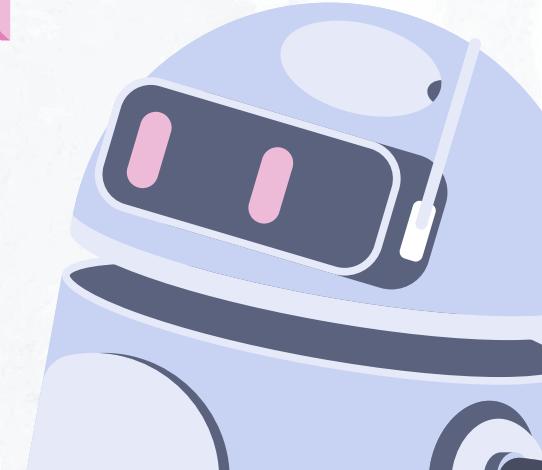


CSD3185

Group Project

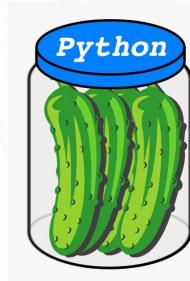
Image Classifier
Done by Team 5

Chiok Wei Wen Gabriel
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:



[christiansafka/img2vec](#)

🔥 Use pre-trained models in PyTorch to extract vector embeddings for any image

8

Contributors

2

Issues

564

Stars

91

Forks



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

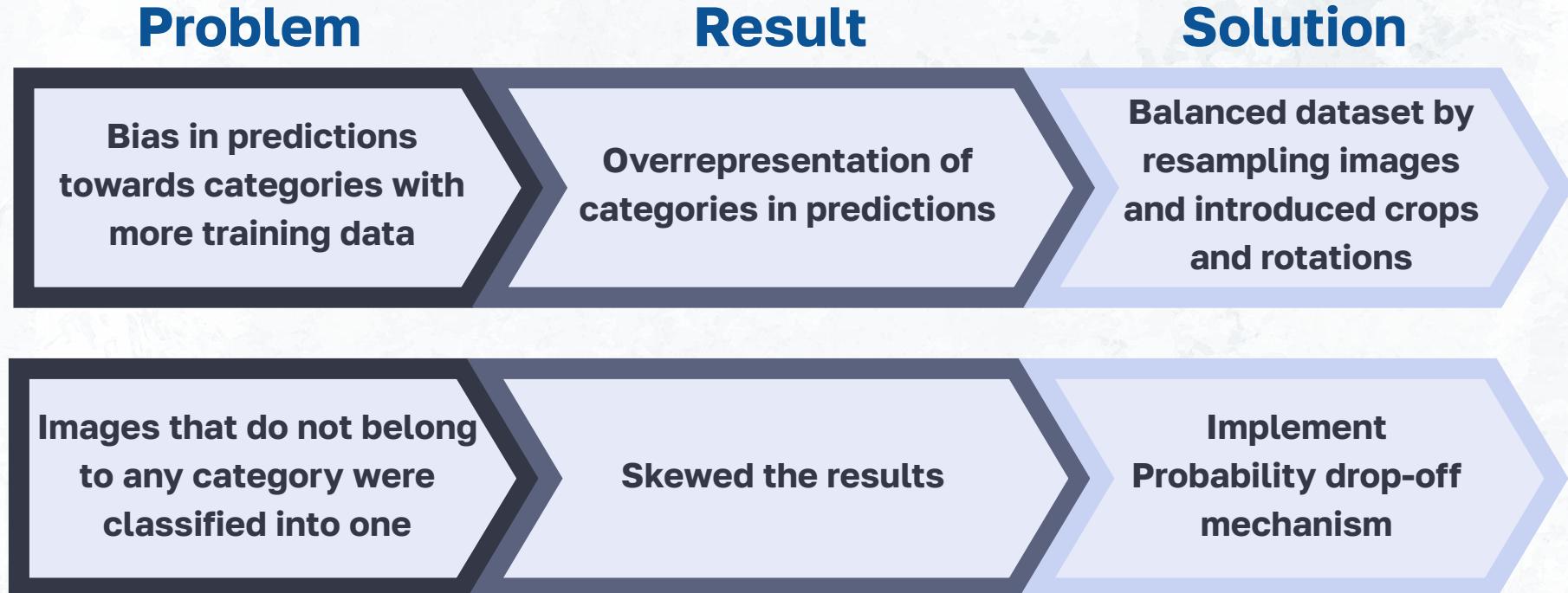
Converted images to grayscale to remove color data



