

# CIFAR-2 Image Recognition

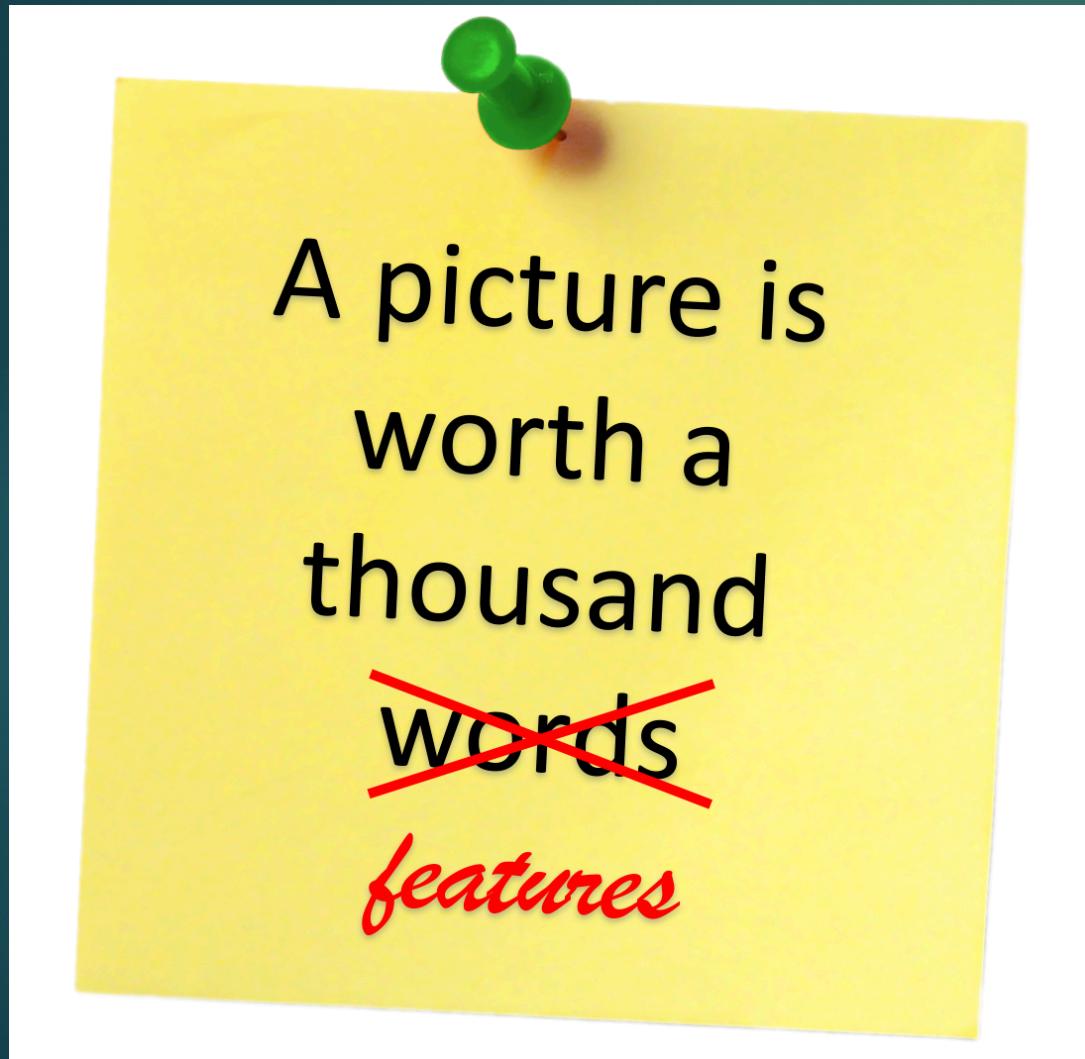
ANIL SIMON

ASHIM BHATTARAI

SARDHENDU MISHRA

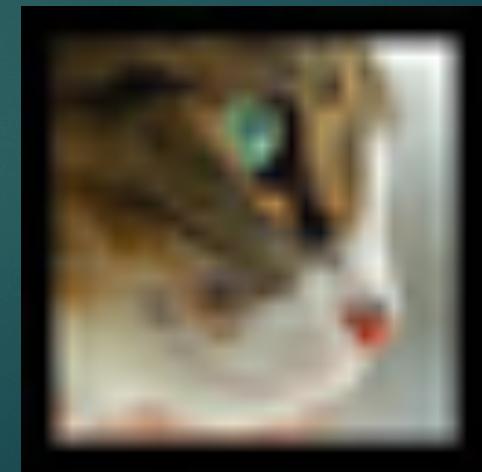
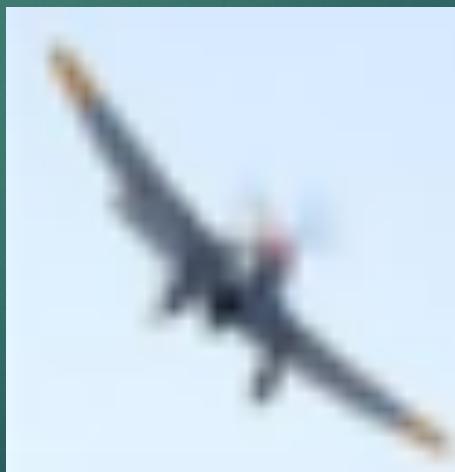
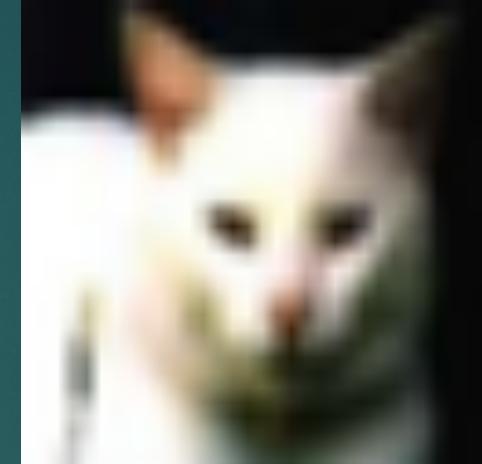
RODRIGO TEJEIDA ESTRADA

# Image Recognition



- Whale Tracking.
- Identify Distracted Drivers.
- Virus Classification.
- Detect Diabetic Retinopathy.
- Detect Lung Cancer.
- And more.

