

Machine Learning

Machine Learning Problems and Algorithms

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Goal of This Unit

- In this unit the four types of machine learning problems and algorithms are explained:
 - Classification/recognition/categorization
 - Regression
 - Clustering
 - Dimension reduction

References

- "Chapter 1 Introduction," Introduction to Machine Learning, 3rd, E. Alpaydin, MIT Press, 2014.
- "Chapter 1 Introduction," Machine Learning - An Algorithmic Perspective, S. Marsland, Chapman and Hall/CRC, 2009.

Contents

- Introduction
- Types of Machine Learning Algorithms
- ML Examples for Image Problems
- Feature Extraction
- Discussions

INTRODUCTION

Four Kinds of ML Problems

	<i>Supervised Learning</i>	<i>Unsupervised Learning</i>
<i>Discrete</i>	classification or categorization	clustering
<i>Continuous</i>	regression	dimensionality reduction

EX1: Scene Categorization / Image Classification

- Is this a kitchen?



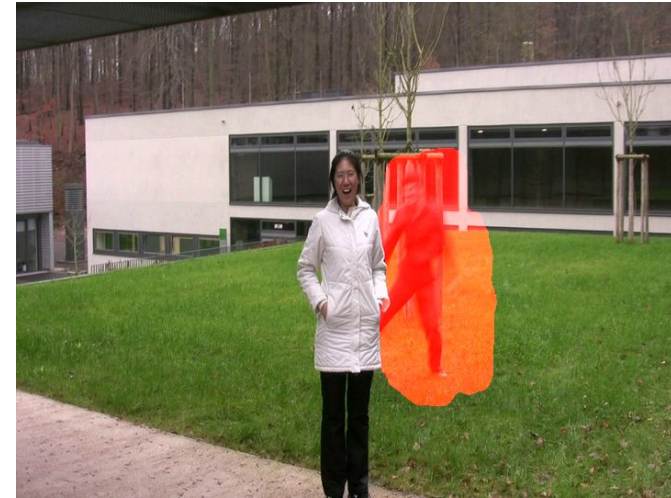
EX2: Image Segmentation



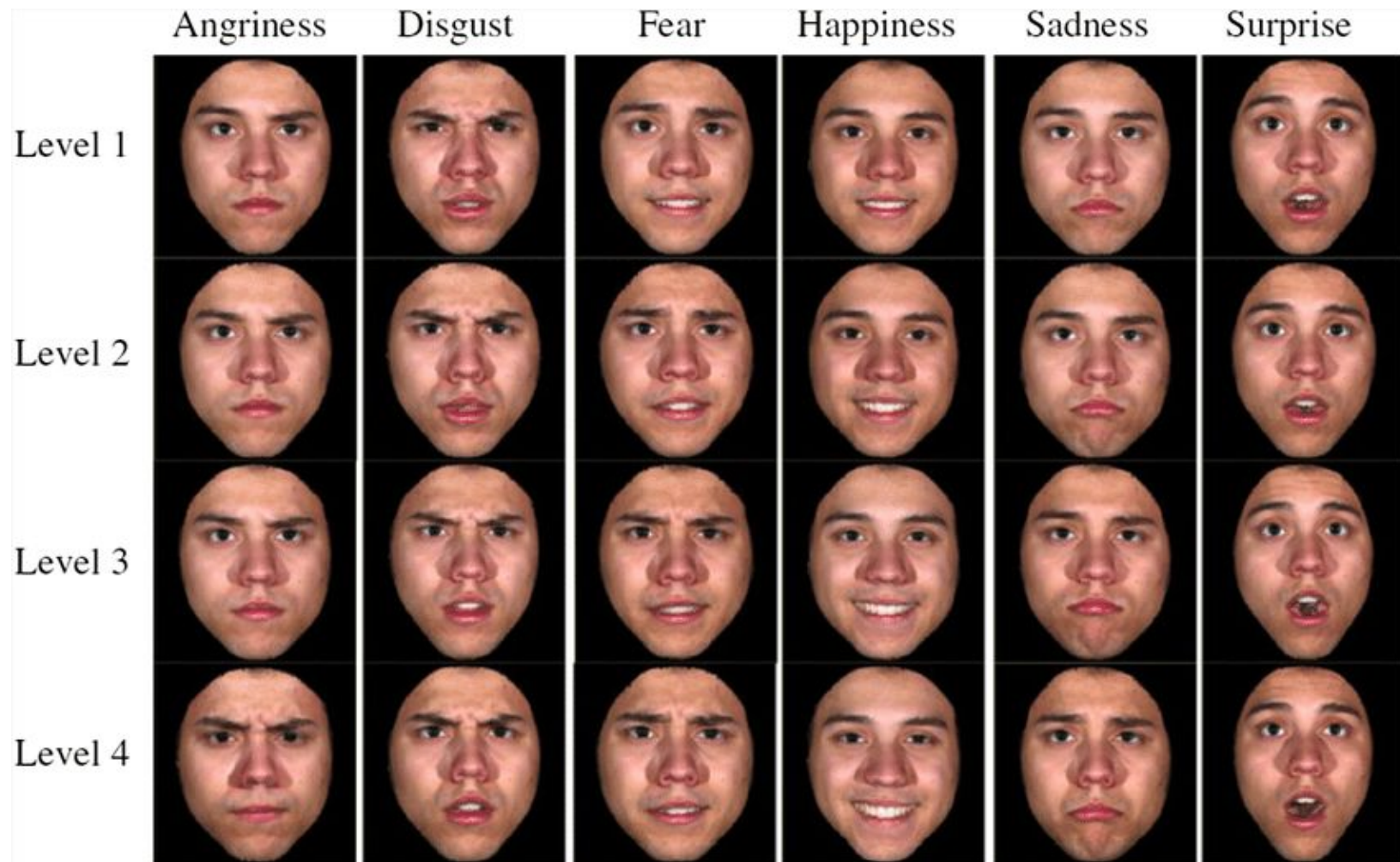
EX3: Image Superresolution



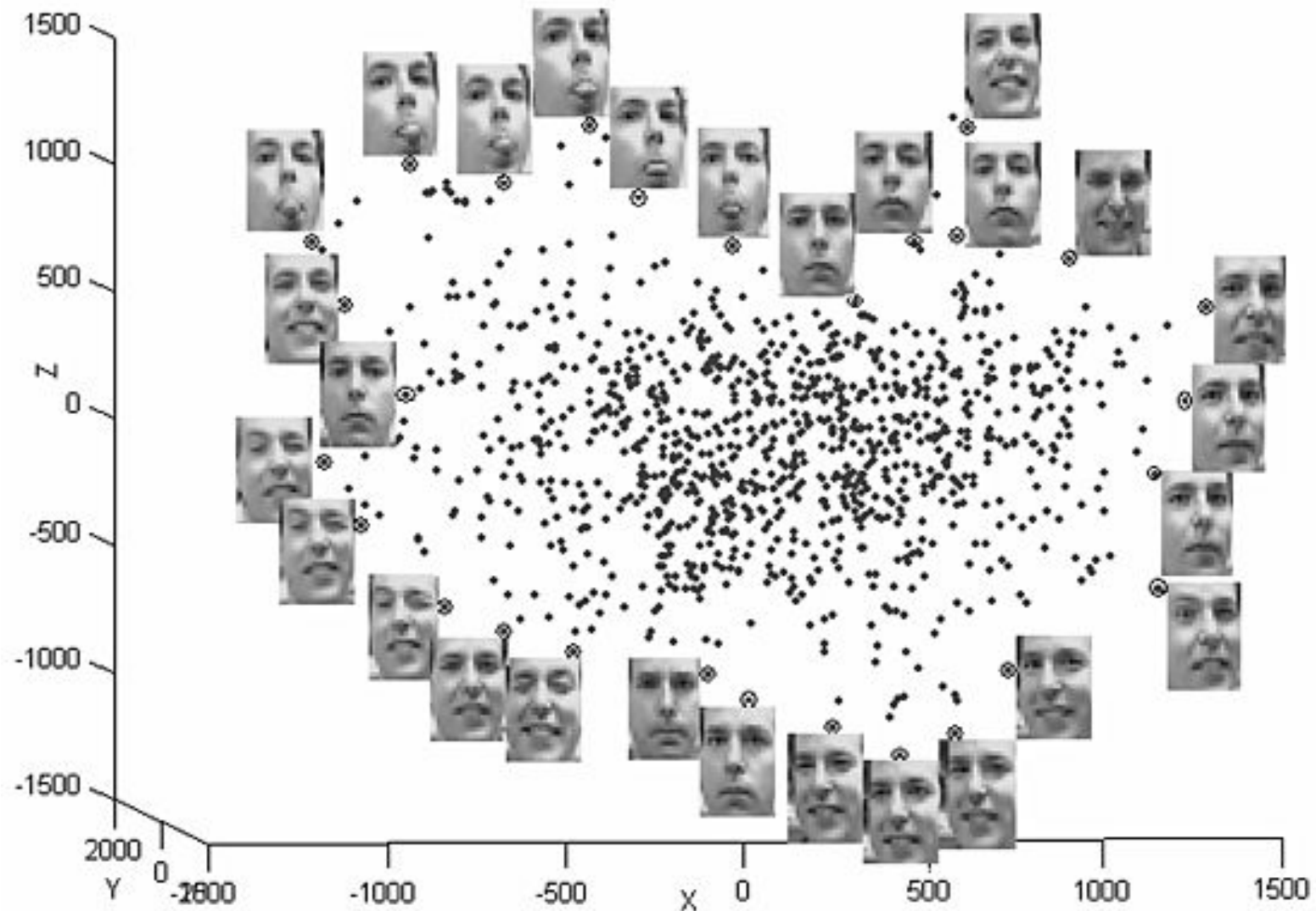
EX4: Inpainting



EX5: Face Expression Recognition



EX6: Face Expression Analysis



Machine Learning as a Function

- Apply a **prediction function** f to a **feature representation** x of the image to get the desired output y :

$$f(\text{apple image}) = \text{"apple"}$$

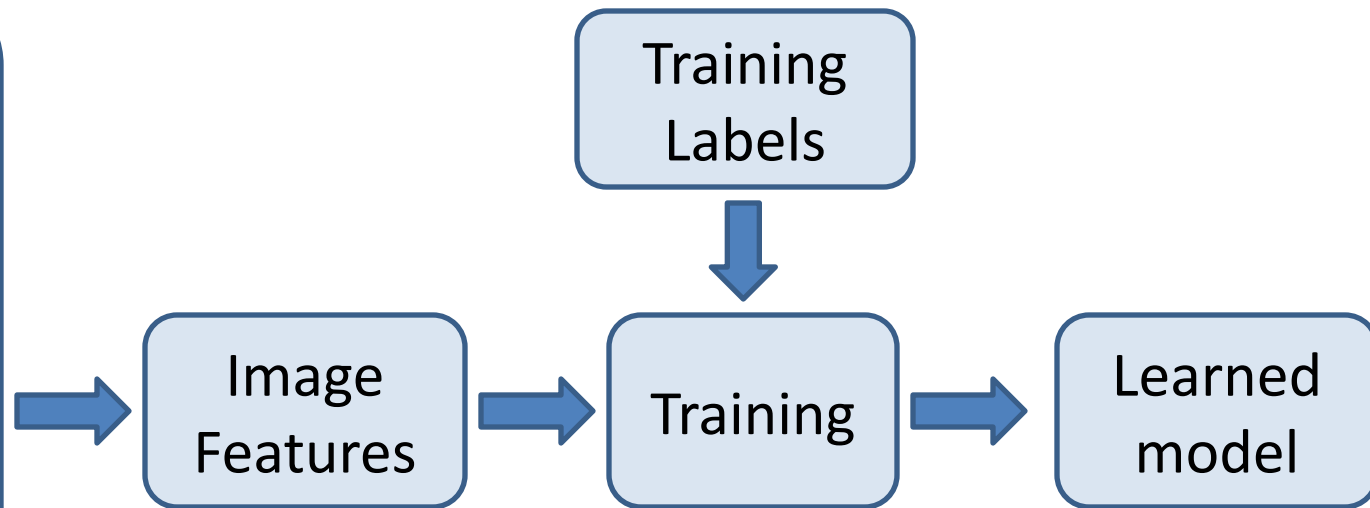
$$f(\text{tomato image}) = \text{"tomato"}$$

$$f(\text{cow image}) = \text{"cow"}$$

Machine Learning Steps

Training

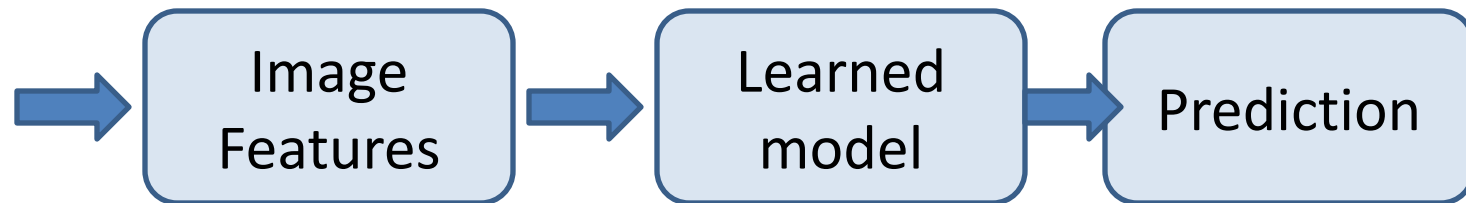
Training
Images



Testing



Test Image



Machine Learning Steps

Training

Training
Images



(x, y)

Training
Labels

Image
Features

Training

Learned
