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Technology:Innovation:Business

Machine Learning Applications using Python

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For questions, write us on ai@madblocks.tech



Agenda

Session – 1: Machine Learning Intro

- What is Machine Learning?
- Applications of Machine Learning
- Types of Machine Learning
- Tools and Pre-Requisites

Session – 2: Data Analysis using Python

- MQTT (Data Collection)
- Data Cleaning
- Data Pre-Processing
- Data Analysis Examples

Session – 3: Building Crop Prediction System

- Dataset Review
- Data Cleaning
- Model Training
- API Creation and Demo





Session – 1: Intro to Machine Learning

Machine Learning

“Learning is any process by which a system improves performance from experience.”

- Herbert Simon

Definition by Tom Mitchell (1998):

Machine Learning is the study of algorithms that

- improve their performance P
- at some task T
- with experience E .

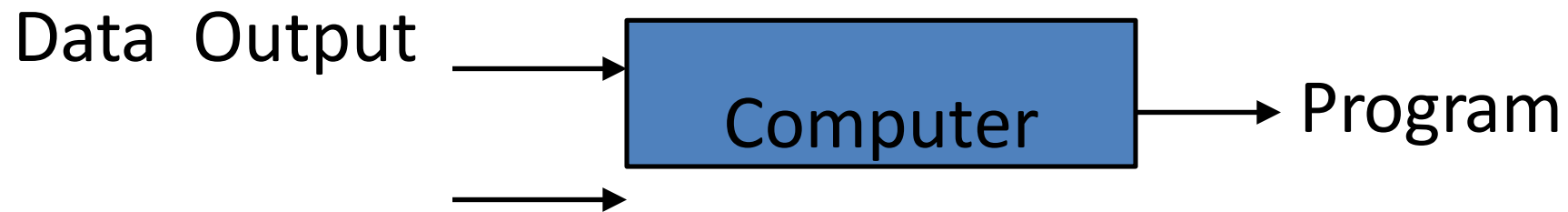
A well-defined learning task is given by $\langle P, T, E \rangle$.



Traditional Programming



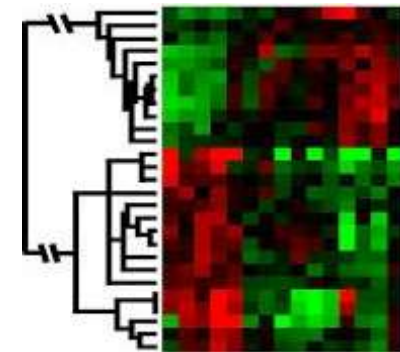
Machine Learning



When Do We Use Machine Learning?

ML is used when:

- Human expertise does not exist (navigating on Mars)
- Humans can't explain their expertise (speech recognition)
- Models must be customized (personalized medicine)
- Models are based on huge amounts of data (genomics)



Some more examples of tasks that are best solved by using a learning algorithm

- Recognizing patterns:
 - Facial identities or facial expressions
 - Handwritten or spoken words
 - Medical images
- Generating patterns:
 - Generating images or motion sequences
- Recognizing anomalies:
 - Unusual credit card transactions
 - Unusual patterns of sensor readings in a nuclear power plant
- Prediction:
 - Future stock prices or currency exchange rates

Sample Applications

- Web search
- Computational biology
- Finance
- E-commerce
- Space exploration
- Robotics
- Information extraction
- Social networks
- Debugging software
- Healthcare

“Machine Learning: Field of study that gives computers the ability to learn without being explicitly programmed.” -Arthur Samuel (1959)



Defining the Learning Task

Improve on task T, with respect to performance metric P, based on experience E

T: Playing checkers

P: Percentage of games won against an arbitrary opponent

E: Playing practice games against itself

T: Recognizing hand-written words

P: Percentage of words correctly classified

E: Database of human-labeled images of handwritten words

T: Driving on four-lane highways using vision sensors

P: Average distance traveled before a human-judged error

E: A sequence of images and steering commands recorded while observing a human driver

