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Paper details :

Link : <https://iopscience.iop.org/article/10.1088/1742-6596/1447/1/012021/pdf>

Paper name : Multi-class Image Classification Using Deep Learning Algorithm

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year of publication : 2021

The dataset used : For the task of training and validation for classification considered four classes from the PASCAL VOC 2007 data-set. Many computer researchers and scientific papers use PASCAL 2007 as data-set for training and validation. PASCAL 2007 data-set is small-scale with 20 object classes. The PASCAL 2007 data-set has text files label every image as the image name and image class. All images are JPG format

Results : The following experimental results are obtained by deep learning framework CAFFE. The model has been trained and accuracy from the three training cycles is the accuracy of (87 %, 93 %, and 92.7 %) with an average accuracy value of 90.9 %

General Information on the selected dataset:

Link of dataset: <https://www.kaggle.com/datasets/ayushv322/animal-classification>

total number of samples : 4000 images

number of classes and their labels: four classes [Zebra _ elephant _ Rhino _ Buffalo]

Implementation details:

- A) A CSV file has been created to contain an index for all images from classes [Zebra_elephant_bufflao_Rhino]
 - B) With python scripting code three CSV files have been generated to index the chosen images and partition images randomly into four equal subset [G1 – G2 – G3 - test] contain the images name, image classes
 - C) Each subset contain equal number of images (250 image for each class) and the total number is 1000.
 - D) The cross-validation technique has been used in training and validation. In the first training and validation cycle, the subsets G1 and G2 have been used for training the model and the subset G3 has been used for the validation
 - E) In the second training and validation cycle subsets G1 and G3 have been used for training and the subset G2 are used for the validation process. In the third training and validation cycle, subsets G2 and G3 have been used for the training process and subset G1 for the validation process.
- After each cycle we test the algorithm with (test data) and see the accuracy and try to develop it.

Architecture of model

```
model = Sequential()
model.add(Conv2D(32, (3,3), input_shape= (100,100,3)))
model.add(Activation("relu"))
model.add(MaxPooling2D())

model.add(Conv2D(32, (3,3),))
model.add(Activation("relu"))
model.add(MaxPooling2D())

model.add(Conv2D(64, (3,3),))
model.add(Activation("relu"))
model.add(MaxPooling2D())
model.add(Flatten())
model.add(Dense(1024))
model.add(Activation("relu"))
model.add(Dropout(0.5))
model.add(Dense(4))#output
model.add(Activation("softmax"))
```