model

$f(x)=y$

Testing



Test Image

x

Image
Features

$f(x)=y$

Learned
model

y

Prediction

Machine Learning as a Function

$$y = f(x)$$

The diagram illustrates the function $y = f(x)$. Below the equation, three labels are positioned: 'output' under y , 'prediction function' under f , and 'Image feature' under x . Red arrows point from each label to its corresponding symbol in the equation: an arrow from 'output' to y , an arrow from 'prediction function' to f , and an arrow from 'Image feature' to x .

- **Training:** given a *training set* of labeled examples $\{(x_1, y_1), \dots, (x_N, y_N)\}$, estimate the prediction function f by minimizing the prediction error on the training set
- **Testing:** apply f to a never before seen *test example* x and output the predicted value $y = f(x)$

Machine Learning Algorithms

Supervised Learning

Unsupervised Learning

Discrete
Continuous

classification or
categorization

clustering

regression

dimensionality
reduction

Image Problems vs. ML Algorithms

- Scene categorization
- Image Segmentation
- Image superresolution
- Image inpainting
- Face expression recognition
- Face expression analysis

	<i>Supervised Learning</i>	<i>Unsupervised Learning</i>
<i>Discrete</i>	classification or categorization	clustering
<i>Continuous</i>	regression	dimensionality reduction

Supervised vs. Unsupervised Learning

Discrete vs. Continuous Outputs

TYPES OF ML ALGORITHMS

Classification

Learning with output label
Output label is discrete

	<i>Supervised Learning</i>	<i>Unsupervised Learning</i>
<i>Discrete</i>	classification or categorization	clustering
<i>Continuous</i>	regression	dimensionality reduction

Classification

Classes:

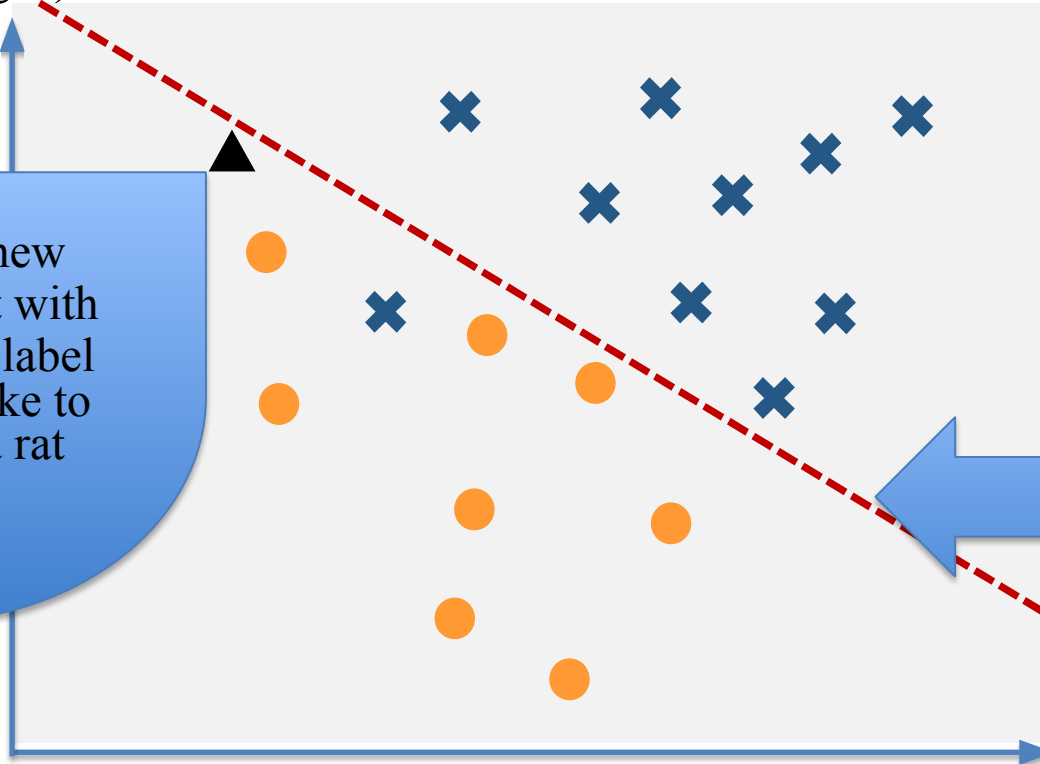
✕ = Rat
● = Mole

One feature
(e.g. weight)