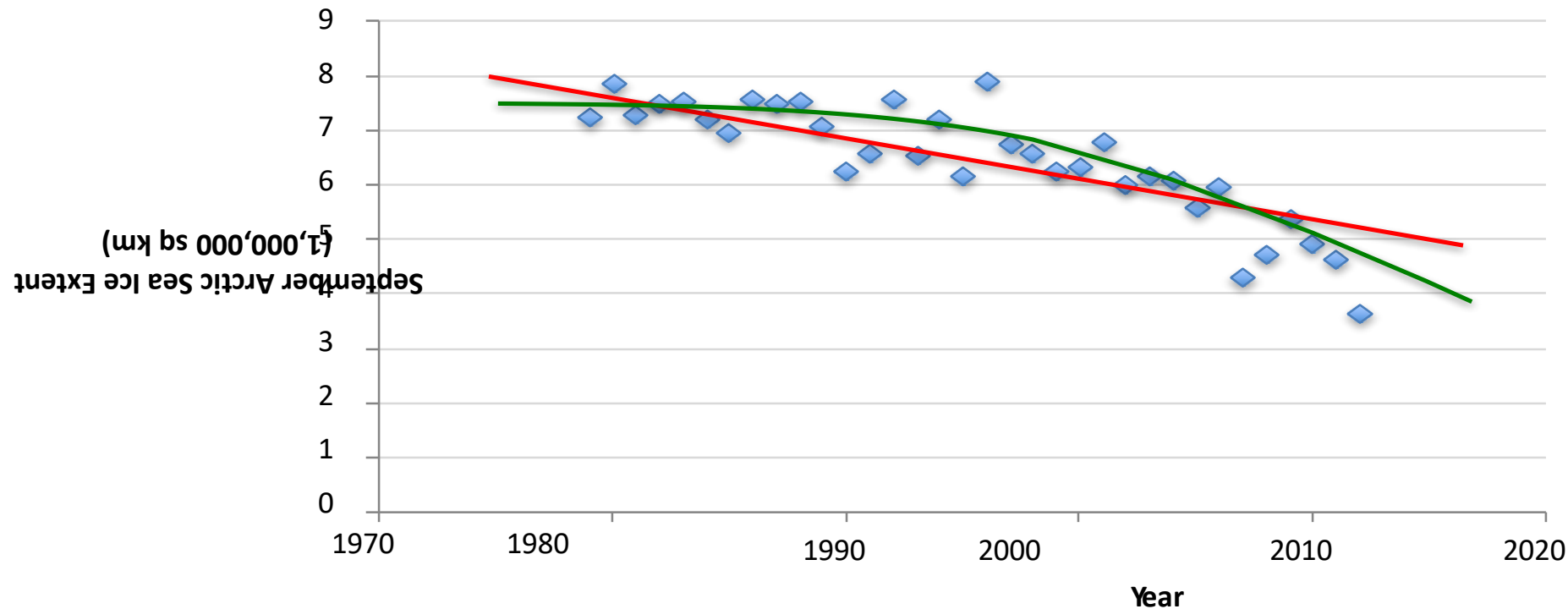
Types of Learning

Types of Learning

- **Supervised (inductive) learning**
 - Given: training data + desired outputs (labels)
- **Unsupervised learning**
 - Given: training data (without desired outputs)
- **Semi-supervised learning**
 - Given: training data + a few desired outputs
- **Reinforcement learning**
 - Rewards from sequence of actions

