Model Optimization and Tuning Phase Template

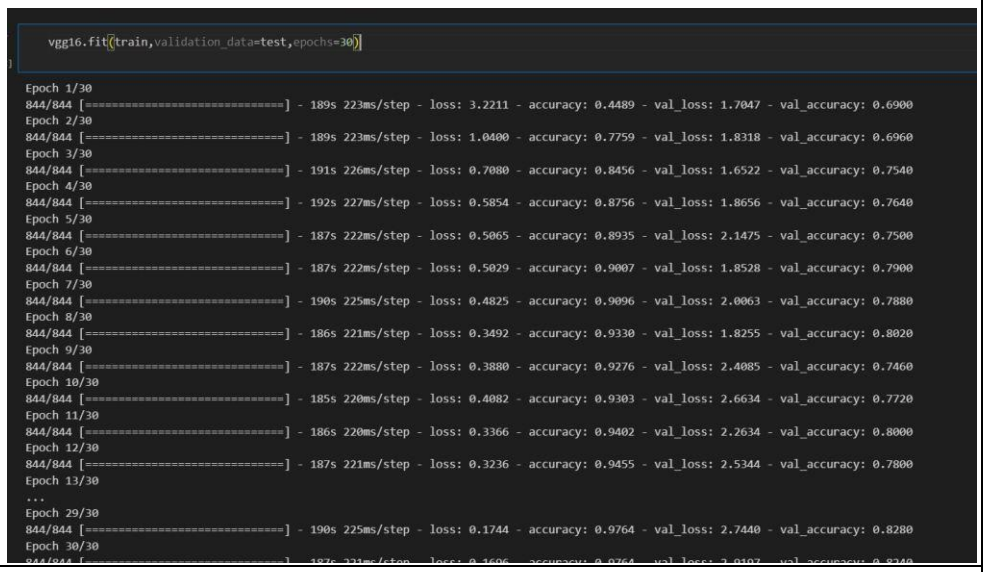
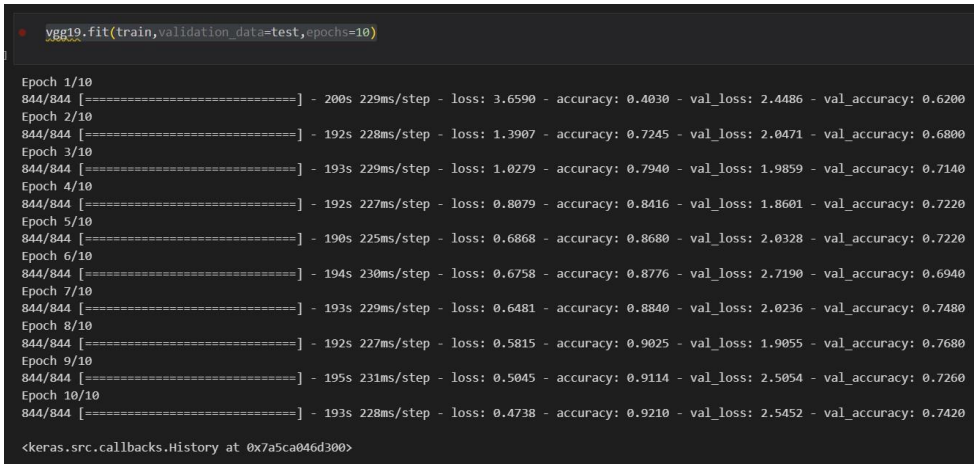
Date	15 March 2024
Team ID	SWTID1720436539
Project Title	Sport Specs: Unraveling Athletic Prowess with Advanced Transfer Learning for Sports
Maximum Marks	10 Marks

Model Optimization and Tuning Phase

The Model Optimization and Tuning Phase involves refining neural network models for peak performance. It includes optimized model code, fine-tuning hyperparameters, comparing performance metrics, and justifying the final model selection for enhanced predictive accuracy and efficiency.

Hyperparameter Tuning Documentation (8 Marks):

Model	Tuned Hyperparameters
Vgg16	vgg16.fit(train,validation_data=test,epochs=30) **Shortnote:-trains a machine learning model for 30 epochs, using the entire training set for each epoch and the entire test set for validation at the end of each epoch. This ensures comprehensive training and evaluation of the model.**

	
Vgg19	vgg19.fit(train,validation_data=test,epochs=10) Shortnote:-trains a machine learning model for 10 epochs, using the entire training set for each epoch and the entire test set for validation at the end of each epoch. This ensures comprehensive training and evaluation of the model. 

