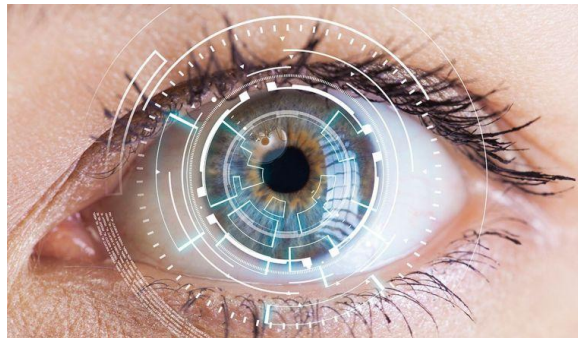# Computer Vision

APPLICATIONS
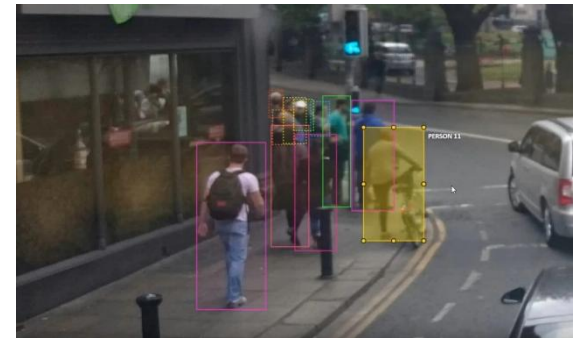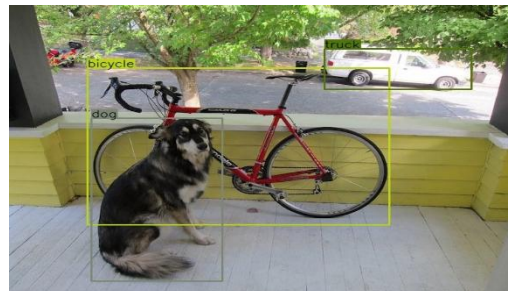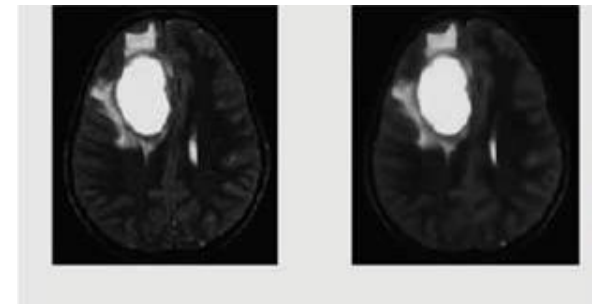

Self driving cars

Biometric Recognition

Image Annotation

Face Recognition

Object Detection

Tumour detection

DIGITAL IMAGE

A Digital image is a 2D matrix made up of small box units called pixel.

The numerical value for each pixel depicts its intensity.
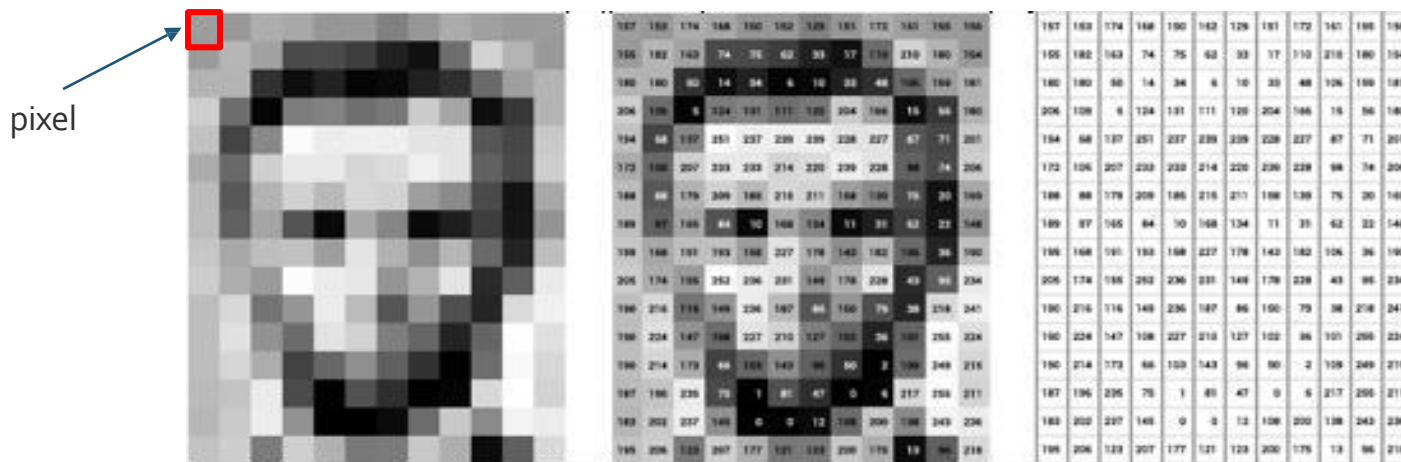
The intensity can range between 0-255.