- This is a new data point with unknown label
- We would like to know if it is a rat or a mole

One possible boundary – this would make our new data point a “mole”

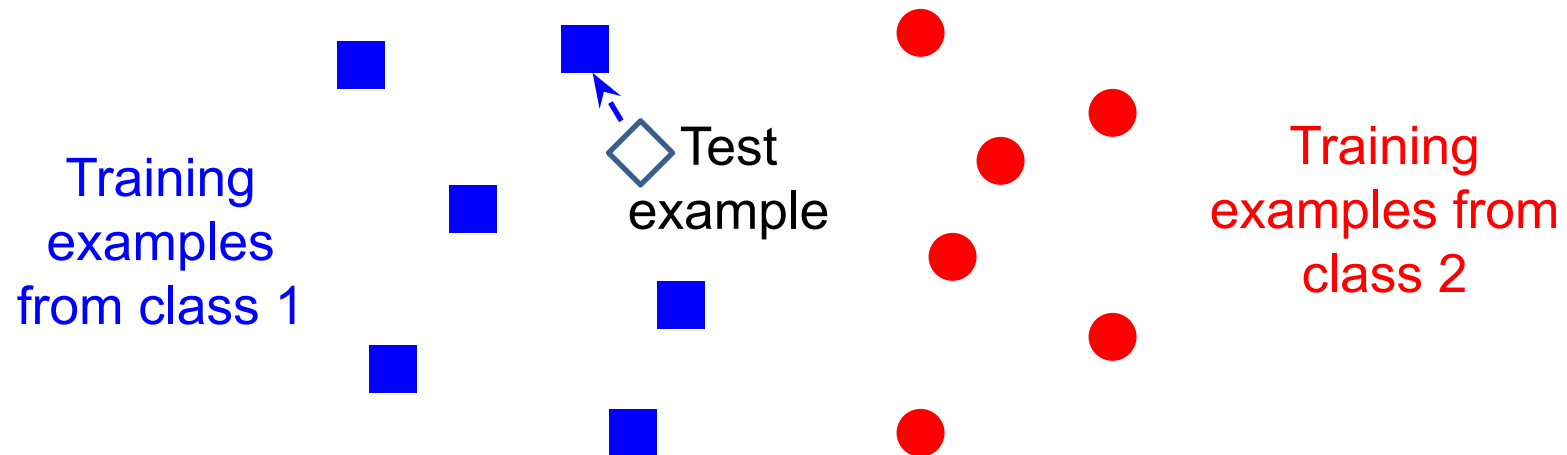
Another feature
(e.g. body length)



Classifiers: Nearest Neighbor

$f(x)$ = label of the training example nearest to x

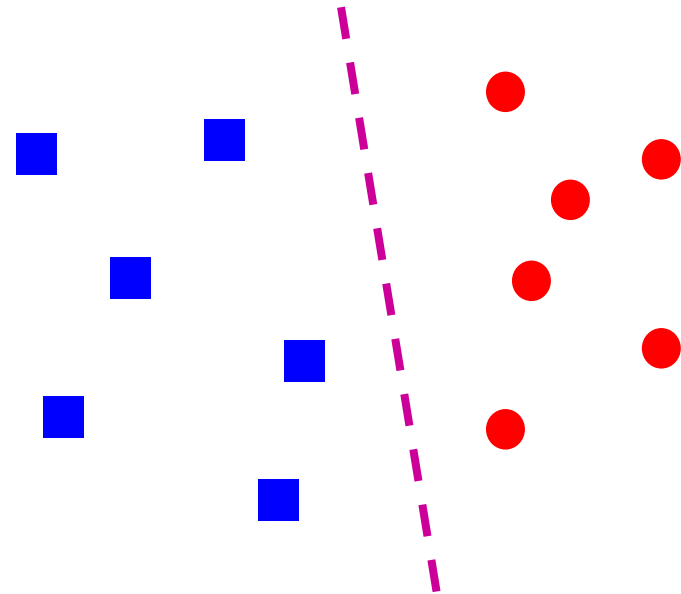
- All we need is a distance function for our inputs
- No training required!



Classifiers: Linear

Find a *linear function* to separate the classes:

$$\begin{cases} f(\mathbf{x}) = \mathbf{w} \cdot \mathbf{x} + b > 0 \\ f(\mathbf{x}) = \mathbf{w} \cdot \mathbf{x} + b < 0 \end{cases}$$



$$f(\mathbf{x}) = \text{sgn}(\mathbf{w} \cdot \mathbf{x} + b)$$

Algorithms to Find Classifier $f(x)$

- **K-nearest neighbor**
- **Decision tree**
- **SVM**
- **Logistic regression**
- **Neural networks**
- **Deep neural networks**
- **Naïve Bayes, Bayesian network**
- **Randomized Forests**
- ...

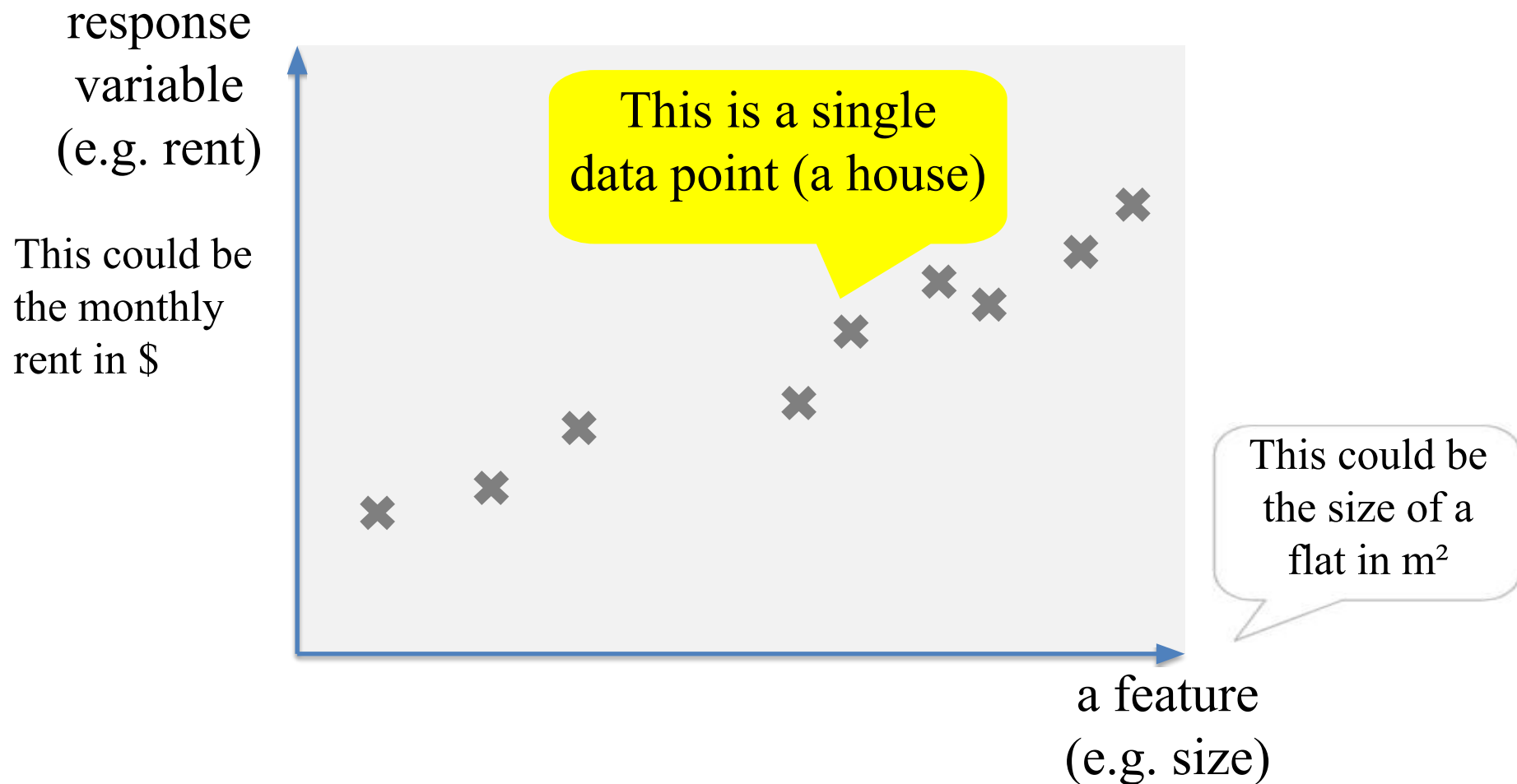
Regression

	<i>Supervised Learning</i>	<i>Unsupervised Learning</i>
<i>Discrete</i>	classification or categorization	clustering
<i>Continuous</i>	regression	dimensionality reduction