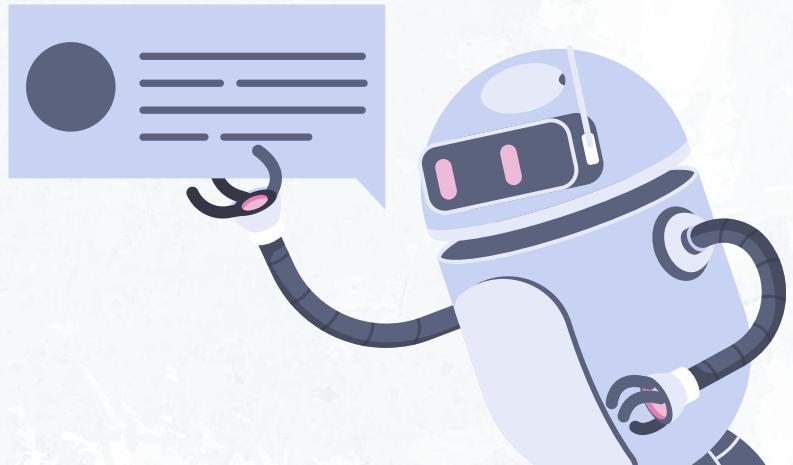
Histogram Equalization

Modified image contrast to spread the most common intensity values

Problems Encountered



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 |

| weights | Accuracy |
|----------|--------------|
| uniform | 0.8468518519 |
| distance | 0.8254355926 |

| algorithm | Accuracy |
|-----------|--------------|
| auto | 0.8468518519 |
| ball_tree | 0.8568345852 |
| kd_tree | 0.8109934528 |
| brute' | 0.7998200019 |

| p | Accuracy |
|-----------|--------------|
| Manhattan | 0.8513345852 |
| Euclidean | 0.8575482388 |

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

Random Forest Results

| n_estimators | Accuracy |
|--------------|--------------|
| 10 | 0.8858642079 |
| 100 | 0.9509638274 |
| 500 | 0.8957218936 |
| 1000 | 0.8924398746 |

| max_depth | Accuracy |
|-----------|--------------|
| 1 | 0.7875652901 |
| 5 | 0.8775215498 |
| 10 | 0.8875783057 |
| 20 | 0.8854691285 |

| min_samples_split | Accuracy |
|-------------------|--------------|
| 2 | 0.8775573929 |
| 10 | 0.8975632846 |
| 20 | 0.8957186349 |
| 50 | 0.8775718635 |

| max_features | Accuracy |
|--------------|--------------|
| sqrt | 0.8895673245 |
| log2 | 0.9057218936 |

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

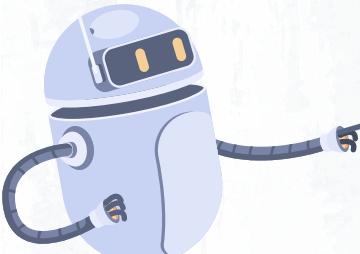
SVM excelled in managing the image data

Generalization

Results in effective classification

Efficiency in Multi-class Classification

Proved more adaptable and efficient



Thanks! →

Chiok Wei Wen Gabriel

Yin Shengkai

Jed Goh Yujie

Seow Kai Jun

Dennys Tay Khaj Tjong

Huang Wei Jhin

CREDITS: This presentation template was created by [Slidesgo](#) and includes icons by [Flaticon](#), infographics & images by [Freepik](#) and content by [Eliana Delacour](#)