pixel

Fig 1. Digital Image (Black and white) Representation

COLOR IMAGES

Colored image consist of three channels: Red, Green and Blue.

Each channel intensity value may range between 0 and 255.

Hence, each pixel is represented by three values. Different intensity values from these channels defines the object's color.
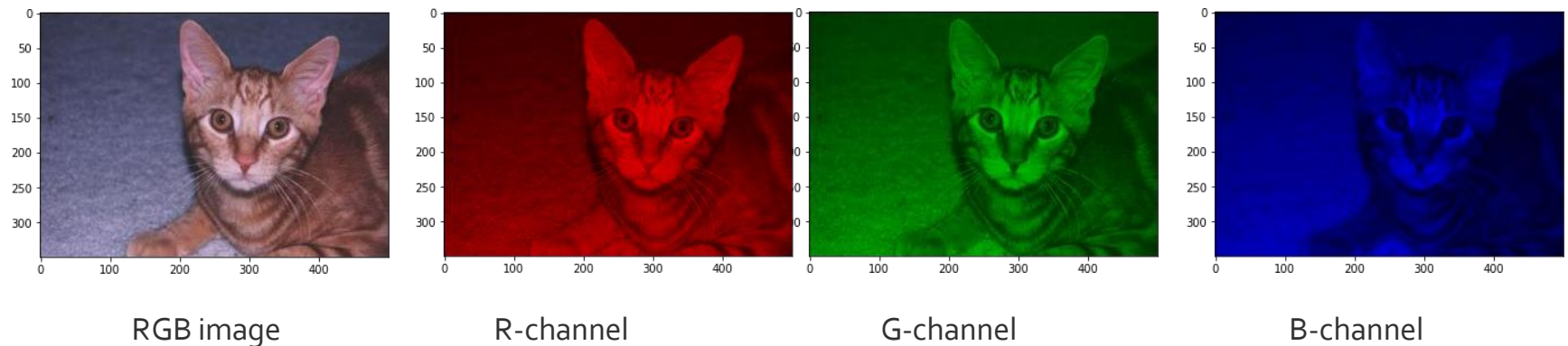

