

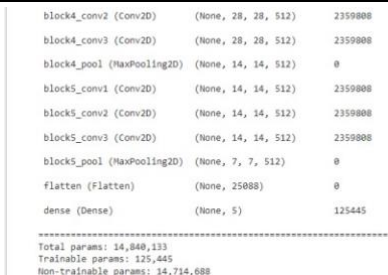
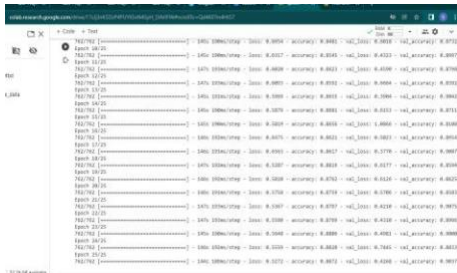
## Project Development Phase

### Model Performance Test

Date	22 November 2022
Team ID	PNT2022TMID49396
Project Name	Digital Naturalist – AI Enabled Tools For Biodiversity Researchers.
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	Total params: 14,840,133 Trainable params: 125,445 Non-trainable params: 14,714,688	 <pre> block4_conv2 (Conv2D)      (None, 28, 28, 512)      2359000 block4_conv3 (Conv2D)      (None, 28, 28, 512)      2359000 block4_pool (MaxPooling2D) (None, 14, 14, 512)      0 block5_conv1 (Conv2D)      (None, 14, 14, 512)      2359000 block5_conv2 (Conv2D)      (None, 14, 14, 512)      2359000 block5_conv3 (Conv2D)      (None, 14, 14, 512)      2359000 block5_pool (MaxPooling2D) (None, 7, 7, 512)        0 flatten (Flatten)          (None, 25088)             0 dense (Dense)              (None, 5)                 125445 ----- Total params: 14,840,133 Trainable params: 125,445 Non-trainable params: 14,714,688                     </pre>
2.	Accuracy	Training Accuracy -88.72  Validation Accuracy – 90.37	 <p>The screenshot shows a training log with columns for Epoch, Loss, and Accuracy. The training accuracy starts at approximately 88.72% and stabilizes around 90.37%, while the validation accuracy remains consistently high, fluctuating between 90.37% and 90.45%.</p>

## Model Summary

```
[18] model = Model(inputs=vgg16.input, outputs=prediction)

model.summary()
```

Model: "model"

Layer (type)	Output Shape	Param #
input_1 (InputLayer)	[None, 224, 224, 3]	0
block1_conv1 (Conv2D)	(None, 224, 224, 64)	1792
block1_conv2 (Conv2D)	(None, 224, 224, 64)	36928
block1_pool (MaxPooling2D)	(None, 112, 112, 64)	0
block2_conv1 (Conv2D)	(None, 112, 112, 128)	73856
block2_conv2 (Conv2D)	(None, 112, 112, 128)	147584
block2_pool (MaxPooling2D)	(None, 56, 56, 128)	0
block3_conv1 (Conv2D)	(None, 56, 56, 256)	295168
block3_conv2 (Conv2D)	(None, 56, 56, 256)	590080
block3_conv3 (Conv2D)	(None, 56, 56, 256)	590080
block3_pool (MaxPooling2D)	(None, 28, 28, 256)	0
block4_conv1 (Conv2D)	(None, 28, 28, 512)	1180160
block4_conv2 (Conv2D)	(None, 28, 28, 512)	2359808

```
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block2_pool (MaxPooling2D) (None, 56, 56, 128) 0
block3_conv1 (Conv2D) (None, 56, 56, 256) 295168
block3_conv2 (Conv2D) (None, 56, 56, 256) 590080
block3_conv3 (Conv2D) (None, 56, 56, 256) 590080
block3_pool (MaxPooling2D) (None, 28, 28, 256) 0
block4_conv1 (Conv2D) (None, 28, 28, 512) 1180160
block4_conv2 (Conv2D) (None, 28, 28, 512) 2359808
block4_conv3 (Conv2D) (None, 28, 28, 512) 2359808
block4_pool (MaxPooling2D) (None, 14, 14, 512) 0
block5_conv1 (Conv2D) (None, 14, 14, 512) 2359808
block5_conv2 (Conv2D) (None, 14, 14, 512) 2359808
block5_conv3 (Conv2D) (None, 14, 14, 512) 2359808
block5_pool (MaxPooling2D) (None, 7, 7, 512) 0
flatten (Flatten) (None, 25088) 0
dense (Dense) (None, 5) 125445

Total params: 14,840,133
Trainable params: 125,445
Non-trainable params: 14,714,688
```

## Accuracy

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RAM  
Disk

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```
[ ] model.compile(  
    loss='categorical_crossentropy',  
    optimizer='adam',  
    metrics=['accuracy']  
)  
  