# CIFAR-2 - Object Recognition in Images



# Preprocessing and Initial Analysis

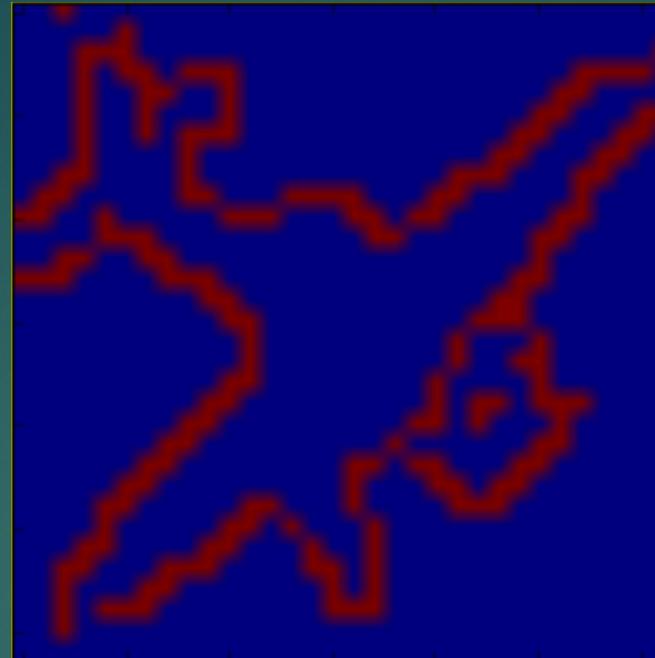
- File Handling:
  - Convert images into a single NumPy nd-array (matrix)
    - Image is  $32 \times 32 \times 3$  array of RGB channels
    - Each row in the nd-array represents one image
  - Input Matrix:  $10,000 \times 3072$ ; Outcome Array (labels):  $10,000 \times 1$
  - Basically each pixel is a “feature”

