

Supervised Learning

CSC 461: Machine Learning

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Supervised Learning Setup

Components of (supervised) learning

- ▶ Input space \mathcal{X}
- ▶ Output space \mathcal{Y}
- ▶ Data instance $x \in \mathcal{X}, y \in \mathcal{Y}$
✓ is a pair (x,y)
- ▶ Data $\{(x_1, y_1), \dots, (x_n, y_n)\}$
✓ is a set of data instances
- ▶ Hypothesis $g : \mathcal{X} \mapsto \mathcal{Y}, g \in \mathcal{H}$

Example

Problem:

automatically tagging email messages as spam (1) or ham (0)

Input Space?

assume every email is represented as a fixed-length vector of 10 features

Output Space

Data

- Samples are assumed to be **independent and identically distributed** from the same probability distribution (**i.i.d**)

$$\mathcal{D} = \{(x_1, y_1), \dots, (x_n, y_n)\}$$

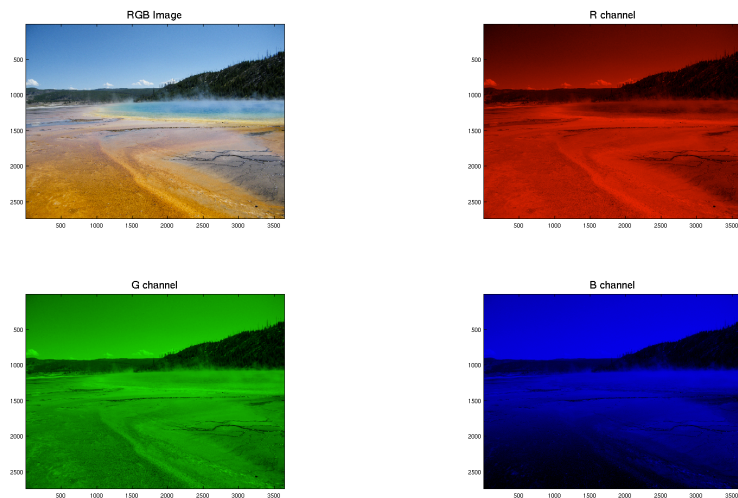
in general $\mathcal{X} = \mathbb{R}^d$

$$(x_i, y_i) \sim P_{\text{unknown}}$$

What are Feature Vectors?

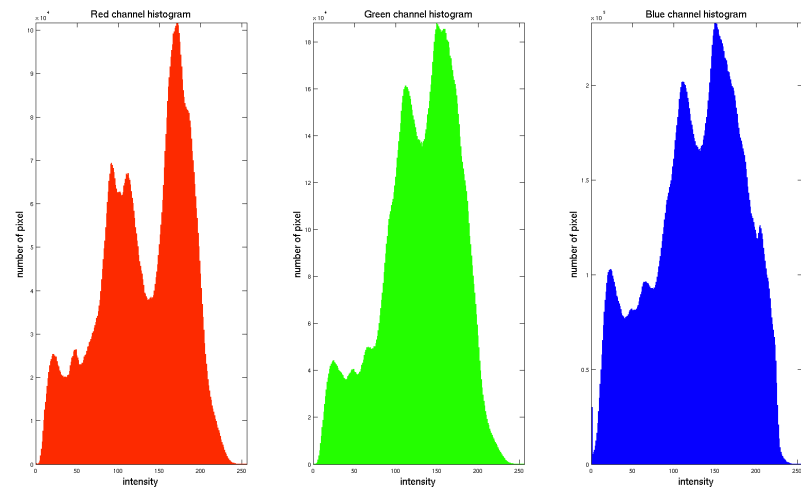
$$(x_i, y_i) \sim P_{\text{unknown}}$$

Feature Vectors



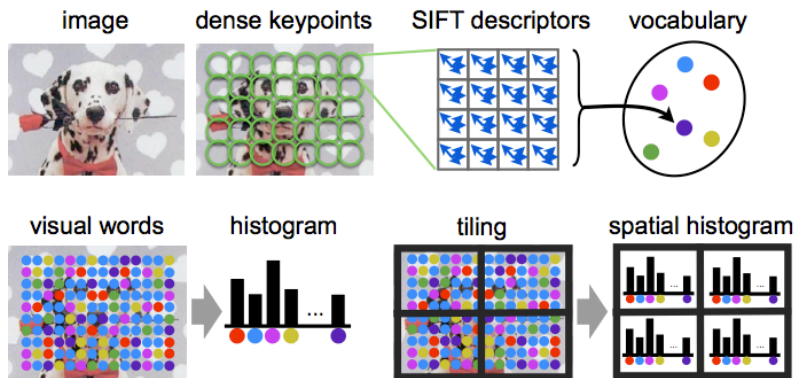
http://www.sci.utah.edu/~acoste/nou/Image/project1/Arthur_COSTE_Project_1_report.html

Feature Vectors



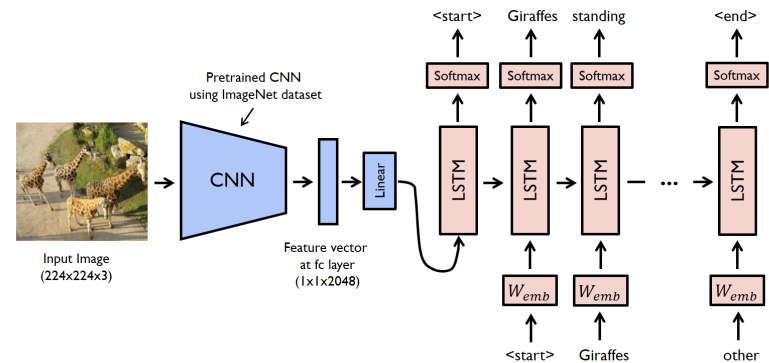
http://www.sci.utah.edu/~acoste/nou/Image/project1/Arthur_COSTE_Project_1_report.html

Feature Vectors



<https://www.di.ens.fr/willow/teaching/recvis13/assignment2/>

Feature Vectors



<https://www.analyticsvidhya.com/blog/2018/04/solving-an-image-captioning-task-using-deep-learning/>

Supervised learning

Binary classification

$$\mathcal{Y} = \{0, 1\}$$

$$\mathcal{Y} = \{-1, +1\}$$

Multiclass classification

$$\mathcal{Y} = \{0, 1, \dots, k-1\}$$

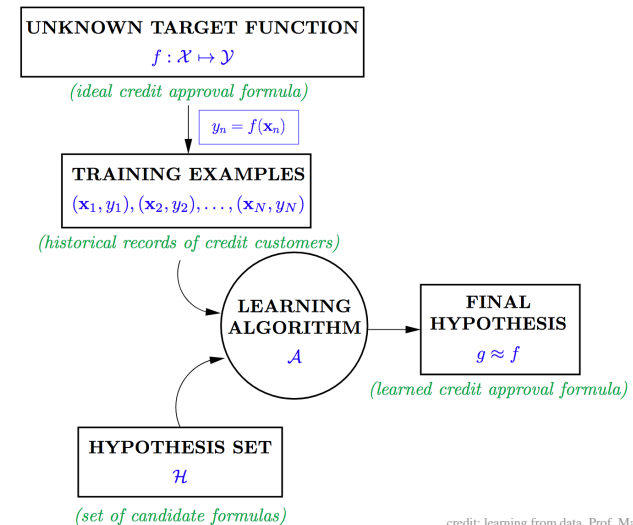
Regression

$$\mathcal{Y} = \mathbb{R}$$

Structure prediction

structured objects

Learning setup



credit: learning from data, Prof. Malik Magdon-Ismail

Example

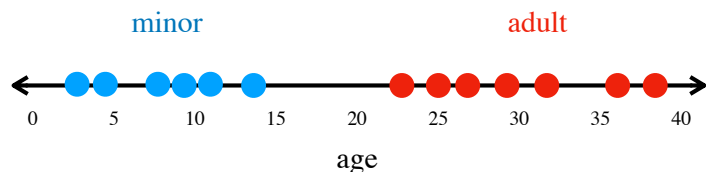
$h_1 \in \mathcal{H}$

$h_2 \in \mathcal{H}$

...

can you define the hypothesis space?

how to pick a hypothesis that makes you happy?



Loss Functions

▸ 0/1 Loss $\mathcal{L}_{0/1}(h, \mathcal{D}) = \frac{1}{n} \sum_{(x_i, y_i) \in \mathcal{D}} I(h(x_i) \neq y_i)$

indicator function

▸ Squared Loss $\mathcal{L}_{sq}(h, \mathcal{D}) = \frac{1}{n} \sum_{(x_i, y_i) \in \mathcal{D}} (h(x_i) - y_i)^2$

▸ Absolute Loss $\mathcal{L}_{abs}(h, \mathcal{D}) = \frac{1}{n} \sum_{(x_i, y_i) \in \mathcal{D}} |h(x_i) - y_i|$

Consider this classifier ...

$$g(x) = \begin{cases} y_i & \text{if } x = x_i \text{ for } (x_i, y_i) \in \mathcal{D} \\ y_1 & \text{otherwise} \end{cases}$$

what is the 0/1 loss for training data?
square loss? absolute loss?

are you happy with this classifier? why?

What is the goal of (supervised) learning?

- A function (**classifier**) that best approximates target function

For $g \in \mathcal{H}$ and $\forall (x_i, y_i) \sim P$, we want $g(x) \approx f(x)$

search and optimization (to minimize expected loss)

Role of Training and Testing

Example on Iris Dataset (next lecture)

From the UCI ML repository ...

Iris Data Set

Download: [Data Folder](#), [Data Set Description](#)

Abstract: Famous database; from Fisher, 1936



Data Set Characteristics:	Multivariate	Number of Instances:	150	Area:	Life
Attribute Characteristics:	Real	Number of Attributes:	4	Date Donated	1988-07-01
Associated Tasks:	Classification	Missing Values?	No	Number of Web Hits:	2096826

Attribute Information:

- sepal length in cm
- sepal width in cm
- petal length in cm
- petal width in cm

Class Information:

- Iris Setosa
- Iris Versicolour
- Iris Virginica

Example using Iris dataset

https://colab.research.google.com/drive/1pDkSHtAG1A2kCKrPY1CNDtVdKj1f_wK3