| RGB image | R-channel | G-channel | B-channel |

Fig. 2. Colored image and its R,G and B channels.
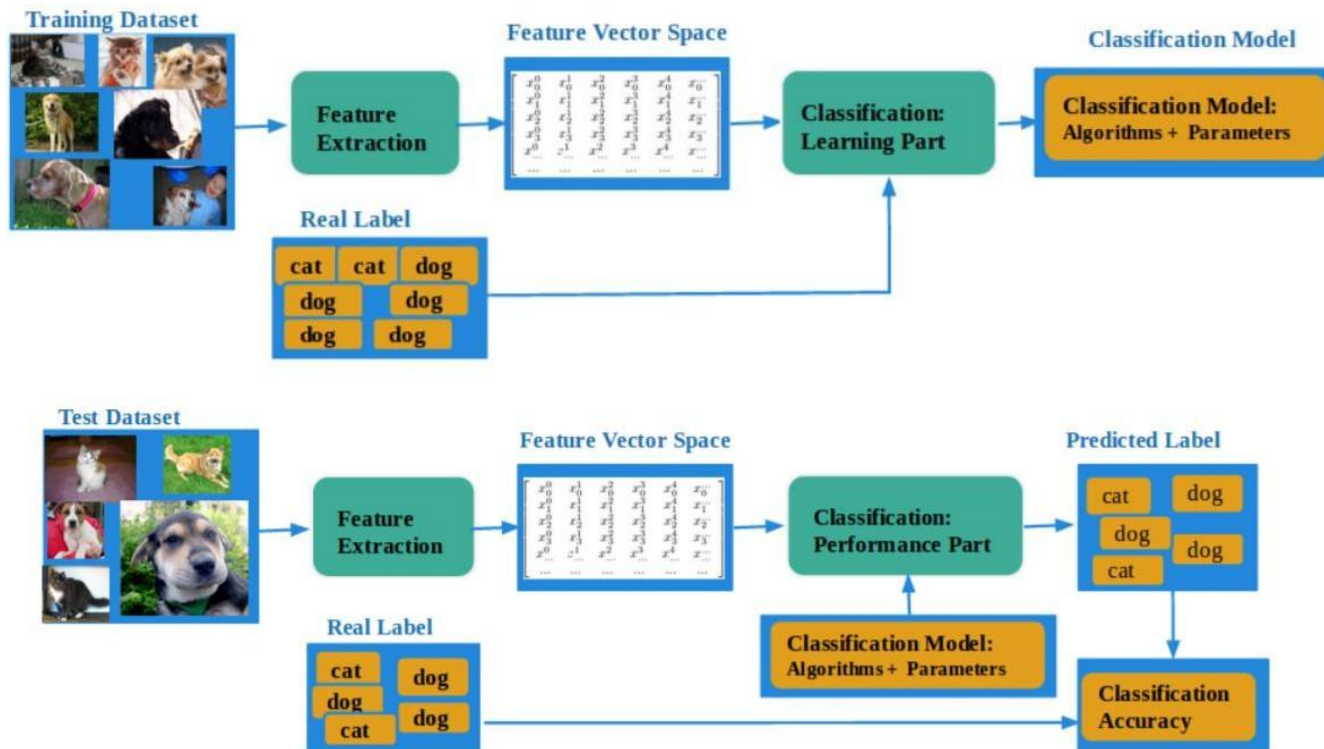
IMAGE CLASSIFICATION FRAMEWORK

Fig 4. Image classification Framework (a) Training framework (b) testing framework.

PYTHON LIBRARIES

OpenCV- Open computer vision used for image processing and feature calculation.

Matplotlib-Plotting figures and display images

OS- accessing files and folders.

Numpy-Numerical Computations.

Tensorflow-Neural Network Implementations, feature normalization and one hot encoding.

Scikit-Learn-Splitting the data into training and testing datasets.

https://colab.research.google.com/drive/1DCoYR7DI7aKDDnh1rfWWOedzRdAg0Z51#scrollTo=3iy5Ln-d0PFI

DATASET PREPARATION

Data set preparation or image preprocessing consist of three steps:

1. Reading the images.

2. Resizing the images.

3. Transforming images to the features and assigning (one hot)labels.

4. Dividing the data to training and testing data set.

FEATURE CALCULATION

1. Gray-scale image pixel intensity values.

2. Object of interest selection gray-scale image pixel intensity values.

3. Scale invariant feature transform (SIFT) Features calculation.

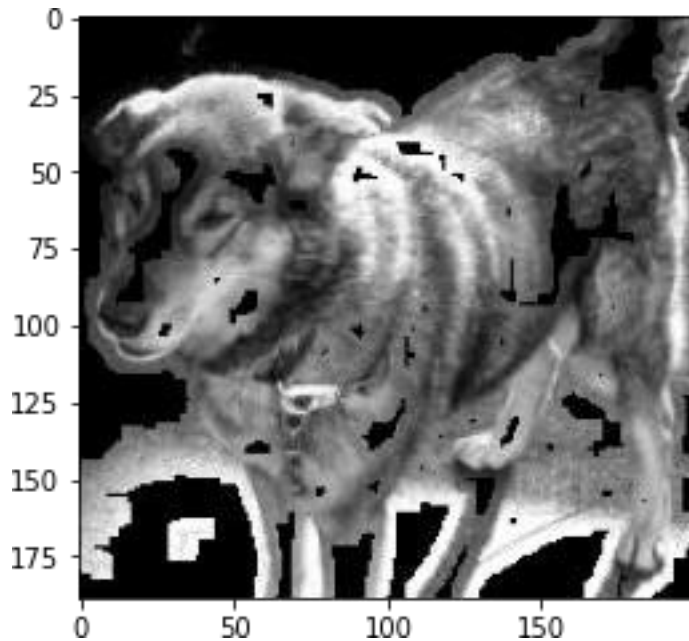4. Speed Up Robust Feature (SURF) Feature calculation.

