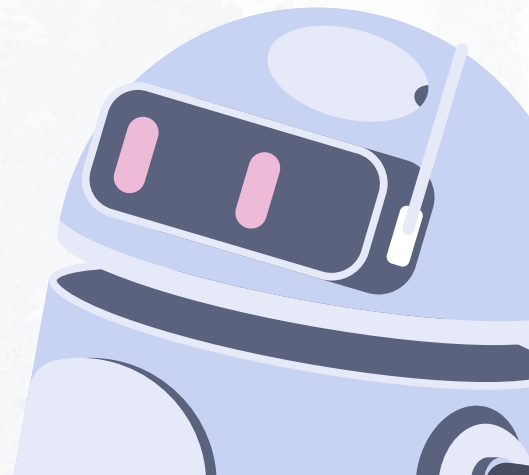


CSD3185

Group Project

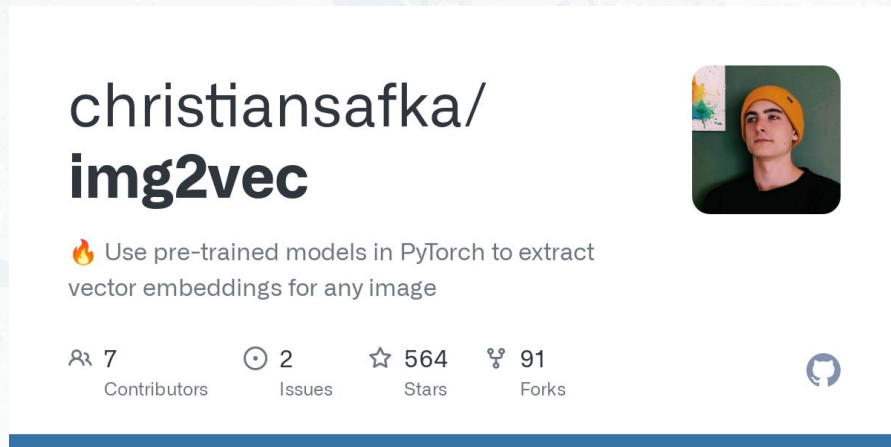
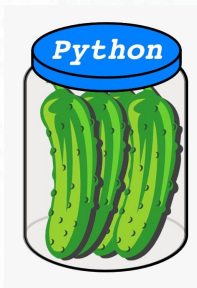
Image Classifier
Done by Team 5

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Yin Shengkai
Jed Goh Yujie
Seow Kai Jun
Dennys Tay Khaj Tjong
Huang Wei Jhin



Introduction

- **Image classifier** is an application to classify images into designated folders automatically
- Utilizes the following libraries:



Data Collection

Data consists of animal images which are retrieved from Kaggle

Images are split into 4 folders:

- **Training Folder** - used to store images for training
- **Validation Folder** - used to store images for validating models accuracy
- **Testing Folder** - used to test the model classification ability
- **Output Folder** - used to store images based on the model predictions

kaggle



Models Used



KNN

Appreciated for its simplicity and its direct approach



Random Forest

Used for its robust handling of complex data



SVM

Stands out for its ability to perform well in high dimension spaces

Pre-processing

Resizing

All images were **standardized** to 224x224 pixels without alpha channel. **Streamlined** feature extraction but also **improved** computation times.

Normalisation

Ensures a uniform feature space, facilitating more **accurate** distance measurements and slightly **boosting** the accuracy of models

Flipping

Added **flipped** versions of image to **augment** our dataset and **enhance** model robustness

Brightness and Contrast Adjustment

Modified the brightness and contrast of images across the board which **enhanced** model accuracy

Feature Extraction



Edge Detection

Accentuated the contours within images to spotlight the shapes and boundaries of objects