▶ r = model.fit(xtrain,steps_per_epoch =762 ,epochs = 25, validation_data = xtest,validation_steps =762)
```

Epoch 1/25  
762/762 [=====] - 156s 193ms/step - loss: 0.9431 - accuracy: 0.7389 - val\_loss: 0.9223 - val\_accuracy: 0.7678  
Epoch 2/25  
762/762 [=====] - 148s 195ms/step - loss: 0.7747 - accuracy: 0.7904 - val\_loss: 0.5969 - val\_accuracy: 0.8266  
Epoch 3/25  
762/762 [=====] - 148s 194ms/step - loss: 0.7529 - accuracy: 0.8118 - val\_loss: 0.5721 - val\_accuracy: 0.8503  
Epoch 4/25  
762/762 [=====] - 145s 191ms/step - loss: 0.7151 - accuracy: 0.8199 - val\_loss: 0.6260 - val\_accuracy: 0.8404  
Epoch 5/25  
762/762 [=====] - 146s 191ms/step - loss: 0.7190 - accuracy: 0.8310 - val\_loss: 0.5615 - val\_accuracy: 0.8543  
Epoch 6/25  
762/762 [=====] - 145s 191ms/step - loss: 0.7020 - accuracy: 0.8381 - val\_loss: 0.4990 - val\_accuracy: 0.8727  
Epoch 7/25  
762/762 [=====] - 146s 191ms/step - loss: 0.6641 - accuracy: 0.8442 - val\_loss: 0.5730 - val\_accuracy: 0.8499  
Epoch 8/25  
762/762 [=====] - 145s 191ms/step - loss: 0.6901 - accuracy: 0.8406 - val\_loss: 0.5268 - val\_accuracy: 0.8768  
Epoch 9/25  
762/762 [=====] - 145s 190ms/step - loss: 0.6654 - accuracy: 0.8481 - val\_loss: 0.6018 - val\_accuracy: 0.8732  
Epoch 10/25  
762/762 [=====] - 145s 190ms/step - loss: 0.6317 - accuracy: 0.8545 - val\_loss: 0.4323 - val\_accuracy: 0.8997  
Epoch 11/25  
762/762 [=====] - 147s 193ms/step - loss: 0.6020 - accuracy: 0.8623 - val\_loss: 0.4590 - val\_accuracy: 0.8798  
Epoch 12/25  
762/762 [=====] - 147s 193ms/step - loss: 0.6093 - accuracy: 0.8592 - val\_loss: 0.6664 - val\_accuracy: 0.8392

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```
[ ] Epoch 10/25  
762/762 [=====] - 145s 190ms/step - loss: 0.6317 - accuracy: 0.8545 - val_loss: 0.4323 - val_accuracy: 0.8997  
Epoch 11/25  
762/762 [=====] - 147s 193ms/step - loss: 0.6020 - accuracy: 0.8623 - val_loss: 0.4590 - val_accuracy: 0.8798  
Epoch 12/25  
762/762 [=====] - 147s 193ms/step - loss: 0.6093 - accuracy: 0.8592 - val_loss: 0.6664 - val_accuracy: 0.8392  
Epoch 13/25  
762/762 [=====] - 145s 191ms/step - loss: 0.5969 - accuracy: 0.8655 - val_loss: 0.3904 - val_accuracy: 0.9042  
Epoch 14/25  
762/762 [=====] - 145s 190ms/step - loss: 0.5879 - accuracy: 0.8681 - val_loss: 0.6153 - val_accuracy: 0.8711  
Epoch 15/25  
762/762 [=====] - 145s 190ms/step - loss: 0.5819 - accuracy: 0.8656 - val_loss: 1.0866 - val_accuracy: 0.8108  
Epoch 16/25  
762/762 [=====] - 146s 192ms/step - loss: 0.6475 - accuracy: 0.8621 - val_loss: 0.5023 - val_accuracy: 0.8954  
Epoch 17/25  
762/762 [=====] - 146s 191ms/step - loss: 0.6563 - accuracy: 0.8617 - val_loss: 0.3770 - val_accuracy: 0.9087  
Epoch 18/25  
762/762 [=====] - 147s 192ms/step - loss: 0.5287 - accuracy: 0.8819 - val_loss: 0.6177 - val_accuracy: 0.8594  
Epoch 19/25  
762/762 [=====] - 146s 192ms/step - loss: 0.5810 - accuracy: 0.8762 - val_loss: 0.6126 - val_accuracy: 0.8625  
Epoch 20/25  
762/762 [=====] - 146s 191ms/step - loss: 0.5758 - accuracy: 0.8759 - val_loss: 0.5706 - val_accuracy: 0.8583  
Epoch 21/25  
762/762 [=====] - 147s 193ms/step - loss: 0.5367 - accuracy: 0.8797 - val_loss: 0.4210 - val_accuracy: 0.9075  
Epoch 22/25  
762/762 [=====] - 147s 193ms/step - loss: 0.5590 - accuracy: 0.8769 - val_loss: 0.4310 - val_accuracy: 0.8966  
Epoch 23/25  
762/762 [=====] - 145s 190ms/step - loss: 0.5648 - accuracy: 0.8806 - val_loss: 0.4981 - val_accuracy: 0.9000  
Epoch 24/25  
762/762 [=====] - 146s 192ms/step - loss: 0.5559 - accuracy: 0.8820 - val_loss: 0.7445 - val_accuracy: 0.8833  
Epoch 25/25  
762/762 [=====] - 144s 189ms/step - loss: 0.5272 - accuracy: 0.8872 - val_loss: 0.4268 - val_accuracy: 0.9037
```