Siamese Network

Load data:

```
from tensorflow.keras.datasets.mnist import load_data

(X_train, y_train), (X_test, y_test) = load_data()

print(X_train.shape)
print(X_test.shape)
print(y_train.shape)
```

Build model CNN

```
from tensorflow.keras.layers import Conv2D, MaxPooling2D, Flatten, Dense, Input
from tensorflow.keras.models import Model
from tensorflow.keras import backend as K
inp = Input(shape=(28, 28, 1))
x = Conv2D(filters=8, kernel_size=3, activation='relu')(inp)
x = MaxPooling2D(pool_size=(2,2))(x)
x = Flatten()(x)
x = Dense(units=32, activation='relu')(x)
x = Dense(units=2)(x)
cnn = Model(inputs=inp, outputs=x)
img1 = Input(shape=(28, 28, 1))
img2 = Input(shape=(28, 28, 1))
f1 = cnn(img1)
f2 = cnn(img2)
d = K.sqrt(K.sum(K.square(f1 - f2), axis=1, keepdims=True))
model = Model(inputs=[img1, img2], outputs=d)
model.summary()
cnn.summary()
def loss(y_true, y_pred):
  proba = K.exp(-K.square(y_pred))
  return -K.mean(y_true * K.log(proba) + (1-y_true) * K.log(1-proba))
def loss1(y_true, y_pred):
  return K.mean(y true * K.square(y pred ) + (1-y true) * K.square(K.maximum(1.0 - y pred, 0)))
model.compile(optimizer='adam',loss=loss1)
```

```
import numpy as np
from matplotlib import pyplot as plt
def generator(X, y, k=8):
 unique labels = np.unique(y)
 while True:
   X1 = []
   X2 = []
   y batch = []
   for label in unique labels:
     label_idx = np.where(y == label)[0]
     other_labels = set(unique_labels) - {label}
      for i in range(k):
       i1 = np.random.choice(label idx)
       i2 = np.random.choice(label idx)
       # il must be different from i2
       while il == i2:
         i2 = np.random.choice(label idx)
       # create positive example
       X1.append(X[i1][:, :, None])
       X2.append(X[i2][:, :, None])
       y batch.append(1.0)
       # create negative example
       i1 = np.random.choice(label idx)
       my_label = np.random.choice(list(other_labels))
       i2 = np.random.choice(list(np.where(y == my label)[θ]))
       X1.append(X[i1][:, :, None])
       X2.append(X[i2][:, :, None])
       y batch.append(0.0)
   yield [np.array(X1) / 255., np.array(X2) / 255.], np.array(y_batch)
#For testing
for pair, y in generator(X_test, y_test):
 print('Batch size: ', len(y))
 idx = np.random.choice(range(len(y)))
 print(pair[0][idx].shape)
 print('Pair label:', y[idx])
 plt.subplot(121)
 plt.imshow(pair[0][idx].reshape(28, 28), cmap='gray')
 plt.subplot(122)
 plt.imshow(pair[1][idx].reshape(28, 28), cmap='gray')
 break
```

Fit model

Visualize learning process

```
plt.plot(history.history['loss'], label='Train', c='b')
plt.plot(history.history['val_loss'], label='Validation', c='r')
plt.legend()
plt.xlabel('Epochs', fontsize=16)
plt.ylabel('Loss', fontsize=16)
```

```
for pair, y in generator(X_test, y_test):
    y_pred = model.predict(pair)
    print('Batch size: ', len(y))
    idx = np.random.choice(range(len(y)))
    print('Pair label:', y[idx])
    print('Distance:', y_pred[idx])

f1 = cnn(pair[0])
    f2 = cnn(pair[1])
    d = np.sqrt(np.sum((f1-f2)**2, axis=1, keepdims=True))
    print("Distance by features:", d[idx])

plt.subplot(121)
    plt.imshow(pair[0][idx].reshape(28, 28), cmap='gray')
    plt.subplot(122)
    plt.imshow(pair[1][idx].reshape(28, 28), cmap='gray')
    break
```

Visualize new feature space

```
f = cnn.predict(X_test/255.)
p=plt.scatter(f[:, 0], f[:, 1], c=y_test, s=1)
plt.colorbar(p)
```

Save model

```
cnn.save('cnn_loss1.h5')
```

Load model and test

```
from tensorflow.keras.models import load_model
m = load_model('cnn_loss1.h5')

fl = m.predict(X_test / 255.)
p=plt.scatter(f1[:, 0], f1[:, 1], c=y_test, s=1)
plt.colorbar(p)
```

Visualize negative distance and positive distance

```
i = 0
y_true = []
y_pred = []
for pair, y in generator(X_test, y_test):
    f1 = cnn(pair[0])
    f2 = cnn(pair[1])
    d = np.sqrt(np.sum((f1-f2)**2, axis=1, keepdims=True))
    y_pred += list(d.ravel())
    y_true += list(y)
    i += 1
    if i > 500:
        break
```

Plot histogram

```
y_pred = np.array(y_pred)
y_true = np.array(y_true)

positive_distances = y_pred[y_true == 1]
negative_distances = y_pred[y_true == 0]

plt.hist(positive_distances, color='r', density=True, bins=20)
plt.hist(negative_distances, color='b', density=True, bins=20)
```

Check report using sklearn

```
thresh = 0.5
y_pred_ = y_pred < thresh</pre>
y_pred_.astype('uint8')
from sklearn.metrics import classification_report
print(classification_report(y_true, y_pred_))
                          recall f1-score
             precision
                                             support
        0.0
                  0.94
                            0.97
                                       0.96
                                                40080
        1.0
                  0.97
                                       0.96
                                                40080
                            0.94
                                      0.96
0.96
   accuracy
                                                80160
                  0.96
                             0.96
  macro avg
                                                80160
weighted avg
                  0.96
                             0.96
                                       0.96
                                                80160
```