

# Animal Classification Using CNN with VGG-16 Architecture

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**Abstract:** Proficient and solid observing of wild animals right at their habitat is fundamental. This venture fosters a calculation to distinguish the animals in untamed life. Since there are enormous number of various animals physically recognizing them can be a troublesome task. This calculation orders animals in light of their pictures so we can screen them all the more effectively. Animal recognition and grouping can assist with forestalling creature vehicle mishaps, follow animal and forestall robbery. This can be accomplished by applying powerful profound learning algorithms. It shows that the proposed procedure positively affect arrangement exactness.

**Keywords:** Animal Classes, CNN, Decision Making, VGG-16.

## I. INTRODUCTION

As of late, because of the hazardous development of advanced substance, programmed grouping of pictures has become one of the most basic difficulties in visual data ordering and recovery frameworks. PC vision is an interdisciplinary and subfield of man-made consciousness that intends to give comparable capacity of human to PC for understanding data from the pictures. A few exploration endeavors were made to defeat these issues, yet these techniques think about the low-level elements of picture natives. Zeroing in on low-level picture elements won't assist with handling the pictures. Picture arrangement is a major issue in PC vision for the many years. If there should be an occurrence of people the picture comprehension, and grouping is done exceptionally simple assignment, yet in the event of PCs it is pricey task. By and large, each picture is made out of set of pixels and every pixel is addressed with various qualities. Hence to store a picture the PC should require more spaces for store information. To order pictures, it should perform larger number of computations. For this it requires frameworks with higher arrangement and really registering power. Progressively to take choices basing on the information is preposterous in light of the fact that it requires some investment for playing out these numerous calculations to give result.

Picture order is a major issue in PC vision for the many years. In the event of people the picture comprehension, and characterization is done extremely simple task, yet if there should arise an occurrence of PCs it is over the top expensive undertaking. By and large, each picture is made out of set of pixels and every pixel is addressed with various qualities. Hence to store a picture the PC should require more spaces for store information. To arrange pictures, it should perform bigger number of estimations. For this it requires frameworks with higher arrangement and really processing power. Progressively to take choices basing on the information is preposterous in light of the fact that it requires some investment for playing out these numerous calculations to give result.

Presently, Deep learning algorithms are giving successful results in computer vision. The Convolutional Neural Network, an AI calculation is being utilized for the picture order. In [1], involves profound learning calculation for group the nature of wood board by utilizing separated surface data from the wood pictures. He additionally made the examination with AI design. CNN is a kind of feed-forward counterfeit neural organization that has been effectively applied to investigations visual pictures. It is motivated by the organic cycles and the neurons are associated as in animal visual cortex. In [2], has talked about programmed acknowledgment dairy cattle pictures utilizing CNN which assists with extricating the important trademark from the cows pictures and Support Vector Machine (SVM) methods is utilized for characterization of those pictures.

The freedom of human mediation in learning channels is great benefit of CNN. CNN is managed profound learning approach which requires huge marked information for training on the network. In the wake of training the model will get familiar with the loads and the exactness of the classifier is gotten to the next level. Then, at that point, a picture is given as information and the classifier shows to which class it has a place with. Google's self-driving vehicle is an original profound gaining project from Google organization is a model for the new advancement in the field of man-made reasoning. For this project the picture information is given as contribution from this present reality and the choices are made basing on the data acquired from the picture. Here the picture grouping is done, and the choices are taken basing on it. In the event that the picture has street the vehicle will go straight, in the event that there is any hindrance like tree or human, the vehicle is halted. Facial acknowledgment project from Facebook in which the photograph of the client is recognized by profound learning. The organization is trained over a few human faces and highlights from face like eyes, ears and nose are gained from the preparation of the organization. The classifier arranges the face in view of the highlights saw in the pictures. In [3], has talked about rearrangement of various types of animal and plant utilizing Unsupervised learning calculation. This rebuild system fundamentally centers the likeness in shape and design of the species shared across various classifications and contrast in the article parts. He likewise proposes the format model for catching the normal part and states of the item parts and Co-event connection of the article design.

