22p-9252-tazmeen-afroz-lab-task

May 3, 2024

[]: import tensorflow as tf

```
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Conv2D, MaxPooling2D, Flatten, Dense
from tensorflow.keras.preprocessing.image import ImageDataGenerator
# Define directory containing your images
train_dir = '/kaggle/input/gender-detection-and-classification-image-dataset/
 ⇔train'
validation_dir = '/kaggle/input/
 ⇒gender-detection-and-classification-image-dataset/test'
img_width, img_height = 224,224
batch_size = 32
# Create image data generators for training and validation
train_datagen = ImageDataGenerator(rescale=1./255)
test_datagen = ImageDataGenerator(rescale=1./255)
train_generator = train_datagen.flow_from_directory(train_dir,_
 →target_size=(img_width, img_height), batch_size=batch_size,
 ⇔class_mode='binary')
validation_generator = test_datagen.flow_from_directory(validation_dir,_
  →target_size=(img_width, img_height), batch_size=batch_size,
  ⇔class_mode='binary')
2024-05-03 09:47:33.977035: E
external/local xla/xla/stream_executor/cuda/cuda_dnn.cc:9261] Unable to register
cuDNN factory: Attempting to register factory for plugin cuDNN when one has
already been registered
2024-05-03 09:47:33.977085: E
external/local_xla/xla/stream_executor/cuda/cuda_fft.cc:607] Unable to register
cuFFT factory: Attempting to register factory for plugin cuFFT when one has
already been registered
2024-05-03 09:47:33.978573: E
external/local_xla/xla/stream_executor/cuda/cuda_blas.cc:1515] Unable to
register cuBLAS factory: Attempting to register factory for plugin cuBLAS when
```

one has already been registered

[]: # Define the CNN model

Found 220 images belonging to 2 classes. Found 80 images belonging to 2 classes.

```
model = Sequential()
model.add(Conv2D(64, (3, 3), activation='relu', input_shape=(img_width,_u

→img_height, 3)))
model.add(MaxPooling2D((2, 2)))
model.add(Conv2D(128, (3, 3), activation='relu'))
model.add(MaxPooling2D((2, 2)))
model.add(Conv2D(32, (3, 3), activation='relu'))
model.add(MaxPooling2D((2, 2)))
model.add(Flatten())
model.add(Dense(64, activation='relu'))
model.add(Dense(1, activation='sigmoid'))
# Compile the model
model.compile(optimizer='adam', loss='binary_crossentropy',_
  →metrics=['accuracy'])
# Train the model
history = model.fit(train_generator, steps_per_epoch=train_generator.samples //_
 ⇔batch_size, epochs=10, validation_data=validation_generator,
 avalidation_steps=validation_generator.samples // batch_size)
# Calculate accuracy
accuracy = model.evaluate(validation_generator)
print("Accuracy: %.2f%%" % (accuracy[1] * 100))
/opt/conda/lib/python3.10/site-
packages/keras/src/layers/convolutional/base_conv.py:99: UserWarning: Do not
pass an `input_shape`/`input_dim` argument to a layer. When using Sequential
models, prefer using an `Input(shape)` object as the first layer in the model
instead.
  super().__init__(
Epoch 1/10
/opt/conda/lib/python3.10/site-
packages/keras/src/trainers/data_adapters/py_dataset_adapter.py:120:
UserWarning: Your `PyDataset` class should call `super().__init__(**kwargs)` in
its constructor. `**kwargs` can include `workers`, `use_multiprocessing`,
`max_queue_size`. Do not pass these arguments to `fit()`, as they will be
ignored.
  self._warn_if_super_not_called()
2/6
               Os 76ms/step -
```

