

# Introduction to Artificial Intelligence (AI) and Machine Learning (ML)

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# Learning objectives

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Terminology of Machine Learning

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The difference between AI and ML

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The different types of machine learning techniques

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Applications of machine learning techniques

# Data Exploration



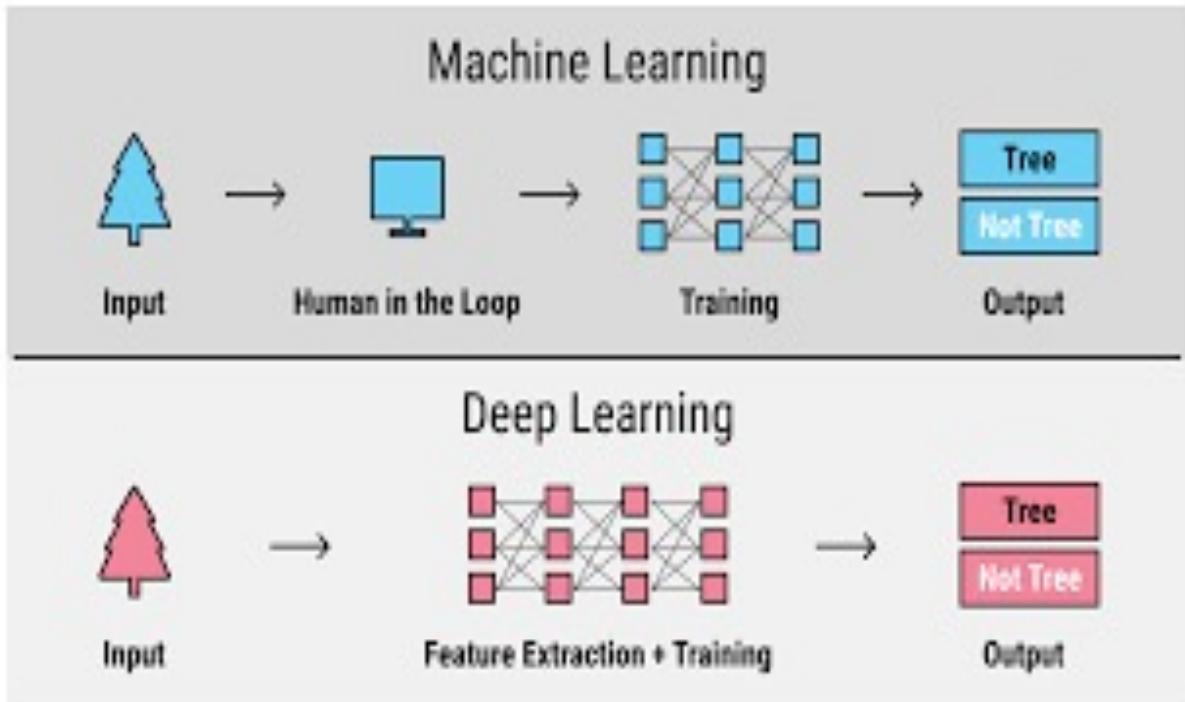
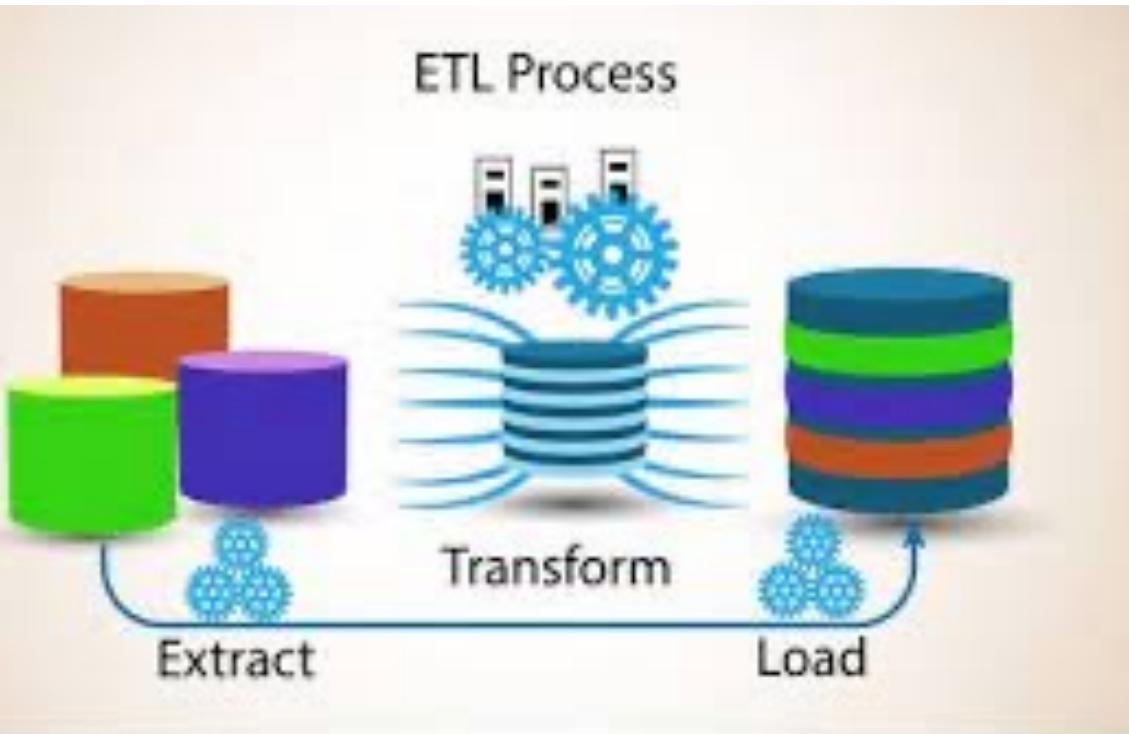
Data manipulation/analysis library  
• Exercises



Data visualization library  
• Exercises



# ENTERPRISE DATA vs MACHINE LEARNING DATA



# Formats of data

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NUMBERS

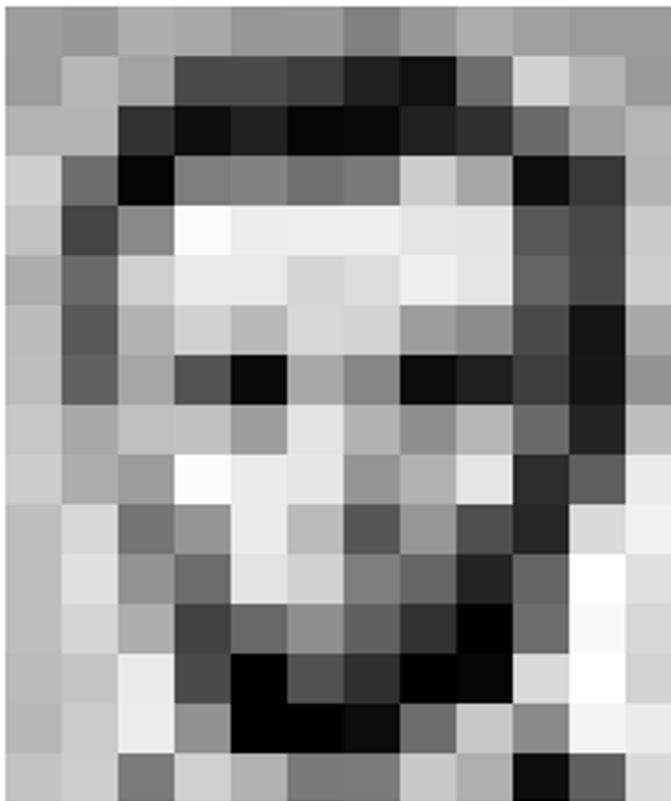


TEXTS



IMAGES

# Image (as seen by computers)



157	153	174	168	150	152	129	151	172	161	155	156
155	182	163	74	75	62	33	17	110	210	180	154
180	180	50	14	34	6	10	33	48	105	159	181
206	109	5	124	131	111	120	204	166	15	56	180
194	68	197	251	237	299	239	228	227	87	71	201
172	105	207	233	233	214	220	239	228	98	74	206
188	88	179	209	185	215	211	158	139	75	20	169
189	97	165	84	10	168	134	11	31	62	22	148
199	168	191	193	158	227	178	143	182	105	36	190
205	174	155	252	236	231	149	178	228	43	95	234
190	216	116	149	236	187	85	150	79	38	218	241
190	224	147	108	227	210	127	102	35	101	255	224
190	214	173	66	103	143	95	50	2	109	249	215
187	196	235	75	1	81	47	0	6	217	255	211
183	202	237	145	0	0	12	108	200	138	243	236
195	206	123	207	177	121	123	200	175	13	96	218

157	153	174	168	150	152	129	151	172	161	155	156
155	182	163	74	75	62	33	17	110	210	180	154
180	180	50	14	34	6	10	33	48	105	159	181
206	109	5	124	131	111	120	204	166	15	56	180
194	68	137	251	237	239	239	228	227	87	71	201
172	105	207	233	233	214	220	239	228	98	74	206
188	88	179	209	185	215	211	158	139	75	20	169
189	97	165	84	10	168	134	11	31	62	22	148
199	168	191	193	158	227	178	143	182	105	36	190
205	174	155	252	236	231	149	178	228	43	95	234
190	216	116	149	236	187	85	150	79	38	218	241
190	224	147	108	227	210	127	102	35	101	255	224
190	214	173	66	103	143	95	50	2	109	249	215
187	196	235	75	1	81	47	0	6	217	255	211
183	202	237	145	0	0	12	108	200	138	243	236
195	206	123	207	177	121	123	200	175	13	96	218