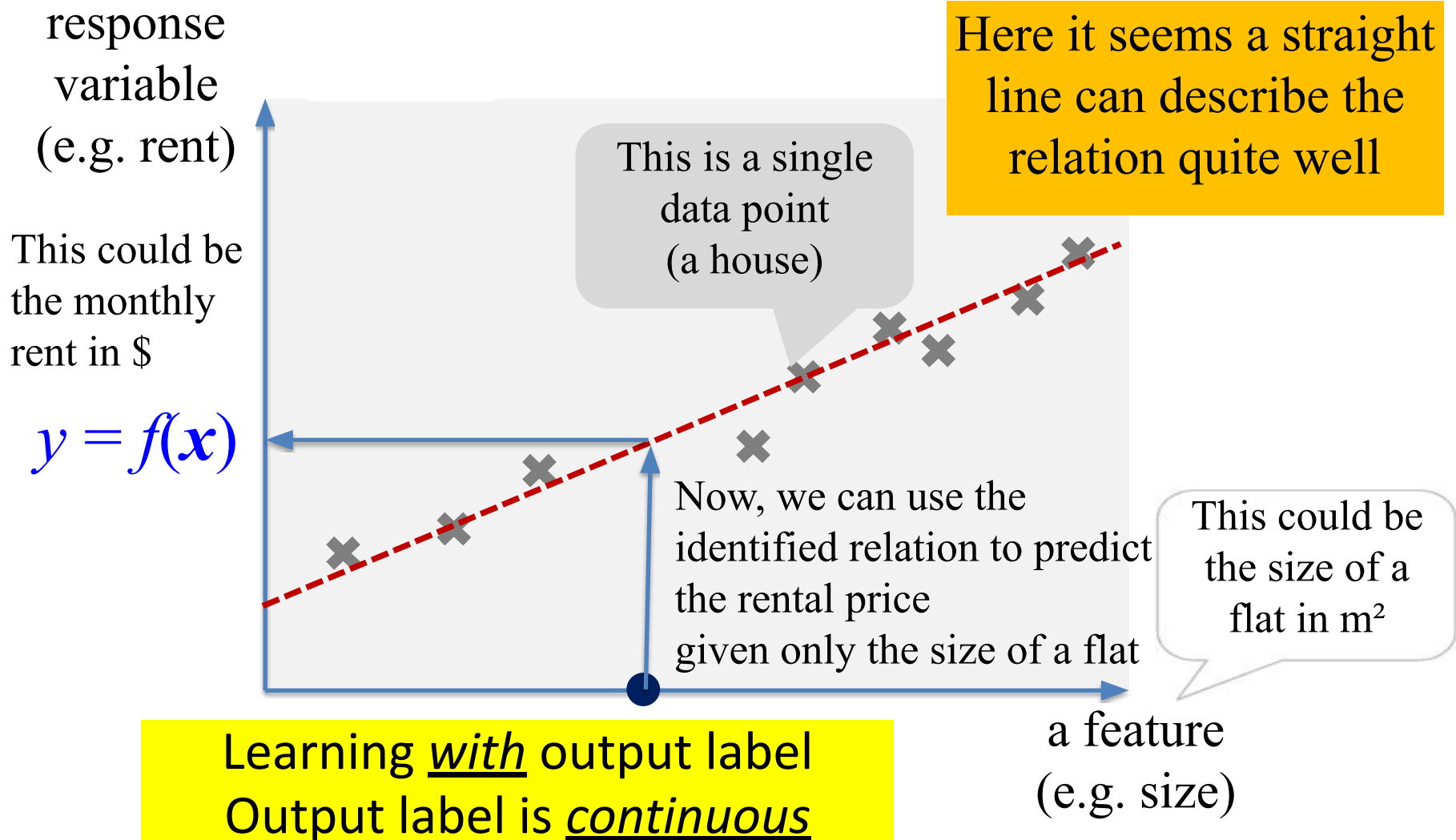
Learning with output label
Output label is continuous

Regression

Rent Price Prediction of Houses



Regression



Regression Algorithms to Find Regressor $f(x)$

- Linear regression
- Weighted regression
- Quadratic regression
- Polynomial regression
- Multivariate regression
- Nonlinear regression
- ...

Clustering

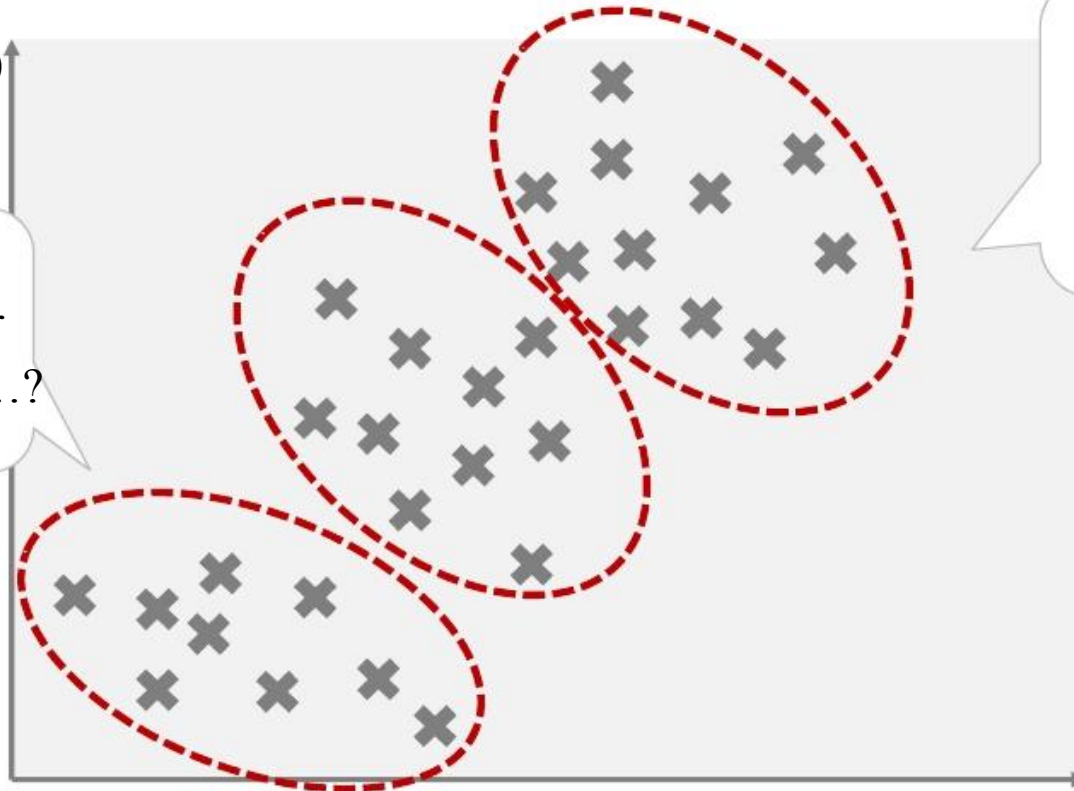
	<i>Supervised Learning</i>	<i>Unsupervised Learning</i>
<i>Discrete</i>	classification or categorization	clustering
<i>Continuous</i>	regression	dimensionality reduction

Learning without output label
Output label is discrete

Clustering

One feature
(e.g. customer age)

Maybe 3 when
we only consider
these 2 features...?



How many
customer groups
should we
distinguish?

Learning without output label
Output label is discrete

Another feature
(e.g. average ticket size)

Clustering Algorithms to Find Cluster $f(x)$

- K-means
 - Iteratively re-assign points to the nearest cluster center
- Agglomerative clustering
 - Start with each point as its own cluster and iteratively merge the closest clusters
- Mean-shift clustering
 - Estimate modes of pdf
- Spectral clustering
 - Split the nodes in a graph based on assigned links with similarity weights

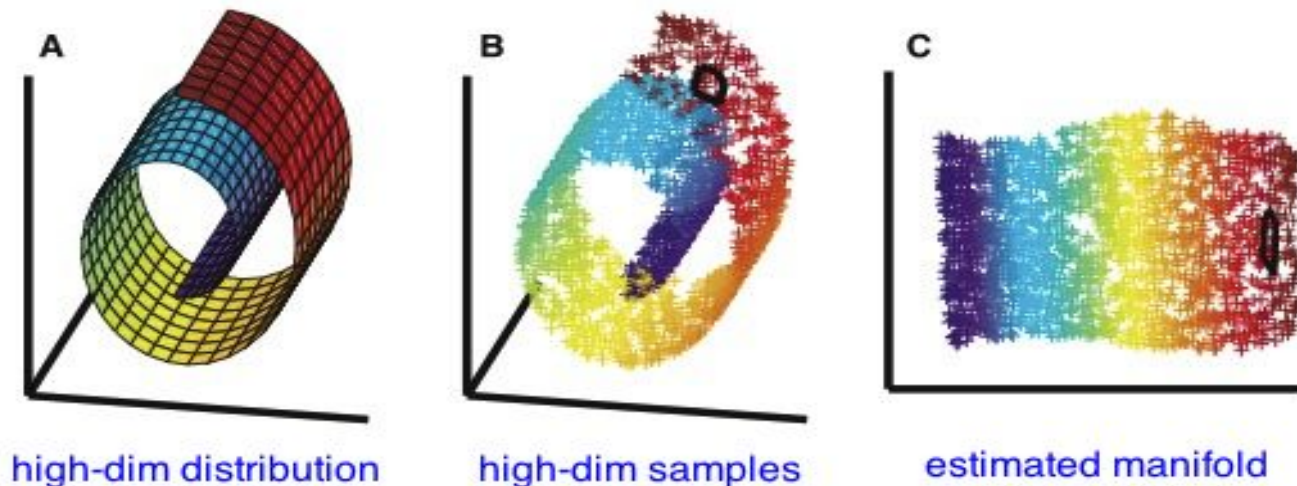
Dimension Reduction

	<i>Supervised Learning</i>	<i>Unsupervised Learning</i>
<i>Discrete</i>	classification or categorization	clustering
<i>Continuous</i>	regression	dimensionality reduction

Learning without output label
Output label is continuous

Dimensionality Reduction

- Reduce high-dimensional data into low-dimensional data for
 - Visualization, analysis, speed improvement



Learning without output label
Output label is continuous

Methods/Algorithms for Dimensionality Reduction

- **PCA (Principal component analysis)**
- ICA (Independent component analysis)
- LLE (Local linear embedding)
- Isomap
- Manifold learning
- ...

Overview of ML Problems

Machine Learning

Supervised Learning

Use the data set where the right answers (labels) are given for each example

Classification

Predict a discrete valued output (e.g. a label or a class)

Regression

Predict a continuous valued output

Unsupervised learning

Use unlabelled data to find peculiarities, similarities or structures

Clustering

Group similar examples into subsets called clusters

Dimensionality reduction

You don't need all the data and need to know the essence of the data

Reinforcement learning

Learn by maximizing your expected reward

ML Algorithms

**There are thousands of machine learning algorithms –
it is impossible to know and understand them all**

Decision trees
K-nearest neighbour (KNN)
Perceptron
Artificial Neural Networks (ANN)
Unsupervised Neural network models
("Restricted Boltzmann machines")
Deep belief networks
Random Forests
Linear Regression
Ordinary least squares (OLS)
Penalised regression
Principal Component Analysis (PCA)
Randomised PCA
Logistic Regression
(Linear / Quadratic) Discriminant Analysis
Support Vector Machines (SVM)
(Linear) Support Vector Classifier (SVC)
Support Vector Regression
Naive Bayes
K-means
Independent Component Analysis (ICA)
Non-negative matrix factorisation (NMF)

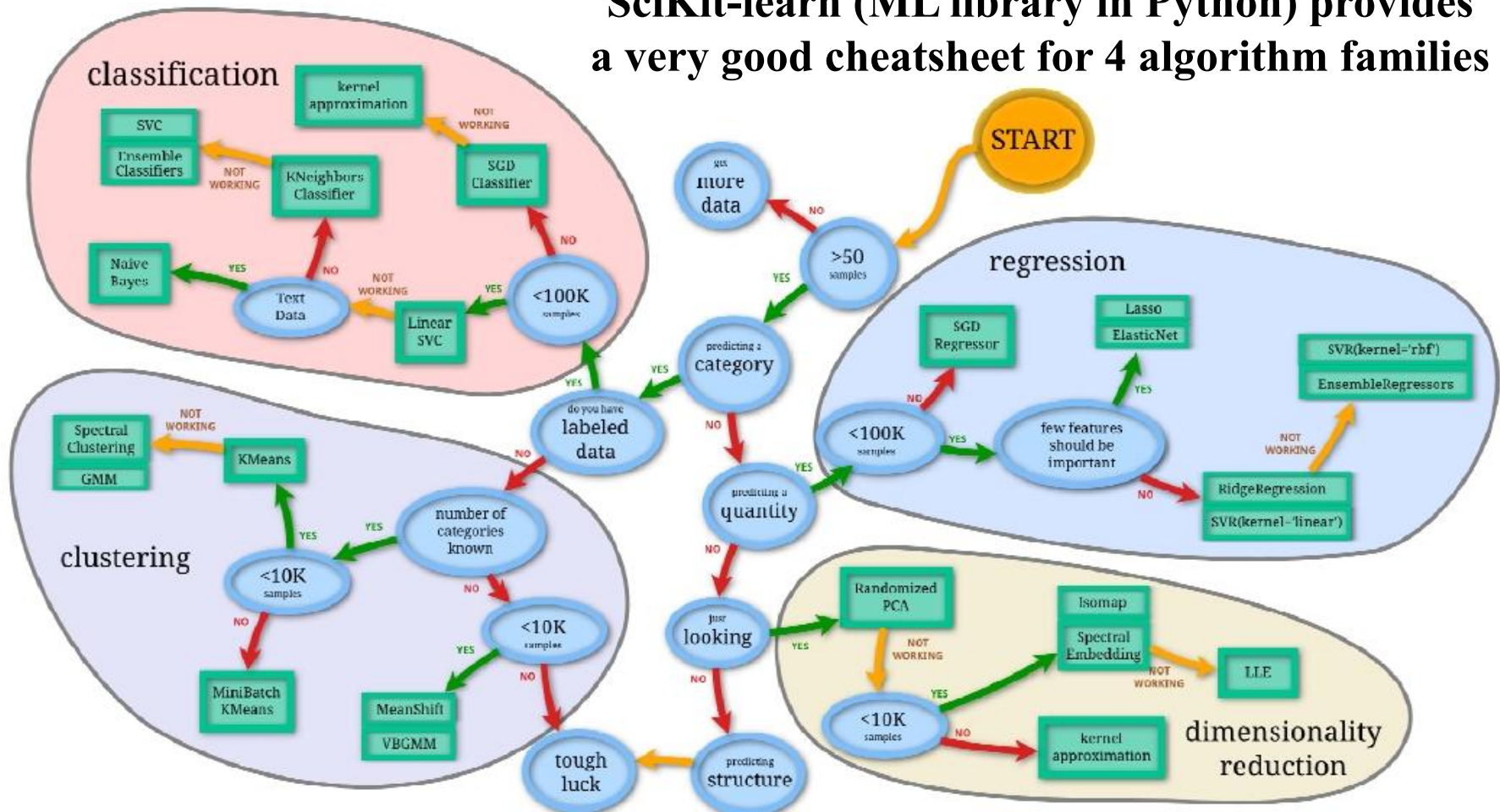
IsoMap
Association analysis
Hidden Markov Model
Kernel Approximation
MeanShift
Recurrent neural networks
Novelty and Outlier Detection
Density Estimation
Gaussian mixture models (GMM)
Manifold learning
Spectral Embedding ("Laplacian
Eigenmaps")
Deep Learning
Locally linear embedding (LLE)
Hessian-based LLE ("Hessian
Eigenmapping")
Multi-dimensional Scaling (MDS)
Bayes nets
Latent linear models
Sparse Bayesian Learning
Gaussian processes
CART

AdaBoost
LogitBoost
Polynomial Regression
State space models
Markov random fields
Convolutional neural networks
Conditional random fields (CRF)
Monte Carlo inference
Markov Chain Monte Carlo (MCMC)
inference
Latent variable models
Latent Dirichlet allocation (LDA)
(Linear) Stochastic gradient descent
(SGD) classifier
Gaussian Naive Bayes Classifier

...and thousands more...

ML Cheatsheet

SciKit-learn (ML library in Python) provides a very good cheatsheet for 4 algorithm families



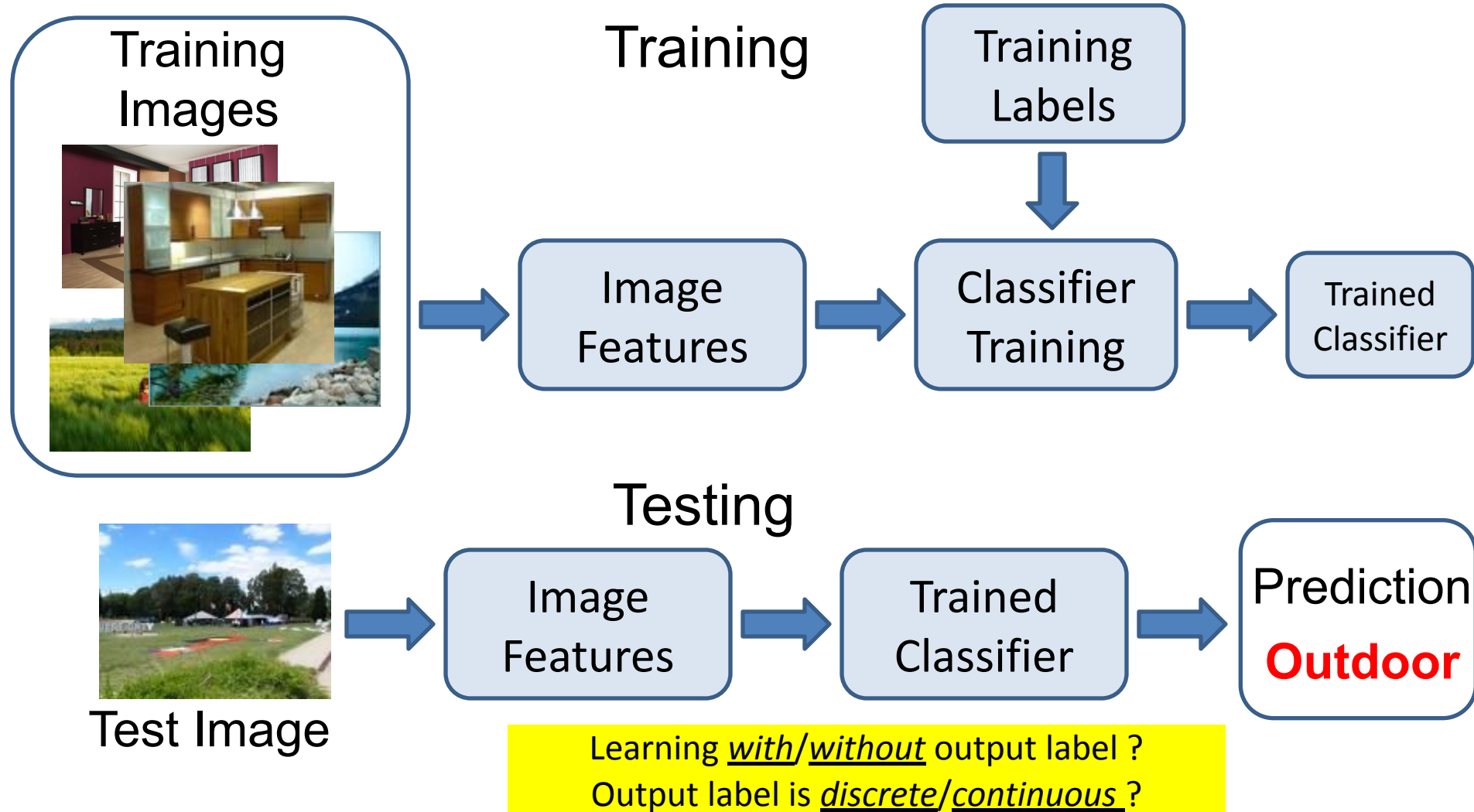
Source: http://scikit-learn.org/stable/tutorial/machine_learning_map/

Image Categorization
Image Segmentation
Image Superresolution
Image Inpainting
Face Recognition & Analysis

ML EXAMPLES FOR IMAGE PROBLEMS

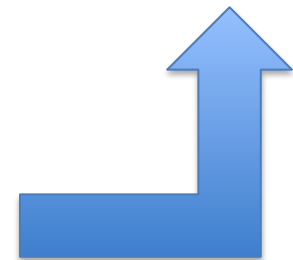
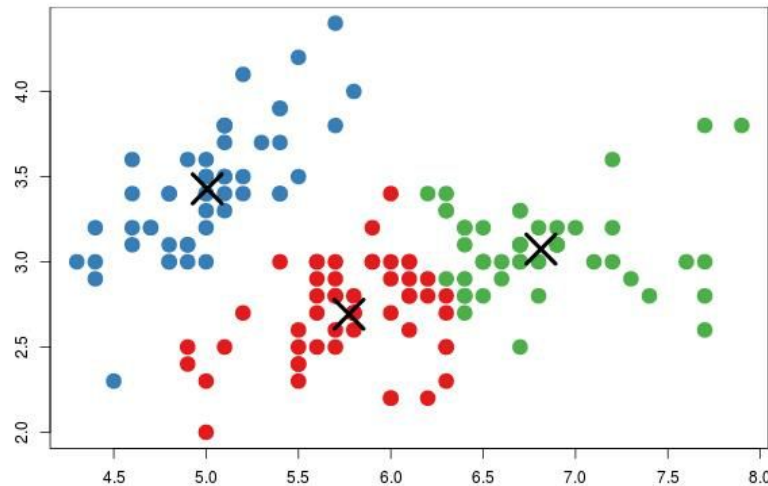
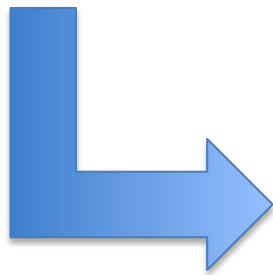
Image Categorization