accuracy: 0.5000 - loss: 0.7118 WARNING: All log messages before absl::InitializeLog() is called are written to STDERR I0000 00:00:1714729691.084101 3103 device_compiler.h:186] Compiled cluster using XLA! This line is logged at most once for the lifetime of the process. W0000 00:00:1714729691.099992 3103 graph_launch.cc:671] Fallback to op-by-op mode because memset node breaks graph update 0s 3s/step accuracy: 0.5067 - loss: 0.7124 W0000 00:00:1714729712.083683 3104 graph_launch.cc:671] Fallback to op-by-op mode because memset node breaks graph update 6/6 52s 5s/step accuracy: 0.5073 - loss: 0.7116 - val_accuracy: 0.4844 - val_loss: 0.6931 Epoch 2/10 1/6 **Os** 107ms/step accuracy: 0.4062 - loss: 0.6942 /opt/conda/lib/python3.10/contextlib.py:153: UserWarning: Your input ran out of data; interrupting training. Make sure that your dataset or generator can generate at least `steps_per_epoch * epochs` batches. You may need to use the `.repeat()` function when building your dataset. self.gen.throw(typ, value, traceback) 6/6 4s 830ms/step accuracy: 0.4062 - loss: 0.6942 - val accuracy: 0.7500 - val loss: 0.6924 Epoch 3/10 6/6 19s 1s/step accuracy: 0.6012 - loss: 0.6929 - val_accuracy: 0.5000 - val_loss: 0.6928 Epoch 4/10 6/6 1s 209ms/step accuracy: 0.4375 - loss: 0.6943 - val_accuracy: 0.6875 - val_loss: 0.6910 Epoch 5/10 19s 1s/step accuracy: 0.6495 - loss: 0.6920 - val accuracy: 0.6250 - val loss: 0.6918 Epoch 6/10 1s 251ms/step accuracy: 0.5938 - loss: 0.6900 - val_accuracy: 0.6250 - val_loss: 0.6912 Epoch 7/10 20s 1s/step accuracy: 0.6292 - loss: 0.6892 - val_accuracy: 0.5625 - val_loss: 0.6893 Epoch 8/10 1s 197ms/step accuracy: 0.6429 - loss: 0.6851 - val_accuracy: 0.3125 - val_loss: 0.6944 Epoch 9/10 19s 975ms/step accuracy: 0.7656 - loss: 0.6777 - val_accuracy: 0.5156 - val_loss: 0.7277 Epoch 10/10