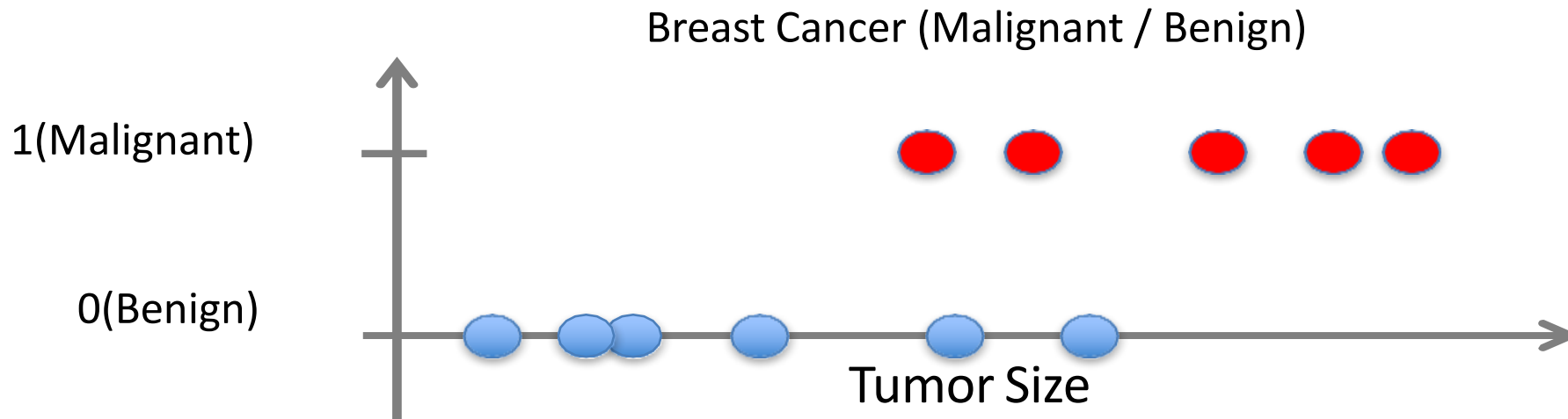
Supervised Learning: Regression

- Given $(x_1, y_1), (x_2, y_2), \dots, (x_n, y_n)$
- Learn a function $f(x)$ to predict y given x
 - y is real-valued == regression



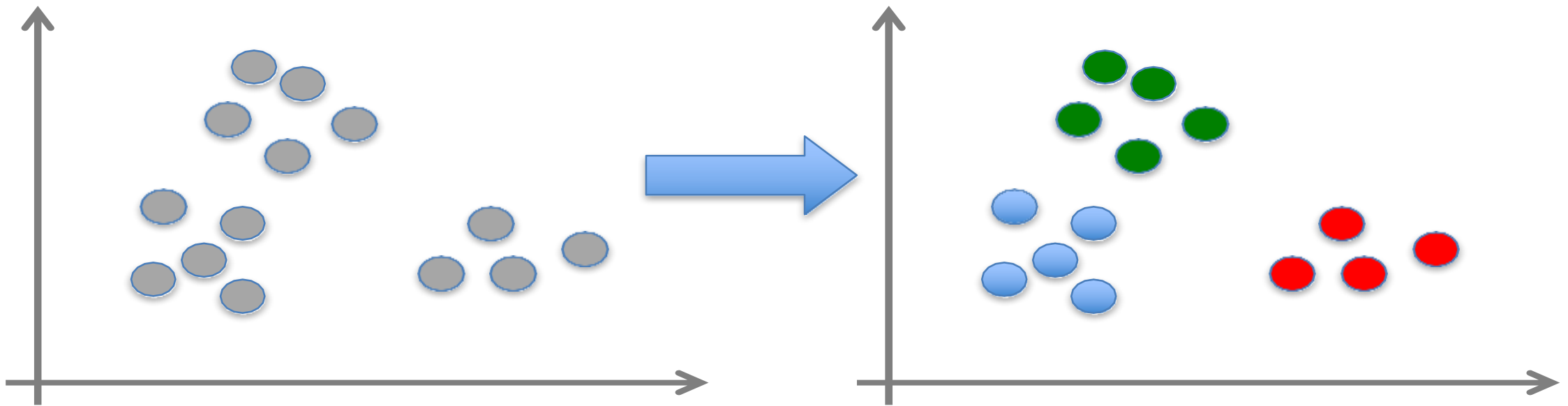
Supervised Learning: Classification

- Given $(x_1, y_1), (x_2, y_2), \dots, (x_n, y_n)$
- Learn a function $f(x)$ to predict y given x
 - y is categorical == classification

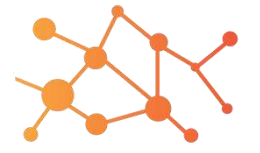


Unsupervised Learning

- Given x_1, x_2, \dots, x_n (without labels)
- Output hidden structure behind the x 's
 - E.g., clustering

