```
In [1]: import cv2
        import os
        import random
        import numpy as np
        from matplotlib import pyplot as plt
        import uuid
        from tensorflow.keras.models import Model
        from tensorflow.keras.layers import Layer, Conv2D, Dense, MaxPooling2D, Input
        import tensorflow as tf
        from tensorflow.keras.metrics import Precision, Recall
In [2]: cwd = os.getcwd()
In [3]: cwd
Out[3]: '/home/kumar.amit1/ondemand/data/sys/myjobs/projects/default/1/data'
In [4]: POS_PATH ='data_crop/positive_crop'
        NEG_PATH ='data_crop/negative_crop'
        ANC PATH ='data crop/anchor crop'
```

Loading and Processing Image

```
In [5]: # Get Image directory
anchor = tf.data.Dataset.list_files(ANC_PATH+'/*.jpg').take(3000)
positive = tf.data.Dataset.list_files(POS_PATH+'/*.jpg').take(3000)
negative = tf.data.Dataset.list_files(NEG_PATH+'/*.jpg').take(3000)
```

2023-04-25 22:24:44.706003: I tensorflow/core/platform/cpu_feature_guard.cc: 142] This TensorFlow binary is optimized with oneAPI Deep Neural Network Lib rary (oneDNN) to use the following CPU instructions in performance-critical operations: SSE4.1 SSE4.2 AVX AVX2 FMA

To enable them in other operations, rebuild TensorFlow with the appropriate compiler flags.

2023-04-25 22:24:45.519464: I tensorflow/core/common_runtime/gpu/gpu_device. cc:1510] Created device /job:localhost/replica:0/task:0/device:GPU:0 with 77 292 MB memory: -> device: 0, name: NVIDIA A100-SXM4-80GB, pci bus id: 0000: bd:00.0, compute capability: 8.0

```
In [6]: # Sacle and Resize
def preprocess(file_path):

    # Read in image from file path
    byte_img = tf.io.read_file(file_path)
    # Load in the image
    img = tf.io.decode_jpeg(byte_img)

# Preprocessing steps - resizing the image to be 100x100x3
img = tf.image.resize(img, (300,300))
# Scale image to be between 0 and 1
img = img / 255.0

# Return image
return img
```

Creating Labels

Train and Test Partiton

```
In [10]: def preprocess_twin(input_img, validation_img, label):
    return(preprocess(input_img), preprocess(validation_img), label)
```

```
In [11]: res = preprocess_twin(*exampple)
         plt.imshow(res[1])
         print(res[2])
         1.0
            0
           50
          100
          150
          200
          250
                           150
                                 200
                                     250
                       100
In [12]: # Build dataloader pipeline
         data = data.map(preprocess_twin)
         data = data.cache()
         data = data.shuffle(buffer_size=10000)
In [13]: # Training partition
         train_data = data.take(round(len(data)*.7))
         train_data = train_data.batch(16)
         train_data = train_data.prefetch(8)
In [14]: #Testing partition
```

Model Engineering

test_data = test_data.batch(16)
test_data = test_data.prefetch(8)

test_data = data.skip(round(len(data)*.7))
test_data = test_data.take(round(len(data)*.3))

Embedding Model

```
In [15]: def form_embedding():
             #Input Layer
             inputlayer = Input(shape=(300,300,3), name='input image')
             #First layer
             c1 = Conv2D(64, (10,10), activation='relu')(inputlayer)
             m1 = MaxPooling2D(64, (2,2), padding='same')(c1)
             # Second Layer
             c2 = Conv2D(128, (7,7), activation='relu')(m1)
             m2 = MaxPooling2D(64, (2,2), padding='same')(c2)
             # Third Layer
             c3 = Conv2D(128, (4,4), activation='relu')(m2)
             m3 = MaxPooling2D(64, (2,2), padding='same')(c3)
             #Final layer
             c4 = Conv2D(256, (4,4), activation='relu')(m3)
             f1 = Flatten()(c4)
             d1 = Dense(4096, activation='sigmoid')(f1)
             return Model(inputs=[inputlayer], outputs=[d1], name='embedding')
```

```
In [16]: embedding=form_embedding()
```

distance Layer

```
In [17]: # Siamese L1 Distance class
class L1Dist(Layer):

    # Init method - inheritance
    def __init__(self, **kwargs):
        super().__init__()

# Magic happens here - similarity calculation
    def call(self, input_embedding, validation_embedding):
        return tf.math.abs(input_embedding - validation_embedding)
```