## Types of data:

- Labelled
- Unlabelled



# Labelled data



Dog



Cat



Dog



Dog



Cat



Cat



Dog



Cat

# Unlabelled Data



# Neural Networks and Biology

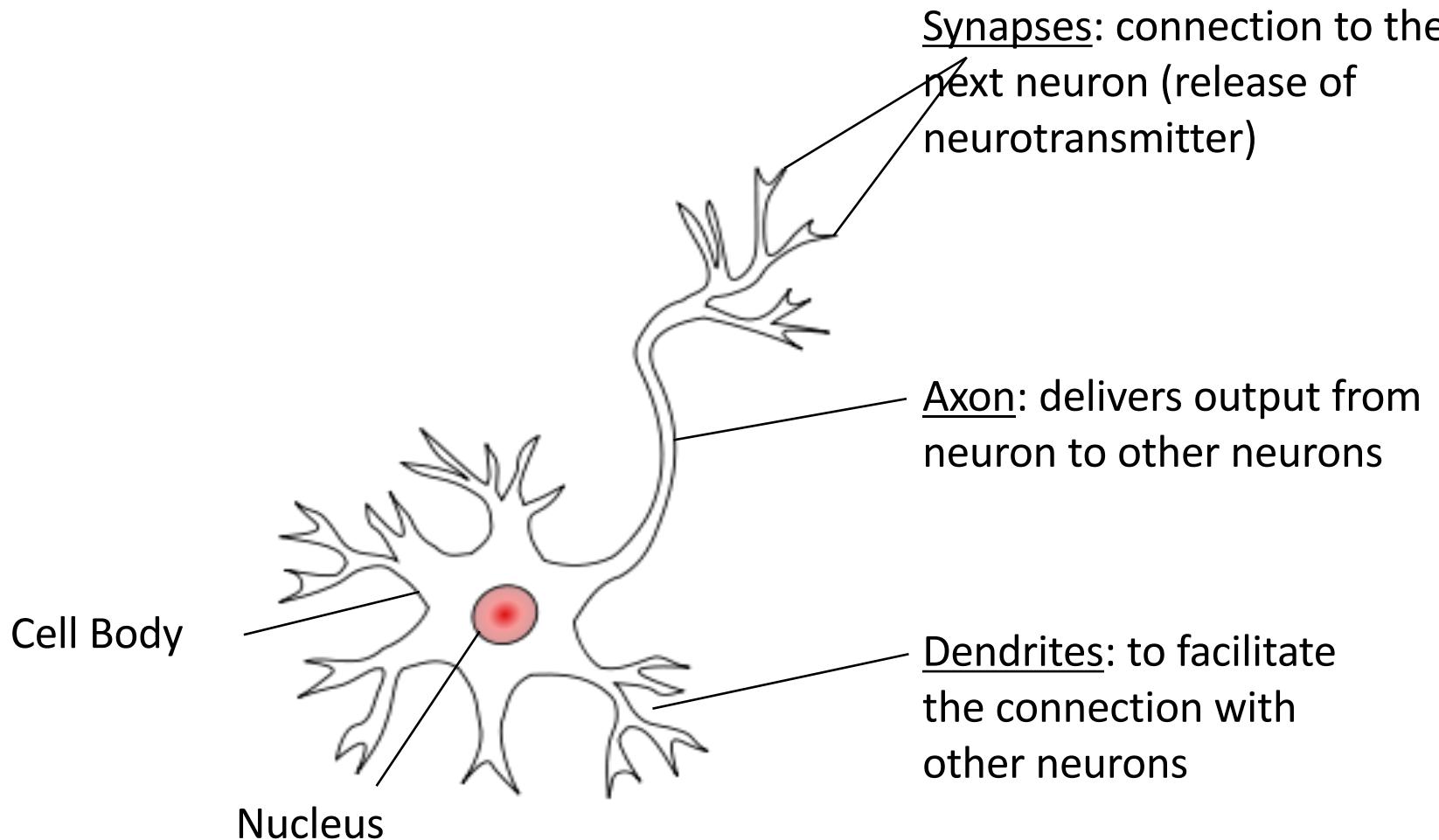
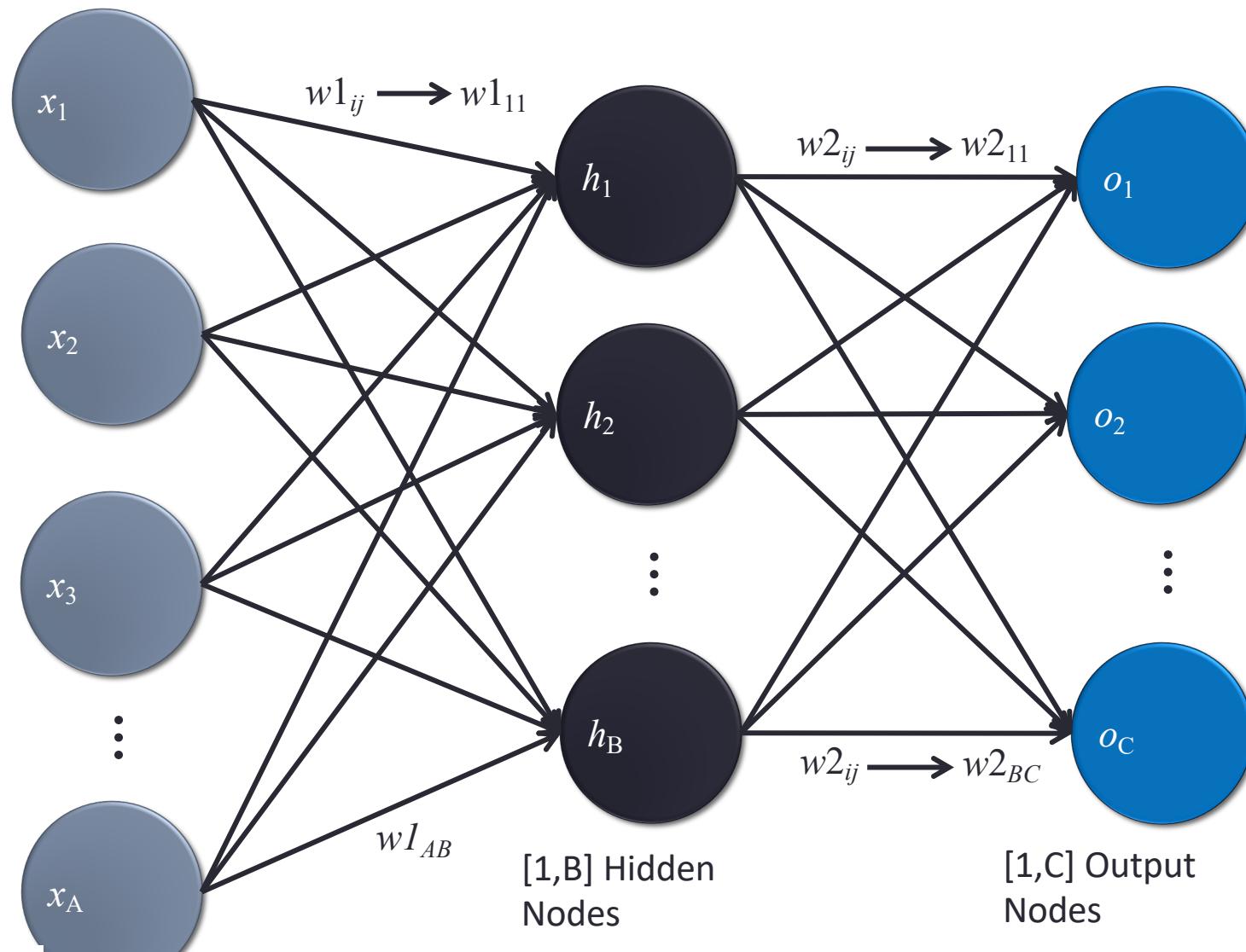
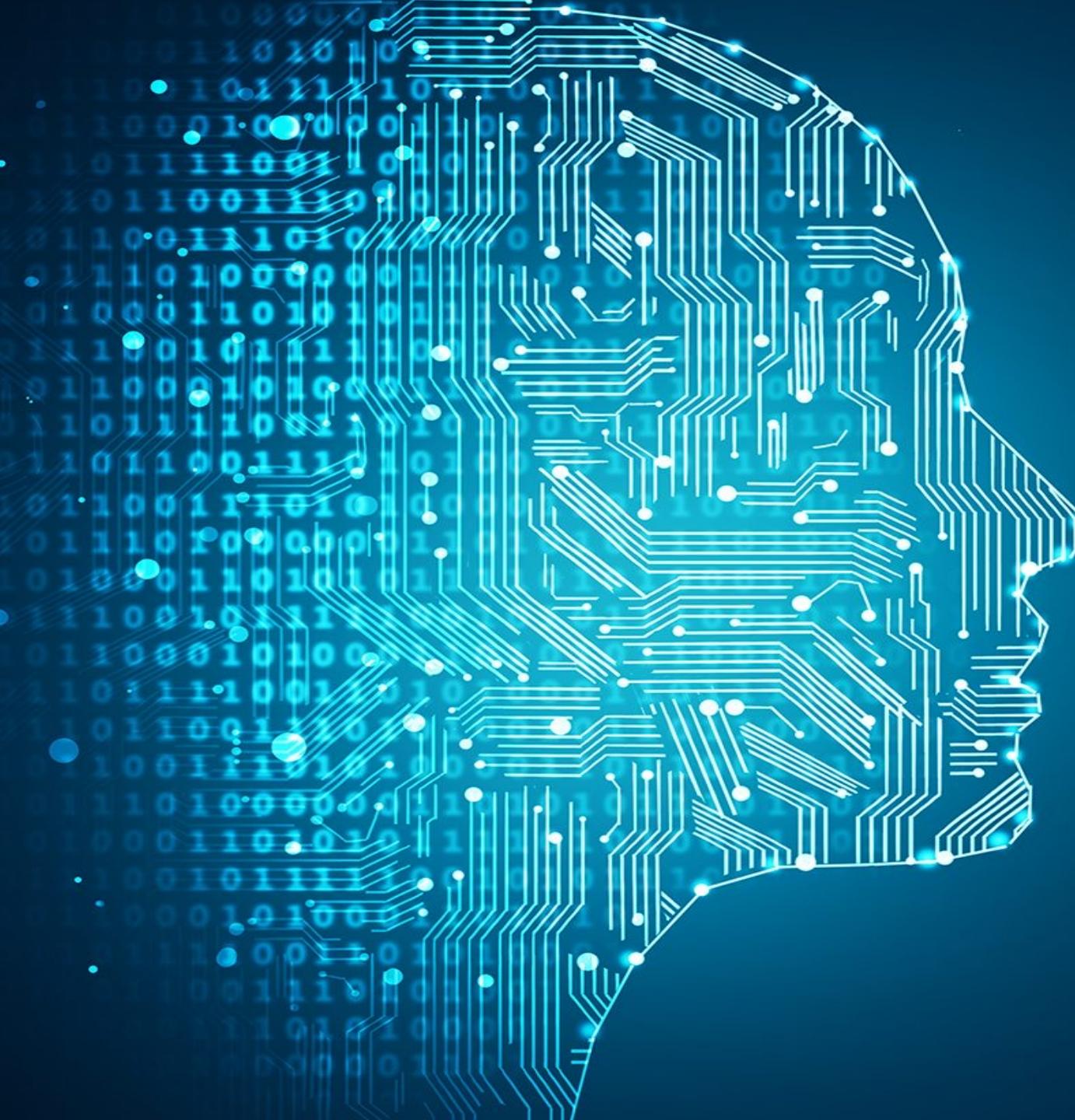