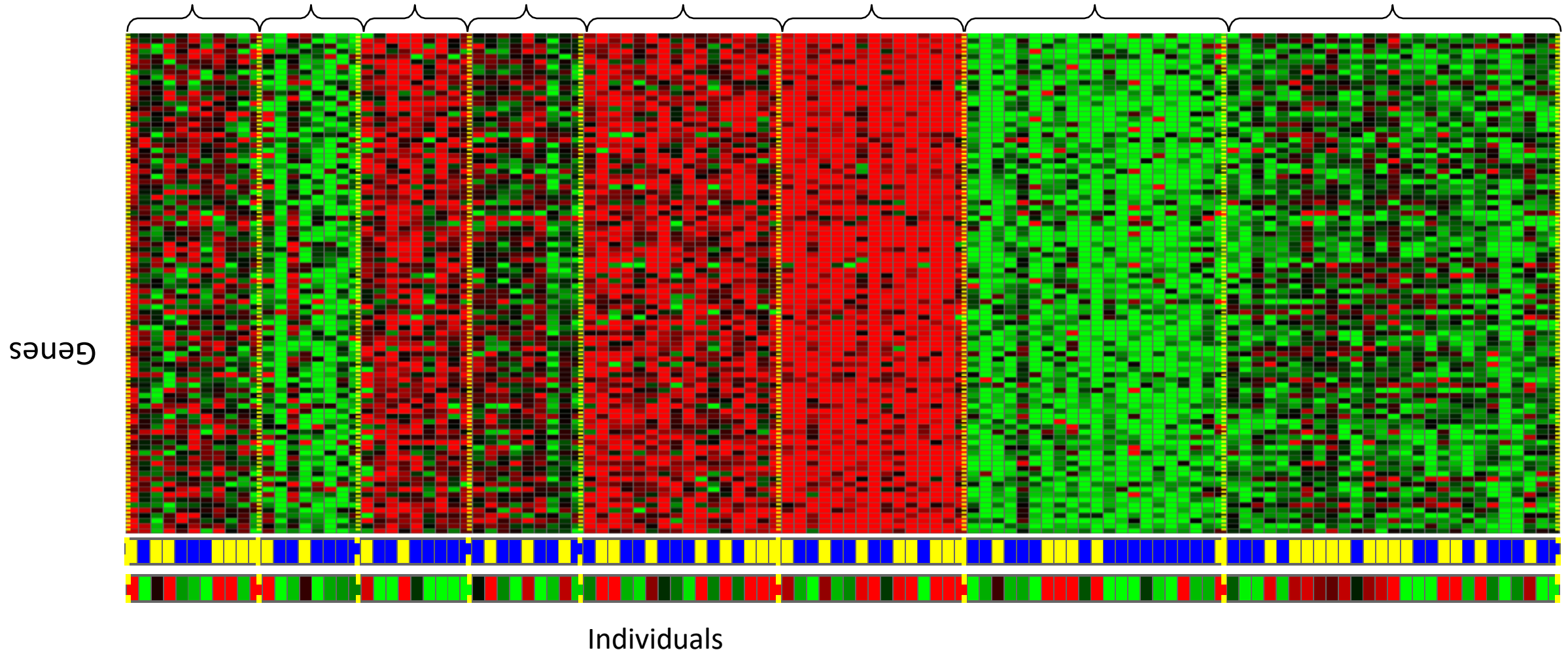


Unsupervised Learning



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Genomics application: group individuals by genetic similarity



Unsupervised Learning



Organize computing clusters



Market segmentation

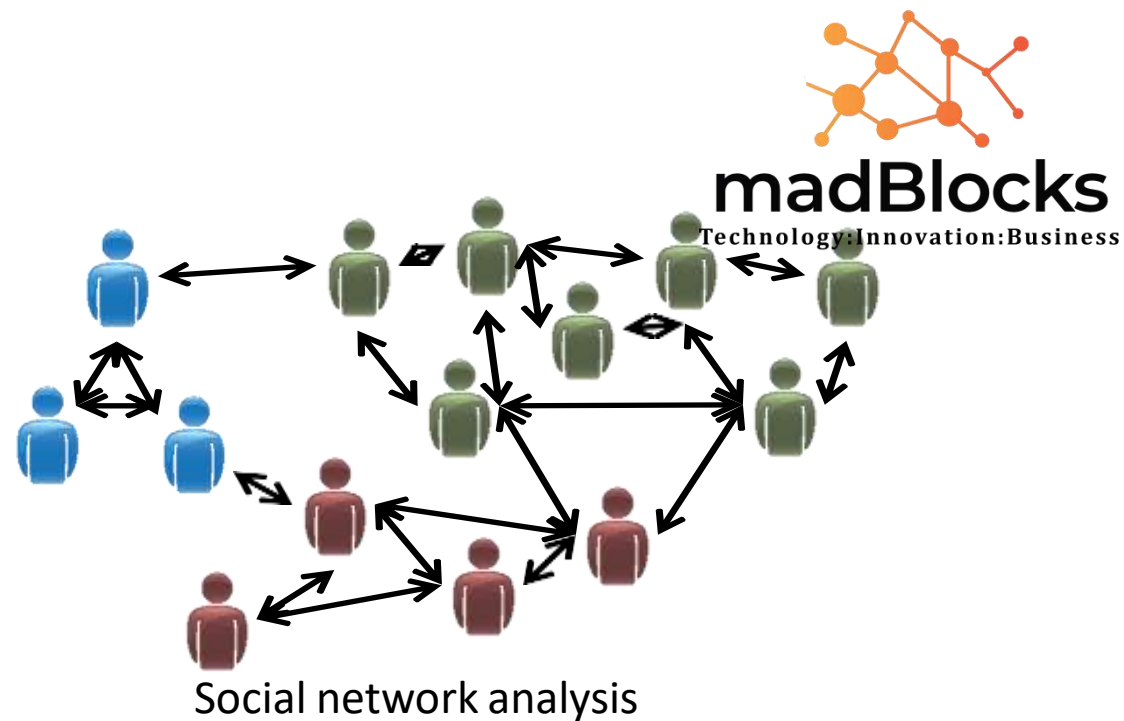


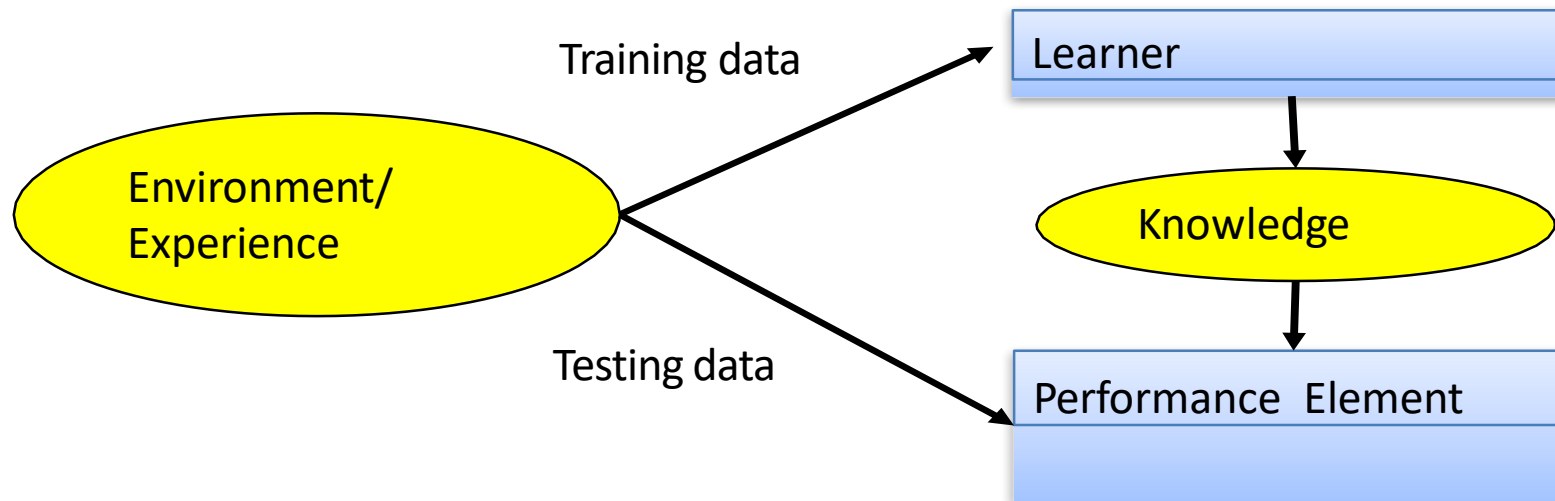
Image credit: NASA/JPL-Caltech/E. Churchwell (Univ. of Wisconsin, Madison)

Astronomical data analysis

Framing a Learning Problem

Designing a Learning System

- Choose the training experience
- Choose exactly what is to be learned
 - i.e. the **target function**
- Choose how to represent the target function
- Choose a learning algorithm to infer the target function from the experience



ML in a Nutshell

- Tens of thousands of machine learning algorithms
 - Hundreds new every year
- Every ML algorithm has three components:
 - **Representation**
 - **Optimization**
 - **Evaluation**

Various Function Representations

- Numerical functions
 - Linear regression
 - Neural networks
 - Support vector machines
- Symbolic functions
 - Decision trees
 - Rules in propositional logic
 - Rules in first-order predicate logic
- Instance-based functions
 - **Nearest-neighbor**
 - Case-based
- Probabilistic Graphical Models
 - Naïve Bayes
 - Bayesian networks
 - Hidden-Markov Models (HMMs)
 - Probabilistic Context Free Grammars (PCFGs)
 - Markov networks



Summary

Pack-Up!