Siamese network

```
In [18]: def make_siamese_model():
    # Anchor image input in the network
    input_image = Input(name='input_img', shape=(300,300,3))

# Validation image in the network
    validation_image = Input(name='validation_img', shape=(300,300,3))

# Combine siamese distance components
    siamese_layer = L1Dist()
    siamese_layer._name = 'distance'
    #distances = siamese_layer(embedding_efB4(input_image), embedding_efB4(vadistances = siamese_layer(embedding(input_image), embedding(validation_image))

# Classification Layer
    classifier = Dense(1, activation='sigmoid')(distances)
    return Model(inputs=[input_image, validation_image], outputs=classifier,
```

```
In [19]: siamese_model = make_siamese_model()
```

In [20]: siamese model.summary() Model: "SiameseNetwork" Layer (type) Output Shape Param # Connected t ______ _____ input_img (InputLayer) [(None, 300, 300, 3) 0 validation_img (InputLayer) [(None, 300, 300, 3) 0 embedding (Functional) (None, 4096) 1008893248 input_img [0][0] validation img[0][0] distance (L1Dist) (None, 4096) embedding [0][0] embedding [1][0] dense_1 (Dense) (None, 1) 4097 distance[0] [0] Total params: 1,008,897,345 Trainable params: 1,008,897,345 Non-trainable params: 0

Training

```
In [21]: # Loss and Optimier
binary_cross_loss = tf.losses.BinaryCrossentropy()
opt = tf.keras.optimizers.Adam(1e-4)

In [22]: # establish checkpoint
checkpoint_dir = './training_checkpoints'
checkpoint_prefix = os.path.join(checkpoint_dir, 'ckpt')
checkpoint = tf.train.Checkpoint(opt=opt, siamese model=siamese model)
```

```
In [23]: # Train Step Function
         @tf.function
         def train_step(batch):
             # Record all of our operations
             with tf.GradientTape() as tape:
                 # Get anchor and positive/negative image
                 X = batch[:2]
                 # Get Label
                 y = batch[2]
                 # Forward pass
                 yhat = siamese_model(X, training=True)
                 # Calculate loss
                 loss = binary_cross_loss(y, yhat)
             print(loss)
             # Calculate gradients
             grad = tape.gradient(loss, siamese_model.trainable_variables)
             # Calculate updated weights and apply to siamese model
             opt.apply_gradients(zip(grad, siamese_model.trainable_variables))
             # Return Loss
             return loss
```

```
In [24]: # Training Loop
         def train(data, EPOCHS):
             # Loop through epochs
             for epoch in range(1, EPOCHS+1):
                 print('\n Epoch {}/{}'.format(epoch, EPOCHS))
                 progbar = tf.keras.utils.Progbar(len(data))
                 # Creating a metric object
                 r = Recall()
                 p = Precision()
                 # Loop through each batch
                 for idx, batch in enumerate(data):
                     # Run train step here
                     loss = train_step(batch)
                     yhat = siamese model.predict(batch[:2])
                      r.update_state(batch[2], yhat)
                     p.update state(batch[2], yhat)
                     progbar.update(idx+1)
                 print(loss.numpy(), r.result().numpy(), p.result().numpy())
                 # Save checkpoints
                 if epoch % 10 == 0:
                     checkpoint.save(file prefix=checkpoint prefix)
```

