

Supervised Learning

Learning from Examples

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In **Supervised Learning**, an agent reads in data in the form **(input, output)** pairs. It learns a **model** (function) that maps input to output.

Formulating the Learning Problem

- **Given:** a dataset that contains data points, described by some attributes x_1, x_2, \dots, x_n .
- Each data point has:
 - ***Some attributes***, referred to as X
 - **A class label**, referred to as y

x1	x2	y

attributes

class labels

Formulating the Learning Problem

- **Required:** Find a function that can predict the class label y of any data point given its attribute values X .

$$y = f(X)$$

- That is, estimate y as a function of x .

We know y & X .
We want to find $f()$

x1	x2	y

attributes class labels

Formulating the Learning Problem

$$y = f(X)$$

There are many options for $f()$.
 $f()$ can be linear or non-linear.

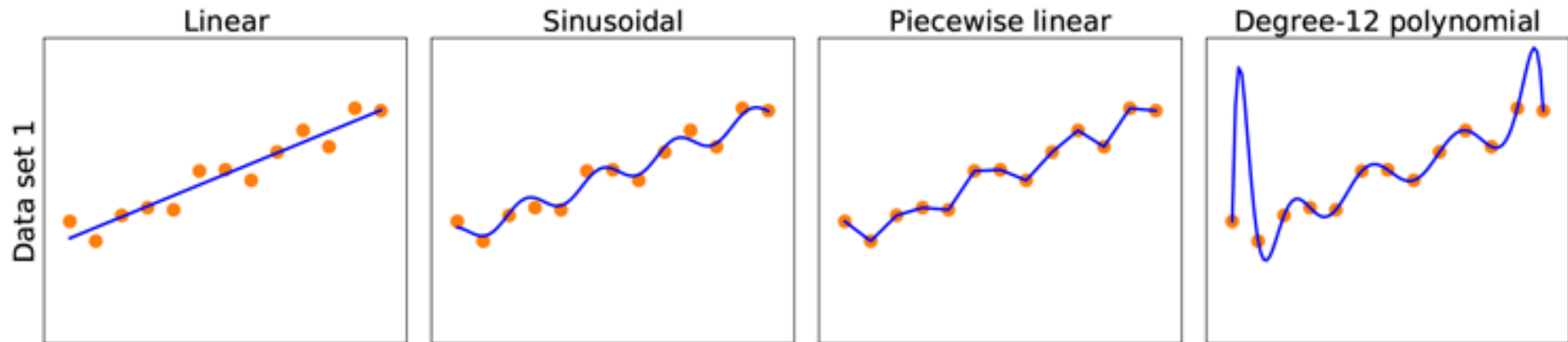


Fig 19.1, Russell & Norvig's Textbook

Example Applications

Object Classification in Images

- Given a dataset of images, classify the objects in the images within a set of ***predefined*** classes.

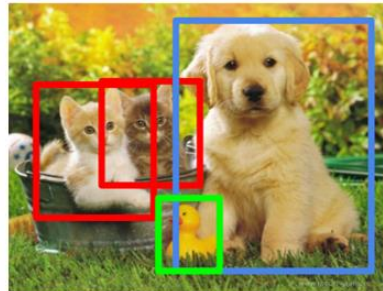
Classification



CAT

Image source: <https://towardsdatascience.com/going-deep-into-object-detection-bed442d92b34>

Object Detection



CAT, DOG, DUCK

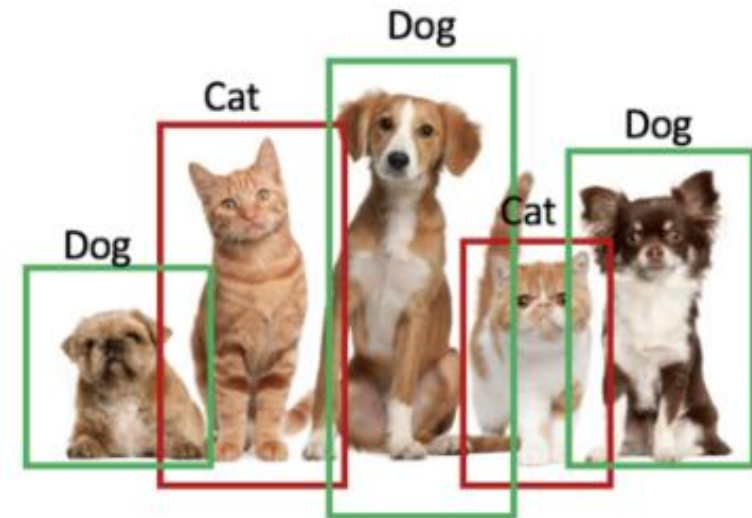


Image source: <https://medium.com/@bluebirb/object-detection-current-popular-techniques-900d36ae9b08>

Object Classification in Images

- Features used for cat/dog classification?
 - Ear shape, face size w.r.t. body size, distance between nose and eyes, etc.