## **II. RELATED WORK/ LITERATURE SURVEY**

### **2.1 Convolutional Neural Network**

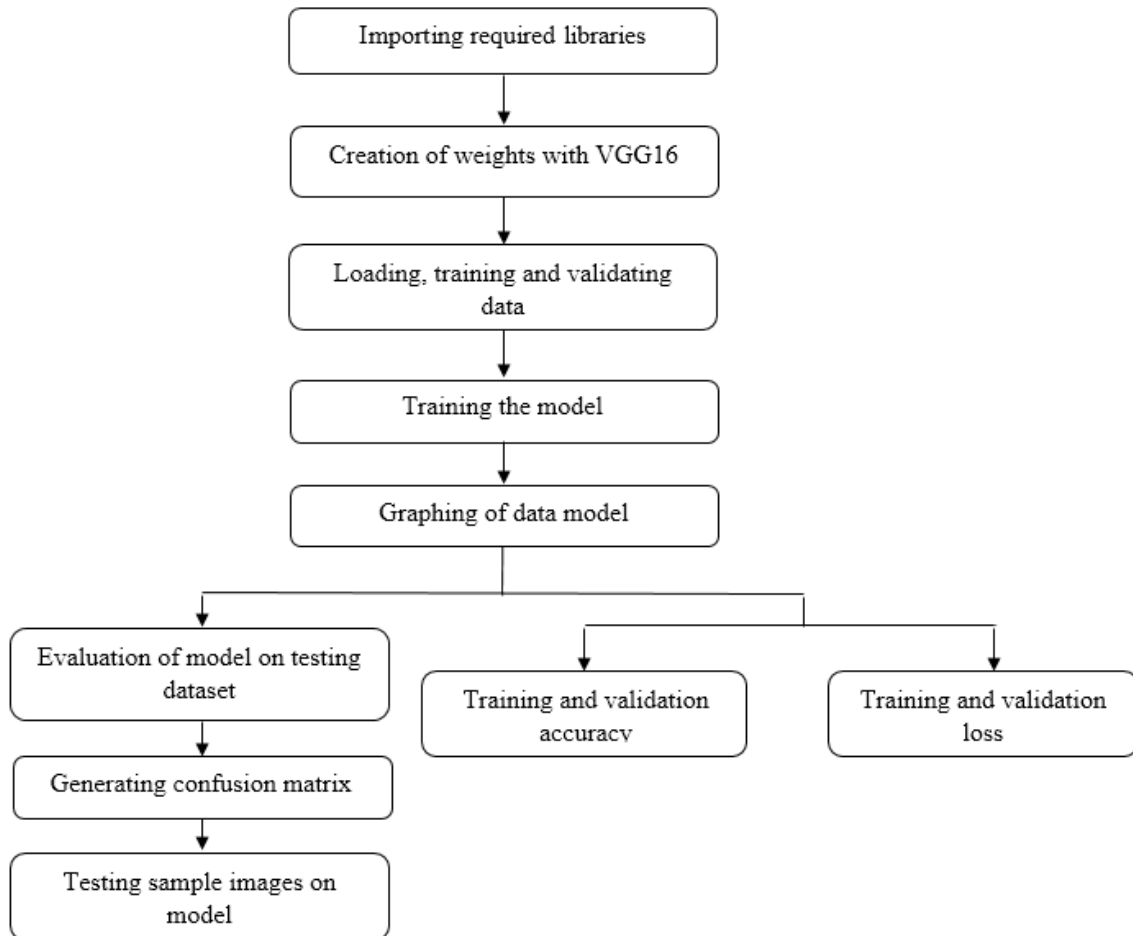
The authors of this paper[4] suggests strategy for robotized submerged fish species order. Well known methodologies accentuation on characterization of fishes outside of water since submerged order conveys a few difficulties, for example, foundation smoothness, mutilation, object segregation, picture quality and impediment. The proposed technique recommended execution of eliminating the smoothness in the dataset. Execution of picture handling before the preparation step assists with killing the submerged snags, soil and non fish bodies from the pictures. The accompanying advance uses Deep Learning approach by execution of Convolutional Neural Network for fish species characterization. Correlation of ReLU, Soft Max and tanh initiation capacities was performed and ReLU initiation work was viewed as profoundly precise.