- We gone through the basic understanding of Machine Learning and applications of ML.
- Machine to get learned can be broadly divided into four types (supervised, unsupervised, semisupervised and reinforcement learning).
- Data Pre-Processing is a very important step on the dataset.

**THANK
YOU!**



Session – 2:

Data Analysis using Python



Data Collection

Data Collection

Let's collect the data from IoT devices:

The data has to be collected from the IoT devices through MQTT Protocol and the received data will be stored in the database server.

The collected data has to be pre-processed for optimizing the model.





MQTT Protocol

Features

Small code footprint,
Ideal if processor or memory
resources are limited,
Ideal if bandwidth is low or
network is unreliable,
Works on top of TCP/IP





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Publisher – Sender,
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Broker and Topic

Broker – Server connecting Pub and Sub
Topic – On which topic the messages are shared
Port – 1883 (Default Broker)
Broker – **Your Gateway IP**





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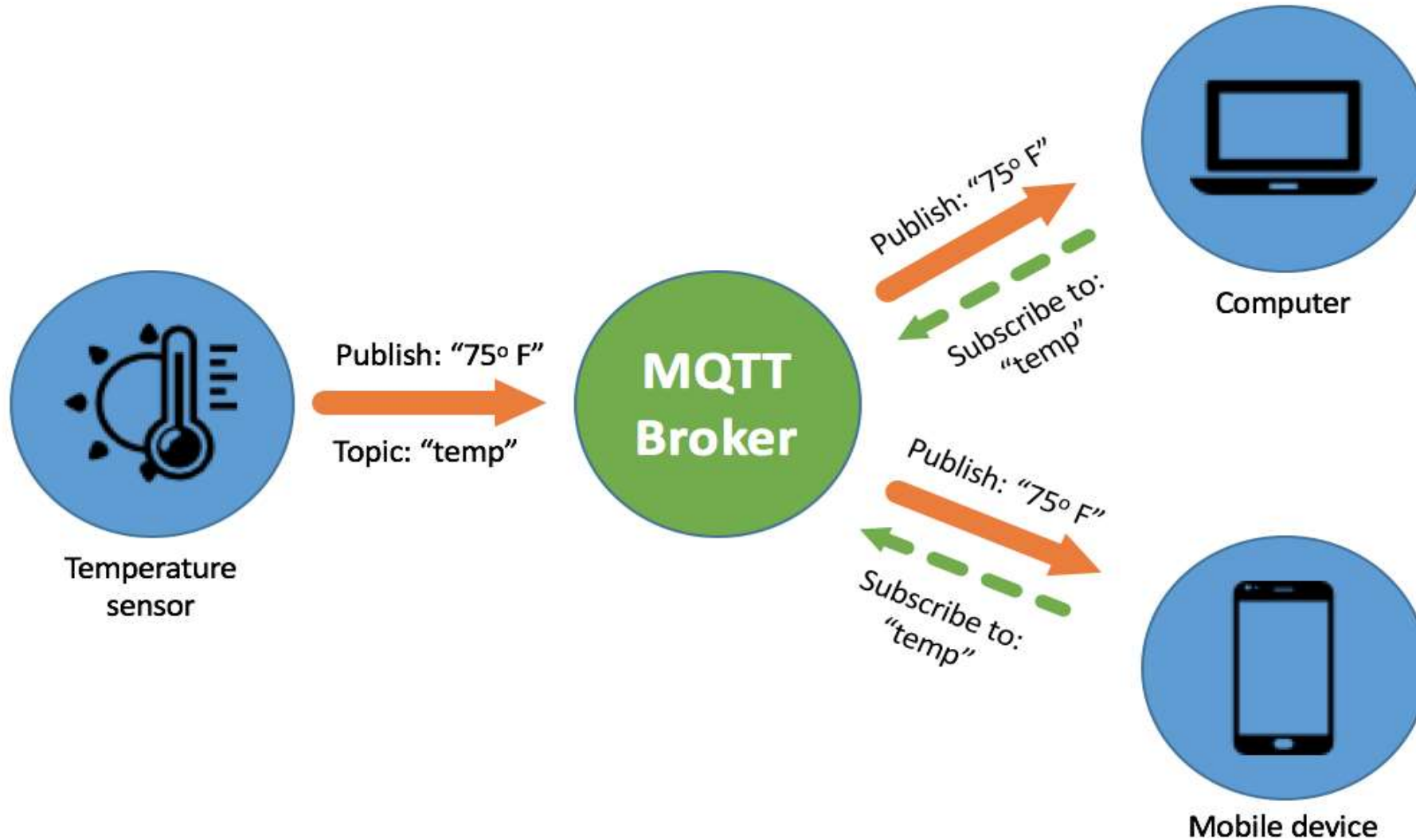
Broker – Server connecting Pub and Sub
Topic – On which topic the messages are shared
Port – 1883 (Default Broker)
Broker – **Your Gateway IP**