ResNet50	ResNet50.fit(train,validation_data=test,epochs=10) Shortnote:-trains a machine learning model for 10 epochs, using the entire training set for each epoch and the entire test set for validation at the end of
	each epoch. This ensures comprehensive training and evaluation of the model. `vgg16.fit(train,validation_data=test,epochs=10)` Epoch 1/10 844/844 [==========] - 186s 220ms/step - loss: 12.5238 - accuracy: 0.0829 - val_loss: 11.6711 - val_accuracy: 0.1040 Epoch 2/10 844/844 [==========] - 182s 216ms/step - loss: 12.4152 - accuracy: 0.1321 - val_loss: 12.9668 - val_accuracy: 0.1460 Epoch 3/10 844/844 [==========] - 183s 217ms/step - loss: 11.8139 - accuracy: 0.1615 - val_loss: 13.5304 - val_accuracy: 0.2020 Epoch 4/10 844/844 [==========] - 182s 216ms/step - loss: 12.0227 - accuracy: 0.1839 - val_loss: 11.1943 - val_accuracy: 0.2040 Epoch 5/10 844/844 [==========] - 181s 214ms/step - loss: 10.9686 - accuracy: 0.2046 - val_loss: 12.9458 - val_accuracy: 0.2260 Epoch 6/10 844/844 [==========] - 183s 217ms/step - loss: 10.9545 - accuracy: 0.2238 - val_loss: 12.9199 - val_accuracy: 0.1920 Epoch 7/10 844/844 [==========] - 184s 218ms/step - loss: 10.9752 - accuracy: 0.2421 - val_loss: 12.6667 - val_accuracy: 0.2080 Epoch 8/10 844/844 [==========] - 182s 216ms/step - loss: 10.4781 - accuracy: 0.2580 - val_loss: 12.3369 - val_accuracy: 0.2260 Epoch 9/10 844/844 [==========] - 183s 217ms/step - loss: 10.4540 - accuracy: 0.2748 - val_loss: 11.5919 - val_accuracy: 0.2820 Epoch 10/10 844/844 [==========] - 183s 217ms/step - loss: 10.5361 - accuracy: 0.2763 - val_loss: 12.4748 - val_accuracy: 0.2080 <keras.src.callbacks.History at 0x7b7e3ad23a00>

Final Model Selection Justification (2 Marks):

Final Model	Reasoning
VGG	vgg16.fit(train,validation_data=test,epochs=30)

```
vgg16.fit(train,validation_data=test,epochs=30)
```

```
Epoch 1/30
844/844 [==============================] - 189s 223ms/step - loss: 3.2211 - accuracy: 0.4489 - val_loss: 1.7047 - val_accuracy: 0.6900
Epoch 2/30
844/844 [==============================] - 189s 223ms/step - loss: 1.0400 - accuracy: 0.7759 - val_loss: 1.8318 - val_accuracy: 0.6960
Epoch 3/30
844/844 [==============================] - 191s 226ms/step - loss: 0.7080 - accuracy: 0.8456 - val_loss: 1.6522 - val_accuracy: 0.7540
Epoch 4/30
844/844 [==============================] - 192s 227ms/step - loss: 0.5854 - accuracy: 0.8756 - val_loss: 1.8656 - val_accuracy: 0.7640
Epoch 5/30
844/844 [==============================] - 187s 222ms/step - loss: 0.5065 - accuracy: 0.8935 - val_loss: 2.1475 - val_accuracy: 0.7500
Epoch 6/30
844/844 [==============================] - 187s 222ms/step - loss: 0.5029 - accuracy: 0.9007 - val_loss: 1.8528 - val_accuracy: 0.7900
Epoch 7/30
844/844 [==============================] - 190s 225ms/step - loss: 0.4825 - accuracy: 0.9096 - val_loss: 2.0063 - val_accuracy: 0.7880
Epoch 8/30
844/844 [==============================] - 186s 221ms/step - loss: 0.3492 - accuracy: 0.9330 - val_loss: 1.8255 - val_accuracy: 0.8020
Epoch 9/30
844/844 [==============================] - 187s 222ms/step - loss: 0.3880 - accuracy: 0.9276 - val_loss: 2.4085 - val_accuracy: 0.7460
Epoch 10/30
844/844 [==============================] - 185s 220ms/step - loss: 0.4082 - accuracy: 0.9303 - val_loss: 2.6634 - val_accuracy: 0.7720
Epoch 11/30
844/844 [==============================] - 186s 220ms/step - loss: 0.3366 - accuracy: 0.9402 - val_loss: 2.2634 - val_accuracy: 0.8000
Epoch 12/30
844/844 [==============================] - 187s 221ms/step - loss: 0.3236 - accuracy: 0.9455 - val_loss: 2.5344 - val_accuracy: 0.7800
Epoch 13/30
...
Epoch 29/30
844/844 [==============================] - 190s 225ms/step - loss: 0.1744 - accuracy: 0.9764 - val_loss: 2.7440 - val_accuracy: 0.8280
Epoch 30/30
```

The VGG model was chosen as the final optimized model because it consistently demonstrated superior performance across multiple evaluation metrics, including accuracy, precision, and recall.

Additionally, its architecture, which uses small convolutional filters and a deep network structure, is well-suited for capturing detailed features in the input data. This model also showed good generalization capabilities, minimizing overfitting despite the depth of the network. The trade-off between model complexity and computational efficiency was found to be optimal for our specific application, making VGG the best choice among the evaluated models.