# DOG BREED IMAGE CLASSIFICATION

-Chris Burger



#### Contents

- 1. Problem Statement
  - 2. EDA/Preprocessing
- 3. Modeling
  - 4. Examples
    - 5. Conclusion and the Future

# Problem<br/>Statement

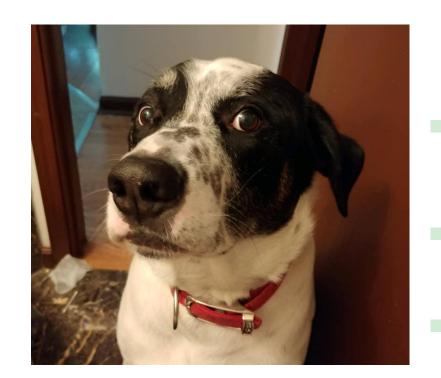
This classifier was made to identify which dog breed is shown in a given picture.

-A classification model to answer the question: "What kind of Dog IS this?!"



## R.I.P. Cooper Michael Jackson Burger



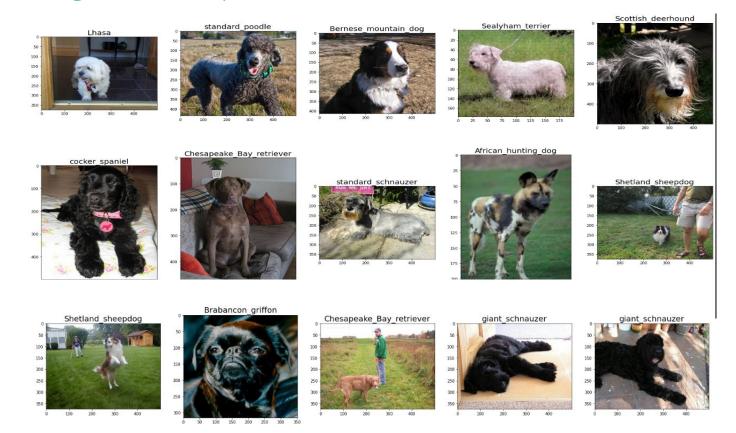


## Dataset information, Pre-Processing and Exploratory Analysis

The dataset used is the 'Stanford Dogs Dataset for Fine-Grained Visual Categorization' found <u>here.</u>

- **20,580** total images of **120** different Dog Breeds (~170 per breed. A Good size of data to work with)
- Reformatted the images as they were different sizes, attempting the dimensions:
  - 28 x 28, 299 x 299 and finally 180 x 180

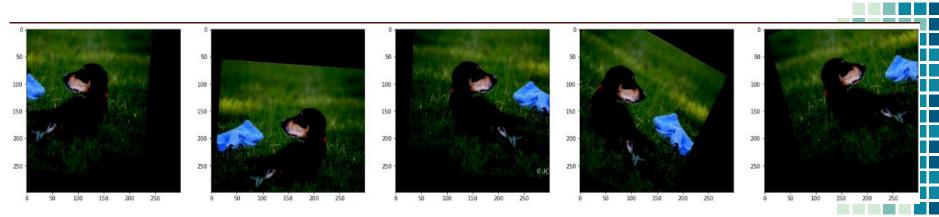
### Image Examples of random breeds:



#### Data Augmentation

Augmenting photos, which basically rotates and flips images, was a good way of creating more data for the models to train off of. (Used for when you need more data)

Below is an example of some augmented versions of one of the photos:





## MODELING

Time to create the best possible CNN Model to determine what breeds of dog are in these images....



#### Results, best parameters, etc.

10 Epochs:

299x299:

85.6% train & 88.6% testing accuracies

180x180:

76.3% training and 77.3% testing accuracies

**Layers:** 

**Inception V3** 

Global Average Pooling 2D

**Flatten** 

**Dropout** 

3 Dense layers (relu, relu and softmax for final layer to pick the breed)

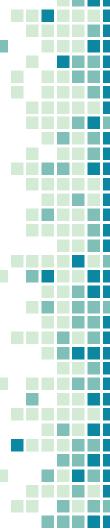
20 Epochs

299x299:

94.2&% training & 92.7% testing accuracies.

