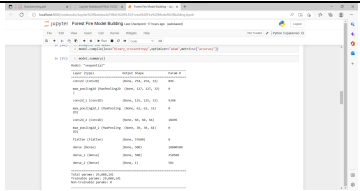
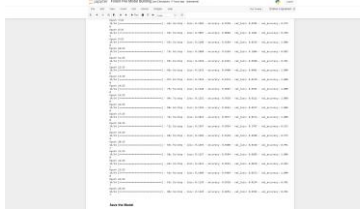


## Project Development Phase Model Performance Test

Date	19 November 2022
Team ID	PNT2022TMID35796
Project Name	Project - Emerging Methods for Early Detection Of Forest Fires
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	Attached below	
2.	Accuracy	Training Accuracy - 0.9654  Validation Accuracy – 0.991	

### Model Summary:

Screenshot:

```
Brainstorming.pdf x Jupyter Notebook/FINAL FILES/ x Forest Fire Model Building - Jup: x +
localhost:8888/notebooks/Jupyter%20Notebook/FINAL%20FILES/Forest%20Fire%20Model%20Building.ipynb
jupyter Forest Fire Model Building Last Checkpoint: 17 hours ago (autosaved) Logout
File Edit View Insert Cell Kernel Widgets Help Not Trusted Python 3 (pykernel)
In [35]: 1 model.compile(loss="binary_crossentropy",optimizer="adam",metrics=["accuracy"])
2 model.summary()
Model: "sequential"
Layer (type) Output Shape Param #
=====
conv2d (Conv2D) (None, 254, 254, 32) 896
max_pooling2d (MaxPooling2D) (None, 127, 127, 32) 0
conv2d_1 (Conv2D) (None, 125, 125, 32) 9248
max_pooling2d_1 (MaxPooling 2D) (None, 62, 62, 32) 0
conv2d_2 (Conv2D) (None, 60, 60, 64) 18496
max_pooling2d_2 (MaxPooling 2D) (None, 30, 30, 64) 0
flatten (Flatten) (None, 57600) 0
dense (Dense) (None, 500) 28800500
dense_1 (Dense) (None, 500) 250500
dense_2 (Dense) (None, 1) 501
=====
Total params: 29,080,141
Trainable params: 29,080,141
Non-trainable params: 0
```

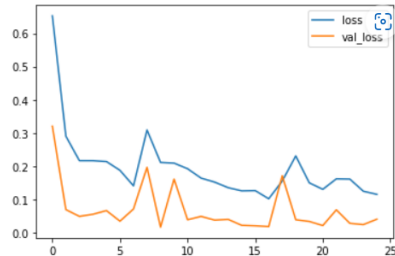
## Values:

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 254, 254, 32)	896
max_pooling2d (MaxPooling2D)	(None, 127, 127, 32)	0
conv2d_1 (Conv2D)	(None, 125, 125, 32)	9248
max_pooling2d_1 (MaxPooling 2D)	(None, 62, 62, 32)	0
conv2d_2 (Conv2D)	(None, 60, 60, 64)	18496
max_pooling2d_2 (MaxPooling 2D)	(None, 30, 30, 64)	0
flatten (Flatten)	(None, 57600)	0
dense (Dense)	(None, 500)	28800500
dense_1 (Dense)	(None, 500)	250500
dense_2 (Dense)	(None, 1)	501
=====		
Total params: 29,080,141		
Trainable params: 29,080,141		
Non-trainable params: 0		

## Accuracy:

```
In [34]: 1 import matplotlib.pyplot as plt
          2 plt.plot(h.history['loss'])
          3 plt.plot(h.history['val_loss'])
          4 plt.legend(['loss', 'val_loss'])
```

Out[34]: <matplotlib.legend.Legend at 0x2aa1e531b80>



Epoch 1/25

14/14 [=====] - 76s 5s/step - loss: 0.6528 - accuracy: 0.6697 - val\_loss: 0.3201 - val\_accuracy: 0.9167

Epoch 2/25

14/14 [=====] - 64s 5s/step - loss: 0.2900 - accuracy: 0.8730 - val\_loss: 0.0686 - val\_accuracy: 0.9750

Epoch 3/25

14/14 [=====] - 62s 4s/step - loss: 0.2161 - accuracy: 0.9145 - val\_loss: 0.0482 - val\_accuracy: 1.0000