Grayscale Conversion

Converted images to grayscale to remove color data



Histogram Equalization

Modified image contrast to spread the most common intensity values

Problems Encountered

Problem

Result

Solution

**Bias in predictions
towards categories with
more training data**

**Overrepresentation of
categories in predictions**

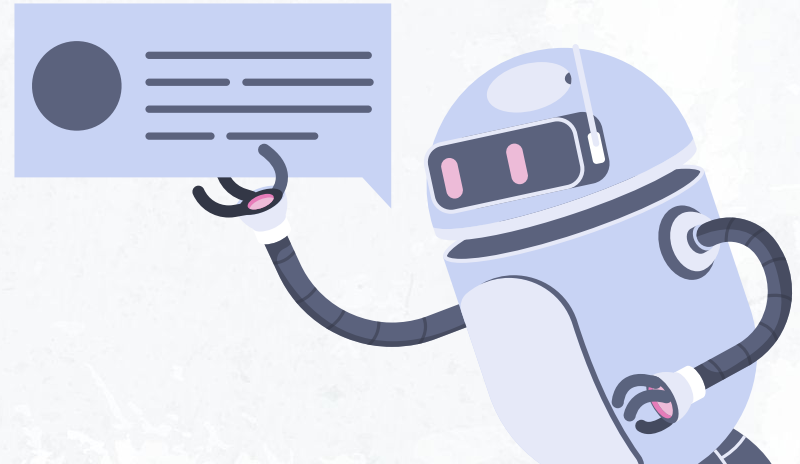
**Balanced dataset by
resampling images
and introduced crops
and rotations**

**Images that do not belong
to any category were
classified into one**

Skewed the results

**Implement
Probability drop-off
mechanism**

Application Demo



Optimisation Methods

Made use of an **automatic** optimisation method through the following steps:

Step 1

Initiate the optimization process by establishing a configuration for each model type

Step 2

Load the training data and labels, then apply StratifiedKFold, and maintain the percentage of samples for each class.

Step 3

Calculate the average accuracy across all folds for each parameter value introduce a size score

Step 4

Select the parameter value that results in the highest combined score for each model

Optimisation Methods

How the Optimisation Method Works

Step 1

Split our data into training and validation sets

Step 2

For settings with predefined choices, test each option one by one

Step 3

For numerical settings, use 3 different settings and repeatedly change their values until the best value is obtained

Step 4

Set our model based on the best numbers or options to get the best performance

KNN Results

n_neighbors	Accuracy
1	0.827962963
3	0.8468518519
5	0.8255925926
20	0.8188888889
50	0.7983512542

p	Accuracy
Manhattan	0.8513345852
Euclidean	0.8575482388

weights	Accuracy
uniform	0.8468518519
distance	0.8254355926

leaf_size	Accuracy
5	0.7909981028
10	0.8198884888
15	0.8275925926
20	0.8275271439
25	0.8625458519
30	0.8572345852

algorithm	Accuracy
auto	0.8468518519
ball_tree	0.8568345852
kd_tree	0.8109934528
brute'	0.7998200019

Random Forest Results

n_estimators	Accuracy	max_depth	Accuracy	min_samples_split	Accuracy	max_features	Accuracy
10	0.8858642079	1	0.7875652901	2	0.8775573929	sqrt	0.8895673245
100	0.9509638274	5	0.8775215498	10	0.8975632846	log2	0.9057218936
500	0.8957218936	10	0.8875783057	20	0.8957186349		
1000	0.8924398746	20	0.8854691285	50	0.8775718635		

SVM Results

C	Accuracy	kernel	Accuracy	gamma	Accuracy	degree	Accuracy
0.1	0.9100153538	linear	0.9114534523	scale	0.9165345236	2	0.9126438624
1	0.9123754788	poly	0.9034564326	0.1	0.2754256272	3	0.9165634524
2.3	0.9169647536	rbf	0.9169647536	1	0.2925732282	4	0.9215773753
10	0.9063457567	sigmoid	0.8474435275	10	0.2757456725	5	0.9365456796

Why SVM?

High-dimensional Data Handling

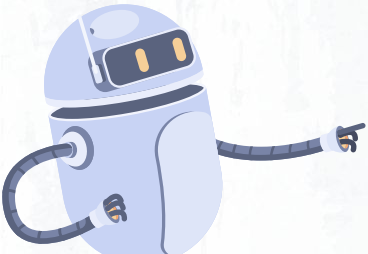
SVM excelled in managing the image data

Generalization

Results in effective classification

Efficiency in Multi-class Classification

Proved more adaptable and efficient



Thanks!



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Huang Wei Jhin

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