## Feature Extraction:

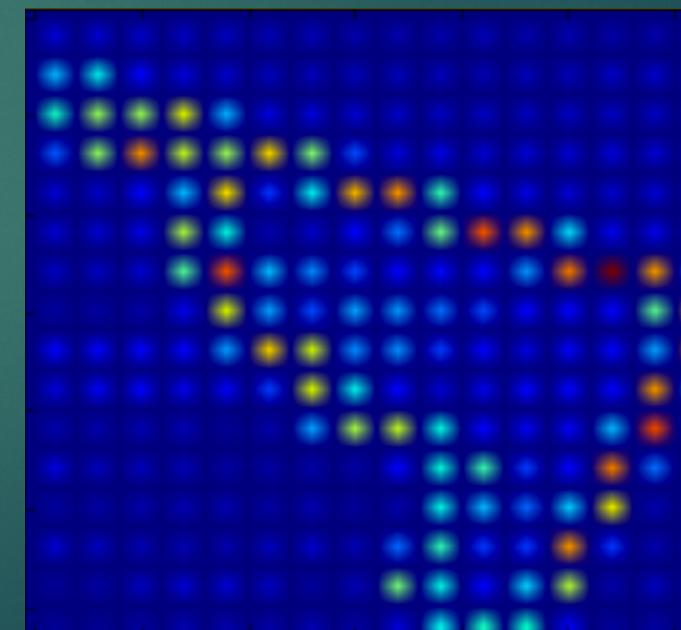
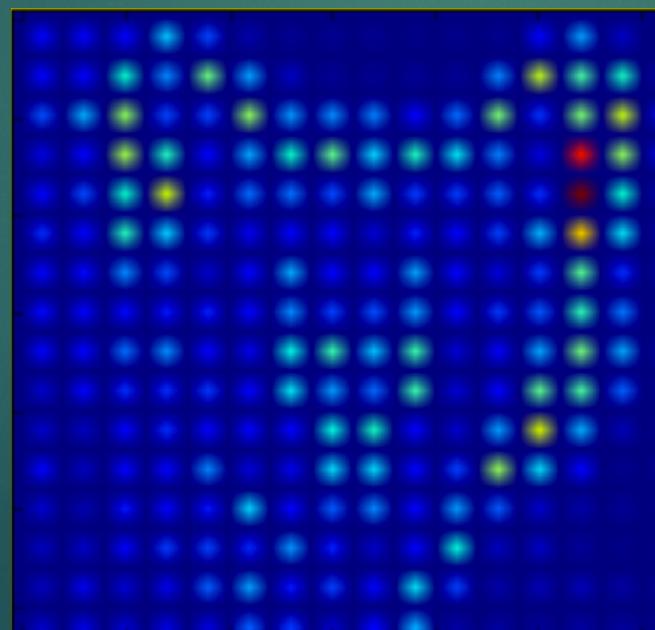
Idea: Extract “useful” feature from images

We extract edges, HOG features, and RGB (above)

