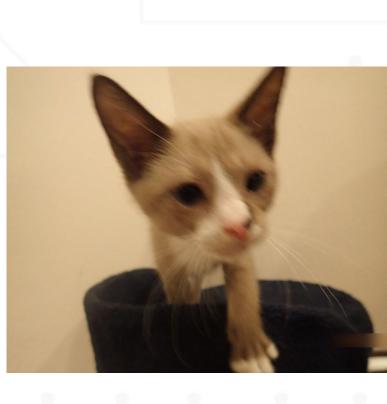
Predicting Pawpularity Score

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Decision surface



Pet Finder.my

Homeless 21083

Objective

images of pets and predict if the photo

will be effective on a pet finder website

• The ML model takes an image of a stray

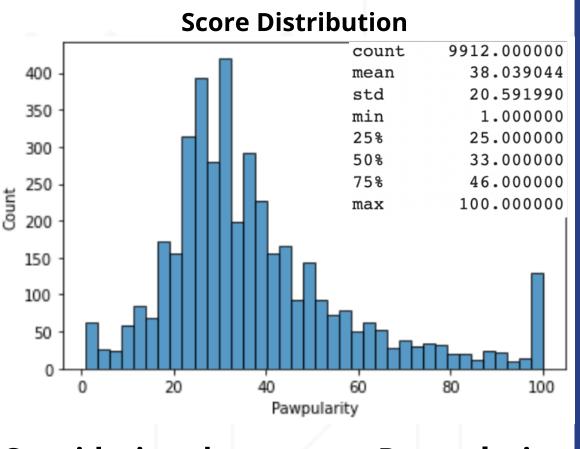
animal and outputs either 'CUTE' or

'NOT CUTE'

Create a ML model that processes

Happy 58401

Pawpularity score: derived from the statistics on the website (i.e. how many times the photo was clicked)



Considering the average Pawpularity score, in order to balance the dataset, we set the Pawpularity score bounds to be:

CUTE ≥ 33 NOT CUTE < 33

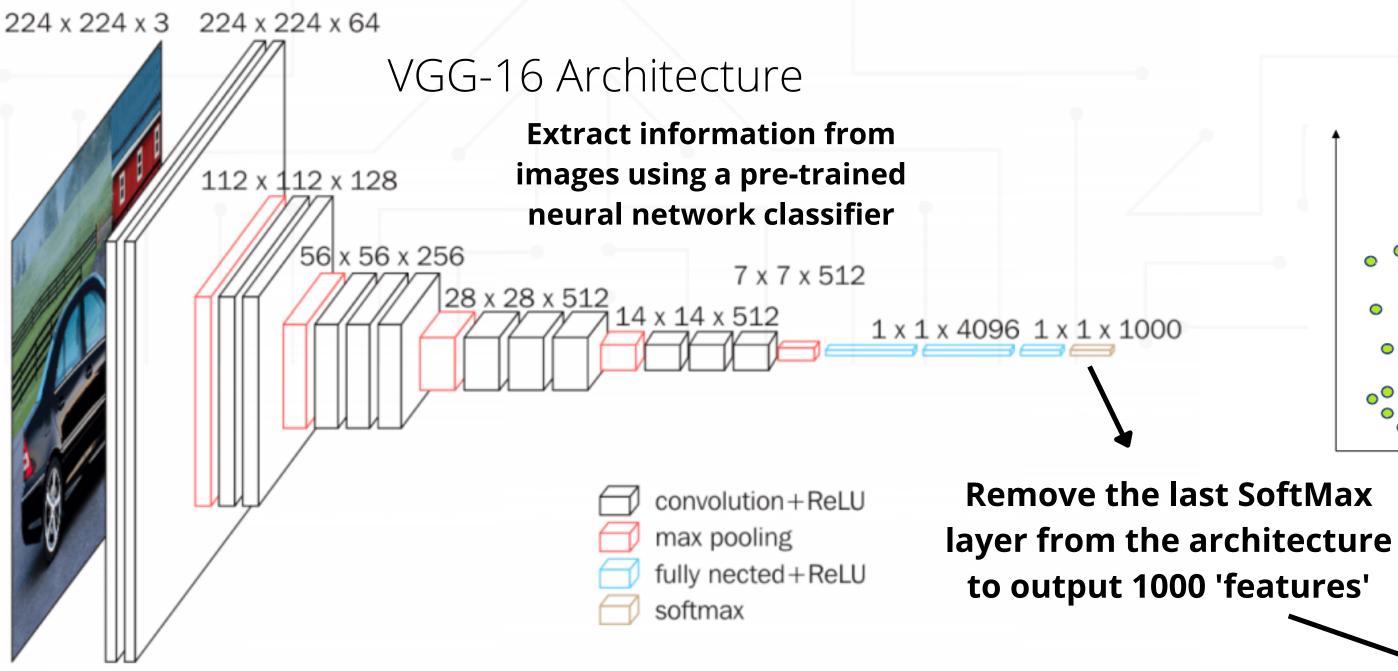
Methodology

- Simplify the problem by turning it into binary classifier ("Cute or not cute?")
- Use a (modified) pre-trained VGG-16 image classifier to extract 'features' from images
- Using the extracted 'features' as input, train a Support Vector Machine (SVM) to predict if the image is 'cute' or 'not cute'
- Then we compared it to another SVM model trained with manually labeled features provided by the websute

After being trained, SVM can accurately tell if the image is be attractive enough to do well on the pet finder website or not **Output: 'CUTE' or 'NOT CUTE'**

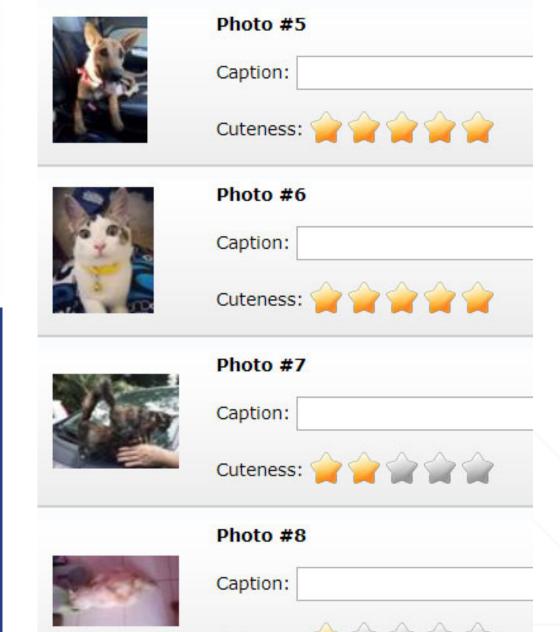
Support Vector Machine (SVM) Classifier

kernel



 $J(\theta_0, \theta_1)$

Cuteness-meter







Use the 1000 features extracted from images as inputs and train the SVM classifier

Conclusion & Reflection

- Training SVM with human labelled data resulted an accuracy same as guessing
- Training SVM with deep-learning extracted features was far more accurate and more efficient
- It will be even better if we can convert the model into a regressor that predicts the Pawpularity score (outputting a more precise numerical value instead of a binary output).

Results

00 07 06 05 04 03 02

• Overall, 70% accuracy for the model with deep-learning extracted features and 50% for the model with manually inputted features

accuracy 0.7031 accuracy 0.5048