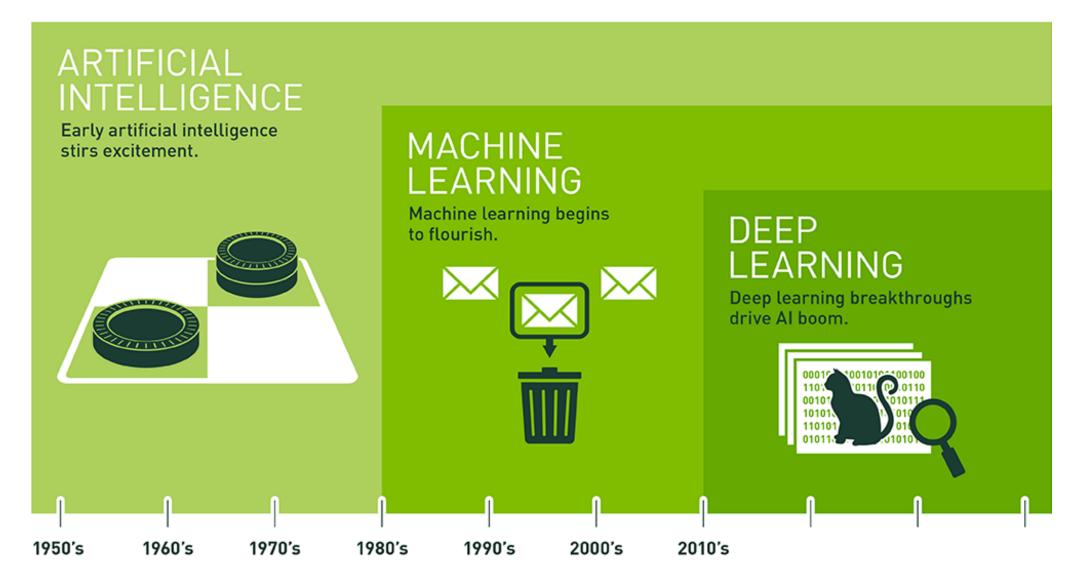
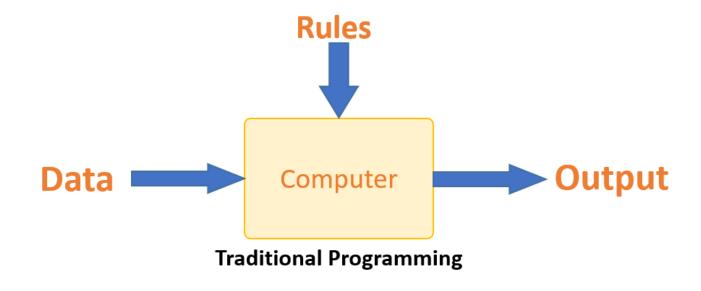
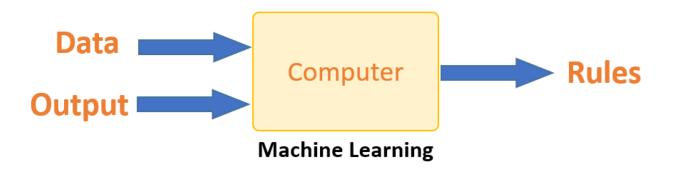
AI, ML and DL



Traditional programming vs machine learning

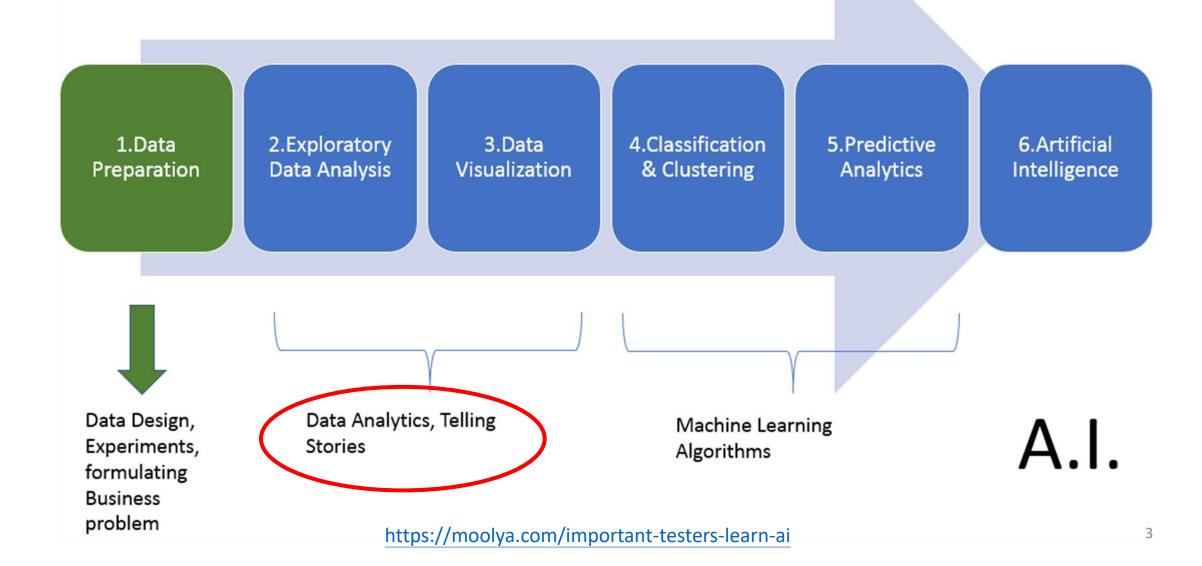




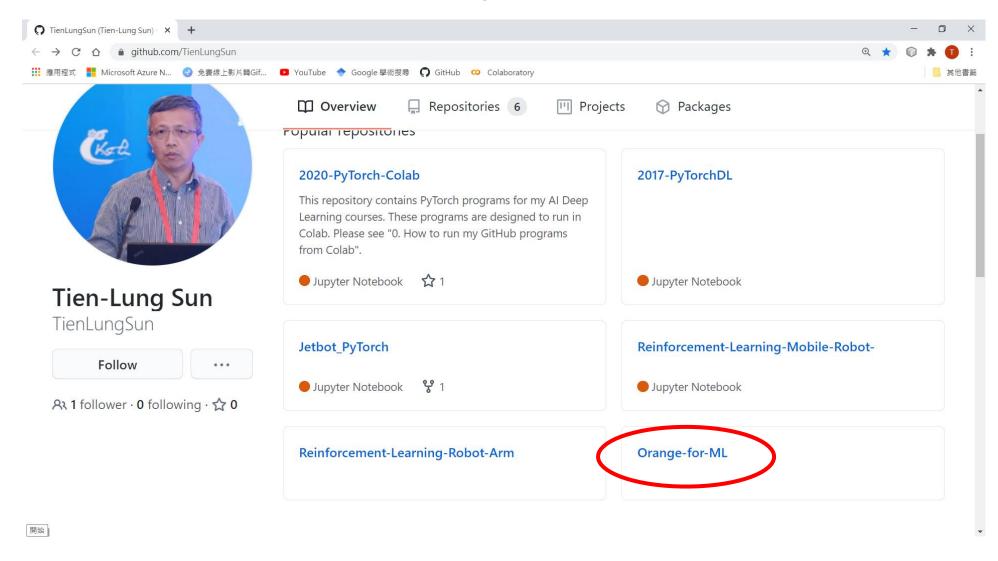
https://www.guru99.com/machine-learning-tutorial.html