Classifier uses features to compare images



Canny Edges



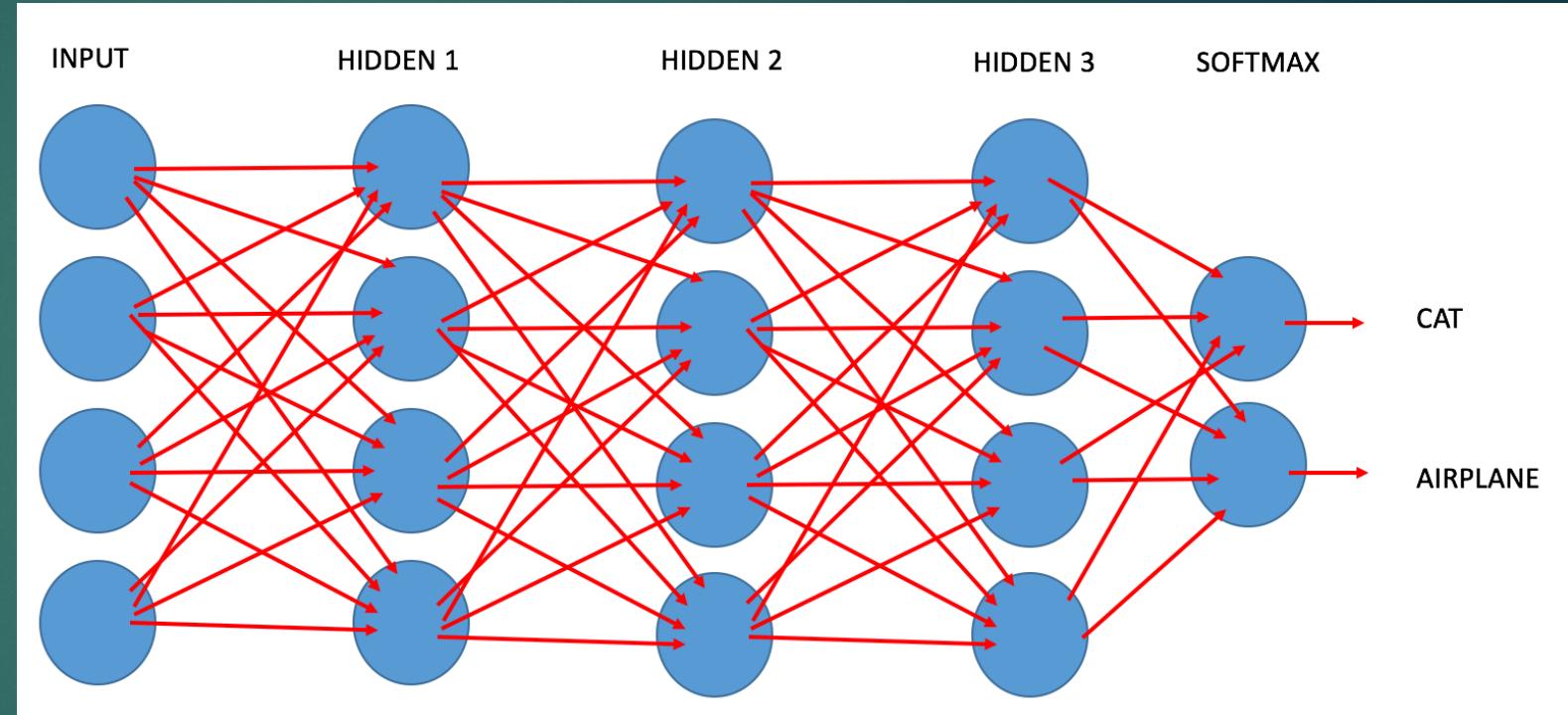
HOG Features

# Models

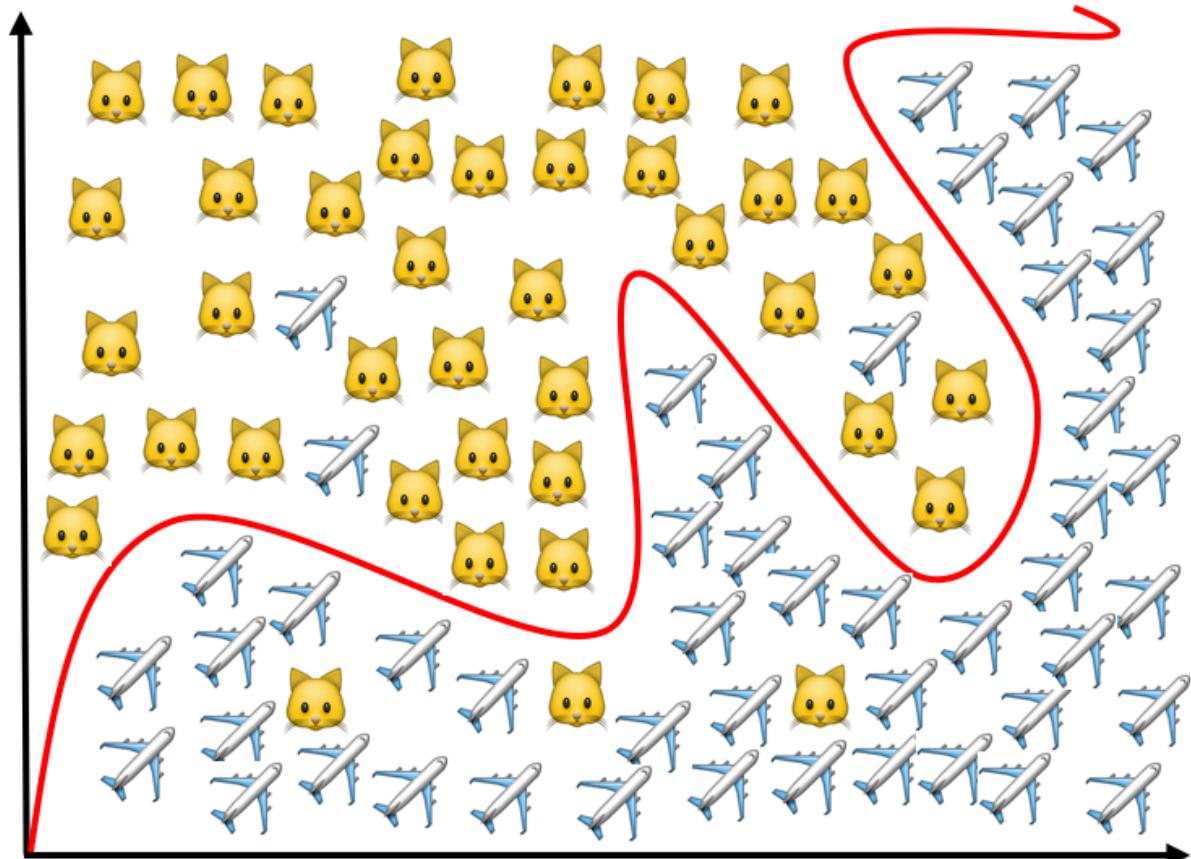


# Neural Network

- Four layer dense network: 1024, 1280, 1280 units
- Learning Rate = 0.006
- Dropout: 0.75, 0.5, 0.5
- Average: 87.2% Accuracy over 10-fold CV

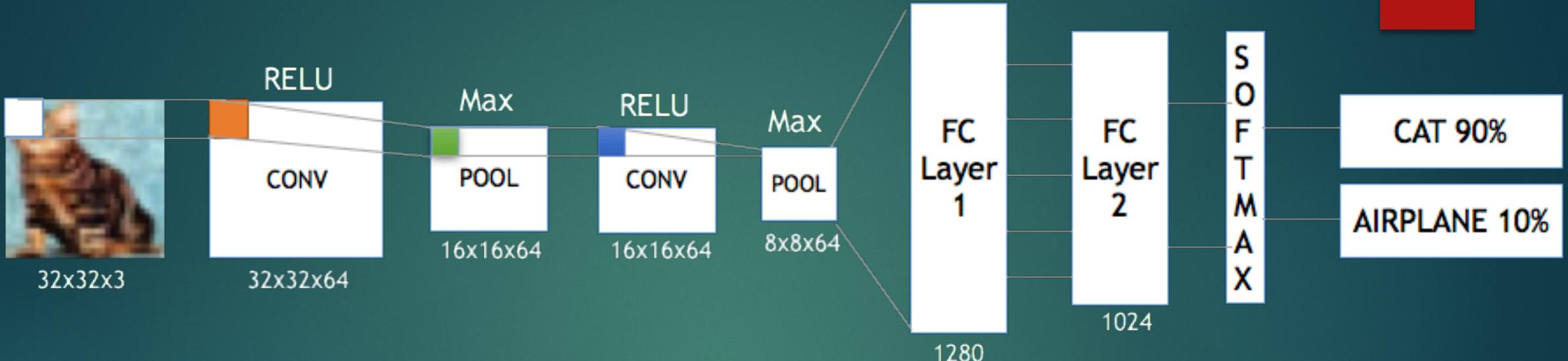


# SVM with RBF kernels



(4x4 kernel)	C = 10 Gamma = 0.1
Average Airplanes Number	444.2
Average Cats Number	445.4
Average Airplane Accuracy	0.8884
Average Cat Accuracy	0.8908
Average Total Accuracy	0.8896

# Convolutional Neural Networks (CNN)



■ Convolution Filter1  
3x3x64

■ Max Pool Filter1  
2x2

■ Convolution Filter2  
5x5x64

■ Max Pool Filter2  
2x2

## MODEL CONFIGURATION

- Feature: RGB, contrast, brightness, Standardize
- Data Augmentation - Flip Image
- Regularization: Dropout {0.5}
- Optimizer: RMSPROP

## MODEL EVALUATION

### Accuracy

Avg - 10 Fold cv - CAT	Avg - 10 Fold cv - Airplanes	Avg - 10 Fold cv - Overall
87.06	93.26	90.16