### **2.2 VGG16**

The paper[5] examines VGG16 applications for Plant Image Classification alongside Data Augmentation and Transfer Learning, where it utilizes move learning and Convolutional neural organization to group the plant species. Leaf pictures are utilized rather than its bloom partners as its low-level elements like tone, shape, and so forth; and are ordinarily utilized in other plant acknowledgment models. This acts a significant hindrance is we utilize just leaf pictures as a sole element/boundary to arrange/perceive various types of plants. Information Expansion, dropout and move learning can successfully help in diminishing one of Convolutional neural network most computationally bulky issue – over fitting in little datasets. It utilizes a VGG net model which was prepared on Image Net informational collection. Age of more examples to assist with displaying train better is upheld by Data Augmentation. Existing preparing tests go through a few essential changes as well. Information increase guarantees that the model doesn't "see" a similar picture two times during training, in this way diminishing upward also the model is presented to numerous different parts of information. This, at last aides in summing up model better.

## **III. PROPOSED SYSTEM**

Here, we present a system for the arrangement of fauna pictures, which will help biologist and researchers to additional review or potentially further develop territory, natural and eradication designs. Figure 1 shows the proposed plan of model for Fauna Image Order utilizing Convolutional Neural Network. The Amino-10 dataset[6] utilized for the grouping is taken from Kaggle. We are utilizing Convolutional Neural Network with Broken ReLU initiation capacity and VGG16 design for our model. The underlying advance focuses on formation of elements with VGG16 model. Utilization of Image Processing along with Loading, Testing, Training, and Validating the dataset before the preparation step makes a difference to eliminate the noise, obstructions, smoothness and dust from the pictures. The subsequent stage employments Convolutional Neural Network alongside Leaky ReLU to train the model to accurately and definitively group animals classes.



**Figure 1:** Design of proposed neural network

To stay away from the issue of Dying ReLU, where some ReLU neurons basically kick the bucket for all inputs and stay idle regardless info is provided, here no inclination streams and on the off chance that huge number of dead neurons are there in a neural organization its exhibition is impacted. To determine this issue, we utilize what is called Broken ReLU, where incline is changed left of  $x=0$  and hence causing a release and expanding the scope of ReLU. Subsequent to training the model, we chart the model's training and validation exactness and loss to have bits of knowledge regarding how well the model is trained. Lesser the loss, more the exactness. The subsequent stage is to create confusion matrix to have precise insights concerning how accurately the model is trained and classifying, as we can't depend on the precision. Ultimately, we tried our model with test information and viewed it as precisely ordered. Underneath is the nitty gritty clarification of every modules executed.

### 3.1 Importing the Libraries

This module is utilized for bringing in the necessary libraries for the neural network model. We utilized different libraries, for example, Panda library is utilized for giving high execution, simple to-utilize information designs and data investigation, NumPy for numerical and legitimate procedure on clusters can be performed, Keras is intended to empower quick trial and error with profound neural organizations, it centers around being easy to understand, measured, and extensible, and some more.

### 3.2 Creation of Weights/Features with VGG16

A bottleneck record was made to urge the network to pack highlight portrayals to best fit in the accessible space, to get the best loss during while training. We classified the picture dataset into train, Validation and testing and stacked it in the model. We likewise stacked a pre-prepared VGG16 model. This module makes loads/highlights with VGG16 model to finetune the neural organization to perform well with the data. This module makes loads/highlights with VGG16 model to approve and test information. Information quality is essential to guarantee accuracy and dependability. The approve network effectively observed validated pictures which had a place with 6 animal classes in and the test network effectively observed test pictures which have a place with 6 animal classes.