Figure: Structure of a typical neuron

# Neural Network





# What is training?

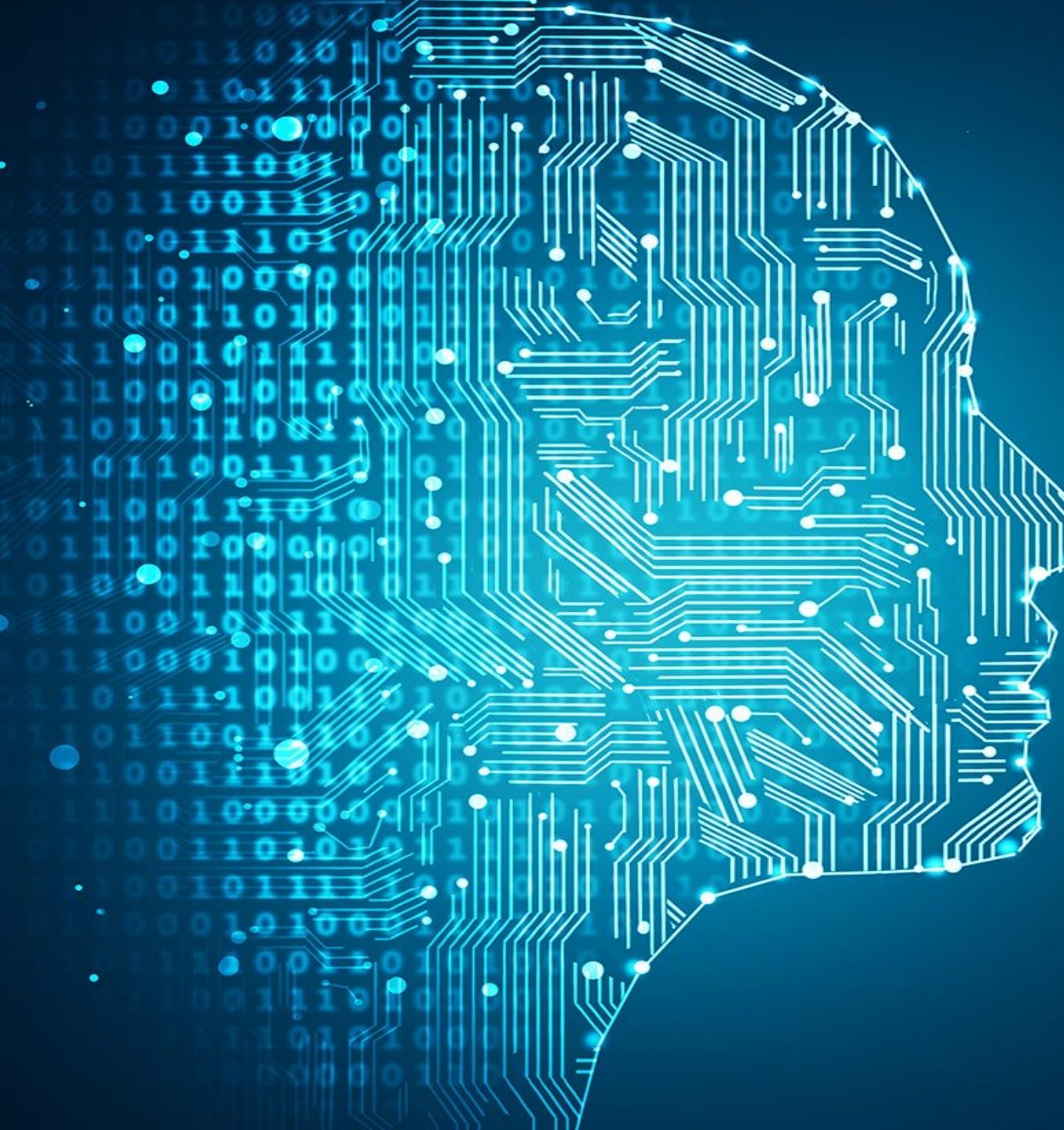
The process used to create our ML model.  
Find a set of weights and biases that have high accuracy.

A blue-toned illustration of a human brain composed of glowing white circuit boards and binary code.

# What is testing?

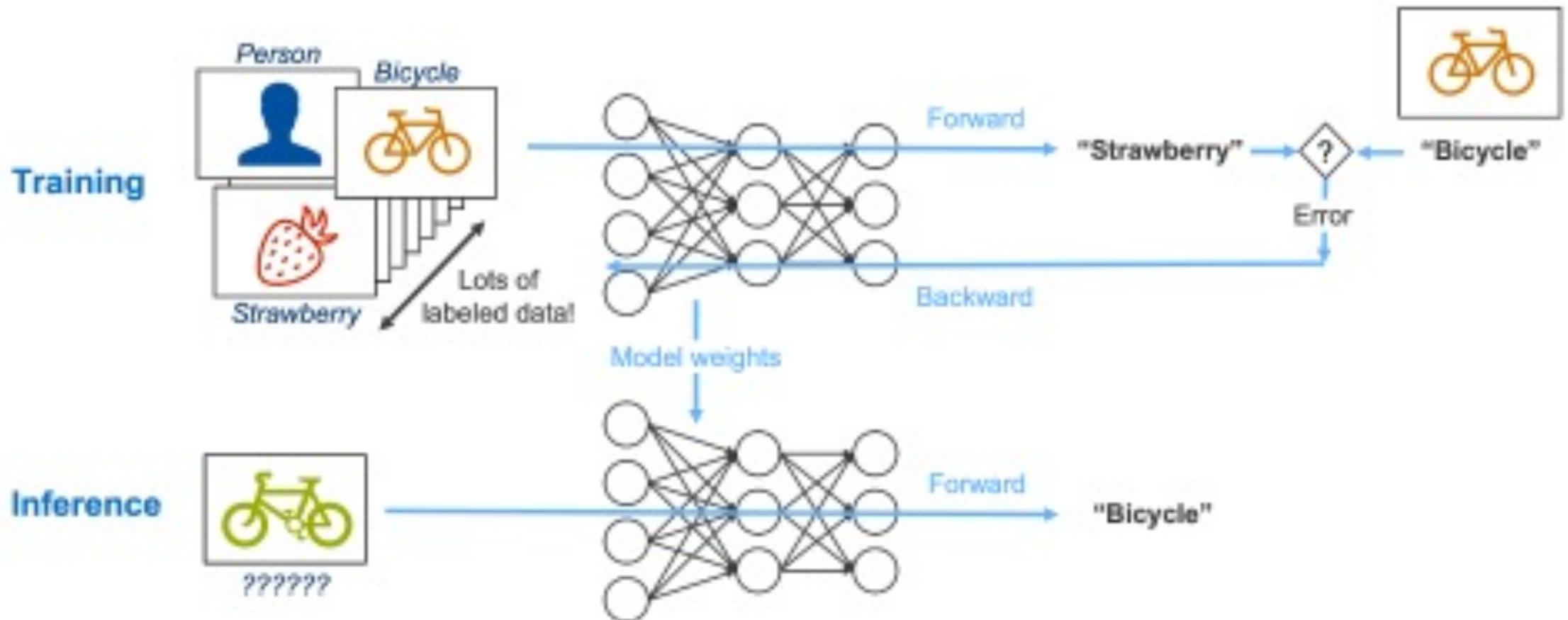
The process used to test our ML model.

Run the model against known outcomes

A blue-toned illustration of a human brain composed of glowing white circuit boards and binary code.

# What is inference?

Running our model on  
live data to produce  
actionable output.