# Logistic Regression



RGB + ZCA + GRAY + STD + EDG+  
HOG

10417

- Total Features

120

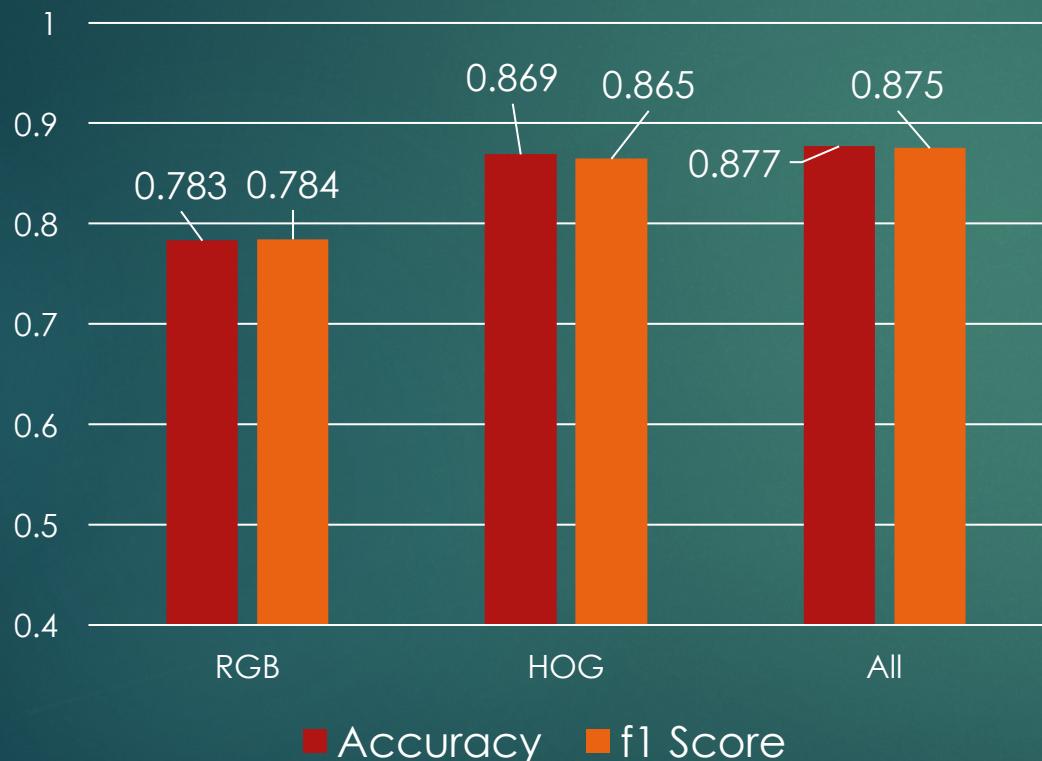
- Selected features

Parameters

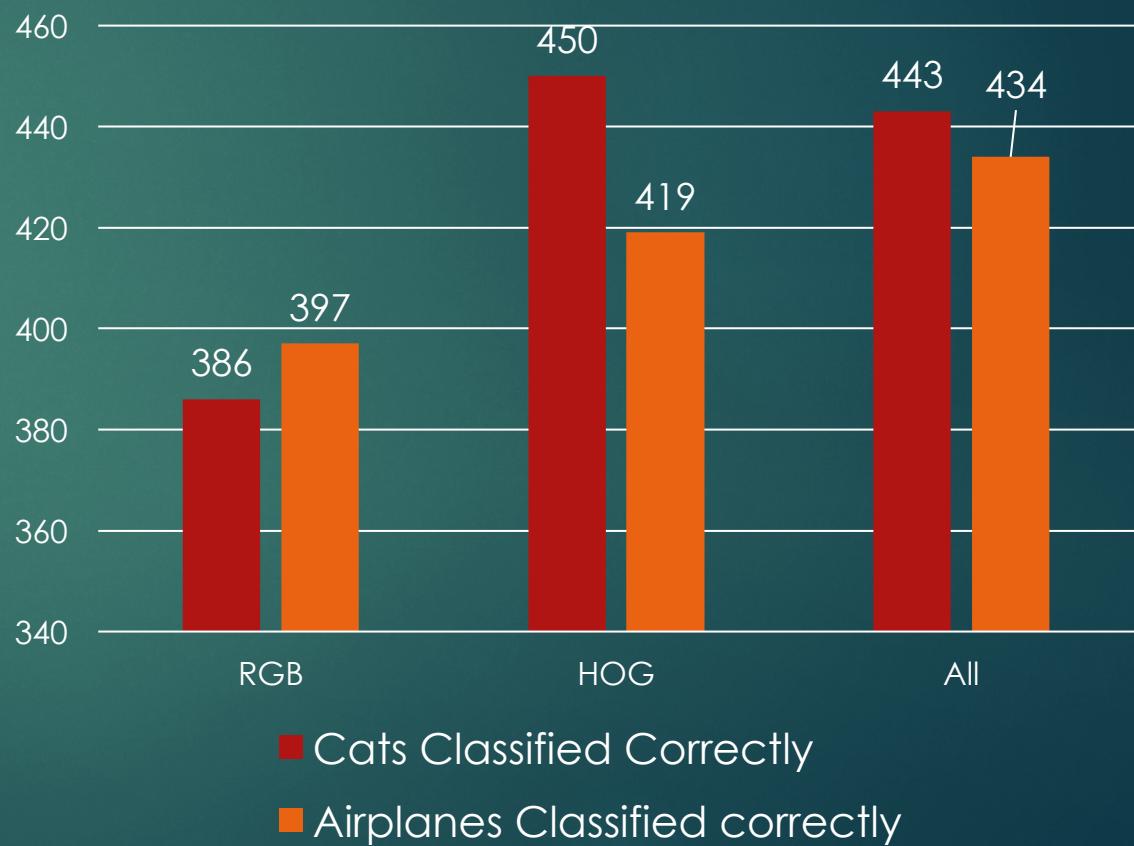
- Regularization = L2
- Regularization Strength = 0.1

