



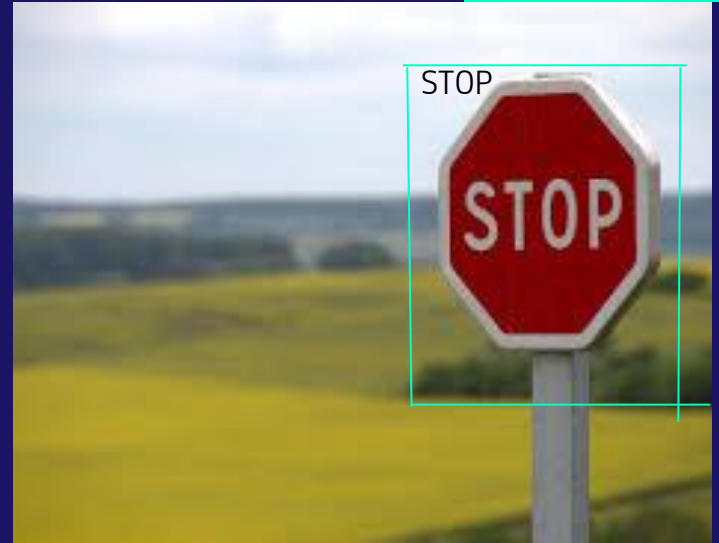
TRAFFIC SIGNS CLASSIFICATION

GUIDE - PROF. SWATI HIRA

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INTRODUCTION

- Trending Technology.
- Basic building block of self driving cars.
- Safety and Accident prevention.
- Area of research.





“Self-Driving cars are the natural extension of active safety and obviously something we should do.”

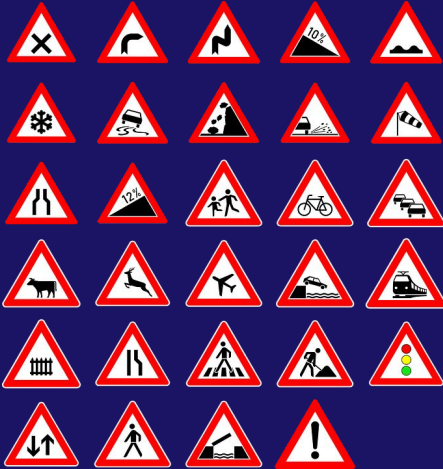
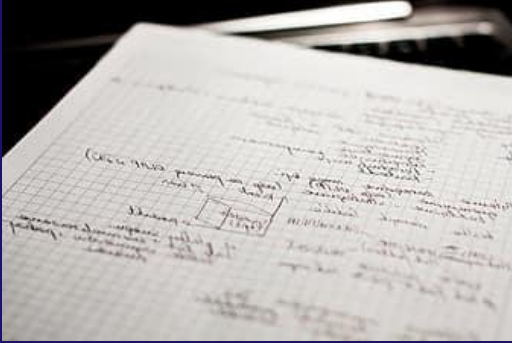
–ELON MUSK



01

Literature Survey and Data Collection

Datasets



GTSRB
Dataset



Research
Papers



Malaysian
Dataset



Dataset for
YOLO-V3

DATASET



900
Images

Prohibitory

Signs which impose restrictions.

Danger

Signs that denote danger.

Mandatory

Regular signs for general purpose.

Others

Pluto is now a dwarf planet



02

Planning and Methodology

TIMELINE

Step 1

Literature Survey and
data collection

Step 2

Planning of project
structure

Step 3

Understanding Yolo-V3
and Darknet
Architecture

Step 4

Learning R-CNN and
object detection

Step 5

Training our custom
data and building a
model

Step 6

Integrating it with a
GUI

TECH-STACK



Python

Base language for coding purpose.



YOLO-V3 and Darknet

For implementing Real time Object Detection.

PyQt5

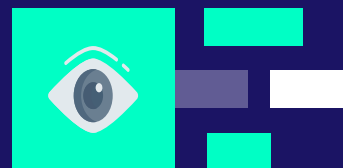
Designing of Graphical User Interface.



0

penCV

For interacting with images, videos and webcam.





03

Yolo-V3 and Darknet Architecture

YOLO-V3

	Type	Filters	Size	Output
	Convolutional	32	3×3	256×256
	Convolutional	64	$3 \times 3 / 2$	128×128
1x	Convolutional	32	1×1	128×128
	Convolutional	64	3×3	
	Residual			
	Convolutional	128	$3 \times 3 / 2$	64×64
2x	Convolutional	64	1×1	64×64
	Convolutional	128	3×3	
	Residual			
	Convolutional	256	$3 \times 3 / 2$	32×32
8x	Convolutional	128	1×1	32×32
	Convolutional	256	3×3	
	Residual			
	Convolutional	512	$3 \times 3 / 2$	16×16
8x	Convolutional	256	1×1	16×16
	Convolutional	512	3×3	
	Residual			
	Convolutional	1024	$3 \times 3 / 2$	8×8
4x	Convolutional	512	1×1	8×8
	Convolutional	1024	3×3	
	Residual			
	Avgpool		Global	
	Connected		1000	
	Softmax			

Figure 1. Architecture of YOLO-3

- YOLO stands for “You Only Look Once” and uses convolutional neural networks (CNN) for object detection.
- When YOLO works it predicts classes’ labels and detects locations of objects at the same time. That is why, YOLO can detect multiple objects in one image.
- YOLO divides image into regions, predicts bounding boxes and probabilities for every such region.
- YOLO also predicts confidence for every bounding box showing information that this particular bounding box actually includes object, and probability of included object in bounding box being a particular class.

