

Project Development Phase Model Performance Test

Date	10 November 2022
Team ID	Team-593093
Project Name	Project – Eye Disease Prediction Using Deep Learning Project
Maximum Marks	10 Marks

S.N o.	Parameter	Values	Screenshot																																																																														
1.	Model Summary	Total params: 20,124,740 Trainable params: 100,356 Non-trainable params: 20,024,384	<div>In [15]:<pre>model.summary()</pre></div> <div>Model: "model"</div> <table><thead><tr><th>Layer (type)</th><th>Output Shape</th><th>Param #</th></tr></thead><tbody><tr><td colspan="3">=====</td></tr><tr><td>input_1 (InputLayer)</td><td>[(None, 224, 224, 3)]</td><td>0</td></tr><tr><td>block1_conv1 (Conv2D)</td><td>(None, 224, 224, 64)</td><td>1792</td></tr><tr><td>block1_conv2 (Conv2D)</td><td>(None, 224, 224, 64)</td><td>36928</td></tr><tr><td>block1_pool (MaxPooling2D)</td><td>(None, 112, 112, 64)</td><td>0</td></tr><tr><td>block2_conv1 (Conv2D)</td><td>(None, 112, 112, 128)</td><td>73856</td></tr><tr><td>block2_conv2 (Conv2D)</td><td>(None, 112, 112, 128)</td><td>147584</td></tr><tr><td>block2_pool (MaxPooling2D)</td><td>(None, 56, 56, 128)</td><td>0</td></tr><tr><td>block3_conv1 (Conv2D)</td><td>(None, 56, 56, 256)</td><td>295168</td></tr><tr><td>block3_conv2 (Conv2D)</td><td>(None, 56, 56, 256)</td><td>590080</td></tr><tr><td>block3_conv3 (Conv2D)</td><td>(None, 56, 56, 256)</td><td>590080</td></tr><tr><td>block3_conv4 (Conv2D)</td><td>(None, 56, 56, 256)</td><td>590080</td></tr><tr><td>block3_pool (MaxPooling2D)</td><td>(None, 28, 28, 256)</td><td>0</td></tr><tr><td>block4_conv1 (Conv2D)</td><td>(None, 28, 28, 512)</td><td>1180160</td></tr><tr><td>block4_conv2 (Conv2D)</td><td>(None, 28, 28, 512)</td><td>2359008</td></tr><tr><td>block4_conv3 (Conv2D)</td><td>(None, 28, 28, 512)</td><td>2359008</td></tr><tr><td>block4_conv4 (Conv2D)</td><td>(None, 28, 28, 512)</td><td>2359008</td></tr><tr><td>block4_pool (MaxPooling2D)</td><td>(None, 14, 14, 512)</td><td>0</td></tr><tr><td>block5_conv1 (Conv2D)</td><td>(None, 14, 14, 512)</td><td>2359008</td></tr><tr><td>block5_conv2 (Conv2D)</td><td>(None, 14, 14, 512)</td><td>2359008</td></tr><tr><td>block5_conv3 (Conv2D)</td><td>(None, 14, 14, 512)</td><td>2359008</td></tr><tr><td>block5_conv4 (Conv2D)</td><td>(None, 14, 14, 512)</td><td>2359008</td></tr><tr><td>block5_pool (MaxPooling2D)</td><td>(None, 7, 7, 512)</td><td>0</td></tr><tr><td>flatten (Flatten)</td><td>(None, 25088)</td><td>0</td></tr><tr><td>dense (Dense)</td><td>(None, 4)</td><td>100356</td></tr></tbody></table> <div>=====</div> <div>Total params: 20,124,740 Trainable params: 