# 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(Dropout(0.5))
model.add(Dropout(0.5))
model.add(Dense(4))#output
model.add(Activation("softmax"))
```

# <u>Summary</u>

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 (MaxPooling2	(None,	23, 23, 32)	0			
conv2d_2 (Conv2D)	(None,	21, 21, 64)	18496			
activation_2 (Activation)	(None,	21, 21, 64)	0			
max_pooling2d_2 (MaxPooling2	(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						

### **Dataset link:**

https://www.kaggle.com/datasets/ayushv322/animal-classification

### **Dataset details:**

There are four classes ( Zebra \_ elephant \_ Rhino \_ Buffalo )

Each class has one thousand of images in it

And the total number is four thousands image

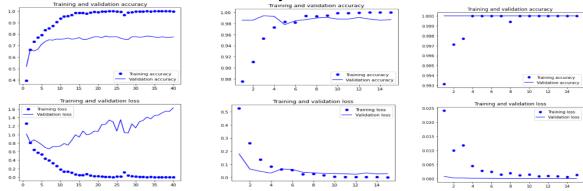
## Implementation details

- 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.
- F) After each cycle we test the algorithm with (test data) and see the accuracy and try to develop it.

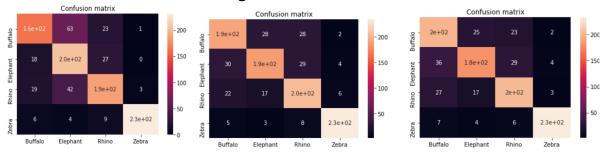
## Results and visualizations

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

#### Train and validation (loss, accuracy):



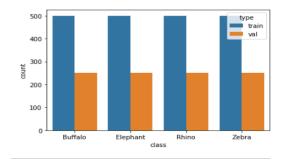
#### Confusion matrix on test images:



#### 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 Elephant	0.77 0.80	0.77 0.75	0.77 0.77	250 250	Buffalo Elephant	0.74 0.80	0.80 0.72	0.77 0.76	250 250
Elephant Rhino Zebra	0.65 0.76 0.98	0.82 0.74 0.92	0.73 0.75 0.95	250 250 250	Rhino Zebra	0.76 0.95	0.82 0.94	0.79 0.94	250 250	Rhino Zebra	0.78 0.96	0.81 0.93	0.79 0.95	250 250 250
accuracy macro avg weighted avg	0.80 0.80	0.79 0.79	0.79 0.79 0.79	1000 1000 1000	accuracy macro avg weighted avg	0.82 0.82	0.82 0.82	0.82 0.82 0.82	1000 1000 1000	accuracy macro avg weighted avg	0.82 0.82	0.82 0.82	0.82 0.82 0.82	1000 1000 1000

#### Images of train and validation:



#### images of test:

