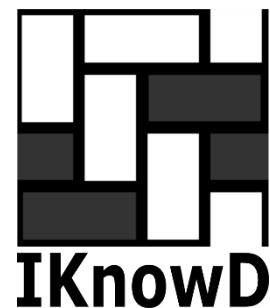




## MADEIRA INTERNATIONAL WORKSHOP IN MACHINE LEARNING



2021

# **IMAGE ANALYSIS USING FEED FORWARD DEEP NEURAL NETWORK**

# 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.

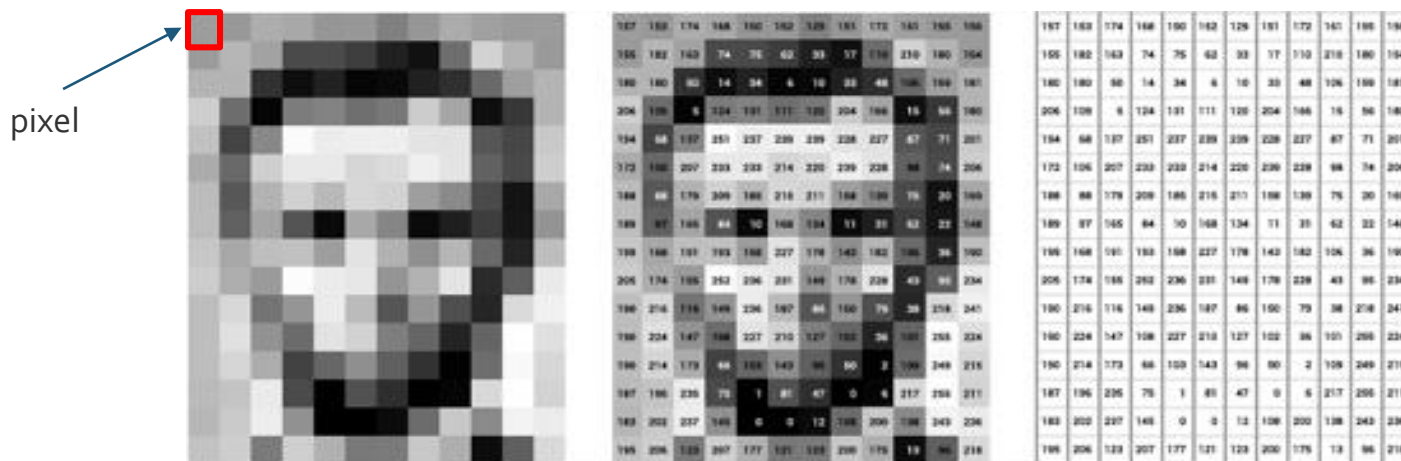


Fig 1. Digital Image (Black and white) Representation

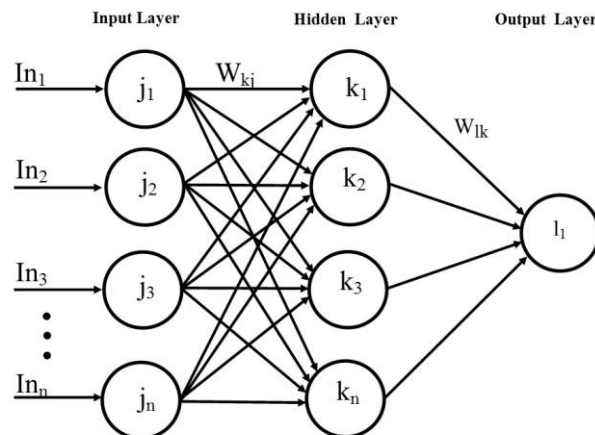
# NEURAL NETWORKS

## Artificial Neural Networks

One Input layer (dependent on the number of features).

One Hidden Layer (Multiple theories, hit and trial).

One Output Layer (Number of classes).