Train The model

```
In [25]: EPOCHS = 5
In [26]: train(train_data, EPOCHS)
         Epoch 1/5
        2023-04-25 22:24:46.997862: I tensorflow/compiler/mlir/mlir graph optimizati
        on_pass.cc:185] None of the MLIR Optimization Passes are enabled (registered
        2)
        2023-04-25 22:24:56.998202: I tensorflow/core/kernels/data/shuffle dataset o
        p.cc:175] Filling up shuffle buffer (this may take a while): 785 of 10000
        2023-04-25 22:24:57.633292: I tensorflow/core/kernels/data/shuffle dataset o
        p.cc:228] Shuffle buffer filled.
        Tensor("binary_crossentropy/weighted_loss/value:0", shape=(), dtype=float32)
        Tensor("binary_crossentropy/weighted_loss/value:0", shape=(), dtype=float32)
        2023-04-25 22:24:58.944432: I tensorflow/stream executor/cuda/cuda dnn.cc:36
        9] Loaded cuDNN version 8201
        2023-04-25 22:24:59.986253: W tensorflow/stream executor/gpu/asm compiler.c
        c:77] Couldn't get ptxas version string: Internal: Running ptxas --version r
        eturned 32512
        2023-04-25 22:25:00.074473: W tensorflow/stream executor/gpu/redzone allocat
        or.cc:314] Internal: ptxas exited with non-zero error code 32512, output:
        Relying on driver to perform ptx compilation.
        Modify $PATH to customize ptxas location.
        This message will be only logged once.
        2023-04-25 22:25:01.890914: I tensorflow/stream_executor/cuda/cuda_blas.cc:1
        760] TensorFloat-32 will be used for the matrix multiplication. This will on
        ly be logged once.
        weighted_loss/value:0", shape=(), dtype=float32)
        45/45 [============= ] - 48s 665ms/step
        0.021557642 0.9854651 0.96033996
         Epoch 2/5
        45/45 [============ ] - 29s 634ms/step
        0.043293778 1.0 1.0
         Epoch 3/5
        45/45 [========= ] - 29s 633ms/step
        1.2546935e-05 0.9971671 0.9971671
         Epoch 4/5
        45/45 [============ ] - 28s 627ms/step
        0.0017192084 0.99712646 0.9942693
         Epoch 5/5
        45/45 [=========== ] - 28s 615ms/step
        0.0038640506 0.9941691 0.98554915
```

Evaluating Model

```
In [27]: # Batch of test data
         test_input, test_val, y_true = test_data.as_numpy_iterator().next()
         y_hat = siamese_model.predict([test_input, test_val])
In [28]: # Post processing the results
         [1 if prediction > 0.5 else 0 for prediction in y_hat ]
         print(y_true)
         [1. 1. 0. 0. 0. 1. 1. 1. 0. 1. 0. 0. 1. 1. 1. 1.]
         Comparision Matrix
In [29]: # Creating a metric object
         m = Recall()
         # Calculating the recall value
         m.update_state(y_true, y_hat)
         # Return Recall Result
         m.result().numpy()
Out[29]: 1.0
In [30]: r = Recall()
         p = Precision()
         for test_input, test_val, y_true in test_data.as_numpy_iterator():
             yhat = siamese_model.predict([test_input, test_val])
             r.update_state(y_true, yhat)
             p.update_state(y_true,yhat)
         print(r.result().numpy(), p.result().numpy())
```

1.0 1.0

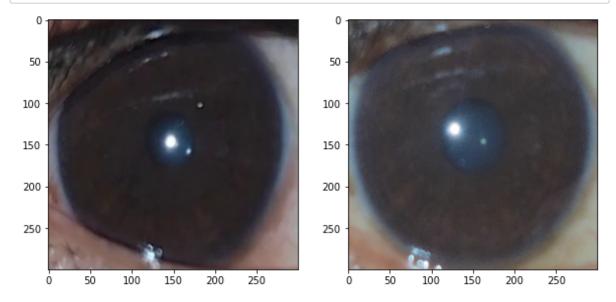
Result

```
In [31]: # Set plot size
    plt.figure(figsize=(10,8))

# Set first subplot
    plt.subplot(1,2,1)
    plt.imshow(test_input[5])

# Set second subplot
    plt.subplot(1,2,2)
    plt.imshow(test_val[5])

# Renders cleanly
    plt.show()
```



Save Model

```
In [ ]: #Save model weights
     #siamese_model.save('siamesemodelv2.h5')
```

WARNING:tensorflow:Compiled the loaded model, but the compiled metrics have yet to be built. `model.compile_metrics` will be empty until you train or ev aluate the model.

2023-04-25 22:27:33.489992: W tensorflow/python/util/util.cc:348] Sets are n ot currently considered sequences, but this may change in the future, so con sider avoiding using them.

In []:	<pre># Make predictions with reloaded model #siamese_model.predict([test_input, test_val])</pre>
In []:	