HI before Al

Interactively visualize and explore your data before Al



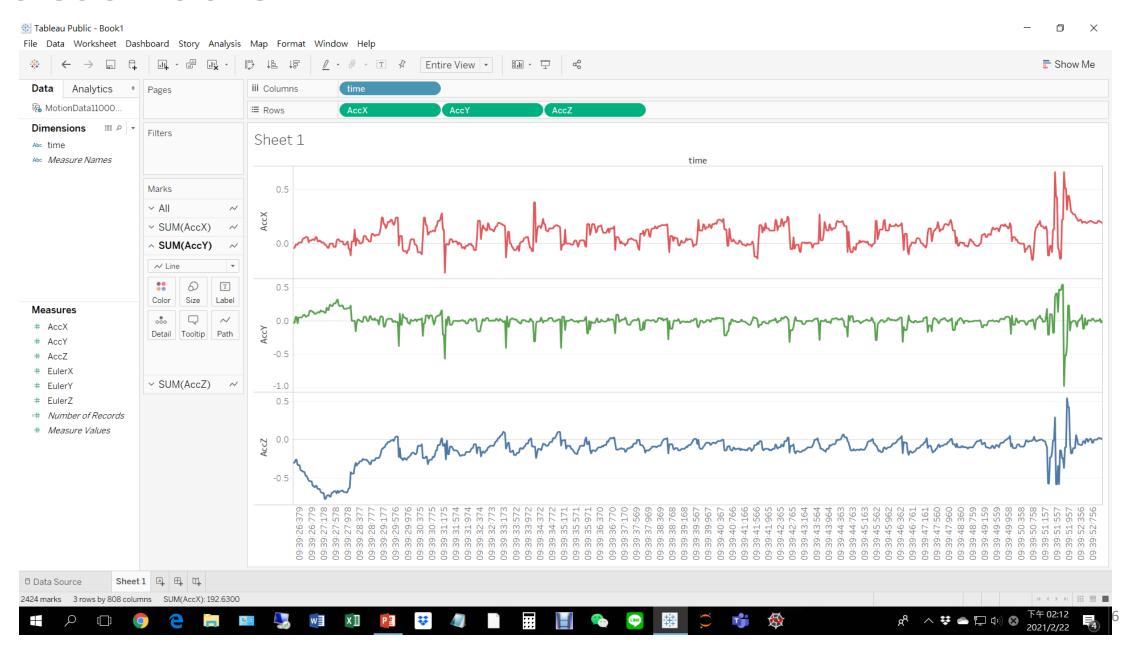
Download data files from my GitHub



Interactively visual exploration – (1) Tableau Public



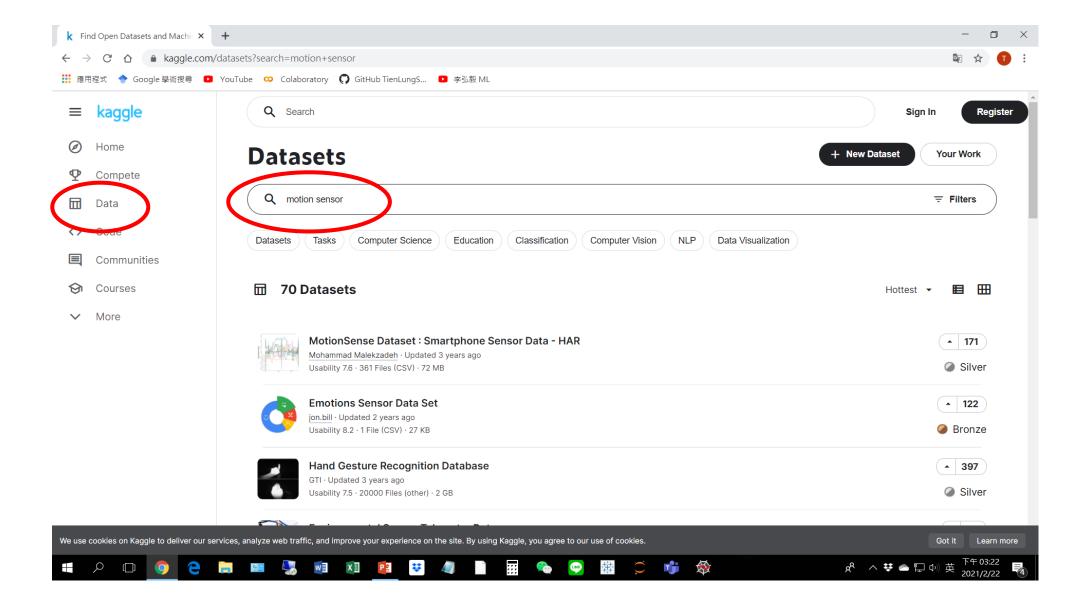
Tableau Public



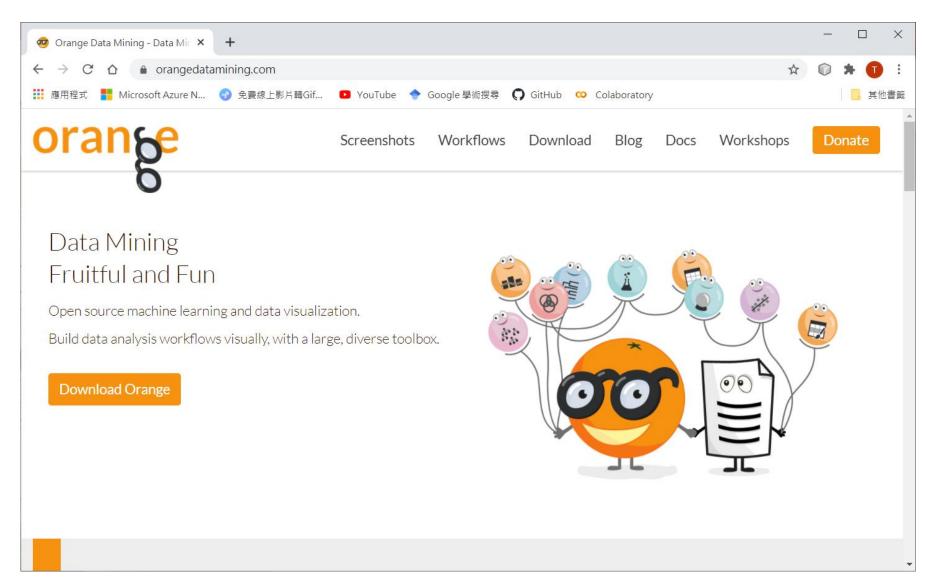