IMAGE FLATTENING

Pooled Feature Map

Flattening

OBJECT OF INTEREST SELECTION
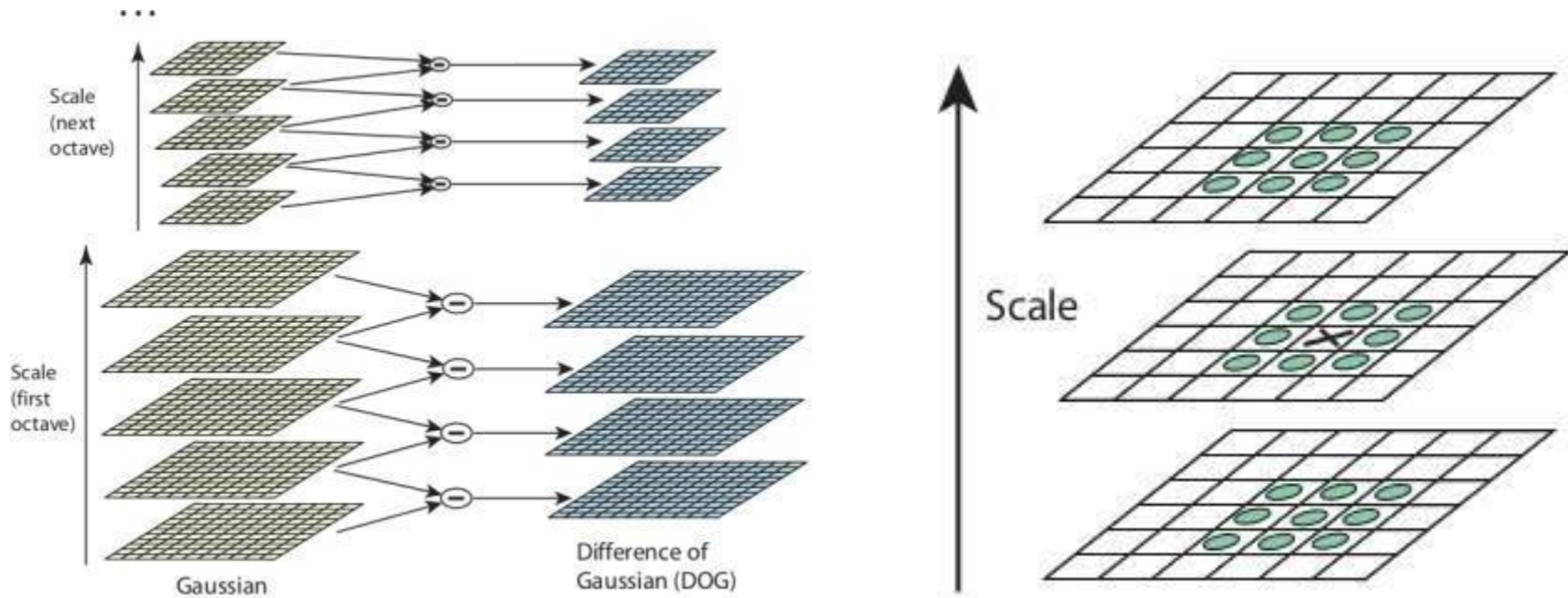
SIFT FEATURES

Gaussian — Difference of Gaussian (DOG)

Scale (next octave)

Scale (first octave)