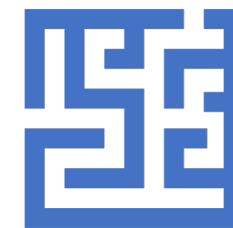
# Common types of Learning



## Supervised learning

We have **labelled data**, and we want to make some prediction

- Regression
- Classification



## Unsupervised learning

We have **unlabeled data**, and we want to make some prediction

- Clustering

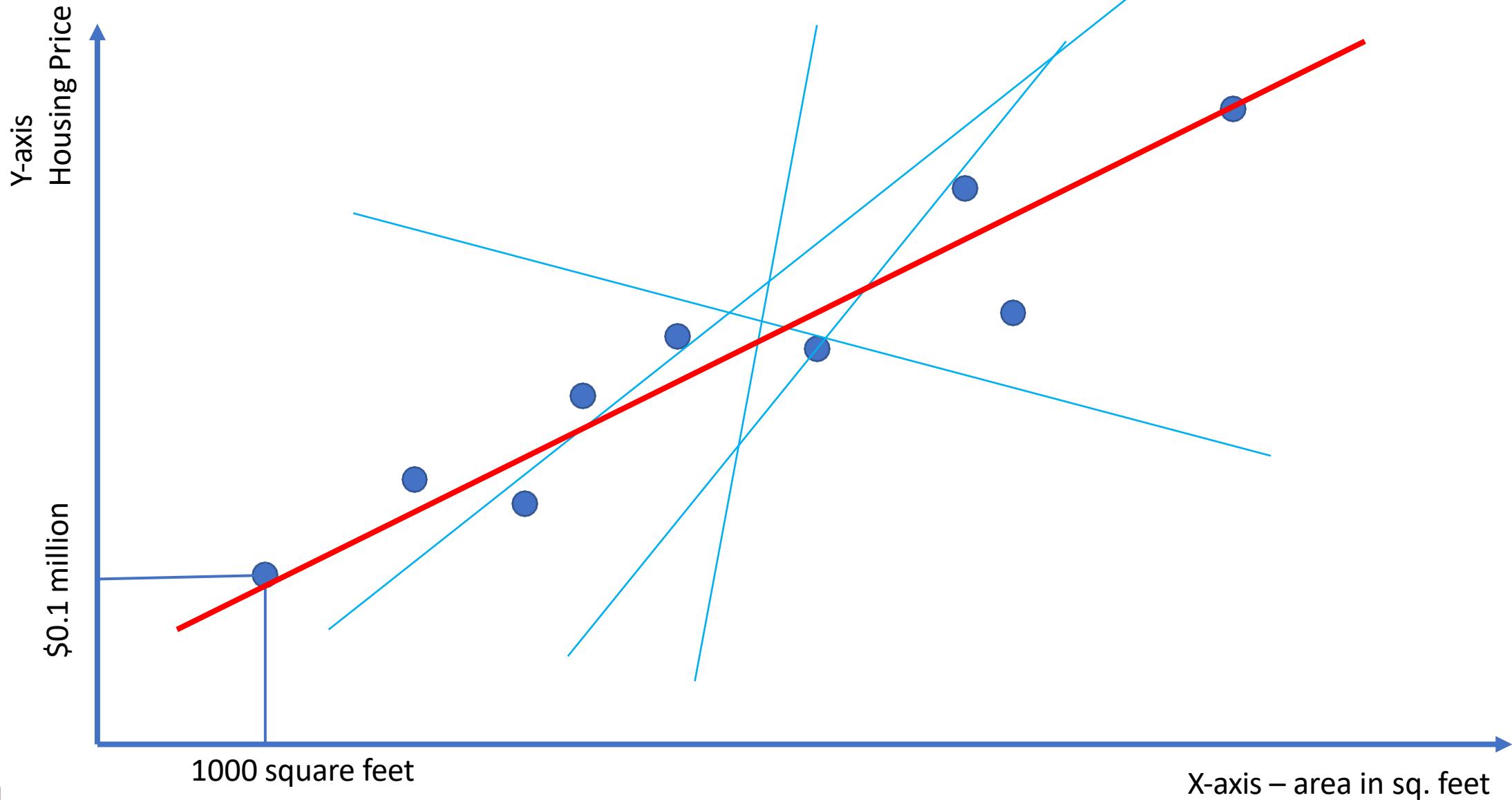
# Supervised learning



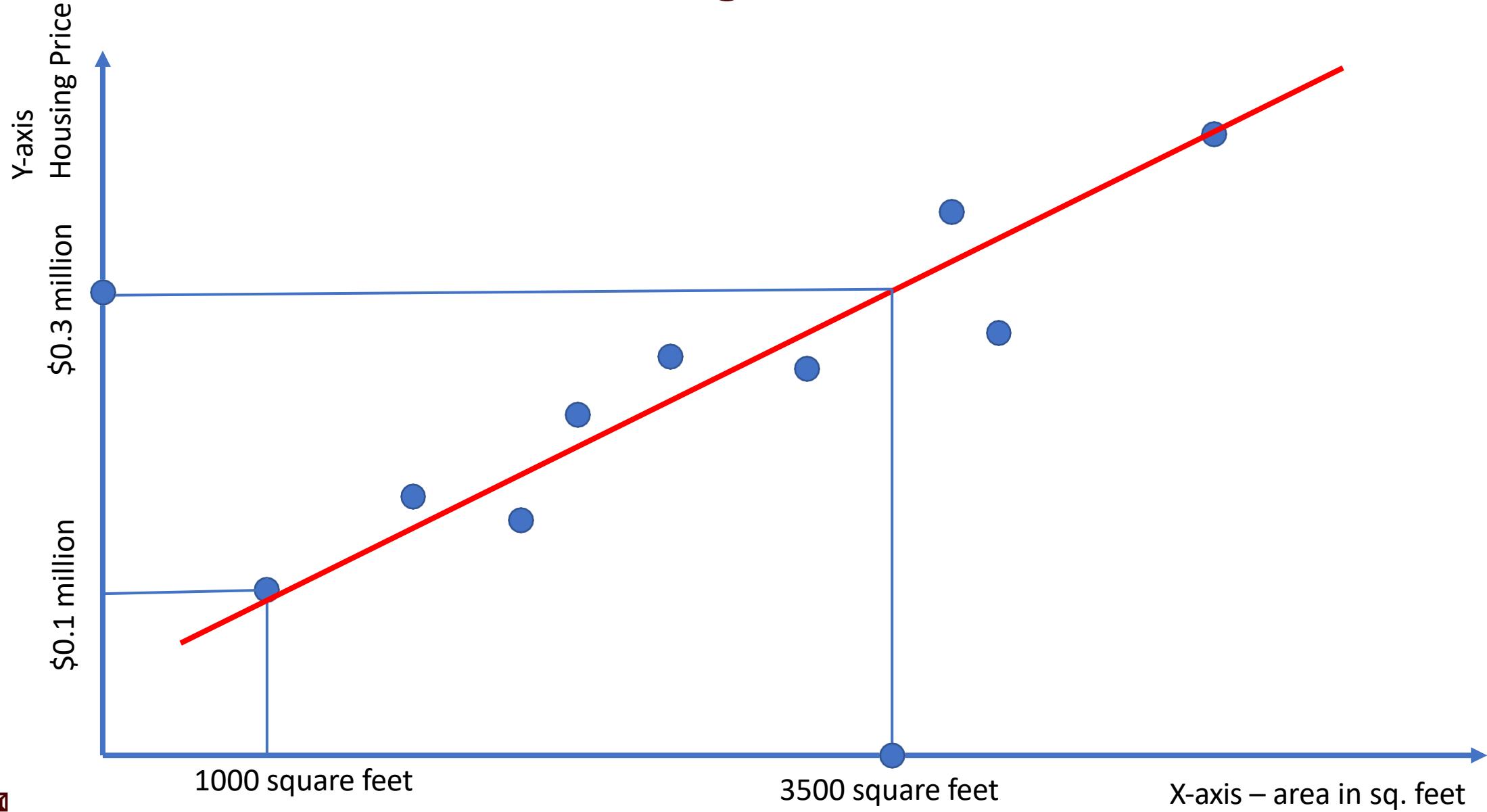
# Regression

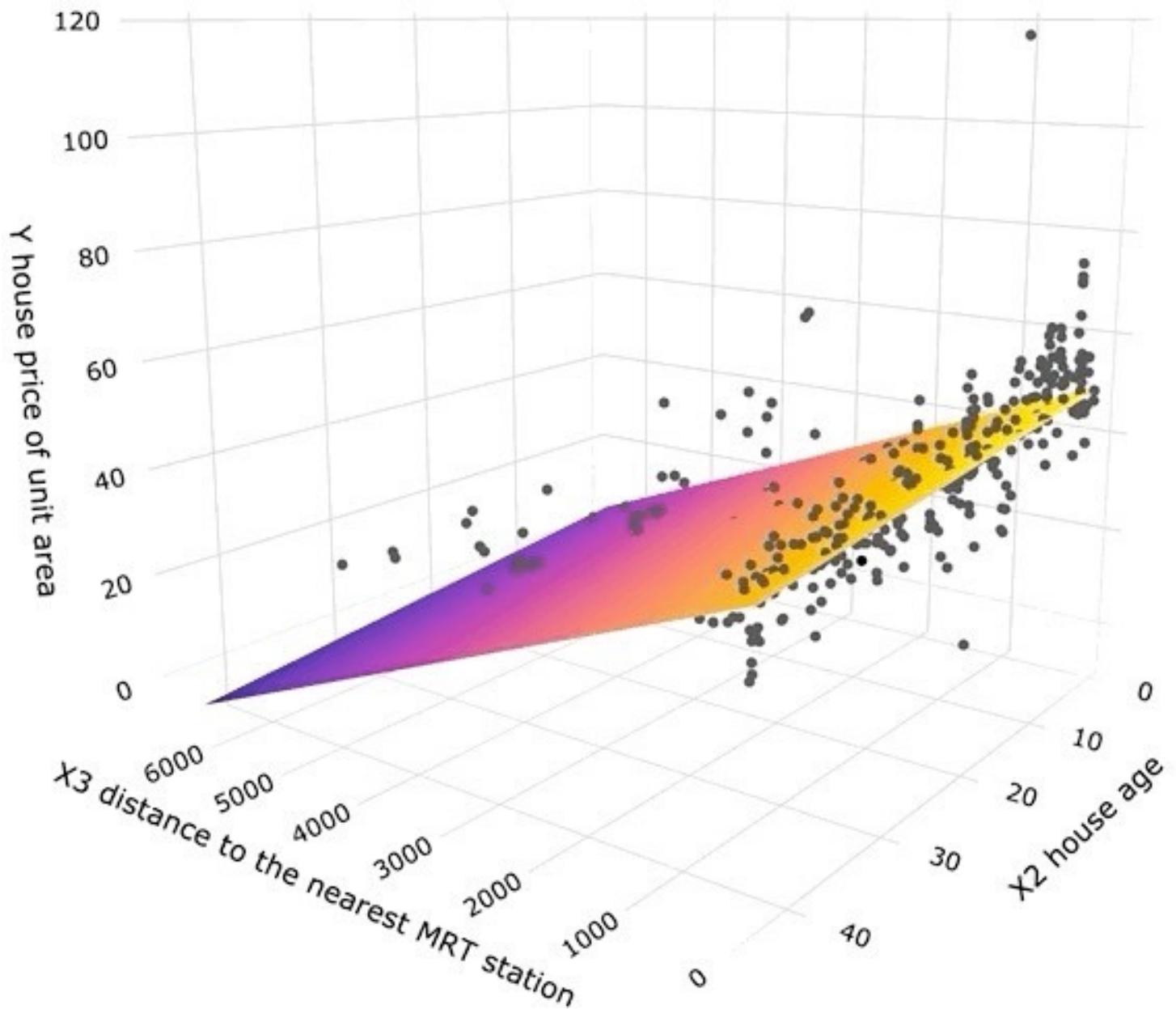


# Regression



# Regression





# Quiz

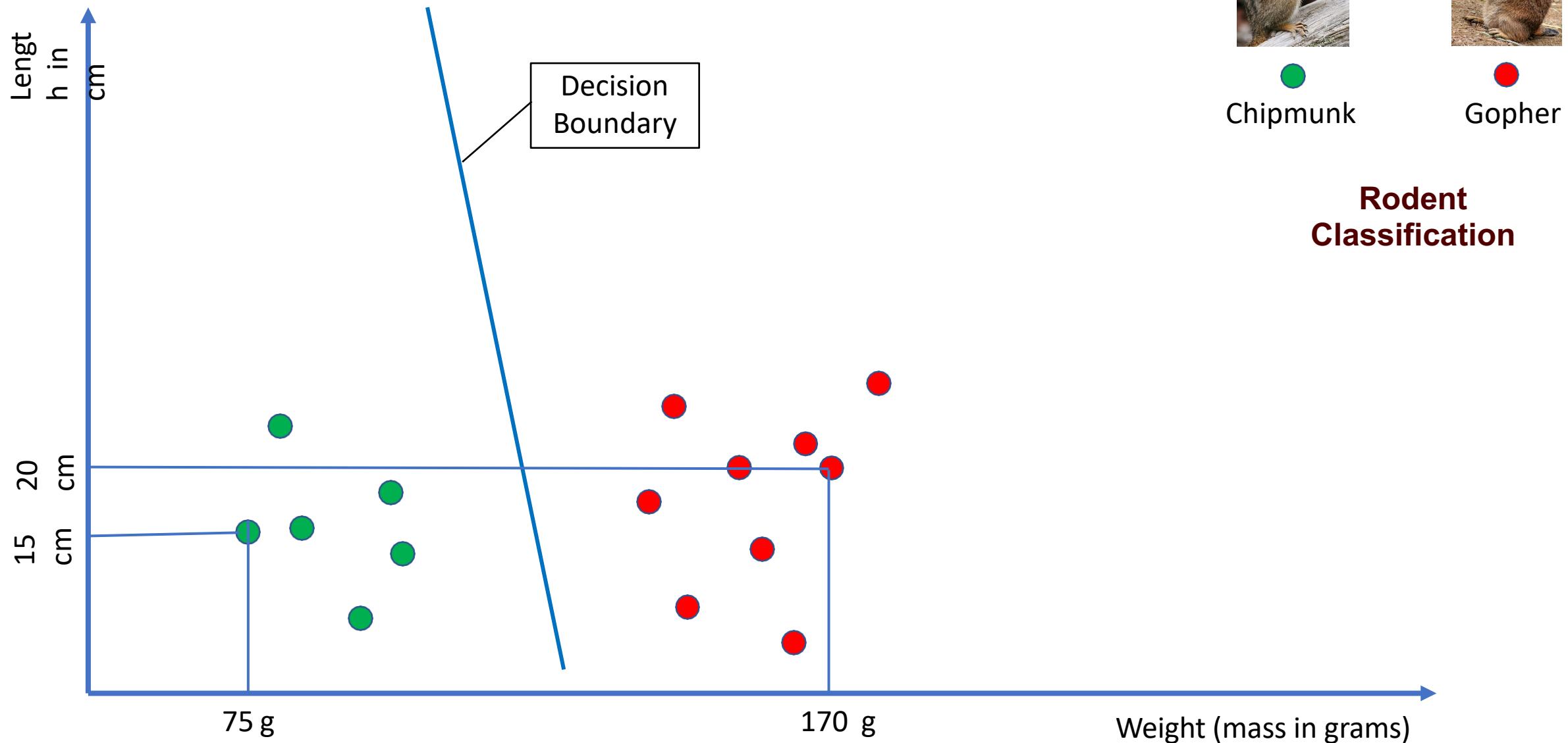
- Which of the following CANNOT be an example of regression?
  - A) Using past data of weather in college station to predict future's weather.
  - B) Predicting prices of stocks using previous month's price data
  - C) Determining if an email is spam or not
  - D) Determining network traffic for today using previous month's data



# Classification



# Classification



Rodent  
Classification

Chipmunk

Gopher



# Quiz

- Which of the following CANNOT be an example of classification?
  - A) Using blood pressure and weight data to determine if a patient is diabetic or not
  - B) Estimating amount of annual rain from previous year's data
  - C) Classifying Pokémon in different types (e.g., fire, ice, poison, electric)
  - D) Determining if an email is spam or not

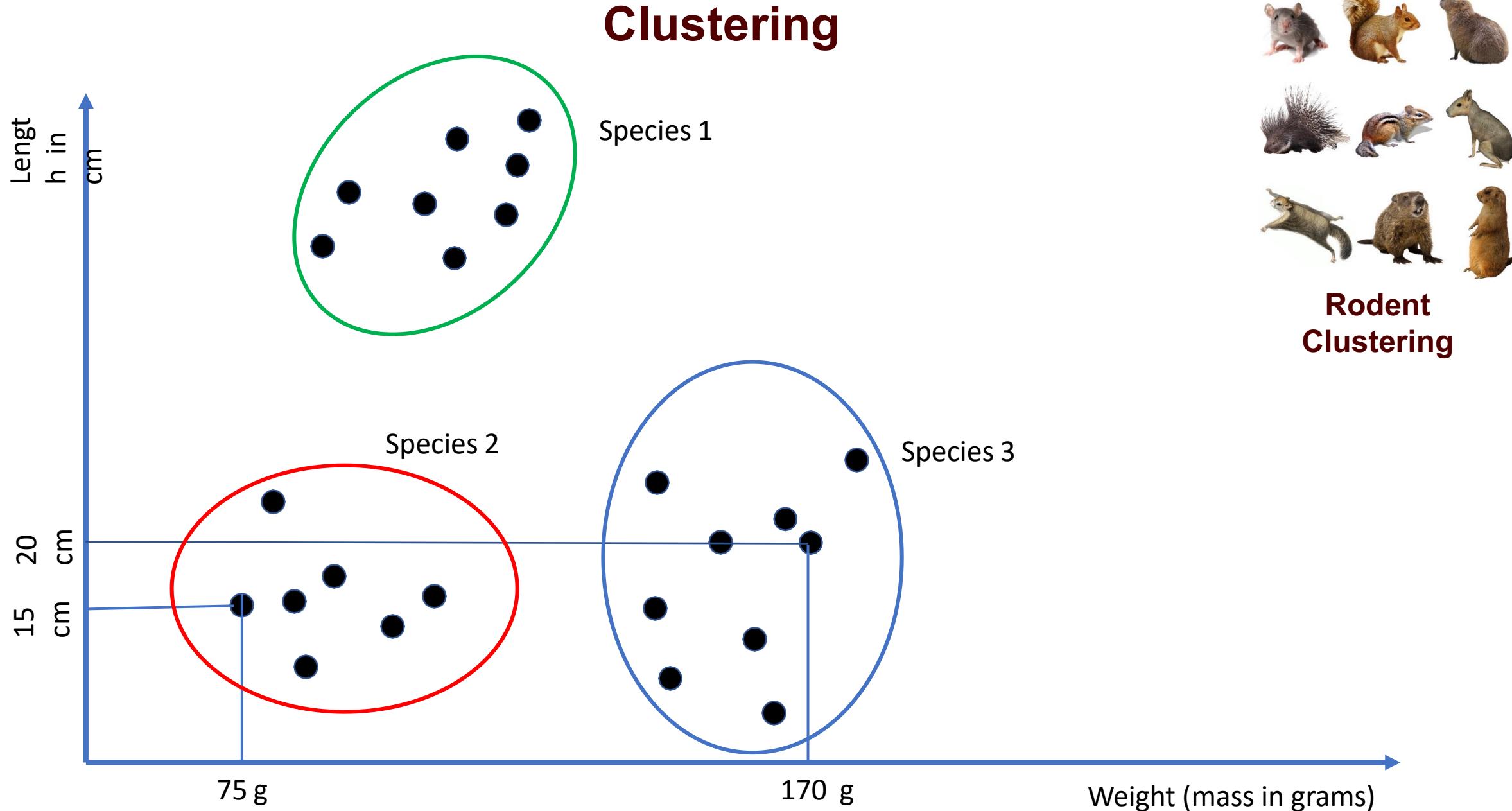