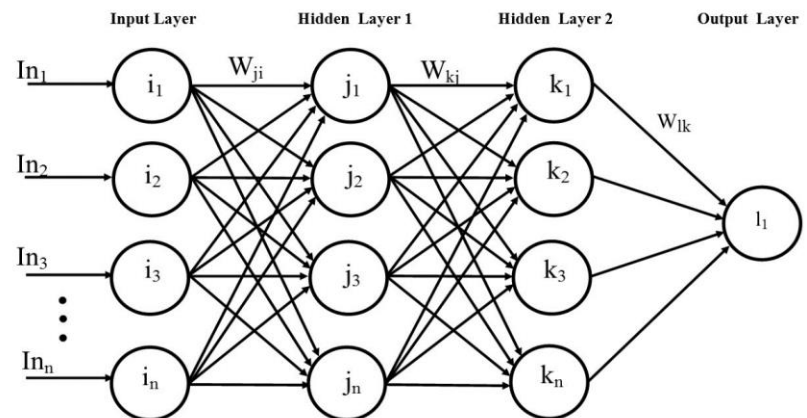


## Deep Neural Networks

One Input layer (dependent on the number of features).

Multiple Hidden Layers (Hierarchical feature learning).

One Output Layer (Number of classes).



# IMAGE CLASSIFICATION FRAMEWORK

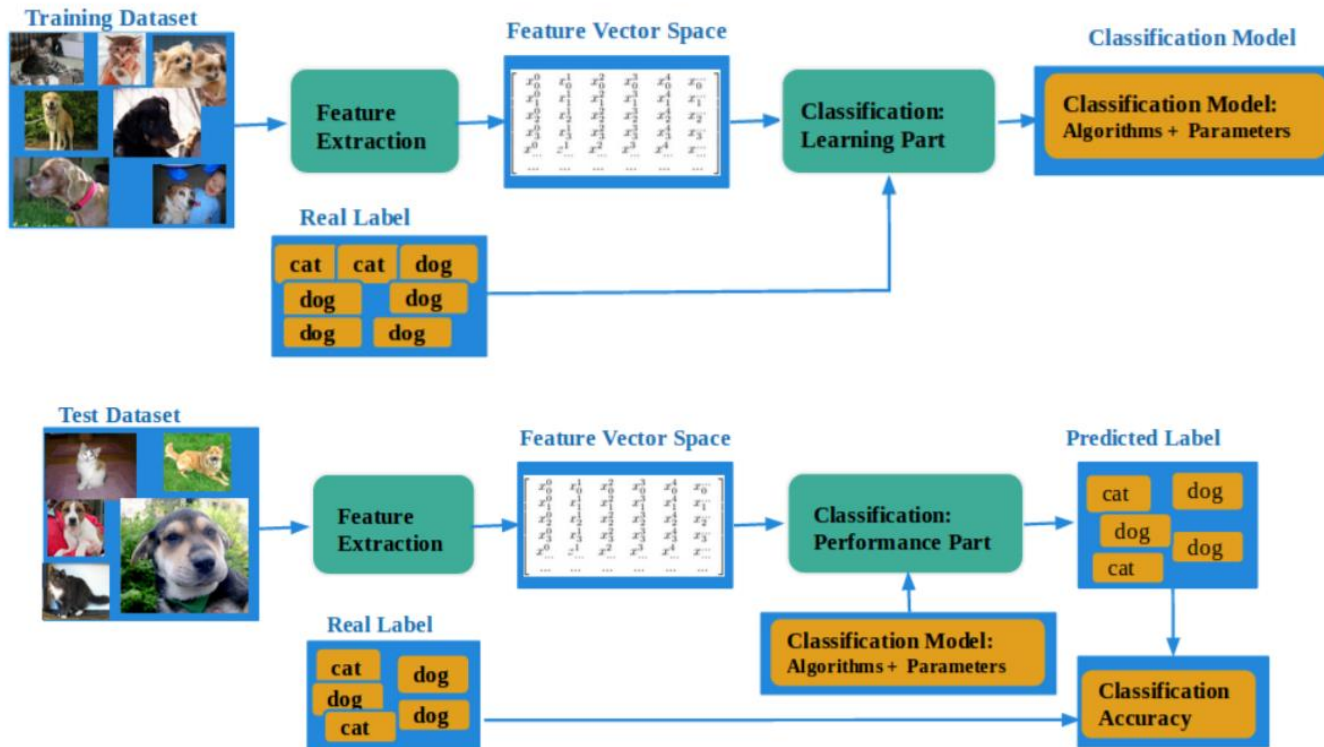


Fig 4. Image classification Framework (a) Training framework (b) testing framework.