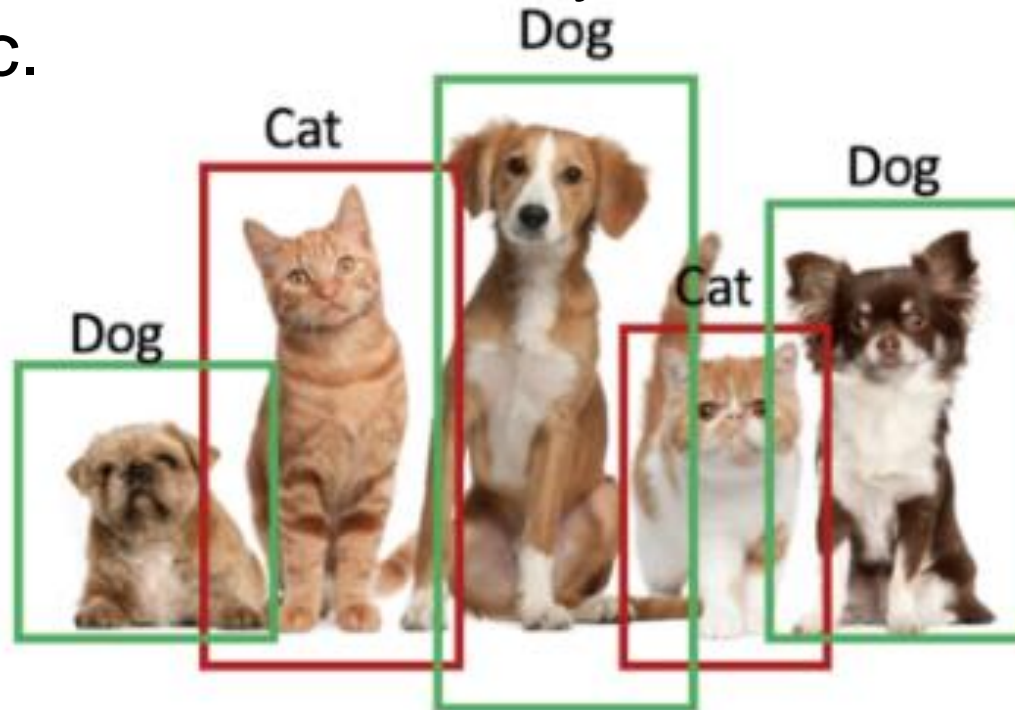


Image source: <https://medium.com/@bluebirb/object-detection-current-popular-techniques-900d36ae9b08>

Tumor Cell Classification

- Given a dataset of images, each containing one or more cells, classify whether they are malignant or benign.
 - Features?