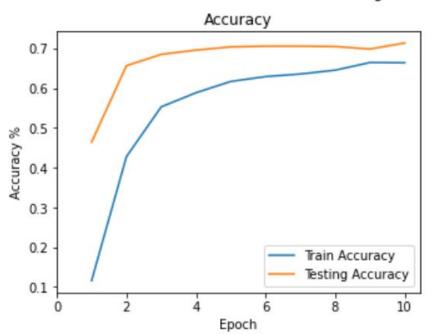
180x180:

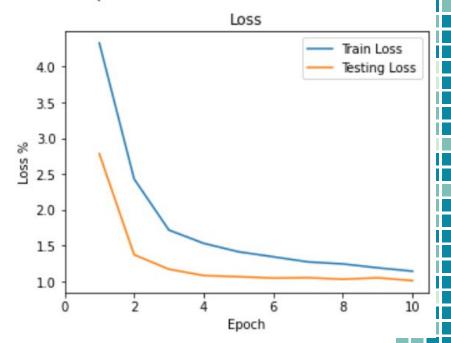
81.92% training and 76.61 testing accuracies



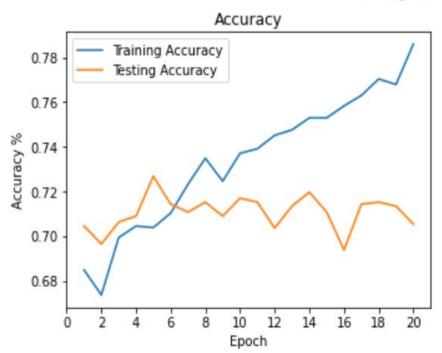
#### Results cont. - only 180 x 180 pixels

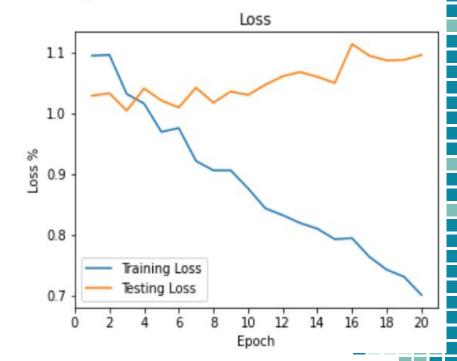
#### Learning Performance Over 10 Epochs





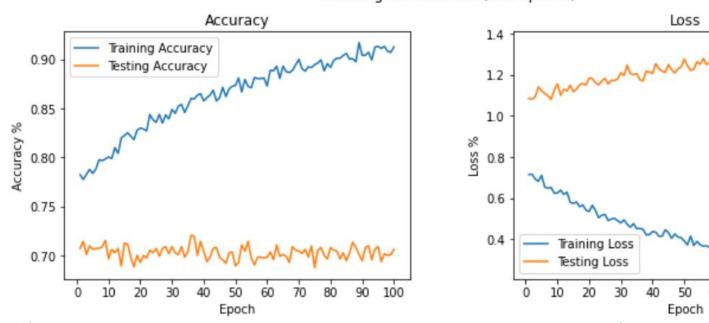
#### Learning Performance (20 Epochs)

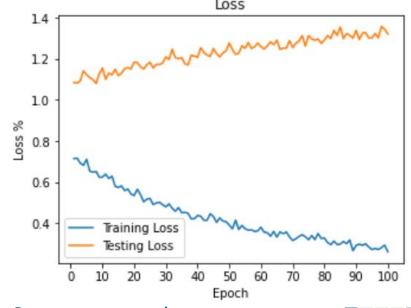




### Finally, at 100 Epochs







(Using the same params - dims of 180 x 180) 91.27% Training and 70.63% testing. Not much improvement/overfit

#### Examples of some predictions & errors made:

Actual: Rhodesian\_ridgeback Predicted: Rhodesian\_ridgel



Actual: English\_foxhound Predicted: Walker\_hound Conf: 0.81851655



Actual: giant\_schnauzer
Predicted: giant\_schnauzer



Actual: Bedlington\_terrier Predicted: Bedlington\_terrier Conf: 0.9978999



Actual: Sealyham\_terrier Predicted: Sealyham\_terrier Conf: 0.9645313



Actual: Maltese\_dog Predicted: Shih-Tzu Conf: 0.9184978



Actual: Brabancon\_griffon Predicted: Brabancon\_griffor Conf: 0.9979175



Actual: Scottish\_deerhound Predicted: Irish\_wolfhound Conf: 0.5985619



Actual: Afghan\_hound Predicted: Afghan\_hound Conf: 0.98769623



Actual: Gordon\_setter
Predicted: Gordon\_setter
Conf: 0.9842406



#### Conclusion/Future

After messing with many different models, InceptionV3 with Global Average Pooling seemed to perform the best.

In the future, I would like to (and will) continue tinker with not only the models/hyperparameters but also different ways of processing the images before sending them through the model, as all of that can have a huge impact.

Also: a Demo, as seen <u>HERE</u> (github user: stormy-ua)



#### THANK YOU FOR YOUR TIME!

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