


Project Development Phase
Model Performance Test

Date	10 November 2022
Team ID	PNT2022TMID34161
Project Name	Project -AI-based localization and classification of skin disease with erythema
Maximum Marks	10 Marks

Model Performance Testing:

Project team shall fill the following information in model performance testing template.

S.No.	Parameter	Values	Screenshot
1	Model Summary		 <p>The screenshot displays the model architecture for a 'sequential' model. It lists the layers and their parameters:</p> <ul style="list-style-type: none"> Layer (type): keras_layer (KerasLayer) Output Shape: multiple Param #: 21802784 dense (Dense): multiple 2049 Total params: 21,804,833 Trainable params: 2,049 Non-trainable params: 21,802,784
2.	Accuracy	Validation Accuracy-	 <p>The screenshot shows the title 'Training the Model'.</p>

```
model_name = f"benign-vs-  
malignant_{batch_size}_{optimizer}"  
tensorboard =  
tf.keras.callbacks.TensorBoard(log_dir=os.path.  
join("logs", model_name))  
# saves model checkpoint whenever we reach  
better weights  
modelcheckpoint =  
tf.keras.callbacks.ModelCheckpoint(model_na  
me + "_{val_loss:.3f}.h5", save_best_only=True,  
verbose=1)
```

```
history = m.fit(train_ds,  
validation_data=valid_ds,  
steps_per_epoch=n_training_samples  
// batch_size,
```

```
validation_steps=n_validation_samples //  
batch_size, verbose=1, epochs=100,  
callbacks=[tensorboard, modelcheckpoint])
```

Train for 31 steps, validate for 2 steps

Epoch 1/100

30/31

[=====>.] -

ETA: 9s - loss: 0.4609 - accuracy: 0.7760

Epoch 00001: val_loss improved from inf to
0.49703, saving model to benign-
vsmalignant_64_rmsprop_0.497.h5

31/31 [=====] -

282s 9s/step - loss: 0.4646 - accuracy: 0.7722 -
val_loss:

0.4970 - val_accuracy: 0.8125

<..SNIPED..>

Epoch 27/100

30/31

[=====>.] -

ETA: 0s - loss: 0.2982 - accuracy: 0.8708

Epoch 00027: val_loss improved from
0.40253 to 0.38991, saving model to
benign-

vsmalignant_64_rmsprop_0.390.h5

31/31 [=====] -

21s 691ms/step - loss: 0.3025 - accuracy:
0.8684 - val_loss: 0.3899 - val_accuracy: 0.8359
<..SNIPED..>

Epoch 41/100

30/31 [=====>.] -

ETA: 0s - loss: 0.2800 - accuracy: 0.8802

Epoch 00041: val_loss did not improve from
0.38991

31/31 [=====] -

21s 690ms/step - loss: 0.2829 - accuracy:
0.8790 - val_loss: 0.3948 - val_accuracy: 0.8281

Epoch 42/100

			<div>30/31 [=====>.] - ETA: 0s - loss: 0.2680 - accuracy: 0.8859 Epoch 00042: val_loss did not improve from 0.38991 31/31 [=====] - 21s 693ms/step - loss: 0.2722 - accuracy: 0.8831 - val_loss: 0.4572 - val_accuracy: 0.8047</div>
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Training Accuracy

```
valid_ds =
valid_ds.map(proces
s_path) train_ds =
train_ds.map(proces
s_path)
# test_ds =
test_ds for
image, label in
train_ds.take(1):
print("Image
shape:",
image.shape)
print("Label:",
label.numpy())
Image shape: (299, 299, 3)
Label: 0
# training
parameter
s
batch_size
= 64
optimizer =
"rmsprop"
def
prepare_for_training(
ds,
cache=True,
batch_size
=64,
shuffle_buffer_size=1

# `prefetch` lets the dataset fetch
batches in the background while
the model # is training.
ds =
ds.prefetch(buffer_size=tf.data
.experimental.AUTOTUNE)
return ds

valid_ds = prepare_for_training(valid_ds,
batch_size=batch_size, cache="valid-
cached-data") train_ds =
prepare_for_training(train_ds,
batch_size=batch_size, cache="train-
cached-data") batch = next(iter(valid_ds))

def
show_batch
(batch):
plt.figure(fi
gsize=(12,1
2)) for n in
range(25):
ax =
plt.subplot(
5,5,n+1)
plt.imshow(
batch[0][n])
```

3.	Confidence Score (Only Yolo Projects)	Class Detected - Confidence Score -	None
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