Applications

Smart Home
Smart City
Smart Healthcare
Smart Agriculture
Smart Industries etc....



MQTT Flow





Work Flow

IoT Devices: Sensory Connected Devices

Protocol: MQTT

Publisher: IoT Devices

Subscriber: Google Colab

Step – 1: Create a Publisher Script

Step – 2: Create a Subscriber Script

Step – 3: Store Data in CSV

Step – 4: Pre-Process the data





Session – 3: Building Crop Prediction System



Problem Statement

The developers team of Anurag University is developing an intelligent crop prediction system for the farming community.

So, let's jump in to support them.

Problem Statement:

Farmers yield various crops as per their knowledge without much study on various parameters which affects the yielding. They follow a void method based on what surrounding neighbors do.

So an engineering group from the university campus wants to develop an assistant to suggest the crop by taking the following parameters into consideration:

1. Temperature
2. Humidity
3. pH
4. rainfall





Explore the Dataset

	temperature	humidity	ph	rainfall	label
0	20.879744	82.002744	6.502985	202.935536	rice
1	21.770462	80.319644	7.038096	226.655537	rice
2	23.004459	82.320763	7.840207	263.964248	rice
3	26.491096	80.158363	6.980401	242.864034	rice
4	20.130175	81.604873	7.628473	262.717340	rice

Each crop has sets of temperature, humidity, ph and rainfall values stored in the dataset.

There were 3100 samples for 31 crops, and now we have to create a model to predict the crop when these values are entered.





Data Pre-Processing

1. Identify the empty fields, if any then try to modify with mean or average of the respective data.
2. Identify the range of data, if the difference is too large then use some standardization methods like Standard Deviation.
3. Identify the categorical data, if found use one-hot encoding and convert them to numerical data.
4. Identify the input variables and target variables from the features of the dataset.
5. Split the data into Training and Testing.





K Nearest Neighbors

KNN can be used for classification — the output is a class membership (predicts a class — a discrete value).