# Unsupervised learning



The background of the image is a deep space scene, filled with numerous small white stars of varying sizes. Interspersed among them are several larger, more luminous clusters of stars, some appearing as bright blue or white points and others as fainter, more diffuse patches. In the upper right quadrant, there is a prominent, multi-colored nebula with shades of blue, purple, and red, emitting a soft glow. The overall atmosphere is dark and mysterious, suggesting the vastness of the universe.

clustering



**Rodent Clustering**



# Quiz

- Which of the following CANNOT be an example of clustering?
  - A) Sorting and making groups of research papers having similar content
  - B) Determining whether a news article is about politics or sports
  - C) Identifying clusters of stars having similar characteristics
  - D) Sorting through subjects of emails and grouping them accordingly



# Quiz

- Which of the following CANNOT be an example of machine learning?  
**Select all that apply.**
  - A) Manually trying out different passwords on your amazon account to check if it works
  - B) Your virtual assistant starts recognizing your voice after first few tries
  - C) Fire alarm goes off when smoke level is more than a specific level
  - D) Sorting through subjects of emails and grouping them accordingly

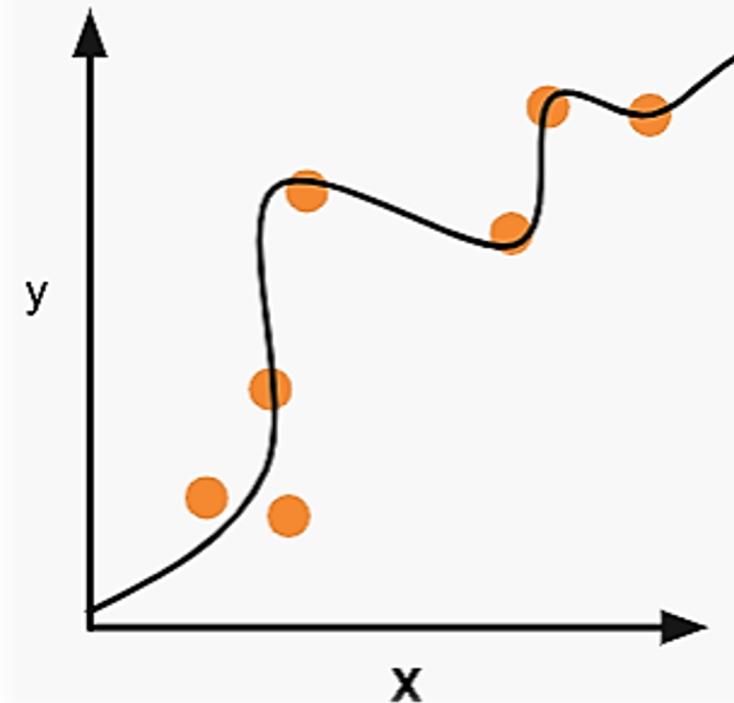


# What is Artificial Intelligence

- **Wikipedia:** intelligence demonstrated by machines as opposed to natural intelligence displayed by animals including humans.
- **Oxford:** the theory and development of computer systems able to perform tasks that normally require human intelligence.
- **IBM:** leverages computers and machines to mimic the problem-solving and decision-making capabilities of the human mind