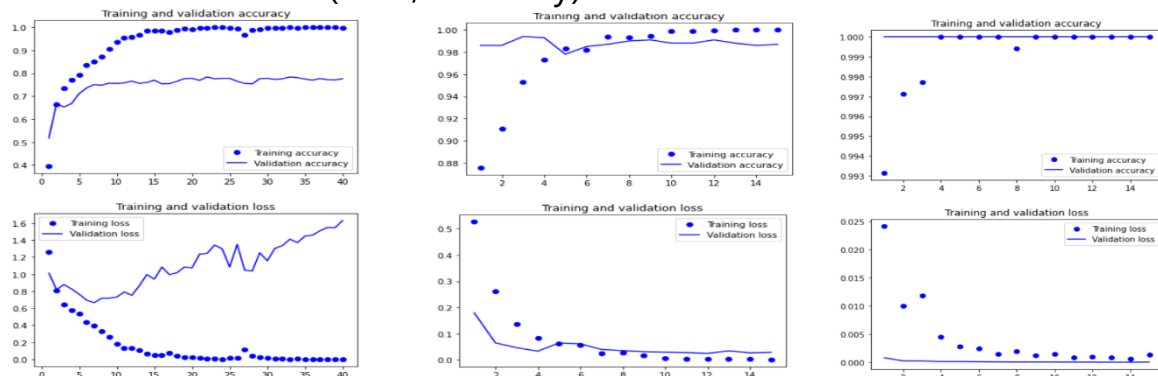
Summary

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 98, 98, 32)	896
activation (Activation)	(None, 98, 98, 32)	0
max_pooling2d (MaxPooling2D)	(None, 49, 49, 32)	0
conv2d_1 (Conv2D)	(None, 47, 47, 32)	9248
activation_1 (Activation)	(None, 47, 47, 32)	0
max_pooling2d_1 (MaxPooling2D)	(None, 23, 23, 32)	0
conv2d_2 (Conv2D)	(None, 21, 21, 64)	18496
activation_2 (Activation)	(None, 21, 21, 64)	0
max_pooling2d_2 (MaxPooling2D)	(None, 10, 10, 64)	0
flatten (Flatten)	(None, 6400)	0
dense (Dense)	(None, 1024)	6554624
activation_3 (Activation)	(None, 1024)	0
dropout (Dropout)	(None, 1024)	0
dense_1 (Dense)	(None, 4)	4100
activation_4 (Activation)	(None, 4)	0
Total params: 6,587,364		
Trainable params: 6,587,364		
Non-trainable params: 0		

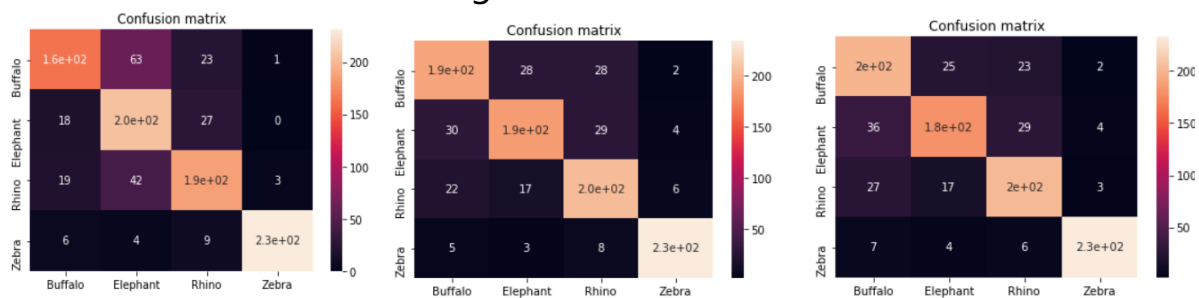
Results and visualizations

Accuracy on test images after cycle 1 =79% ,cycle 2 =81% ,cycle 3 = 81%
Accuracy average is = 80%

Train and validation (loss , accuracy) :



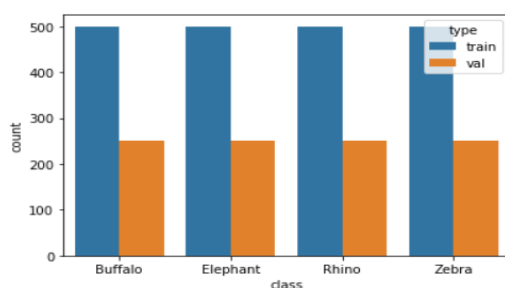
Confusion matrix on test images:



Reports on test images:

	precision	recall	f1-score	support		precision	recall	f1-score	support		precision	recall	f1-score	support
Buffalo	0.79	0.65	0.71	250	Buffalo	0.77	0.77	0.77	250	Buffalo	0.74	0.80	0.77	250
Elephant	0.65	0.82	0.73	250	Elephant	0.80	0.75	0.77	250	Elephant	0.80	0.72	0.76	250
Rhino	0.76	0.74	0.75	250	Rhino	0.76	0.82	0.79	250	Rhino	0.78	0.81	0.79	250
Zebra	0.98	0.92	0.95	250	Zebra	0.95	0.94	0.94	250	Zebra	0.96	0.93	0.95	250
accuracy			0.79	1000	accuracy			0.82	1000	accuracy			0.82	1000
macro avg	0.80	0.79	0.79	1000	macro avg	0.82	0.82	0.82	1000	macro avg	0.82	0.82	0.82	1000
weighted avg	0.80	0.79	0.79	1000	weighted avg	0.82	0.82	0.82	1000	weighted avg	0.82	0.82	0.82	1000

Images of train and validation:



images of test :