Epoch 4/25

14/14 [=====] - 63s 4s/step - loss: 0.2161 - accuracy: 0.9053 - val\_loss: 0.0545 - val\_accuracy: 0.9833

Epoch 5/25

14/14 [=====] - 67s 5s/step - loss: 0.2135 - accuracy: 0.9145 - val\_loss: 0.0656 - val\_accuracy: 0.9917

Epoch 6/25

14/14 [=====] - 66s 5s/step - loss: 0.1870 - accuracy: 0.9169 - val\_loss: 0.0337 - val\_accuracy: 0.9917

Epoch 7/25

14/14 [=====] - 64s 4s/step - loss: 0.1402 - accuracy: 0.9330 - val\_loss: 0.0705 - val\_accuracy: 0.9750

Epoch 8/25

14/14 [=====] - 65s 5s/step - loss: 0.3087 - accuracy: 0.8868 - val\_loss: 0.1961 - val\_accuracy: 0.9583

Epoch 9/25

14/14 [=====] - 65s 5s/step - loss: 0.2108 - accuracy: 0.9284 - val\_loss: 0.0156 - val\_accuracy: 1.0000

Epoch 10/25

14/14 [=====] - 79s 6s/step - loss: 0.2088 - accuracy: 0.9169 - val\_loss: 0.1604 - val\_accuracy: 0.8833

Epoch 11/25

14/14 [=====] - 85s 6s/step - loss: 0.1918 - accuracy: 0.9192 - val\_loss: 0.0380 - val\_accuracy: 0.9917

Epoch 12/25

14/14 [=====] - 86s 6s/step - loss: 0.1635 - accuracy: 0.9400 - val\_loss: 0.0481 - val\_accuracy: 1.0000  
Epoch 13/25  
14/14 [=====] - 87s 6s/step - loss: 0.1515 - accuracy: 0.9376 - val\_loss: 0.0370 - val\_accuracy: 1.0000  
Epoch 14/25  
14/14 [=====] - 75s 5s/step - loss: 0.1348 - accuracy: 0.9607 - val\_loss: 0.0391 - val\_accuracy: 1.0000  
Epoch 15/25  
14/14 [=====] - 94s 7s/step - loss: 0.1252 - accuracy: 0.9538 - val\_loss: 0.0212 - val\_accuracy: 1.0000  
Epoch 16/25  
14/14 [=====] - 80s 6s/step - loss: 0.1259 - accuracy: 0.9561 - val\_loss: 0.0197 - val\_accuracy: 1.0000  
Epoch 17/25  
14/14 [=====] - 74s 5s/step - loss: 0.1013 - accuracy: 0.9677 - val\_loss: 0.0174 - val\_accuracy: 1.0000  
Epoch 18/25  
14/14 [=====] - 72s 5s/step - loss: 0.1547 - accuracy: 0.9654 - val\_loss: 0.1707 - val\_accuracy: 0.9333  
Epoch 19/25  
14/14 [=====] - 68s 5s/step - loss: 0.2305 - accuracy: 0.9238 - val\_loss: 0.0380 - val\_accuracy: 0.9750  
Epoch 20/25  
14/14 [=====] - 66s 5s/step - loss: 0.1493 - accuracy: 0.9400 - val\_loss: 0.0328 - val\_accuracy: 0.9917  
Epoch 21/25  
14/14 [=====] - 68s 5s/step - loss: 0.1297 - accuracy: 0.9584 - val\_loss: 0.0205 - val\_accuracy: 1.0000  
Epoch 22/25  
14/14 [=====] - 64s 5s/step - loss: 0.1613 - accuracy: 0.9515 - val\_loss: 0.0678 - val\_accuracy: 0.9833  
Epoch 23/25  
14/14 [=====] - 61s 4s/step - loss: 0.1606 - accuracy: 0.9469 - val\_loss: 0.0272 - val\_accuracy: 1.0000  
Epoch 24/25  
14/14 [=====] - 63s 5s/step - loss: 0.1239 - accuracy: 0.9538 - val\_loss: 0.0234 - val\_accuracy: 0.9917  
Epoch 25/25  
14/14 [=====] - 88s 6s/step - loss: 0.1145 - accuracy: 0.9654 - val\_loss: 0.0396 - val\_accuracy: 0.9917