# Comparison of Feature sets

Performance with different feature sets

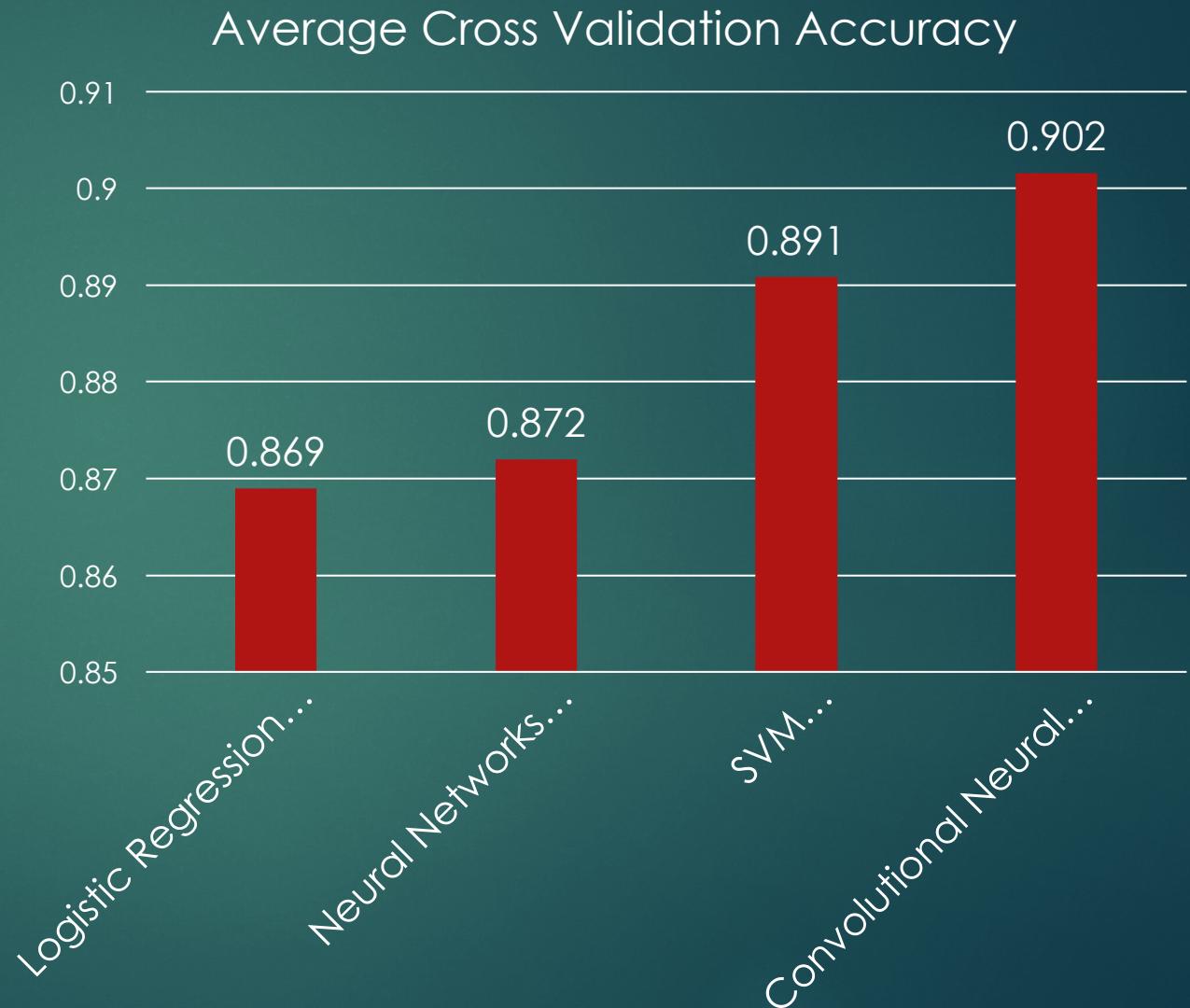


Average number of correctly predicted instances



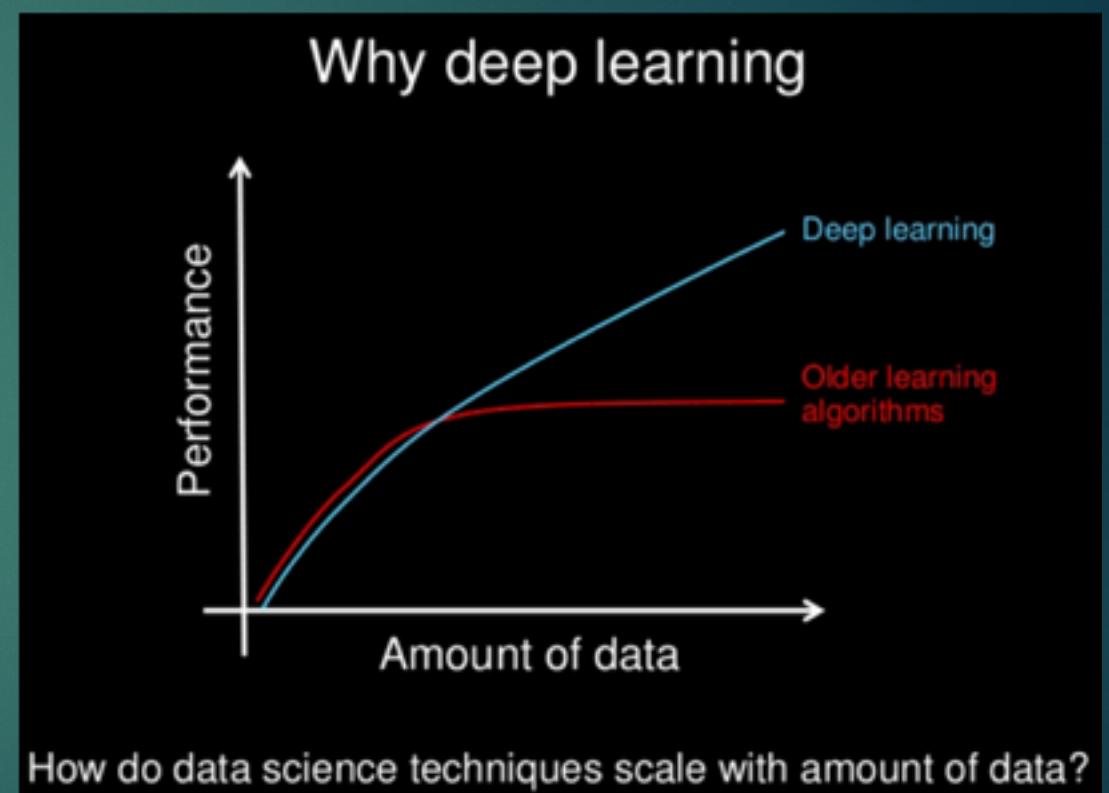
# Comparison of Model Performance

- Complexity of features used
  - Time taken for Extraction
- Training time
- Accuracy of predictions
- Scope for improvement/tuning



# Lessons Learned.

- Keep Calm and lower your learning rate.
- Don't be too proud to use widely accepted model architecture.
- Preference of a Simple solution over a Complex one. (CNN - Total parameters = 6,563,586)
- Image Recognition + Big Data + GPU = Conv Nets.



# Possible Next Steps.

- Identify misclassified images - check for common patterns
- Implement Inception modules for CNN
- Use Ensembles for different classes.
- Tune parameters and run the models for all the 10 class.

Thank  
You!!