Scale

https://docs.opencv.org/master/da/df5/tutorial_py_sift_intro.html

SURF FEATURES

L_{yy}

L_{xy}

can do match

can do match

no matching
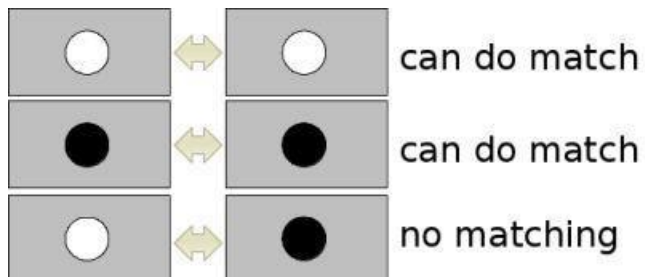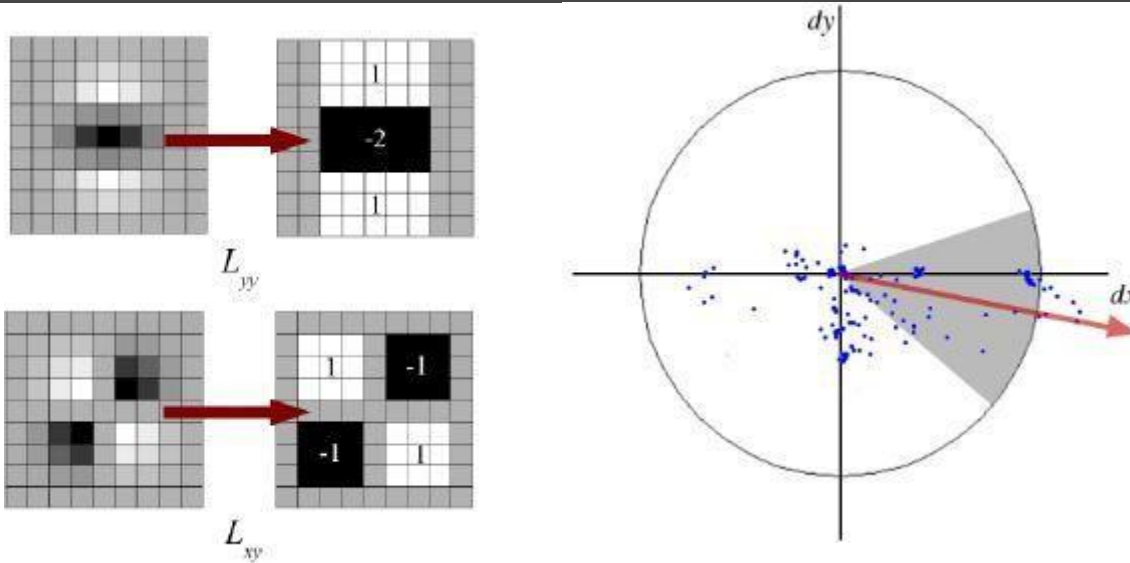
https://docs.opencv.org/4.5.2/df/dd2/tutorial_py_surf_intro.html

ONE-HOT ENCODING

The labels are fed to the neural network architecture as one-hot encoded labels. This can also be considered as binary representations of categorical labels for each sample.

In this type of encoding ,the sample belonging to particular class is assigned a value 1 and, 0 for other classes. For instance, imagine following are the labels for dog-cat classification.

Categories	Cat	Dog
Cat	1,	0
Dog	0,	1
Cat	1,	0
Dog	0,	1
Cat	1,	0

NEURAL NETWORK ARCHITECTURE

Input Layer-(Features dependent)

Hidden layers (8) –containing 2-2 layers of 256,128,64 and 32 neuron units respectively.

Output Layer- 2 neuron units.

The architecture is same for all scenarios.

NEURAL NETWORK PARAMETERS

Input Layer-(Features dependent)

Hidden layers (8) -256(2),128(2),64(2),32(2) neuron units.

Output Layer- 2 neuron units.

Dataset: Training 80%,validation 33% of training data, Testing 20%.

Optimizing Algorithm-Adam. **Loss**

function-Binary-cross entropy.

Metrics-Accuracy.

THANK YOU