logvar):
        eps = tf.random.normal(shape=mean.shape)
        return eps * tf.exp(logvar * .5) + mean

    def decode(self, z, apply_sigmoid=False):
        logits = self.generative_net(z)

```

```

    if apply_sigmoid:
        probs = tf.sigmoid(logits)
        return probs

    return logits

```

```

In [17]: optimizer = tf.keras.optimizers.Adam(1e-4)

def log_normal_pdf(sample, mean, logvar, raxis=1):
    log2pi = tf.math.log(2. * np.pi)
    return tf.reduce_sum(
        -.5 * ((sample - mean) ** 2. * tf.exp(-logvar) + logvar + log2pi),
        axis=raxis)

@tf.function
def compute_loss(model, x):
    mean, logvar = model.encode(x)
    z = model.reparameterize(mean, logvar)
    x_logit = model.decode(z)

    cross_ent = tf.nn.sigmoid_cross_entropy_with_logits(logits=x_logit, labels=x)
    logpx_z = -tf.reduce_sum(cross_ent, axis=[1, 2, 3])
    logpz = log_normal_pdf(z, 0., 0.)
    logqz_x = log_normal_pdf(z, mean, logvar)
    return -tf.reduce_mean(logpx_z + logpz - logqz_x)

@tf.function
def compute_apply_gradients(model, x, optimizer):
    with tf.GradientTape() as tape:
        loss = compute_loss(model, x)
    gradients = tape.gradient(loss, model.trainable_variables)
    optimizer.apply_gradients(zip(gradients, model.trainable_variables))

```

```

In [18]: epochs = 100
latent_dim = 50
num_examples_to_generate = 16

# keeping the random vector constant for generation (prediction) so
# it will be easier to see the improvement.
random_vector_for_generation = tf.random.normal(
    shape=[num_examples_to_generate, latent_dim])
model = CVAE(latent_dim)

```



```
In [19]: def generate_and_save_images(model, epoch, test_input):
    predictions = model.sample(test_input)
    fig = plt.figure(figsize=(4,4))

    for i in range(predictions.shape[0]):
        plt.subplot(4, 4, i+1)
        plt.imshow(predictions[i, :, :, 0], cmap='gray')
        plt.axis('off')

    # tight_layout minimizes the overlap between 2 sub-plots
    plt.savefig('image_at_epoch_{:04d}.png'.format(epoch))
    plt.show()
```

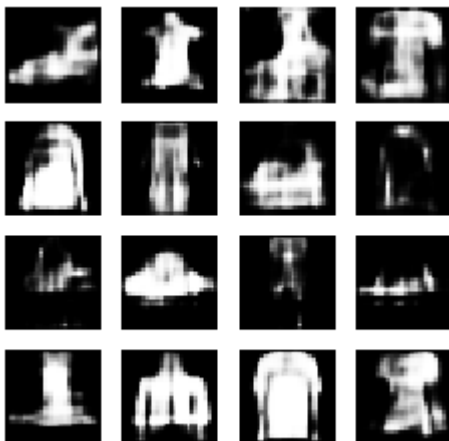
```
In [20]: generate_and_save_images(model, 0, random_vector_for_generation)

for epoch in range(1, epochs + 1):
    start_time = time.time()
    for train_x in train_dataset:
        compute_apply_gradients(model, train_x, optimizer)
    end_time = time.time()

    if epoch % 1 == 0:
        loss = tf.keras.metrics.Mean()
        for test_x in test_dataset:
            loss(compute_loss(model, test_x))
        elbo = -loss.result()
        display.clear_output(wait=False)
        print('Epoch: {}, Test set ELBO: {}, '
              'time elapse for current epoch {}'.format(epoch,
                                                          elbo,
                                                          end_time - start_time))

        generate_and_save_images(
            model, epoch, random_vector_for_generation)
```

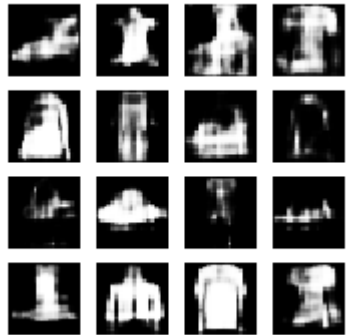
Epoch: 100, Test set ELBO: -121.53059387207031, time elapse for current epoch 2
7.841289043426514



```
In [22]: def display_image(epoch_no):
    return PIL.Image.open('image_at_epoch_{:04d}.png'.format(epoch_no))
```

```
In [23]: plt.imshow(display_image(epochs))  
plt.axis('off')# Display images
```

```
Out[23]: (-0.5, 287.5, 287.5, -0.5)
```



```
In [24]: anim_file = 'cvae2.gif'  
  
with imageio.get_writer(anim_file, mode='I') as writer:  
    filenames = glob.glob('image*.png')  
    filenames = sorted(filenames)  
    last = -1  
    for i,filename in enumerate(filenames):  
        frame = 2*(i**0.5)  
        if round(frame) > round(last):  
            last = frame  
        else:  
            continue  
        image = imageio.imread(filename)  
        writer.append_data(image)  
    image = imageio.imread(filename)  
    writer.append_data(image)  
  
import IPython  
if IPython.version_info >= (6,2,0,''):  
    display.Image(filename=anim_file)
```