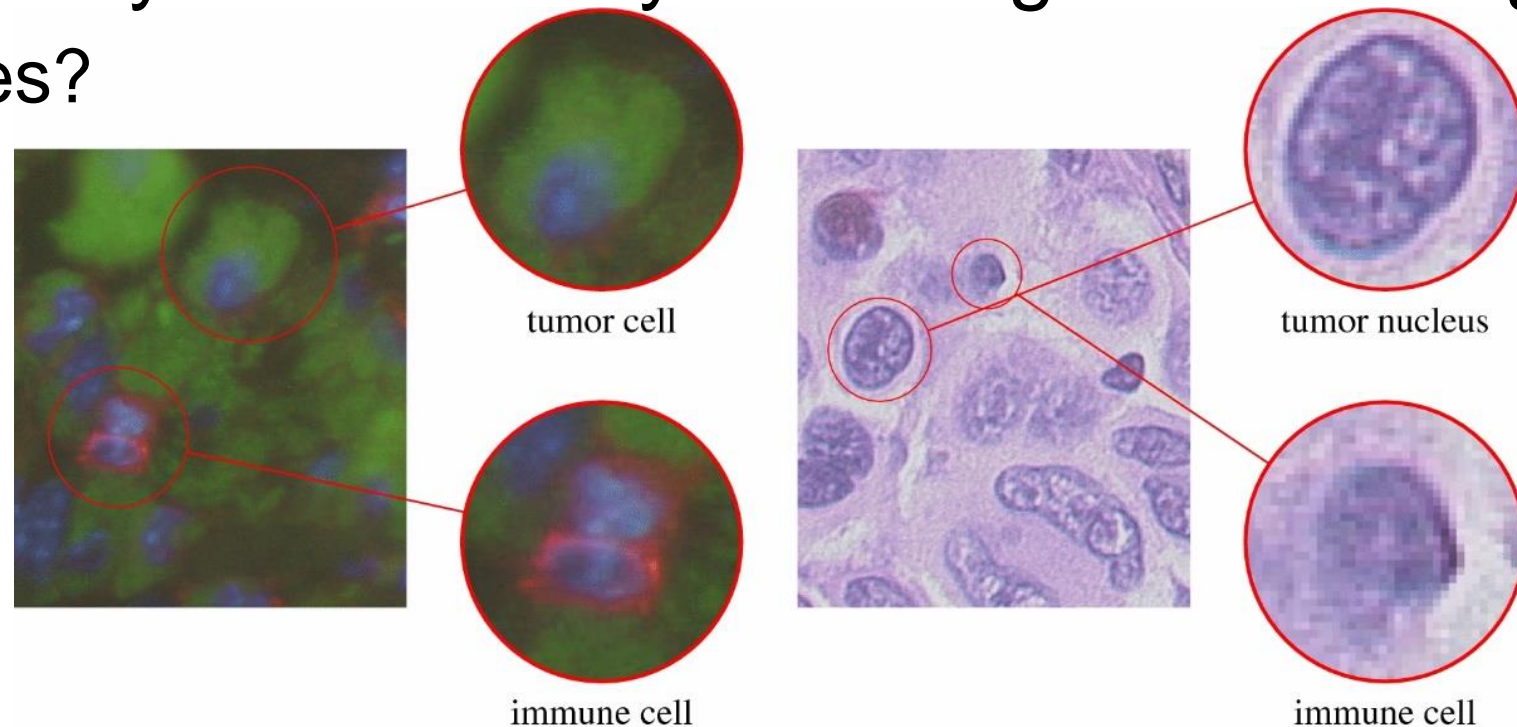
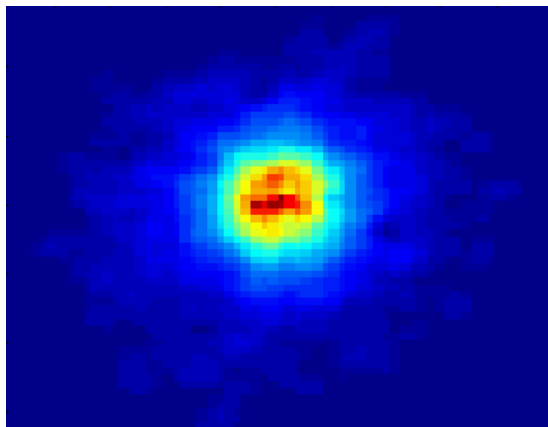


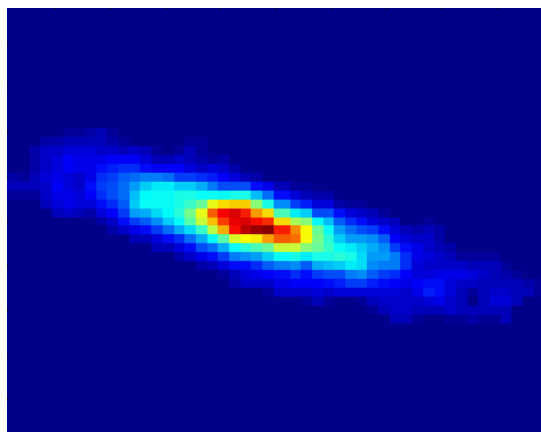
Image source: <https://link.springer.com/article/10.1007/s11548-019-01919-z>
Asmaa Elbadrawy – ASU – 2021

Galaxy Classification

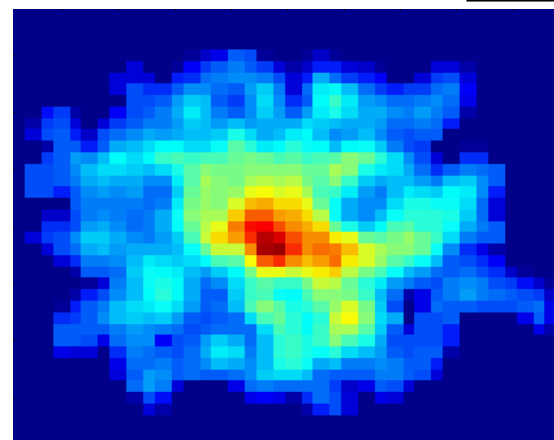
Early



Intermediate



Late



Class:

- Stages of Formation

Attributes:

- Image features,
- Characteristics of light waves received, etc.

Courtesy: <http://aps.umn.edu>

Data Size:

- 72 million stars, 20 million galaxies
- Object Catalog: 9 GB
- Image Database: 150 GB

Predicting Stock prices given market conditions.



