# CS5242 Dog Breed Classification

Liu lu, Kaiyu, Liuhuo, Jiang Liu

April 16, 2019

## Outline

- Background
- 2 Data preprocess
  - Train, validation and test data preparation
  - Data feature visualization
  - Data augmentation
- Model building
  - Basic models' comparison
  - Model analysis
  - Hyperparameter tuning
- 4 Conclusion

# Background

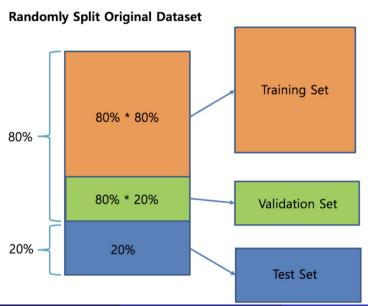
## Homeless dog

Approximately millions of stray dogs per year worldwide, to help them find a home and improve the work efficiency for dog adoption center staff by updating the manual process of dog classification to an automated process

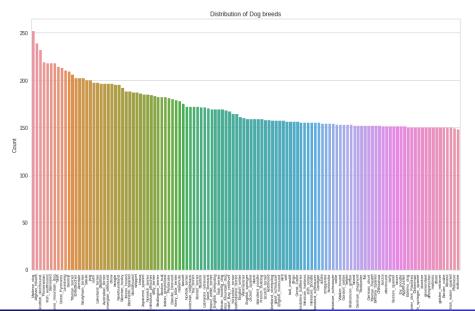
### Dataset Analysis

- approximate 20000 pictures
- 120 class of dog breed http://vision.stanford.edu/aditya86/ImageNetDogs/
- about 200 pictures of each class

# Data Preparation



## Data Feature Visualization



## Data Augmentation

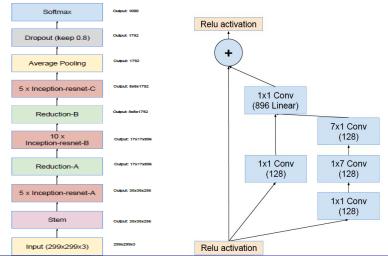
- rotation
- shear
- channel shift
- width shift
- height shift
- zoom range
- horizontal flip
- vertical flip

# Model Comparison

Mode	training accuracy	val accuracy	test accuracy
VGG16	0.2401	0.2247	0.2119
ResNet50	0.9867	0.5272	0.5344
Xception	0.9985	0.6486	0.6568
InceptionV3	0.9744	0.5833	0.6011
InceptionResNetV2	0.9951	0.7231	0.7261

# InceptionResnetV2 Model

- Add residual connection to Inception
- Compared to V1, V2 is larger and more precise



# Transfer learning

### Last layer:

- GlobalAveragePooling2D()(x)
- Dense(number)(x)
- BatchNormalization()(x)
- Activation('relu')(x)
- Dropout(0.5)(x)
- Dense(numberClasses, activation='softmax')(x)

## Fine tuning

#### Version 1

```
model.add(GlobalAveragePooling2D())
model.add(layers.Dense(1024, activation = "relu"))
model.add(layers.Dense(120, activation = "softmax"))
```

accuracy (input size = 200): train: 0.992 val: 0.7346 test: 0.7430

#### Version 2

```
model.add(GlobalAveragePooling2D())
model.add(layers.Dense(1024))
model.add(BatchNormalization())
model.add(Activation("relu"))
model.add(layers.Dense(120, activation = "softmax"))
```

```
accuracy (input size = 200): train: 0.99976 val: 0.66323 test: 0.6640 accuracy (input size = 299): train: 0.9983 val: 0.8224 test: 0.8207
```

## Fine tuning

### Version 3

```
model.add(BatchNormalization())
model.add(Dropout(0.001))
model.add(GlobalMaxPooling2D())
model.add(layers.Dense(512, activation = "relu"))
model.add(BatchNormalization())
model.add(Dropout(0.001))
model.add(layers.Dense(120, activation = "softmax"))
```

```
accuracy (input size = 200): train: 0.9951 val: 0.7231 test: 0.7262 accuracy (input size = 299): train: 0.9981 val: 0.8221 test: 0.8141
```

#### Version 4

```
model.add(Flatten())
model.add(Dropout(0.001))
model.add(layers.Dense(120, activation = "relu"))
model.add(BatchNormalization())
model.add(Dropout(0.001))
model.add(layers.Dense(120, activation = "relu"))
model.add(BatchNormalization())
model.add(Dropout(0.001))
model.add(layers.Dense(120, activation = "softmax"))
```

```
accuracy (input size = 200): train: 0.9542 val: 0.7668 test: 0.7774 accuracy (input size = 299): train: 0.9983 val: 0.7647 test: 0.7602
```

### Result

```
Epoch 88818: ReduceLROnPlateau reducing learning rate to 1.56249996852793e-86.
Epoch 00018: val_acc did not improve from 0.82235
Epoch 00019: val_acc did not improve from 0.82235
Epoch 20/20
Epoch 88828: ReduceLROnPlateau reducing learning rate to 7.81249988263965e-87.
Epoch 00020: val_acc did not improve from 0.82235
# accuracy metric 1
from keras.metrics import categorical_accuracy
result = model.evaluate_generator(test_gen_steps = test_steps)
print(model.metrics_names)
print(result)
['loss', 'acc']
[0.7886696782161937, 0.8286997885127353]
```

## Conclusion

### Challenges

- Unforeseeable hyper-parameter tuning result
- Evaluation difficulty due to long duration to train model

## Future works to improve model accuracy

- Add boosting
  - Combine different models
  - Tune hyperparameter better

# Thank you

Question?