# Train a linear regression model



# Jupyter Notebooks on UArizona HPC with Python.

ood.hpc.arizona.edu

The screenshot shows a file manager interface with two panes. The left pane displays a tree view of the 'Home Directory' containing various folders like 'Documents', 'MNIST\_data', 'Phenix', etc. The right pane shows a list of files in the directory '/home/u13/chrisreidy/'. Red arrows point from the top right towards the 'A-z Rename/Move' and 'Upload' buttons in the toolbar. The file list includes:

name	size	modified date
singularity	dir	07/02/2019
target	dir	10/01/2020
test	dir	06/16/2020
testtest	dir	04/30/2018
tutorial	dir	08/30/2018
1node_1V100.log	81.10kb	08/12/2020
Chirp.ipynb	44.77kb	11/27/2017
ML-HPC.ipynb	45.46kb	09/16/2021
Matlab-engine.ipynb	1.31kb	01/07/2019
Ocelote-Announce.pbs	1.95kb	08/02/2017
Ocelote-Announce.rtf	1.95kb	08/02/2017
README.binary	2.21kb	12/20/2018
Trinity.ipynb	8.09kb	11/27/2017
Untitled.ipynb	555b	09/16/2021
Untitled4.ipynb	55.68kb	09/16/2021



# Jupyter Notebooks on UArizona HPC with Python

The screenshot shows the UArizona Research Technologies HPC Systems web interface. At the top, there is a navigation bar with links for Files, Jobs, Clusters, Interactive Apps, and a user icon. The 'Interactive Apps' menu is currently open, displaying a list of available applications categorized into Desktops, GUIs, and Servers.

**Desktops:**

- Interactive Desktop

**GUIs:**

- ABAQUS GUI
- ANSYS Workbench GUI
- MATLAB GUI
- Mathematica GUI

**Servers:**

- Jupyter Notebook
- RStudio Server

A message in the top right corner states: "Please NOTE: 'windfall' jobs will be restarted if pre-empted by a 'standard'". Below the navigation bar, a success message says "Session was successfully deleted." The main content area shows the user is viewing "My Interactive Sessions".



# Jupyter Notebooks on UArizona HPC with Python

Interactive Apps

Desktops
Interactive Desktop
GUIs
ABAQUS GUI
ANSYS Workbench GUI
MATLAB GUI
Mathematica GUI
Servers
<b>Jupyter Notebook</b>
RStudio Server

**Jupyter Notebook**

This app will launch a [Jupyter](#) server using [Python](#) on a UAz cluster.

**Cluster**

Ocelote Cluster

**Run Time**

1

Enter maximum number of wall clock hours the job is allowed to run.

**Core count on a single node**

1

Enter the number of cores on a single node that the job is allowed to use.

**Memory per core**

6

Enter the number of Gigabytes of RAM needed per core.

**Special Options**

Enter node specific requirements, if any.

**PI Group**

chrisreidy

Enter an HPC PI group to be charged for time used.



# Jupyter Notebooks on UArizona HPC with Python

The screenshot shows the UArizona HPC interface. On the left, there is a sidebar titled "Interactive Apps" with options: Desktops, Interactive Desktop, GUIs, ABAQUS GUI, ANSYS Workbench GUI, MATLAB GUI, Mathematica GUI, and Servers. On the right, a session card for "Jupyter Notebook (169042)" is displayed. The card includes the host (">\_i14n0.ocelote.hpc.arizona.edu"), creation date ("Created at: 2021-09-16 18:54:53 MST"), time remaining ("Time Remaining: 59 minutes"), session ID ("Session ID: 46d680cf-27d0-45cb-a479-ef00c6ebb44a"), and a "Connect to Jupyter" button. A red arrow points from the "Connect to Jupyter" button towards the bottom-left of the slide.



# Jupyter Notebooks on UArizona HPC with Python

The screenshot shows the Jupyter Notebook interface. At the top, there is a navigation bar with tabs for "Files", "Running", and "Clusters". On the right side of the header, there are "Quit" and "Logout" buttons. Below the header, a message says "Select items to perform actions on them." A file list is displayed, showing various directories and files. A red arrow points from the text "Notebook" in the context menu to the "Python 3" option. The context menu also includes options for "Upload", "New", and "Terminal".

Files

Running Clusters

Select items to perform actions on them.

0 /

Name ↴

Notebook:

- Bash
- Julia 0.6.1
- Julia 1.0.0
- Python 3** (highlighted by a red arrow)
- R
- chrisreidy
- jupytertest

Other:

- Text File
- Folder
- Terminal

ciml-summer-institute-2021

conda

cuda

data

Documents

egl

eyra

git

hello-world

hpl

8 months ago

a year ago



# Jupyter Notebooks on UArizona HPC with Python

jupyter ML-HPC Last Checkpoint: 2 hours ago (autosaved)  Logout

File Edit View Insert Cell Kernel Widgets Help Trusted Python 3

In [1]: `import pandas as pd`

In [2]: `import numpy as np`

In [3]: `import matplotlib.pyplot as plt`

In [4]: `from sklearn.linear_model import LinearRegression`

In [5]: `from sklearn.model_selection import train_test_split`

In [6]: `# Load dat and view the first 5 rows`  
`data = pd.read_excel("king_county_house_data.xls")`

In [7]: `data.head(5)`

Out[7]:

	id	date	price	bedrooms	bathrooms	sqft_living	sqft_lot	floors	waterfront	view	...	grade	sqft
0	7129300520	20141013T000000	221900	3	1.00	1180	5650	1.0	0	0	...	7	
1	6414100192	20141209T000000	538000	3	2.25	2570	7242	2.0	0	0	...	7	
2	5631500400	20150225T000000	180000	2	1.00	770	10000	1.0	0	0	...	6	
3	2487200875	20141209T000000	604000	4	3.00	1960	5000	1.0	0	0	...	7	
4	1954400510	20150218T000000	510000	3	2.00	1680	8080	1.0	0	0	...	8	

5 rows × 21 columns



# Jupyter Notebooks on HPC

ood.hpc.arizona.edu

ocelote / 2 hours / 1 core / 6 mem / standard queue / chrisreidy

## Accessing files for the exercises

ssh [netid@hpc.arizona.edu](mailto:netid@hpc.arizona.edu)

shell

ocelote

mkdir intro-to-hpc

cd intro-to-hpc

<https://ua-researchcomputing-hpc.github.io/Intro-to-HPC/>

Then Accessing Workshop Files and cut / paste the section starting “wget”

(old method: cp /xdisk/chrisreidy/workshops/\* .)

Choice #1: Cut and paste commands into Jupyter from .txt file

Choice #2: Run the Notebook .ipynb file

Choice #3: Type in the commands. Syntax is very important



# Train a linear regression model

- Import libraries

```
# Import libraries

import pandas as pd
import numpy as np
import matplotlib.pyplot as plt

from sklearn.linear_model import LinearRegression
from sklearn.model_selection import train_test_split
```

- Use Pandas to load the data and view the first 5 rows

```
# Load data and view the first 5 rows
data = pd.read_excel("king_county_house_data.xlsx")

data.head(5)
```

	id	date	price	bedrooms	bathrooms	sqft_living	sqft_lot	floors	yr_built	yr_renovated	lat	long
0	7129300520	20141013T000000	221900	3	1.00	1180	5650	1.0	2014	0	47.6	-122.3
1	6414100192	20141209T000000	538000	3	2.25	2570	7242	2.0	2014	0	47.6	-122.3
2	5631500400	20150225T000000	180000	2	1.00	770	10000	1.0	2015	0	47.6	-122.3
3	2487200875	20141209T000000	604000	4	3.00	1960	5000	1.0	2014	0	47.6	-122.3
4	1954400510	20150218T000000	510000	3	2.00	1680	8080	1.0	2015	0	47.6	-122.3



# Train a linear regression model

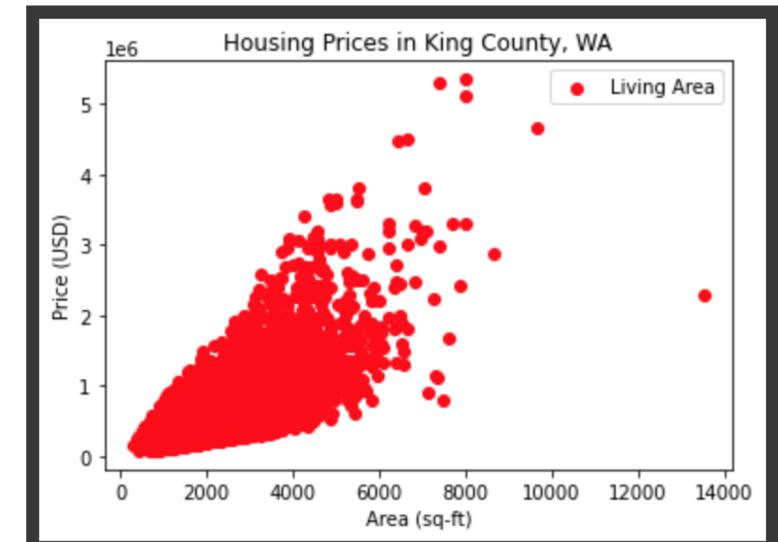
- Choose the columns from the data
- Split the data into train and test sets
- Visualize the train set

```
space = data['sqft_living']
price = data['price']

# Change X into 2D array
X = np.array(space).reshape(-1, 1)
Y = np.array(price)

# Split data into train sets and test sets
X_train,X_test,Y_train,Y_test = train_test_split(X,Y,test_size=1/3,random_state=0)
```

```
# Visualize training set
plt.scatter(X_train,Y_train,color="red",label="Living Area")
plt.title("Housing Prices in King County, WA")
plt.xlabel("Area (sq-ft)")
plt.ylabel("Price (USD)")
plt.legend()
plt.show()
```