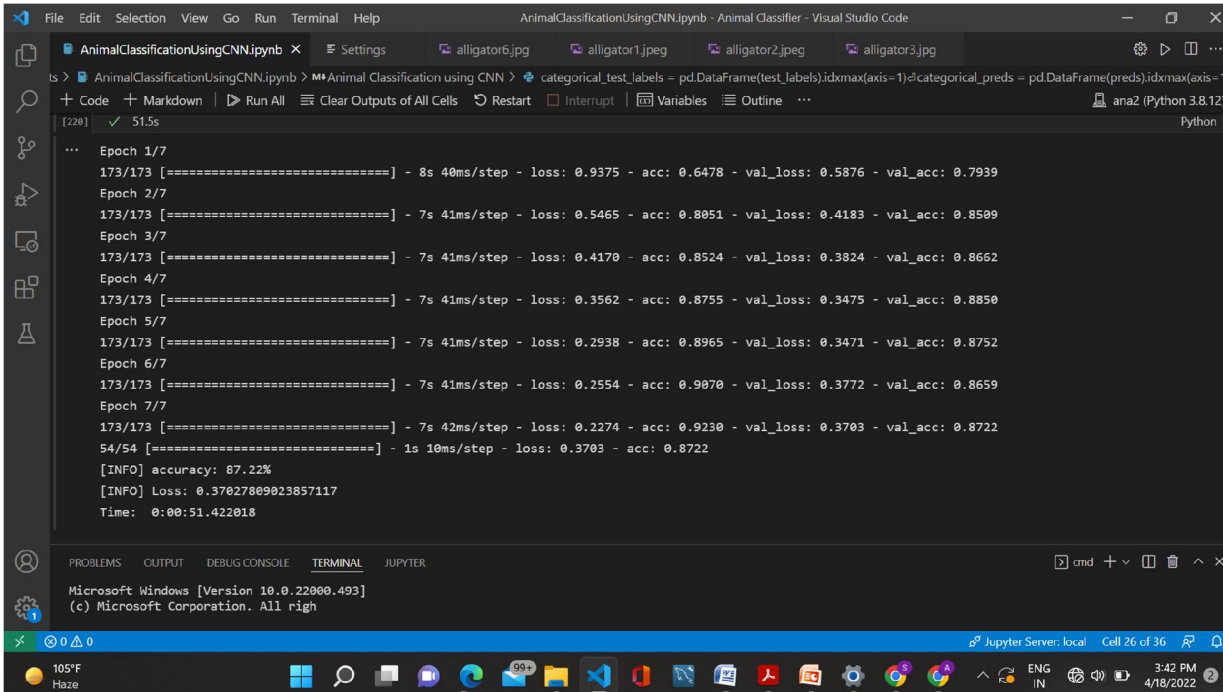
### 3.3 Loading Training, Validating, and Testing Data

This module loads training, testing and validate dataset for testing the model. Preparing information is the genuine dataset that we use to train the model. The neural organization model "notifies" and "learns" on its own from the training information. Testing data is the example of information that is utilized to give a fair assessment of the best last model on the training dataset. validated information is the example information that is utilized to give a fair assessment of a model on the training information while tuning model hyper parameters.

The assessment becomes more one-sided on the validation dataset is consolidated into the model arrangement. The training network effectively validated train images. The testing network effectively got train pictures which contain 6 animal classes. The validation network effectively observed validated images from the trained 6 animal classes.