The Restaurant Waiting Problem

- Should we wait for a table at a restaurant?
- $y \rightarrow$ Boolean variable deciding whether wait = True/False
- $X \rightarrow$ 10 different attributes

Th

- S
- y
- X

1. *Alternate*: whether there is a suitable alternative restaurant nearby.
2. *Bar*: whether the restaurant has a comfortable bar area to wait in.
3. *Fri/Sat*: true on Fridays and Saturdays.
4. *Hungry*: whether we are hungry right now.
5. *Patrons*: how many people are in the restaurant (values are *None*, *Some*, and *Full*).
6. *Price*: the restaurant's price range (\$, \$\$, \$\$\$).
7. *Raining*: whether it is raining outside.
8. *Reservation*: whether we made a reservation.
9. *Type*: the kind of restaurant (French, Italian, Thai, or burger).
10. *WaitEstimate*: host's wait estimate: 0–10, 10–30, 30–60, or >60 minutes.

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The Restaurant Waiting Problem

- There are thousands of possible values for the 10 attributes.

Example	Input Attributes										Output
	<i>Alt</i>	<i>Bar</i>	<i>Fri</i>	<i>Hun</i>	<i>Pat</i>	<i>Price</i>	<i>Rain</i>	<i>Res</i>	<i>Type</i>	<i>Est</i>	<i>WillWait</i>
x₁	<i>Yes</i>	<i>No</i>	<i>No</i>	<i>Yes</i>	<i>Some</i>	<i>\$\$\$</i>	<i>No</i>	<i>Yes</i>	<i>French</i>	<i>0–10</i>	<i>y₁ = Yes</i>
x₂	<i>Yes</i>	<i>No</i>	<i>No</i>	<i>Yes</i>	<i>Full</i>	<i>\$</i>	<i>No</i>	<i>No</i>	<i>Thai</i>	<i>30–60</i>	<i>y₂ = No</i>
x₃	<i>No</i>	<i>Yes</i>	<i>No</i>	<i>No</i>	<i>Some</i>	<i>\$</i>	<i>No</i>	<i>No</i>	<i>Burger</i>	<i>0–10</i>	<i>y₃ = Yes</i>
x₄	<i>Yes</i>	<i>No</i>	<i>Yes</i>	<i>Yes</i>	<i>Full</i>	<i>\$</i>	<i>Yes</i>	<i>No</i>	<i>Thai</i>	<i>10–30</i>	<i>y₄ = Yes</i>
x₅	<i>Yes</i>	<i>No</i>	<i>Yes</i>	<i>No</i>	<i>Full</i>	<i>\$\$\$</i>	<i>No</i>	<i>Yes</i>	<i>French</i>	<i>>60</i>	<i>y₅ = No</i>
x₆	<i>No</i>	<i>Yes</i>	<i>No</i>	<i>Yes</i>	<i>Some</i>	<i>\$\$</i>	<i>Yes</i>	<i>Yes</i>	<i>Italian</i>	<i>0–10</i>	<i>y₆ = Yes</i>
x₇	<i>No</i>	<i>Yes</i>	<i>No</i>	<i>No</i>	<i>None</i>	<i>\$</i>	<i>Yes</i>	<i>No</i>	<i>Burger</i>	<i>0–10</i>	<i>y₇ = No</i>
x₈	<i>No</i>	<i>No</i>	<i>No</i>	<i>Yes</i>	<i>Some</i>	<i>\$\$</i>	<i>Yes</i>	<i>Yes</i>	<i>Thai</i>	<i>0–10</i>	<i>y₈ = Yes</i>
x₉	<i>No</i>	<i>Yes</i>	<i>Yes</i>	<i>No</i>	<i>Full</i>	<i>\$</i>	<i>Yes</i>	<i>No</i>	<i>Burger</i>	<i>>60</i>	<i>y₉ = No</i>
x₁₀	<i>Yes</i>	<i>Yes</i>	<i>Yes</i>	<i>Yes</i>	<i>Full</i>	<i>\$\$\$</i>	<i>No</i>	<i>Yes</i>	<i>Italian</i>	<i>10–30</i>	<i>y₁₀ = No</i>
x₁₁	<i>No</i>	<i>No</i>	<i>No</i>	<i>No</i>	<i>None</i>	<i>\$</i>	<i>No</i>	<i>No</i>	<i>Thai</i>	<i>0–10</i>	<i>y₁₁ = No</i>
x₁₂	<i>Yes</i>	<i>Yes</i>	<i>Yes</i>	<i>Yes</i>	<i>Full</i>	<i>\$</i>	<i>No</i>	<i>No</i>	<i>Burger</i>	<i>30–60</i>	<i>y₁₂ = Yes</i>

Fig 19.2, Russell & Norvig's Textbook