100,356 Non-trainable params: 20,024,384</div>	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_conv4 (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)	2359008	block4_conv3 (Conv2D)	(None, 28, 28, 512)	2359008	block4_conv4 (Conv2D)	(None, 28, 28, 512)	2359008	block4_pool (MaxPooling2D)	(None, 14, 14, 512)	0	block5_conv1 (Conv2D)	(None, 14, 14, 512)	2359008	block5_conv2 (Conv2D)	(None, 14, 14, 512)	2359008	block5_conv3 (Conv2D)	(None, 14, 14, 512)	2359008	block5_conv4 (Conv2D)	(None, 14, 14, 512)	2359008	block5_pool (MaxPooling2D)	(None, 7, 7, 512)	0	flatten (Flatten)	(None, 25088)	0	dense (Dense)	(None, 4)	100356
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2.	Accuracy	<div>Training Accuracy - 0.8983</div> <div>Validation Accuracy -0.8130</div>	<div>history=model.fit(trainset,validation_data=testset,epochs=50,steps_per_epoch=len(trainset),validation_steps=len(testset))</div> <div>Epoch 1/50 53/53 [=====] - 91s 1s/step - loss: 1.0707 - accuracy: 0.6189 - val_loss: 0.6095 - val_accuracy: 0.7645 Epoch 2/50 53/53 [=====] - 68s 1s/step - loss: 0.5714 - accuracy: 0.7767 - val_loss: 0.6958 - val_accuracy: 0.6994 Epoch 3/50 53/53 [=====] - 68s 1s/step - loss: 0.5256 - accuracy: 0.7933 - val_loss: 0.5836 - val_accuracy: 0.8213 Epoch 4/50 53/53 [=====] - 67s 1s/step - loss: 0.4917 - accuracy: 0.7951 - val_loss: 0.6498 - val_accuracy: 0.7408 Epoch 5/50 53/53 [=====] - 67s 1s/step - loss: 0.4712 - accuracy: 0.8149 - val_loss: 0.5934 - val_accuracy: 0.7645 Epoch 6/50 53/53 [=====] - 67s 1s/step - loss: 0.4310 - accuracy: 0.8301 - val_loss: 0.5497 - val_accuracy: 0.7751 Epoch 7/50 53/53 [=====] - 67s 1s/step - loss: 0.4203 - accuracy: 0.8351 - val_loss: 0.4655 - val_accuracy: 0.8189 Epoch 8/50 53/53 [=====] - 66s 1s/step - loss: 0.5172 - accuracy: 0.7963 - val_loss: 0.5633 - val_accuracy: 0.7728 Epoch 9/50 53/53 [=====] - 66s 1s/step - loss: 0.4127 - accuracy: 0.8324 - val_loss: 0.4334 - val_accuracy: 0.8402 Epoch 10/50 53/53 [=====] - 67s 1s/step - loss: 0.4135 - accuracy: 0.8363 - val_loss: 0.4537 - val_accuracy: 0.8462 Epoch 11/50 53/53 [=====] - 67s 1s/step - loss: 0.4090 - accuracy: 0.8405 - val_loss: 0.4435 - val_accuracy: 0.8268 Epoch 12/50 53/53 [=====] - 67s 1s/step - loss: 0.3711 - accuracy: 0.8493 - val_loss: 0.5505 - val_accuracy: 0.7929 Epoch 13/50 53/53 [=====] - 67s 1s/step - loss: 0.3483 - accuracy: 0.8591 - val_loss: 0.4632 - val_accuracy: 0.8343 Epoch 14/50 53/53 [=====] - 67s 1s/step - loss: 0.3573 - accuracy: 0.8559 - val_loss: 0.4453 - val_accuracy: 0.8367 Epoch 15/50 53/53 [=====] - 67s 1s/step - loss: 0.3588 - accuracy: 0.8514 - val_loss: 0.4199 - val_accuracy: 0.8462 Epoch 16/50 53/53 [=====] - 67s 1s/step - loss: 0.3258 - accuracy: 0.8692 - val_loss: 0.6269 - val_accuracy: 0.7574 Epoch 17/50 53/53 [=====] - 67s 1s/step - loss: 0.3678 - accuracy: 0.8482 - val_loss: 0.4263 - val_accuracy: 0.8414 Epoch 18/50 53/53 [=====] - 67s 1s/step - loss: 0.3512 - accuracy: 0.8633 - val_loss: 0.3987 - val_accuracy: 0.8568 Epoch 19/50 53/53 [=====] - 65s 1s/step - loss: 0.2917 - accuracy: 0.8823 - val_loss: 0.5970 - val_accuracy: 0.7680 Epoch 20/50 53/53 [=====] - 65s 1s/step - loss: 0.3805 - accuracy: 0.8514 - val_loss: 0.3609 - val_accuracy: 0.8757 Epoch 21/50 53/53 [=====] - 66s 1s/step - loss: 0.3198 - accuracy: 0.8760 - val_loss: 0.3978 - val_accuracy: 0.8604 Epoch 22/50 53/53 [=====] - 67s 1s/step - loss: 0.2961 - accuracy: 0.8805 - val_loss: 0.5506 - val_accuracy: 0.8142 Epoch 23/50 53/53 [=====] - 66s 1s/step - loss: 0.2692 - accuracy: 0.8921 - val_loss: 0.5955 - val_accuracy: 0.7882 Epoch 24/50 53/53 [=====] - 65s 1s/step - loss: 0.3054 - accuracy: 0.8763 - val_loss: 0.3774 - val_accuracy: 0.8734 Epoch 25/50 53/53 [=====] - 67s 1s/step - loss: 0.2613 - accuracy: 0.8953 - val_loss: 0.3771 - val_accuracy: 0.8675 Epoch 26/50 53/53 [=====] - 66s 1s/step - loss: 0.2670 - accuracy: 0.8918 - val_loss: 0.5199 - val_accuracy: 0.8047 Epoch 27/50 53/53 [=====] - 65s 1s/step - loss: 0.3260 - accuracy: 0.8698 - val_loss: 0.4162 - val_accuracy: 0.8604 Epoch 28/50 53/53 [=====] - 66s 1s/step - loss: 0.2852 - accuracy: 0.8823 - val_loss: 0.4423 - val_accuracy: 0.8521 Epoch 29/50 53/53 [=====] - 66s 1s/step - loss: 0.2588 - accuracy: 0.8995 - val_loss: 0.4975 - val_accuracy: 0.8414 Epoch 30/50 53/53 [=====] - 66s 1s/step - loss: 0.2876 - accuracy: 0.8873 - val_loss: 0.3928 - val_accuracy: 0.8769 Epoch 31/50 53/53 [=====] - 66s 1s/step - loss: 0.2659 - accuracy: 0.8977 - val_loss: 0.4968 - val_accuracy: 0.8260 Epoch 32/50 53/53 [=====] - 65s 1s/step - loss: 0.2979 - accuracy: 0.8835 - val_loss: 0.5791 - val_accuracy: 0.8107 Epoch 33/50 53/53 [=====] - 66s 1s/step - loss: 0.2894 - accuracy: 0.8891 - val_loss: 0.4767 - val_accuracy: 0.8497 Epoch 34/50 53/53 [=====] - 66s 1s/step - loss: 0.2589 - accuracy: 0.9007 - val_loss: 0.3831 - val_accuracy: 0.8769 Epoch 35/50 53/53 [=====] - 66s 1s/step - loss: 0.2543 - accuracy: 0.9001 - val_loss: 0.4896 - val_accuracy: 0.8296 Epoch 36/50 53/53 [=====] - 66s 1s/step - loss: 0.2654 - accuracy: 0.8959 - val_loss: 0.3568 - val_accuracy: 0.8698 Epoch 37/50 53/53 [=====] - 66s 1s/step - loss: 0.2505 - accuracy: 0.8983 - val_loss: 0.5416 - val_accuracy: 0.8150</div>
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