### 3.4 Training & Graphing the Accuracy and Loss

We are graphing the training and validation accuracy and loss for every age. During an epoch, the misfortune work is determined across each data item thing and it is ensured to give the quantitative loss measure at the given age and Indicating curve across cycles just gives the loss on a subset of the whole dataset. The model assess work tells how well the machine we just made can forecast against inconspicuous information. Model assessment states how well the machine we just made can anticipate against inconspicuous information. There are two incredible techniques to perceive how well our machine can foresee or group. One of them is the classified matrix and the other is confusion matrix.



```

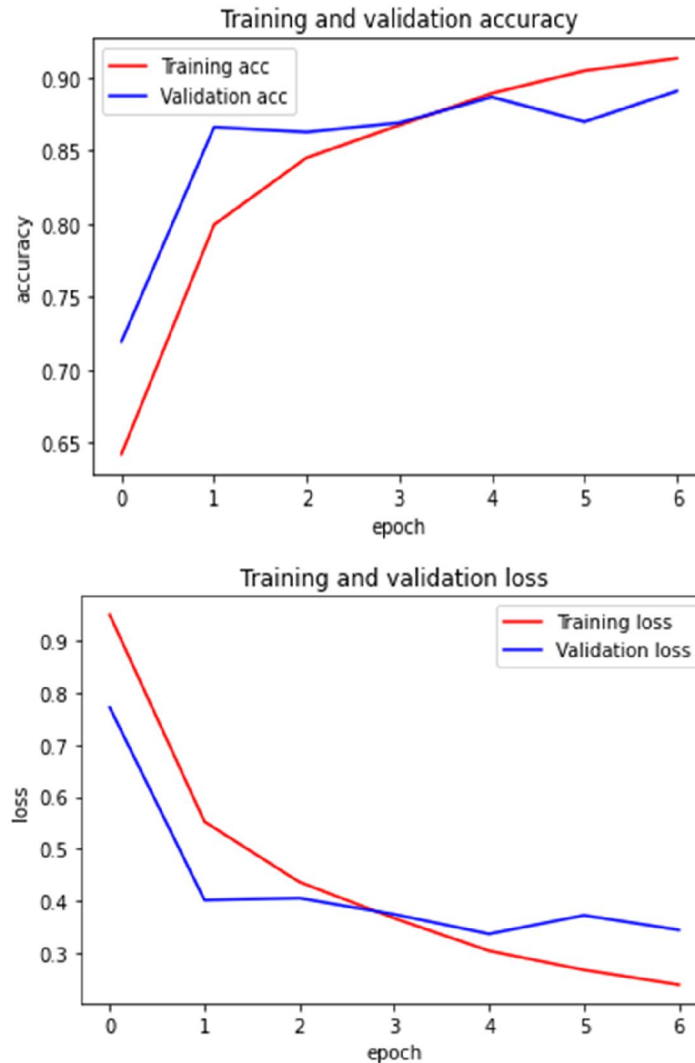
AnimalClassificationUsingCNN.ipynb - Animal Classifier - Visual Studio Code
File Edit Selection View Go Run Terminal Help
AnimalClassificationUsingCNN.ipynb x Settings alligator6.jpg alligator1.jpeg alligator2.jpeg alligator3.jpg
ls > AnimalClassificationUsingCNN.ipynb > Animal Classification using CNN > categorical_test_labels = pd.DataFrame(test_labels).idxmax(axis=1); categorical_preds = pd.DataFrame(preds).idxmax(axis=1)
+ Code + Markdown | Run All | Clear Outputs of All Cells | Restart | Interrupt | Variables | Outline ... ana2 (Python 3.8.12)
[220] ✓ 51.5s Python

...
Epoch 1/7
173/173 [=====] - 8s 40ms/step - loss: 0.9375 - acc: 0.6478 - val_loss: 0.5876 - val_acc: 0.7939
Epoch 2/7
173/173 [=====] - 7s 41ms/step - loss: 0.5465 - acc: 0.8051 - val_loss: 0.4183 - val_acc: 0.8509
Epoch 3/7
173/173 [=====] - 7s 41ms/step - loss: 0.4170 - acc: 0.8524 - val_loss: 0.3824 - val_acc: 0.8662
Epoch 4/7
173/173 [=====] - 7s 41ms/step - loss: 0.3562 - acc: 0.8755 - val_loss: 0.3475 - val_acc: 0.8850
Epoch 5/7
173/173 [=====] - 7s 41ms/step - loss: 0.2938 - acc: 0.8965 - val_loss: 0.3471 - val_acc: 0.8752
Epoch 6/7
173/173 [=====] - 7s 41ms/step - loss: 0.2554 - acc: 0.9070 - val_loss: 0.3772 - val_acc: 0.8659
Epoch 7/7
173/173 [=====] - 7s 42ms/step - loss: 0.2274 - acc: 0.9230 - val_loss: 0.3703 - val_acc: 0.8722
54/54 [=====] - 1s 10ms/step - loss: 0.3703 - acc: 0.8722
[INFO] accuracy: 87.22%
[INFO] Loss: 0.37027809023857117
Time: 0:00:51.422018

PROBLEMS OUTPUT DEBUG CONSOLE TERMINAL JUPYTER
Microsoft Windows [Version 10.0.22000.493]
(c) Microsoft Corporation. All rights reserved.

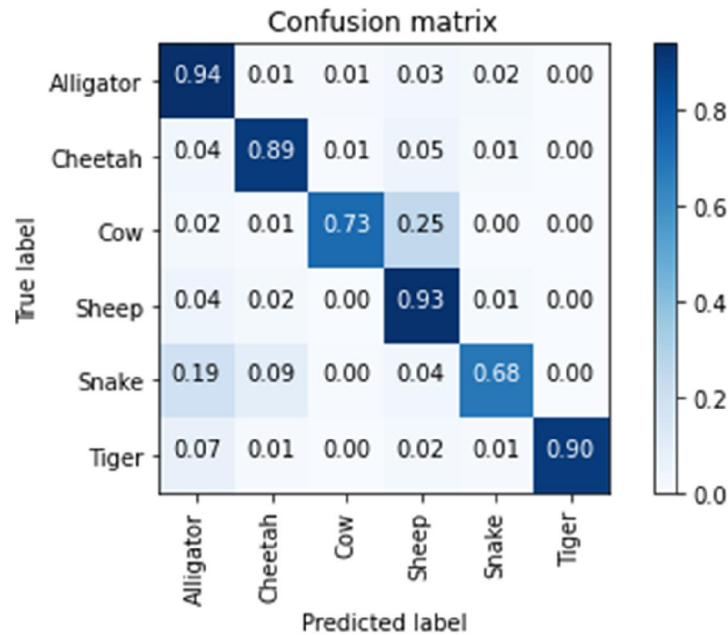
Jupyter Server: local Cell 26 of 36
105°F Haze 3:42 PM 4/18/2022