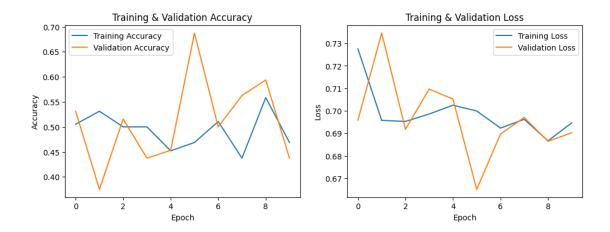
```
6/6
                    2s 282ms/step -
    accuracy: 0.4688 - loss: 0.6916 - val_accuracy: 0.5000 - val_loss: 0.7670
    3/3
                    6s 2s/step -
    accuracy: 0.5273 - loss: 0.8085
    Accuracy: 50.00%
[]:|
         model2 = Sequential()
         model2.add(Conv2D(64, (3, 3), activation='relu', input_shape=(img_width, u
      →img_height, 3)))
         model2.add(MaxPooling2D((2, 2)))
         model2.add(Conv2D(128, (3,3), activation='relu'))
         model2.add(MaxPooling2D((2, 2)))
         model2.add(Conv2D(256, (3,3), activation='relu'))
         model2.add(MaxPooling2D((2, 2)))
         model2.add(Conv2D(512, (3,3), activation='relu'))
         model2.add(MaxPooling2D((2, 2)))
         # Flattening tensors
         model2.add(Flatten())
         # Fully-Connected Layers
         model2.add(Dense(64, activation='relu'))
         # Output Layer
         model2.add(Dense(1, activation='sigmoid'))
         # Compile the model
     model2.compile(optimizer='adam', loss='binary_crossentropy',_
      →metrics=['accuracy'])
     # Train the model
     history = model2.fit(train_generator, steps_per_epoch=train_generator.samples //
      → batch_size, epochs=10, validation_data=validation_generator, u
      avalidation_steps=validation_generator.samples // batch_size)
     # Calculate accuracy
     accuracy = model2.evaluate(validation_generator)
     print("Accuracy: %.2f%%" % (accuracy[1] * 100))
    Epoch 1/10
    6/6
                    38s 2s/step -
    accuracy: 0.4193 - loss: 1.2403 - val_accuracy: 0.5156 - val_loss: 0.6935
```

```
Epoch 2/10
    6/6
                    16s 575ms/step -
    accuracy: 0.4643 - loss: 0.7001 - val accuracy: 0.4375 - val loss: 0.6986
    Epoch 3/10
    6/6
                    20s 1s/step -
    accuracy: 0.4189 - loss: 0.6965 - val_accuracy: 0.5000 - val_loss: 0.6930
    6/6
                    1s 230ms/step -
    accuracy: 0.4688 - loss: 0.6930 - val_accuracy: 0.5625 - val_loss: 0.6930
    Epoch 5/10
    6/6
                    20s 1s/step -
    accuracy: 0.4701 - loss: 0.6935 - val_accuracy: 0.5312 - val_loss: 0.6931
    Epoch 6/10
    6/6
                    1s 150ms/step -
    accuracy: 0.6562 - loss: 0.6927 - val_accuracy: 0.5625 - val_loss: 0.6924
    Epoch 7/10
    6/6
                    20s 1s/step -
    accuracy: 0.5236 - loss: 0.6932 - val accuracy: 0.5312 - val loss: 0.6926
    Epoch 8/10
    6/6
                    1s 258ms/step -
    accuracy: 0.5625 - loss: 0.6929 - val_accuracy: 0.3125 - val_loss: 0.6932
    Epoch 9/10
                    21s 1s/step -
    accuracy: 0.4660 - loss: 0.6932 - val_accuracy: 0.4688 - val_loss: 0.6939
    Epoch 10/10
    6/6
                    1s 113ms/step -
    accuracy: 0.4688 - loss: 0.6924 - val_accuracy: 0.6250 - val_loss: 0.6887
                    6s 2s/step -
    accuracy: 0.4922 - loss: 0.6929
    Accuracy: 50.00%
[]:|
         model3 = Sequential()
         model3.add(Conv2D(32, (3, 3), activation='relu', input_shape=(img_width,_u
      →img_height, 3)))
         model3.add(MaxPooling2D((2, 2)))
         model3.add(Conv2D(64, (3,3), activation='relu'))
         model3.add(MaxPooling2D((2, 2)))
         model3.add(Conv2D(128, (3,3), activation='relu'))
         model3.add(MaxPooling2D((2, 2)))
         model3.add(Conv2D(256, (3,3), activation='relu'))
         model3.add(MaxPooling2D((2, 2)))
         # Flattening tensors
         model3.add(Flatten())
```

```
# Fully-Connected Layers
    model3.add(Dense(30, activation='sigmoid'))
    # Output Layer
    model3.add(Dense(1, activation='sigmoid'))
    # Compile the model
model3.compile(optimizer='adam', loss='binary_crossentropy',_
  →metrics=['accuracy'])
# Train the model
history = model3.fit(train_generator, steps_per_epoch=train_generator.samples //
 → batch_size, epochs=10, validation_data=validation_generator,
 yvalidation_steps=validation_generator.samples // batch_size)
# Calculate accuracy
accuracy = model3.evaluate(validation_generator)
print("Accuracy: %.2f%%" % (accuracy[1] * 100))
Epoch 1/10
6/6
               47s 4s/step -
accuracy: 0.5265 - loss: 0.7191 - val_accuracy: 0.5312 - val_loss: 0.6959
Epoch 2/10
6/6
               3s 681ms/step -
accuracy: 0.5312 - loss: 0.6957 - val_accuracy: 0.3750 - val_loss: 0.7346
Epoch 3/10
6/6
               18s 956ms/step -
accuracy: 0.5124 - loss: 0.6958 - val accuracy: 0.5156 - val loss: 0.6918
Epoch 4/10
6/6
               1s 246ms/step -
accuracy: 0.5000 - loss: 0.6986 - val_accuracy: 0.4375 - val_loss: 0.7096
Epoch 5/10
6/6
               20s 1s/step -
accuracy: 0.4465 - loss: 0.7029 - val_accuracy: 0.4531 - val_loss: 0.7052
Epoch 6/10
6/6
               1s 218ms/step -
accuracy: 0.4688 - loss: 0.6999 - val_accuracy: 0.6875 - val_loss: 0.6651
Epoch 7/10
               19s 1s/step -
6/6
accuracy: 0.5105 - loss: 0.6912 - val_accuracy: 0.5000 - val_loss: 0.6897
Epoch 8/10
6/6
               1s 215ms/step -
accuracy: 0.4375 - loss: 0.6962 - val accuracy: 0.5625 - val loss: 0.6970
Epoch 9/10
6/6
               19s 1s/step -
accuracy: 0.5345 - loss: 0.6838 - val_accuracy: 0.5938 - val_loss: 0.6865
```

Training Accuracy: 46.88% Validation Accuracy: 43.75%

```
[]: import matplotlib.pyplot as plt
     plt.figure(figsize=(12, 4))
     plt.subplot(1, 2, 1)
     plt.plot(history.history['accuracy'], label='Training Accuracy')
     plt.plot(history.history['val accuracy'], label='Validation Accuracy')
     plt.title('Training & Validation Accuracy')
     plt.xlabel('Epoch')
     plt.ylabel('Accuracy')
     plt.legend()
     plt.subplot(1, 2, 2)
     plt.plot(history.history['loss'], label='Training Loss')
     plt.plot(history.history['val_loss'], label='Validation Loss')
     plt.title('Training & Validation Loss')
     plt.xlabel('Epoch')
     plt.ylabel('Loss')
     plt.legend()
     plt.show()
```



[]: