

Supervised learning: a sketch

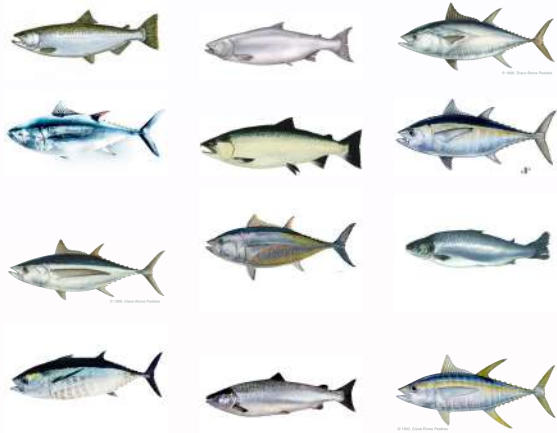
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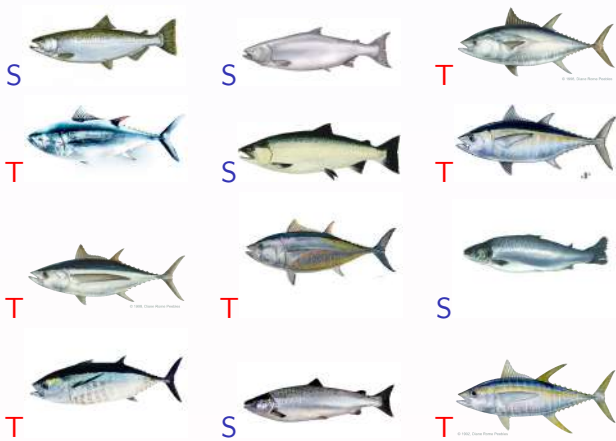


- 1 Objects and categories
- 2 Supervised learning

A set of objects



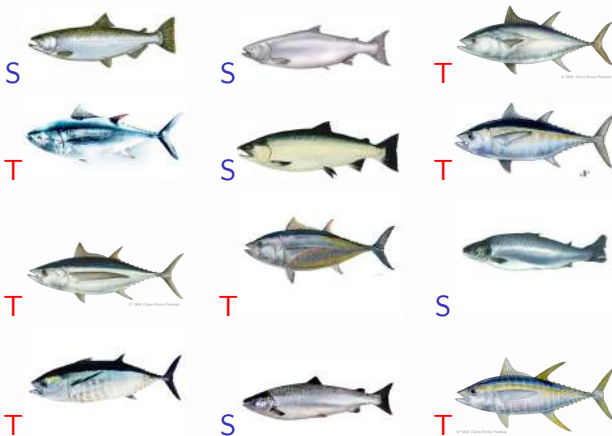
Several categories of objects



S = salmon

T = tuna

Several categories of objects



S = salmon

How to make the difference ?

T = tuna

Description by features

Features, attributes, characteristics

Feature :	Possible values :
length	0-500cm
width	0-100cm
color	{silver, blue, rose, yellow, multicolored}
mean length of fins	0-50cm
has open mouth?	{yes, no}
...	...

Example descriptions

length
width
color
mean length of fins
has open mouth?
...

120cm	110cm	130cm	140cm
50cm	48cm	70cm	71cm
silver	silver	silver	blue
10cm	7cm	20cm	10cm
yes	yes	yes	no
...

Example descriptions

length	120cm	110cm	130cm	140cm
width	50cm	48cm	70cm	71cm
color	silver	silver	silver	blue
mean length of fins	10cm	7cm	20cm	10cm
has open mouth?	yes	yes	yes	no
...

Feature vectors

Feature engineering

Example descriptions


length	120cm	110cm	130cm	140cm
width	50cm	48cm	70cm	71cm
color	silver	silver	silver	blue
mean length of fins	10cm	7cm	20cm	10cm
has open mouth?	yes	yes	yes	no
...

Feature vectors


Feature engineering

- Other representations are possible (trees, etc.)

Description of objects

S 


120cm
50cm
silver
10cm
yes
...

T 


140cm
71cm
blue
10cm
no
...

T 


...
...
...
...
...
...

T 


...
...
...
...
...
...

S 


110cm
48cm
silver
7cm
yes
...

S 


...
...
...
...
...
...

T 


...
...
...
...
...
...

S 


...
...
...
...
...
...

T 


130cm
70cm
silver
20cm
yes
...

T 

...
...
...
...
...
...













S 

...
...
...
...
...
...

T 

...
...
...
...
...
...

Inclusion of the category: description of the solution

	120cm 50cm silver 10cm yes S		110cm 48cm silver 7cm yes S		130cm 70cm silver 20cm yes T
	140cm 71cm blue 10cm no T		...		...
	...		...		...
	...		...		...

Description of the solution

120cm 50cm silver 10cm yes S	110cm 48cm silver 7cm yes S	130cm 70cm silver 20cm yes T
140cm 71cm blue 10cm no T
...
...
...
...
T	T	S
...
...
...
...
T	S	T

Goal

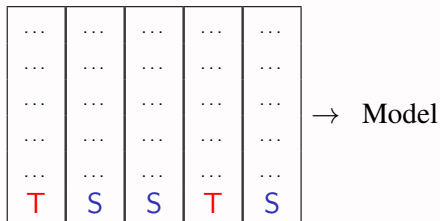
- Given these examples of objects in two categories,
- Learn to determine the category of a new object

Goal

- Given these examples of objects in two categories,
- Learn to determine the category of a new object

Training

Build a model that relates object features to categories



Goal

- Given these examples of objects in two categories,
- Learn to determine the category of a new object

Training

Build a **model** that relates object features to categories

...
...
...
...
...
...
T	S	S	T	S

→ Model

Application (“Inference”)

Apply the **model** to determine the category of a new object based upon its features

...
...
...
...
...
...
?

Model → T