```



### 3.5 Generating Confusion Matrix

We performed confusion matrix as we can't depend on the accuracy. Confusion matrix is utilized to assess the nature level of the result of a classifier. Confusion matrix is an even perception of the model forecasts versus the ground-truth marks. Each line of confusion matrix addresses occurrences in an anticipated class and every section addresses the occasions in a genuine class. The inclining components address the quantity of points for which the anticipated name is equivalent to the genuine mark, while off-corner to corner components are those that are mislabeled by the classifier. The higher the askew upsides of the confusion matrix work the better, showing many right expectations. The NumPy exhibit we made before is put inside an information outline. confusion matrix works best on information outlines. The figures show the confusion matrix with and without standardization by class support size (number of components in each class). This sort of standardization can be fascinating if there should be an occurrence of class awkwardness to have a more visual translation of which class is being misclassified.



### 3.6 Testing Images on Model

At last, the last stage is the testing of the trained model on an example image to check whether the neural network is trained precisely and is working without mistake. The image is fed in the neural network model and the model precisely categorize the animal class. The def read\_image work is telling our machine that it needs to stack the picture, change the size and turn it to an array. The def test\_single\_image work is utilizing transfer learning's forecast model and an iterative capacity to assist with anticipating the image appropriately. The path is where we characterize the image location and lastly the test\_single\_image cell block will print out the eventual outcome, contingent upon the forecast from the another cell block. Our neural network accurately distinguished different animal image with various classes, commotion, and so on and arranged it to the animal class with.

```
test_single_image(path)
[278] ✓ 35s

... [INFO] loading and preprocessing image...
ID: 0, Label: Alligator 0.0%
ID: 1, Label: Cheetah 0.0%
ID: 2, Label: Cow 0.0%
ID: 3, Label: Sheep 0.0%
ID: 4, Label: Snake 100.0%
ID: 5, Label: Tiger 0.0%

Final Decision:
ID: 5, Label: Snakes
```

```
test_single_image(path)
[249] ✓ 0.9s

test_single_image(path)
[250] ✓ 10.1s

... [INFO] loading and preprocessing image...
ID: 0, Label: Alligator 0.0%
ID: 1, Label: Cheetah 100.0%
ID: 2, Label: Cow 0.0%
ID: 3, Label: Sheep 0.0%
ID: 4, Label: Snake 0.0%
ID: 5, Label: Tiger 0.0%

Final Decision:
ID: 2, Label: Cheetah
```





```

[271] ✓ 0.6s

test_single_image(path)
[272] ✓ 4.2s

... [INFO] loading and preprocessing image...
ID: 0, Label: Alligator 0.32%
ID: 1, Label: Cheetah 0.0%
ID: 2, Label: Cow 99.67%
ID: 3, Label: Sheep 0.0%
ID: 4, Label: Snake 0.01%
ID: 5, Label: Tiger 0.0%

Final Decision:
ID: 3, Label: Cow
  