- Classification
- Regression
- Clustering
- Dimension Reduction



Segmentation

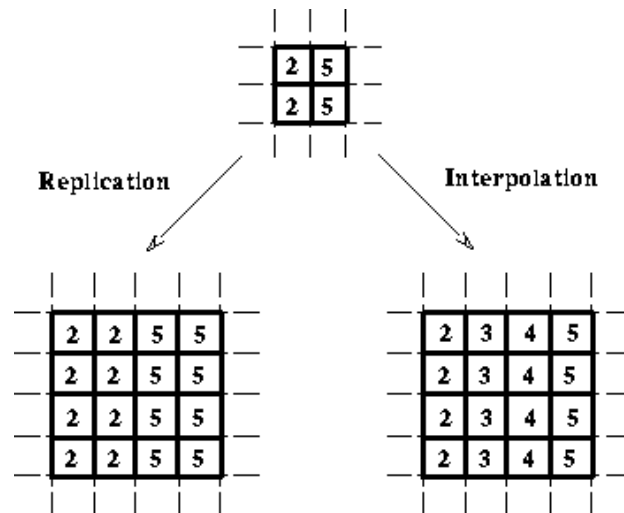
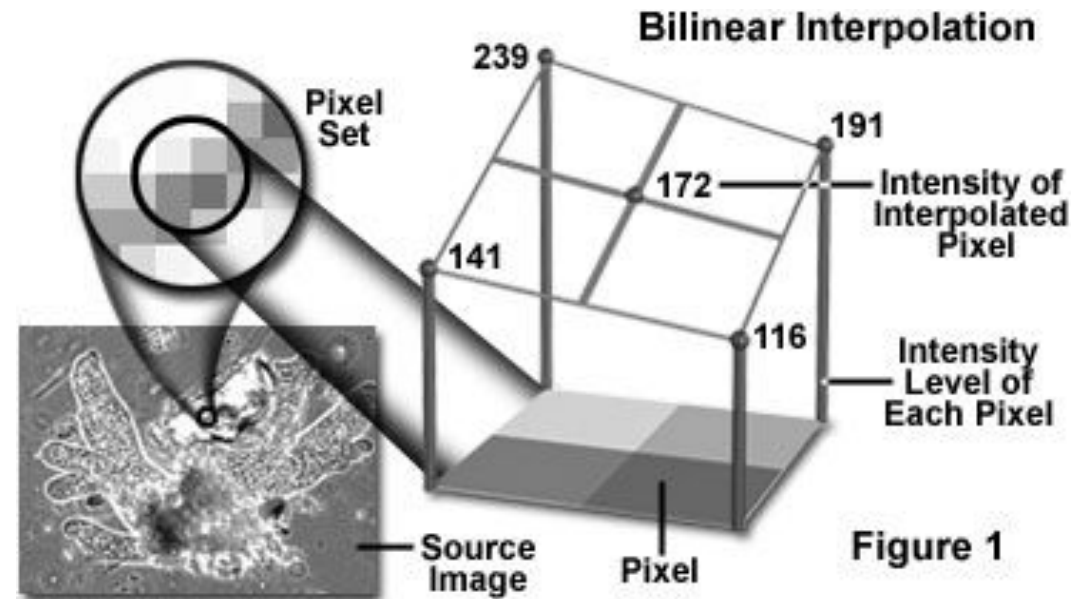
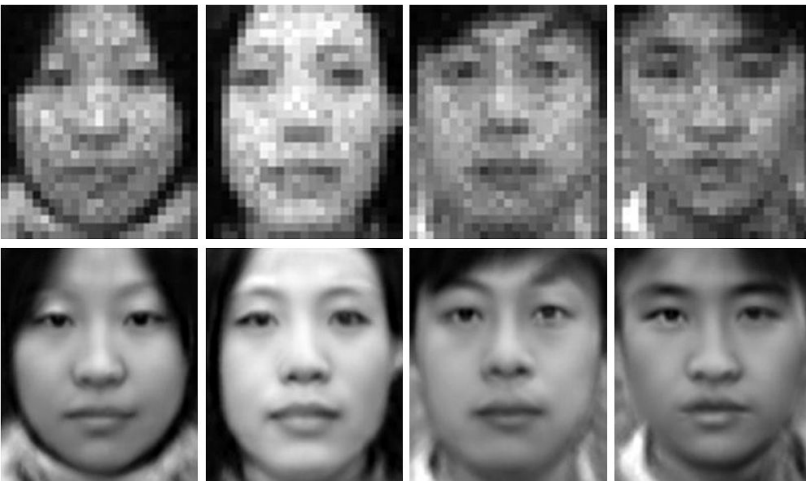
- Classification
- Regression
- Clustering
- Dimension Reduction



Learning with/without output label ?
Output label is discrete/continuous ?

- ☐ Classification
- ☐ Regression
- ☐ Clustering
- ☐ Dimension Reduction

Image Superresolution



Learning with/without output label ?
Output label is discrete/continuous ?

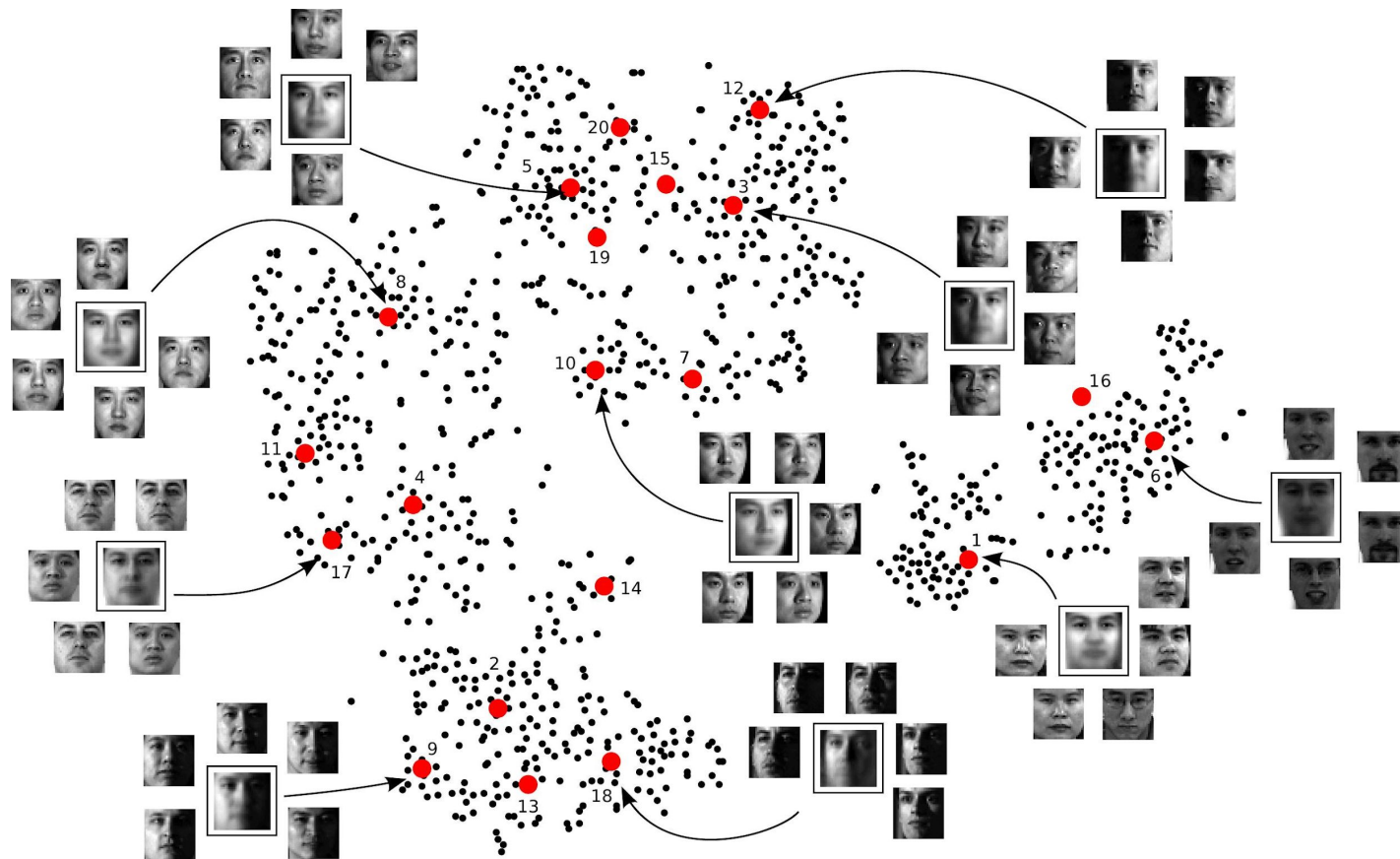
- ☐ Classification
- ☐ Regression
- ☐ Clustering
- ☐ Dimension Reduction

Inpainting



Learning with/without output label ?
Output label is discrete/continuous ?

Face Pose Analysis



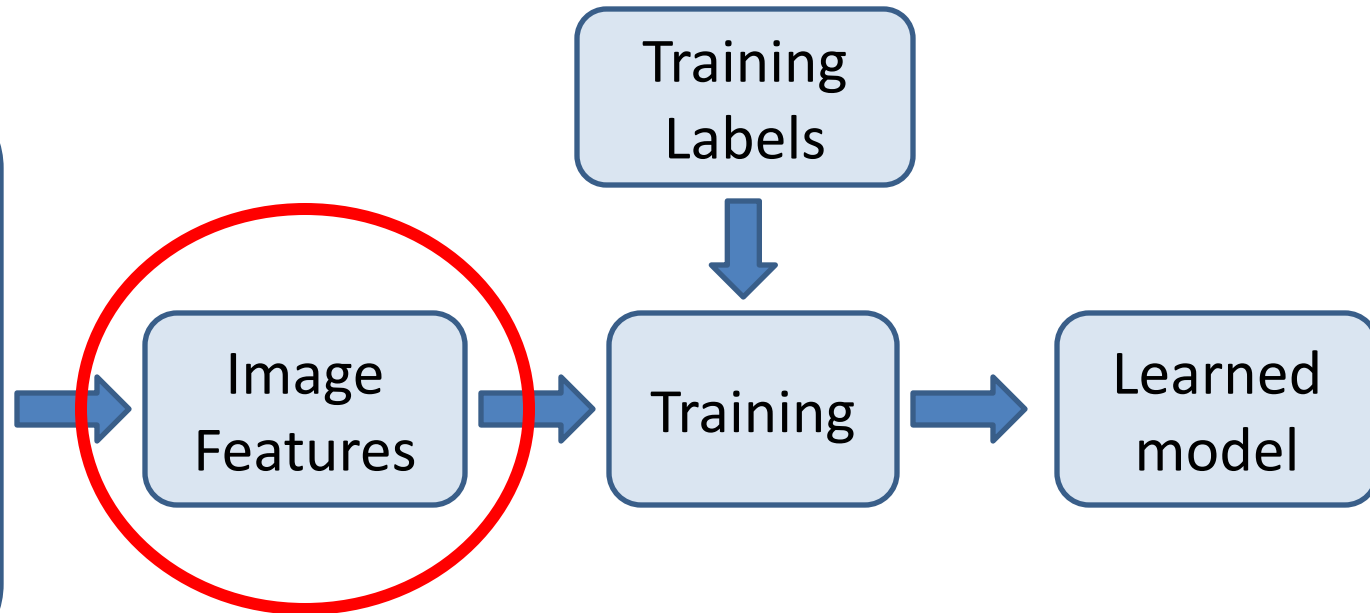
Learning with/without output label ?
Output label is discrete/continuous ?