# 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

1	1	0
4	2	1
0	2	1

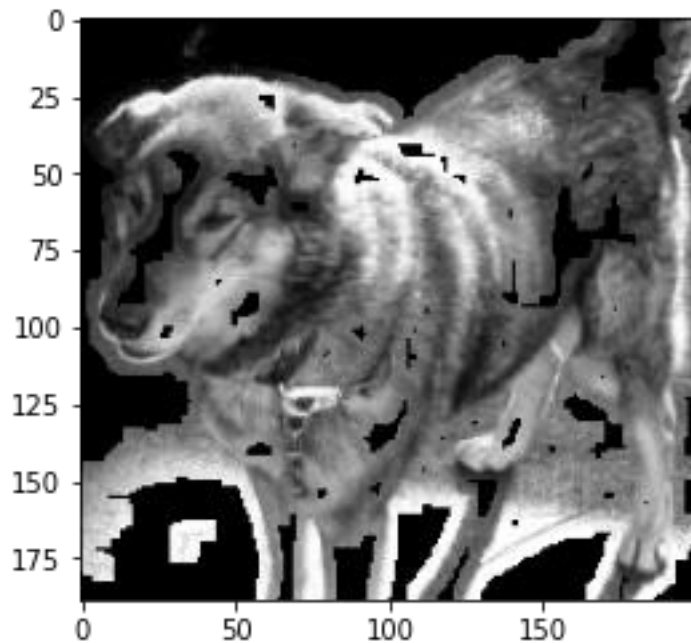
Pooled Feature Map

Flattening



1
1
0
4
2
1
0
2
1

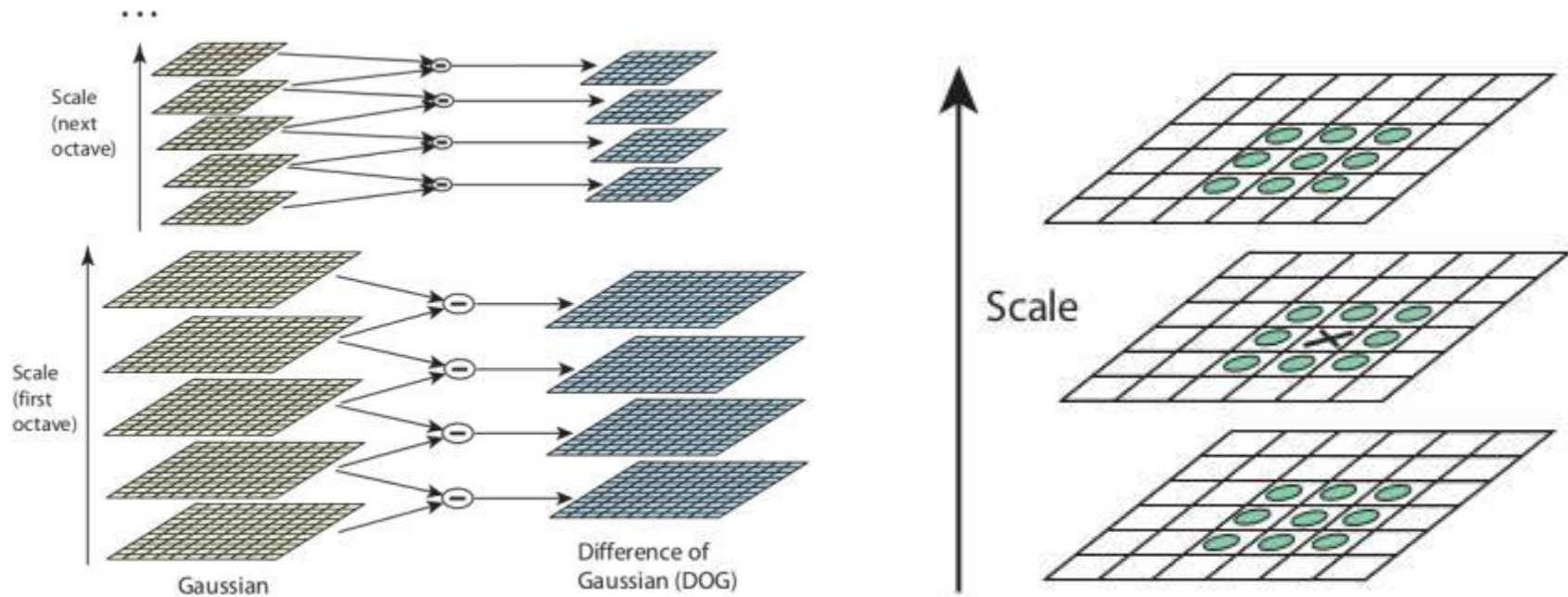
# OBJECT OF INTEREST SELECTION



Flattening

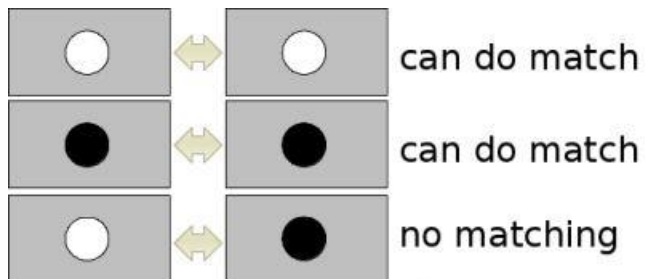
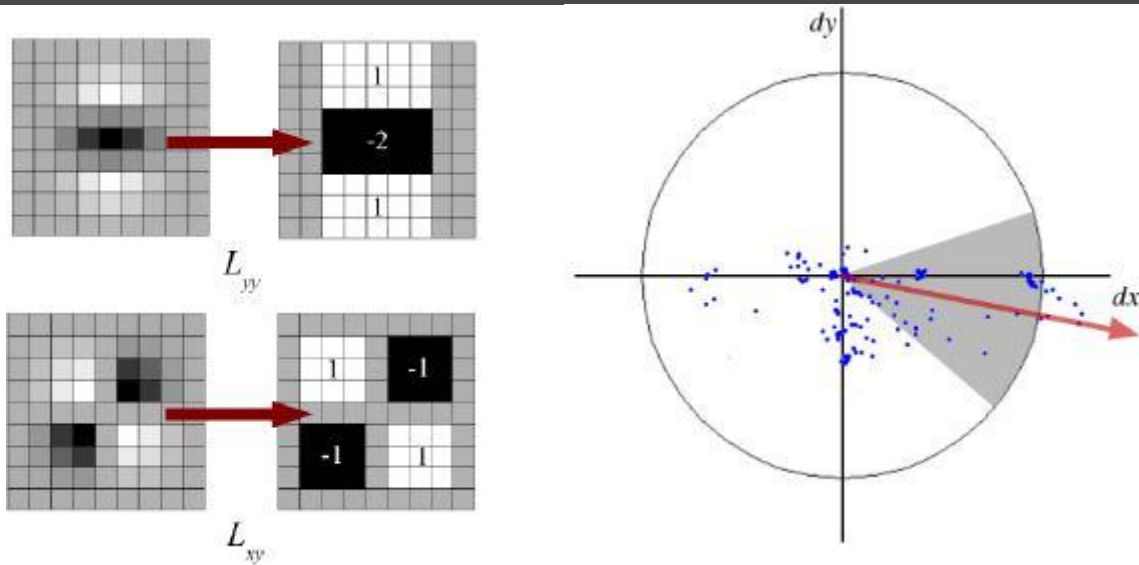
1
1
0
4
2
1
0
2
1

# SIFT FEATURES



[https://docs.opencv.org/master/da/df5/tutorial\\_py\\_sift\\_intro.html](https://docs.opencv.org/master/da/df5/tutorial_py_sift_intro.html)

# SURF FEATURES



[https://docs.opencv.org/4.5.2/df/dd2/tutorial\\_py\\_surf\\_intro.html](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**