```

PROBLEMS OUTPUT DEBUG CONSOLE TERMINAL JUPYTER

Microsoft Windows [Version 10.0.22000.493]  
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```

[275] ✓ 0.1s

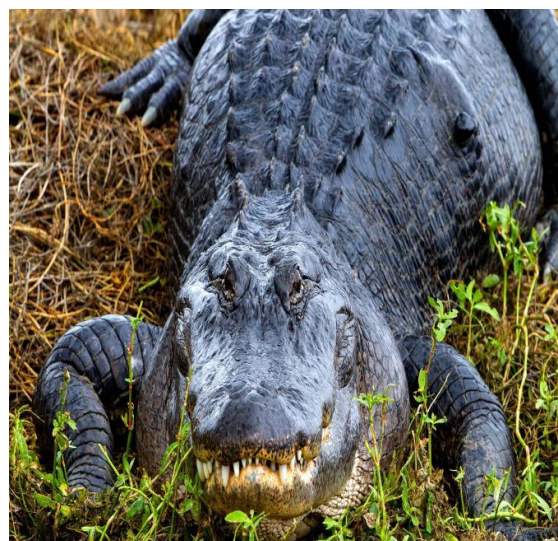
test_single_image(path)
[276] ✓ 10.3s

... [INFO] loading and preprocessing image...
ID: 0, Label: Alligator 84.02%
ID: 1, Label: Cheetah 0.0%
ID: 2, Label: Cow 1.22%
ID: 3, Label: Sheep 13.85%
ID: 4, Label: Snake 0.43%
ID: 5, Label: Tiger 0.47%

Final Decision:
ID: 1, Label: Alligator
  
```

PROBLEMS OUTPUT DEBUG CONSOLE TERMINAL JUPYTER

Microsoft Windows [Version 10.0.22000.493]  
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#### IV. RESULT AND OBSERVATION

The proposed model was coded using Python language and tested in Visual Studio on a dataset that contain 12,984 images from 6 different animal species that belong to different animal classes/kingdom. The model could achieve accuracy of 87.22 for 6 animal classes. The neural network could successfully identify the animal image and classified it to the correct animal class. And some test images are also being classified to their respective classes with an accuracy of 100%. The model successfully detected test 1,675 images from 6 animal classes, 8,640 training images from 6 animal classes, and validated 2,669 images from 6 animal classes.

Animal	Number Of Images
Cheetah	1632
Cow	2272
Alligator	2207
sheep	2083
Tiger	1590
Total	12984

#### V. CONCLUSION

This model utilizes Convolutional Neural Network (CNN) algorithm to distinguish animal. The algorithm characterizes animals productively with a proper accuracy and furthermore the image of the classified animal is shown for a finer outcome. So it will be utilized for different purposes, for example, identifying wild animals going into human territory and to forestall natural life poaching and surprisingly human animal dispute.

#### VI. FUTURE WORK

Fostering a simple yet effective UI for the undertaking for simple use for scientist, photographers, computer researchers, the arrangement exactness, accuracy and decrease in mistakes /errors, training and testing time .The image categorization model can be enhance / upgrade in future, by including low-level elements, for example, shape and spatial area includes separated from upgrading the weights and learning pace of the neural organization. At the point when these enhancements are joined in the arrangement framework, it would help further work on the exhibition and be valuable for applications implied for the accurate grouping framework.

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