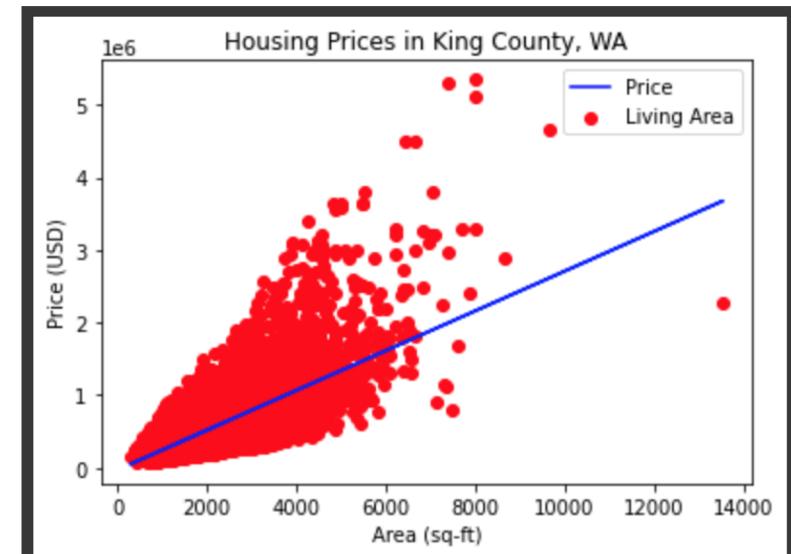


# Train a linear regression model

- Train the model with train set
- Predict on test set
- Visualize the train data and the best fit line

```
# Train  
regressor = LinearRegression()  
regressor.fit(X_train, Y_train)  
  
# Prediction  
y_pred = regressor.predict(X_test)
```

```
# Visualize the data and the bestfit line  
plt.scatter(X_train,Y_train,color="red",label="Living Area")  
plt.title("Housing Prices in King County, WA")  
plt.plot(X_train,regressor.predict(X_train),color="blue",label="Price")  
plt.xlabel("Area (sq-ft)")  
plt.ylabel("Price (USD)")  
plt.legend()  
plt.show()
```



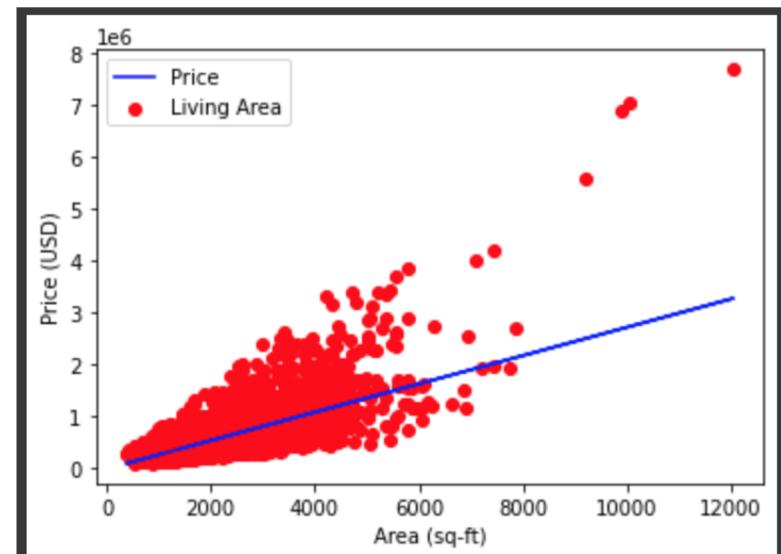
# Train a linear regression model

- Predict the price of a house with a certain area

```
area = 5000  
  
price = regressor.predict([[area]])  
  
print('House of %d sq-ft costs about $%d' % (area, price))  
  
House of 5000 sq-ft costs about $1339969
```

- Visualize the test data

```
# Visualize test set  
plt.scatter(X_test,Y_test,color='red',label="Living Area")  
plt.plot(X_test,regressor.predict(X_test),color="blue",label="Price")  
plt.xlabel("Area (sq-ft)")  
plt.ylabel("Price (USD)")  
plt.legend()  
plt.show()
```

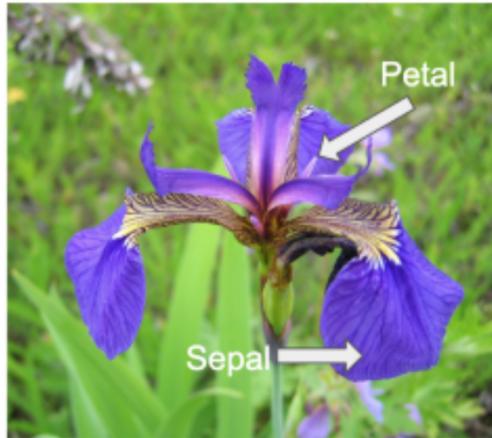


# Build a clustering model for Iris Dataset

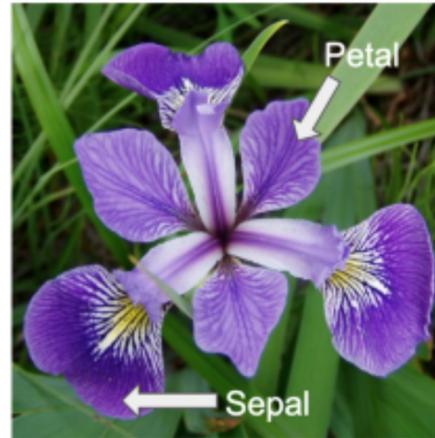


# Build a clustering model – Iris dataset

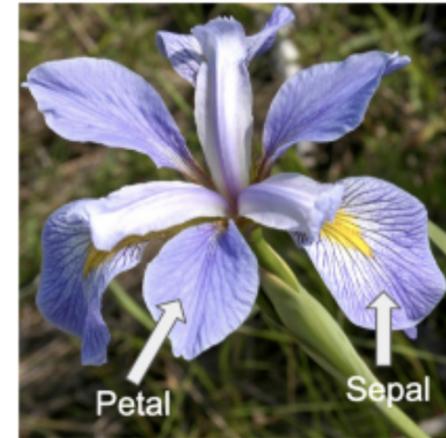
*Iris setosa*



*Iris versicolor*



*Iris virginica*



- Import libraries

```
[1] #import libraries
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
%matplotlib inline
from sklearn.cluster import KMeans
from sklearn.datasets import load_iris
```



# Build a clustering model – Iris dataset

- Load the data

```
[2] iris=load_iris()
iris

{'DESCR': '.. _iris_dataset:\n\nIris plants dataset\n--\n'data': array([[5.1, 3.5, 1.4, 0.2],
   [4.9, 3. , 1.4, 0.2],
   [4.7, 3.2, 1.3, 0.2],
   [4.6, 3.1, 1.5, 0.2],
   [5. , 3.6, 1.4, 0.2],
   [5.4, 3.9, 1.7, 0.4],
   [4.6, 3.4, 1.4, 0.3],
   [5. , 3.4, 1.5, 0.2],
   [4.4, 2.9, 1.4, 0.2],
   [4.9, 3.1, 1.5, 0.1],
```

```
[3] df=pd.DataFrame(data=iris.data, columns=['sepal length','sepal width','petal length','petal width'])
df['target']=pd.Series(iris.target)
df
```

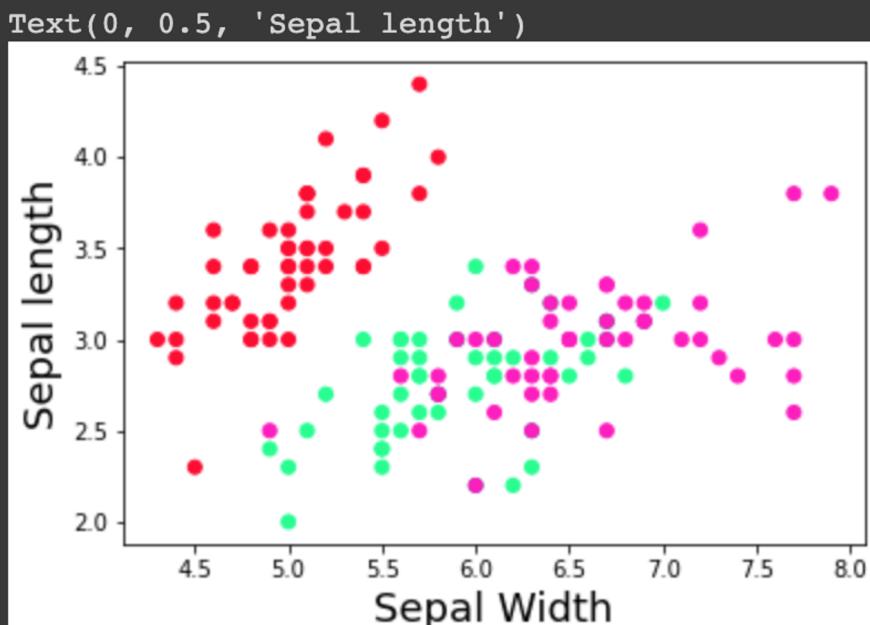
	sepal length	sepal width	petal length	petal width	target
0	5.1	3.5	1.4	0.2	0
1	4.9	3.0	1.4	0.2	0
2	4.7	3.2	1.3	0.2	0
3	4.6	3.1	1.5	0.2	0
4	5.0	3.6	1.4	0.2	0
...	...	...	...	...	...
145	6.7	3.0	5.2	2.3	2



# Build a clustering model – Iris dataset

- Visualize the data

```
[4] plt.scatter(x=df['sepal length'], y=df['sepal width'], c=iris.target, cmap='gist_rainbow')  
  
plt.xlabel('Sepal Width', fontsize=18)  
plt.ylabel('Sepal length', fontsize=18)
```