FEATURE EXTRACTION

Image Features

Training

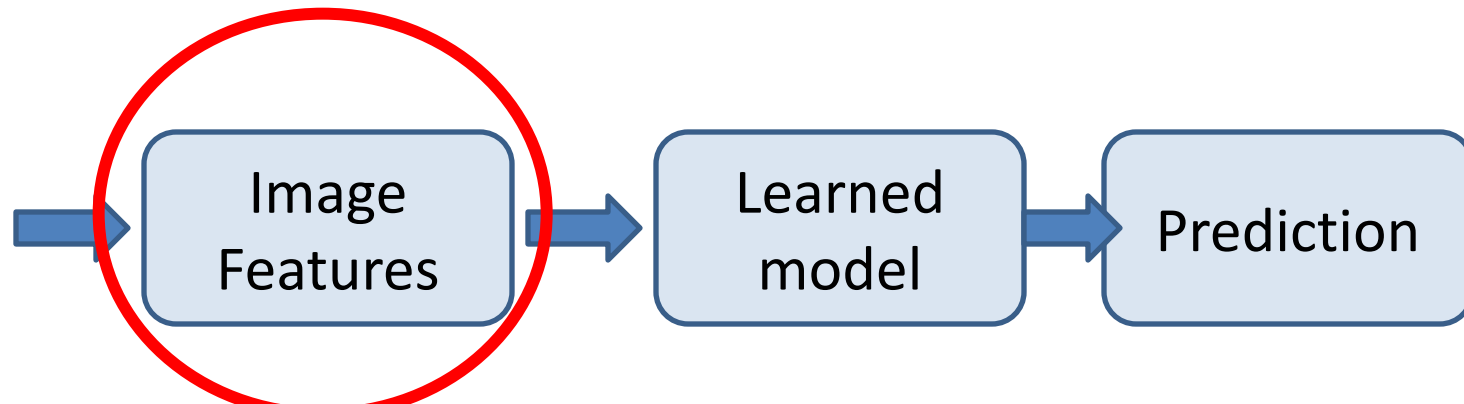
Training
Images



Testing

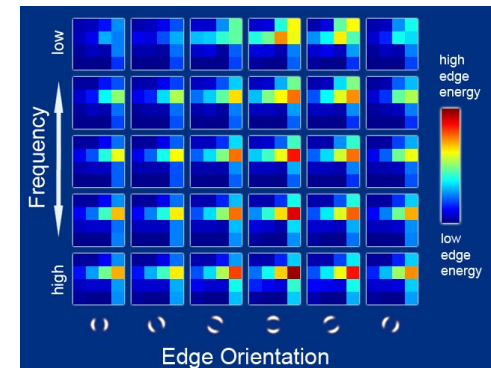
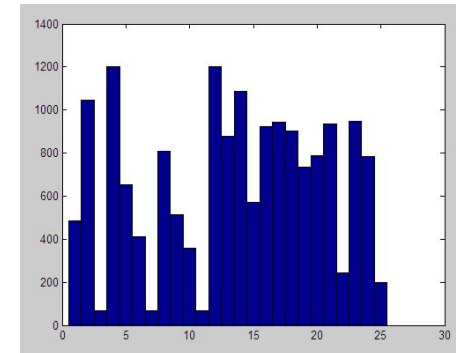


Test Image

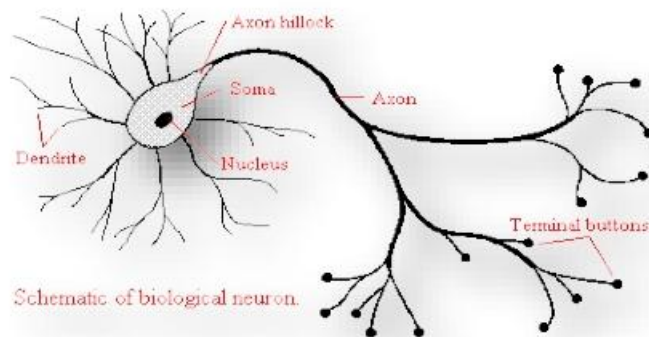


Features: Image Representations

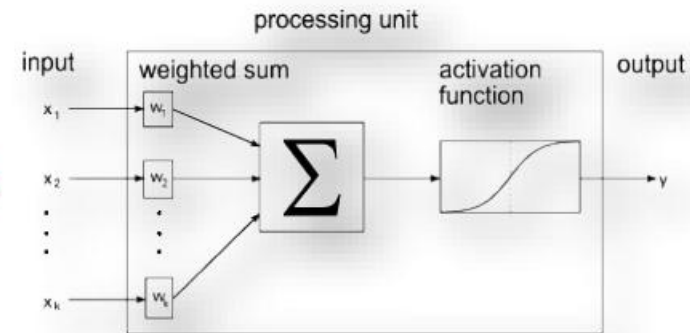
- Raw pixels
- Histograms
- GIST/SIFT descriptors
- ...



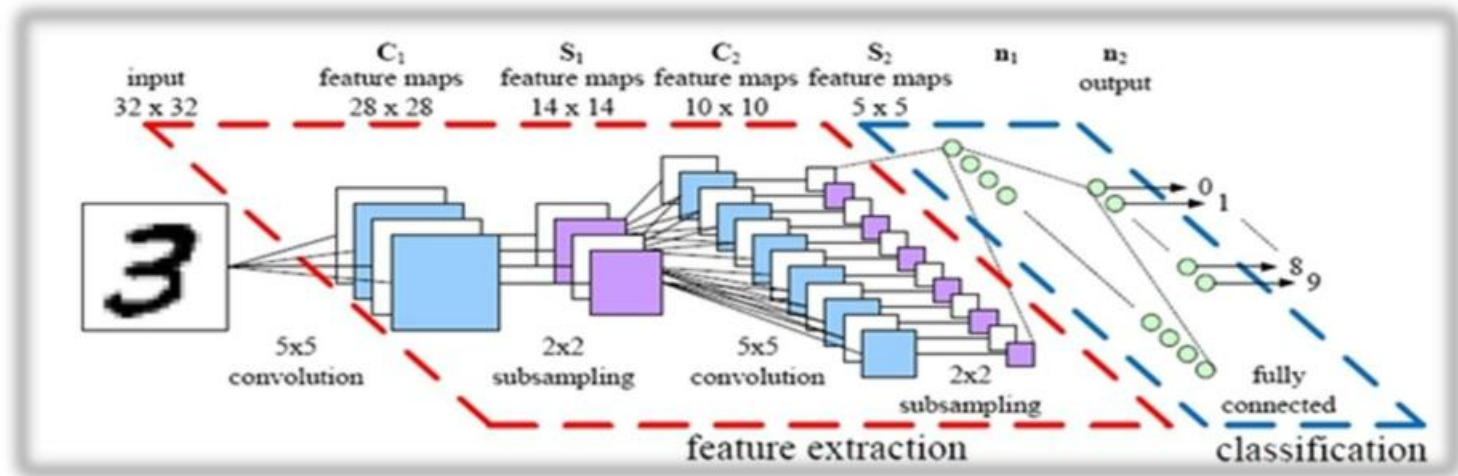
Deep Learning: the New Frontier



Schematic of biological neuron.



Brief history of Neural Nets: born in 50s, dead in 60s 70s, developed in 80s, dead again in 90s, forgotten in 2000, state of the art in 2010



DISCUSSIONS

Which ML Type?

- Face Detection
/ Face Tagging

- Pedestrian Detection



Which ML Type?

- Language Translation
- Google Image Search
- Google Search
- ChatGPT