WHAT IS DARKNET?





04

Preprocessing Dataset

Task at Hand

- Convert Images from .ppm to .jpg
- Convert annotation file to YOLO
- Create .txt file with annotations next to every image that will have the same names as images files.



04

Training Model on custom data

DATA FILE

EVERY DATA FILE CONTAINS:

- Number of classes
- Full path to train.txt file
- Full path to test.txt file
- Full path to classes.names file.
- Full path to location where trained weights are to stored.

CONFIGURATION FILE

- Parameters for Training file
 - Learning Rate
 - Angle and Saturation
 - Exposure and Hue
- Structure of CNN Layers
 - Filters
 - Activation
 - Size and stride
- Last 3 YOLO layers

SOME ADDITIONAL CONFIGURATIONS

- $\text{FILTERS} = (\text{classes} + \text{coordinates} + 1) * \text{masks}$,
 $= (4 + 4 + 1) * 3 = 27$
- $\text{MAX_BATCHES} \rightarrow$ (Total number of iterations for training.)
 $= \text{classes} * 2000$ (but not less than 4000)
 $= 4 * 2000 = 8000$
- $\text{STEPS} \rightarrow$ (Number of iterations after which learning rate will be multiplied by scale factor.)
 $= 80\% \text{ of MAX_BATCHES, } 90\% \text{ of MAX_BATCHES}$
 $= 6400, 7200$
- $\text{BATCH} \rightarrow$ (Number of images that will be processed during One iteration.)
 $= 64$

PRELIMINARY TRAINED
WEIGHTS



TRANSFER LEARNING

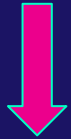


DECREASES
TRAINING TIME

READING RGB IMAGE



GETTING BLOB



LOADING YOLO V3
NETWORK



IMPLEMENTING
FORWARD PASS



DRAWING BOUNDING
BOXES WITH LABELS



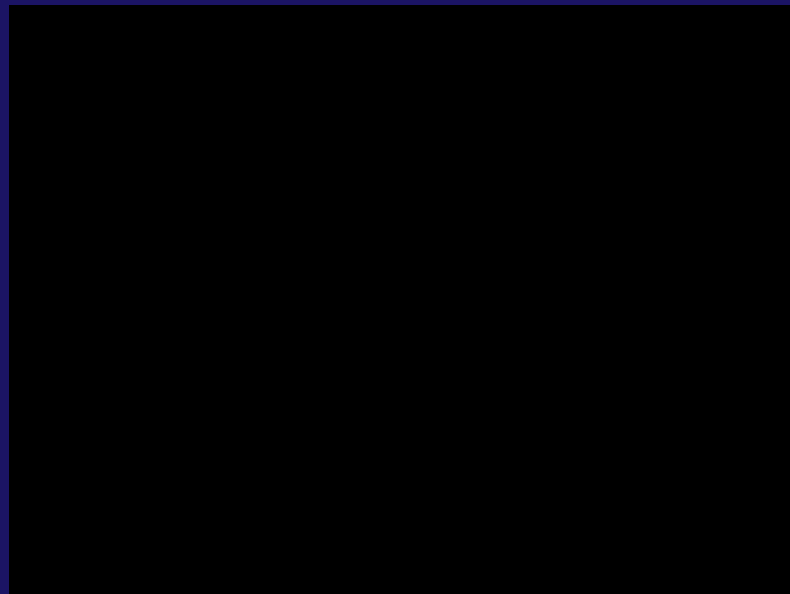
NON-MAXIMUM
SUPPRESSION



GETTING BOUNDING
BOXES

COMPLETE IDEA

SAMPLES





TO-DO

- Integrating the Trained Model with a GUI.
- Increase the number of classes for prediction.
- Train and test the model on Google Colab for Real-time classification.



```
..._setopt(comm, CURLOPT_URL, url);
...err, "Failed to set URL [%s]\n", errorbuf;
...;
..._easy_setopt(comm, CURLOPT_FOLLOWLOCATION,
...= CURLOPT_OK);
...err, "Failed to set redirect option [%s]\n";
...;
...;
..._easy_setopt(comm, CURLOPT_WRITEFUNCTION,
...= CURLOPT_OK);
...err, "Failed to set writer [%s]\n",
```

THANK YOU!

Questions are welcomed...