# Build a clustering model – Iris dataset

- Estimate k with elbow method- first try  $k = 5$

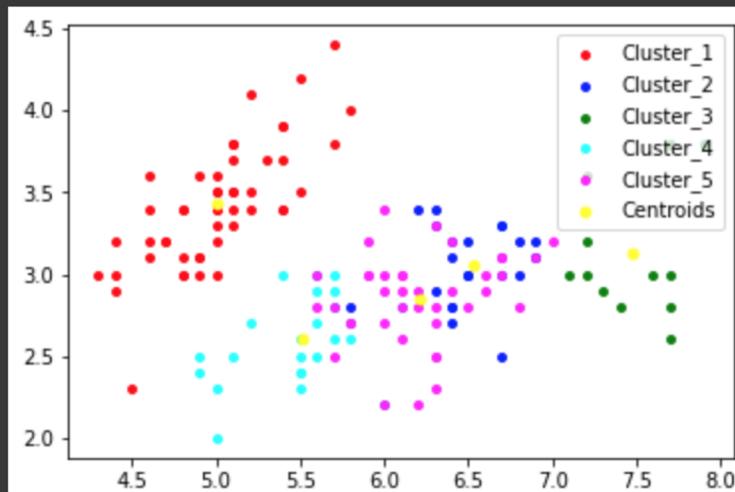


# Build a clustering model – Iris dataset

- Estimate k with elbow method

```
[7] plt.scatter(x[y == 0,0], x[y==0,1], s = 15, c= 'red', label = 'Cluster_1')
    plt.scatter(x[y == 1,0], x[y==1,1], s = 15, c= 'blue', label = 'Cluster_2')
    plt.scatter(x[y == 2,0], x[y==2,1], s = 15, c= 'green', label = 'Cluster_3')
    plt.scatter(x[y == 3,0], x[y==3,1], s = 15, c= 'cyan', label = 'Cluster_4')
    plt.scatter(x[y == 4,0], x[y==4,1], s = 15, c= 'magenta', label = 'Cluster_5')

    plt.scatter(kmeans5.cluster_centers_[:,0], kmeans5.cluster_centers_[:,1], s = 25, c = 'yellow', label = 'Centroids')
    plt.legend()
    plt.show()
```

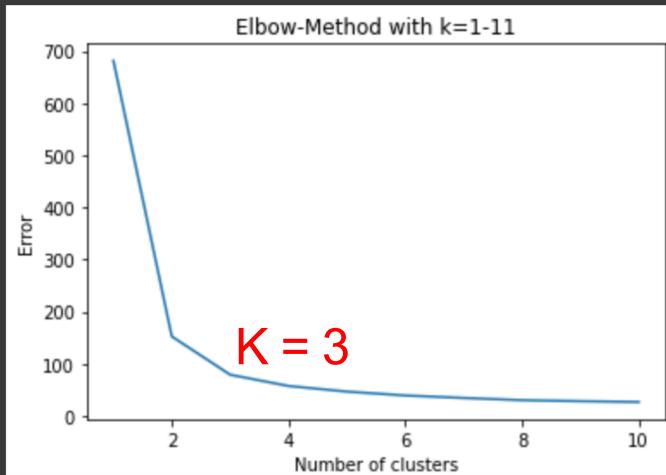


# Build a clustering model – Iris dataset

- Estimate k with elbow method

```
[8] Error = []
    for i in range(1, 11):
        kmeans11 = KMeans(n_clusters = i, init = 'k-means++', max_iter = 300, n_init = 10, random_state = 0).fit(x)
        kmeans11.fit(x)
        Error.append(kmeans11.inertia_)

    import matplotlib.pyplot as plt
    plt.plot(range(1, 11), Error)
    plt.title('Elbow-Method with k=1-11') #within cluster sum of squares
    plt.xlabel('Number of clusters')
    plt.ylabel('Error')
    plt.show()
```

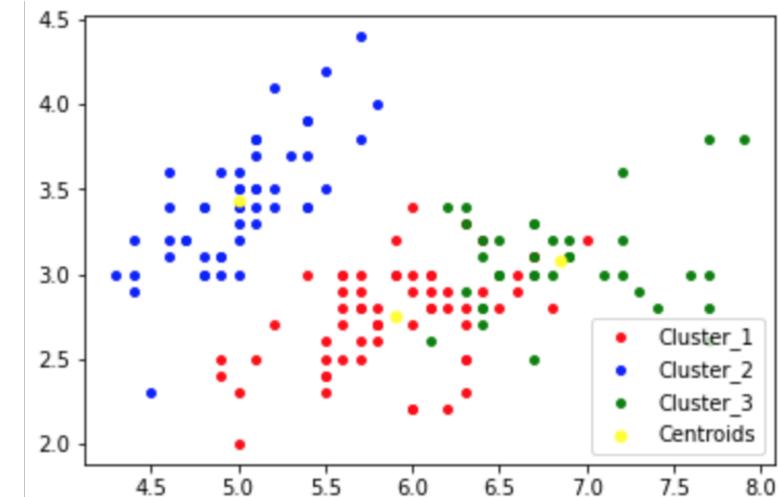


# Build a clustering model – Iris dataset

- Get the optimal  $k = 3$  from the elbow method. Cluster centers

```
[9] kmeans3 = KMeans(n_clusters=3, random_state=21)
y = kmeans3.fit_predict(x)
kmeans3.cluster_centers_

array([[5.9016129 , 2.7483871 , 4.39354839, 1.43387097],
       [5.006      , 3.428      , 1.462      , 0.246      ],
       [6.85       , 3.07368421, 5.74210526, 2.07105263]])
```

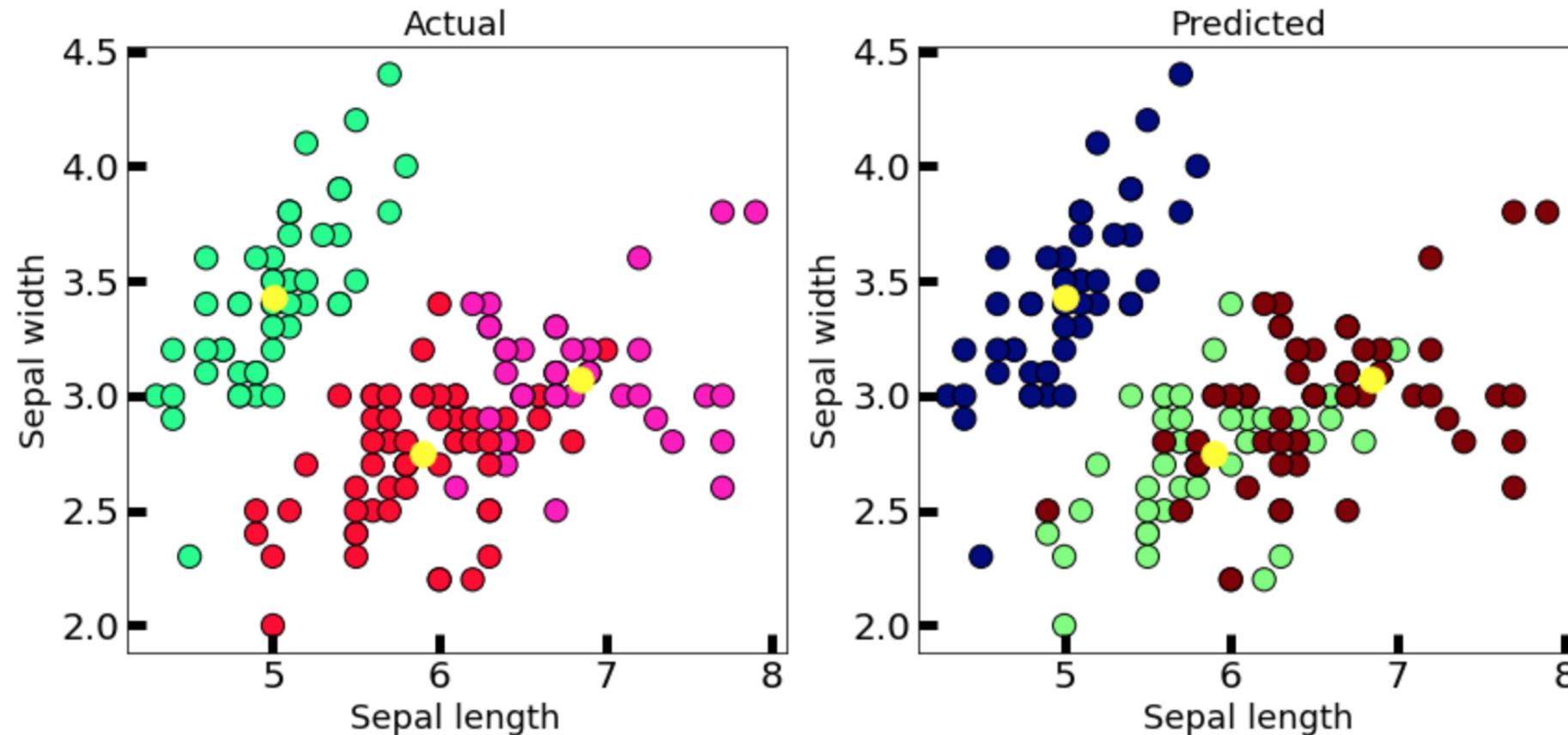


```
[11] plt.scatter(x[y == 0,0], x[y==0,1], s = 15, c= 'red', label = 'Cluster_1')
plt.scatter(x[y == 1,0], x[y==1,1], s = 15, c= 'blue', label = 'Cluster_2')
plt.scatter(x[y == 2,0], x[y==2,1], s = 15, c= 'green', label = 'Cluster_3')
plt.scatter(kmeans3.cluster_centers_[:,0], kmeans3.cluster_centers_[:,1], s = 25, c = 'yellow', label = 'Centroids')
plt.legend()
plt.show()
```



# Build a clustering model – Iris dataset

- Compared the actual and predicted clusters



# Getting help

- HPC documentation [docs.hpc.arizona.edu](https://docs.hpc.arizona.edu)
- Support ticket  
[https://uarizona.servicenow.com/sp?id=sc\\_cat\\_item&sys\\_id=2983102adbd23c109627d90d689619c6&sysparm\\_category=84d3d1acdbc8f4109627d90d6896191f](https://uarizona.servicenow.com/sp?id=sc_cat_item&sys_id=2983102adbd23c109627d90d689619c6&sysparm_category=84d3d1acdbc8f4109627d90d6896191f)
- Office Hours – Wednesday 2-4 PM  
<https://gather.town/app/dVsAprPNBVmI9NpL/hpc-office-hours>
- HPC consulting  
[hpc-consult@list.arizona.edu](mailto:hpc-consult@list.arizona.edu)
- Visualization consulting  
[vislab-consult@list.arizona.edu](mailto:vislab-consult@list.arizona.edu)
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