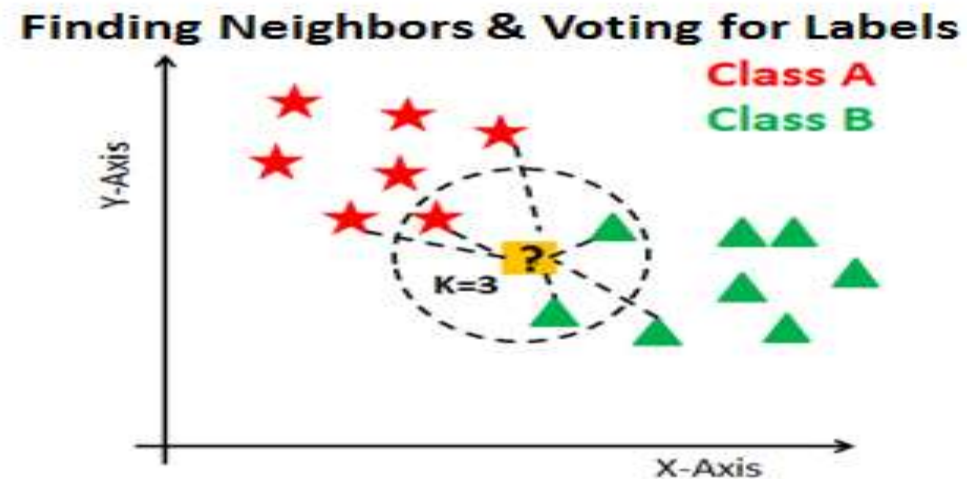
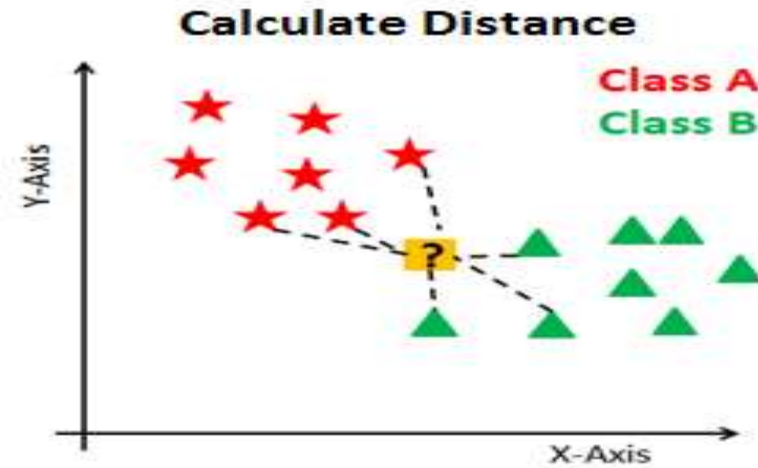
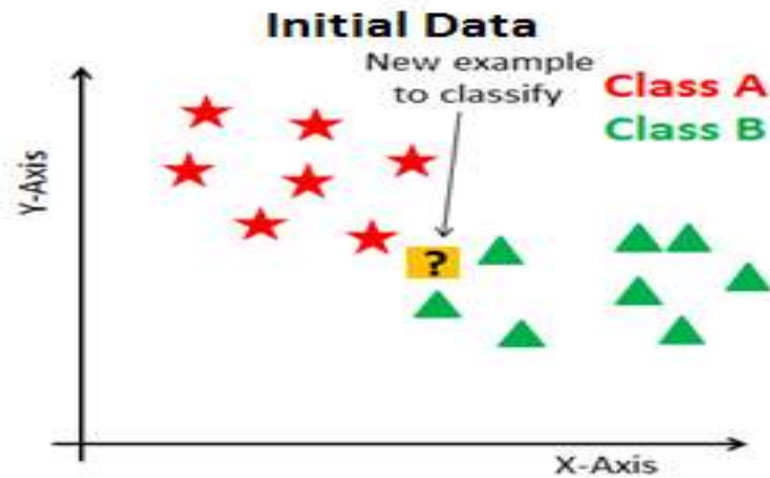
An object is classified by a majority vote of its neighbors, with the object being assigned to the class most common among its k nearest neighbors.

It can also be used for regression — output is the value for the object (predicts continuous values). This value is the average (or median) of the values of its k nearest neighbors.





K Nearest Neighbors



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K Nearest Neighbors

Applications of KNN:

Credit ratings —

Should the bank give a loan to an individual?

Would an individual default on his or her loan?

Is that person closer in characteristics to people who defaulted or did not default on their loans?

In political science —

classifying a potential voter to a “will vote” or “will not vote”, or to “vote Democrat” or “vote Republican”.

More advance examples could include handwriting detection (like OCR), image recognition and even video recognition.





K Nearest Neighbors

Pros:

Easy to implement

Simple algorithm — to explain and understand/interpret

High accuracy (relatively) — it is pretty high

Versatile — useful for classification or regression





K Nearest Neighbors

Cons:

Computationally expensive — because the algorithm stores all of the training data

High memory requirement

Stores all (or almost all) of the training data

Prediction stage might be slow (with big N)

Sensitive to irrelevant features and the scale of the data





Work Flow

Process:

1. Create an object to perform KNN Classification
2. Train the Data
3. Test the Data
4. Calculate the accuracy
5. Create API





Summary

Pack-Up!

- We gone through the creation of a model to predict the crop using the dataset.
- We have understood how to train the model using KNN Classification method.
- We have created a simple API to access the model using Flask and ngrok packages.

**THANK
YOU!**