Practice – Tableau public

- 1. Download and install Tableau Public
- 2. Visualize the motion data file
- 3. Search Kaggle (https://www.kaggle.com/) to find a sensor data file (see next slide)
- 4. Use Tableau public to visualize the data file

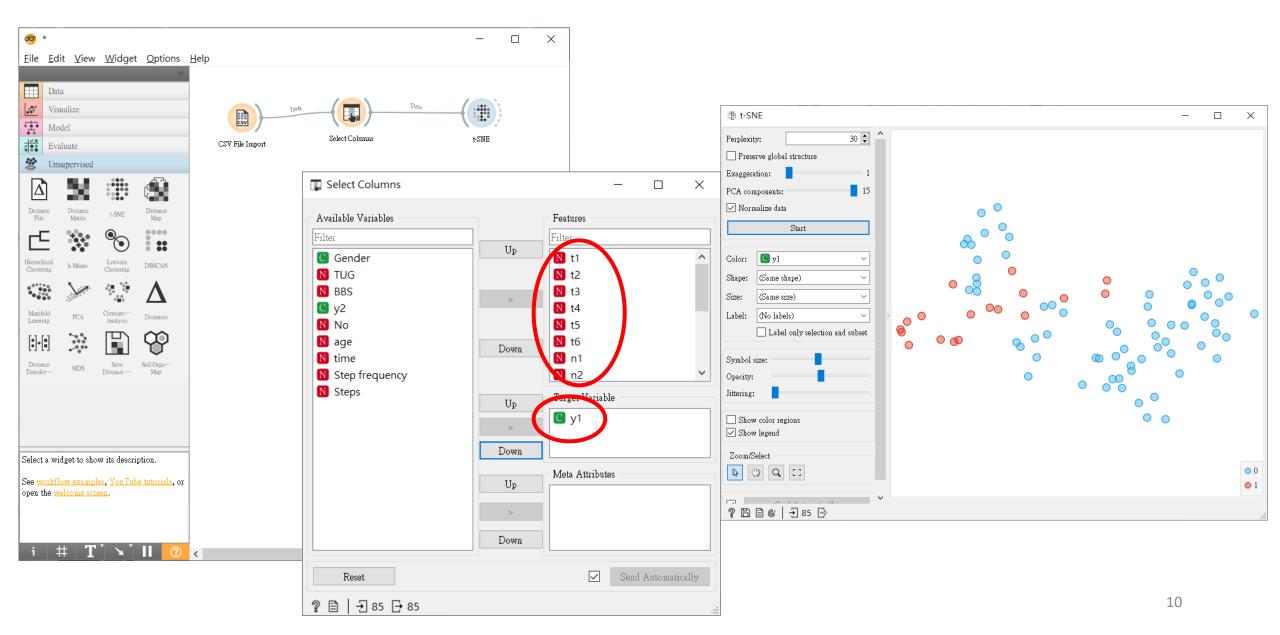
Kaggle



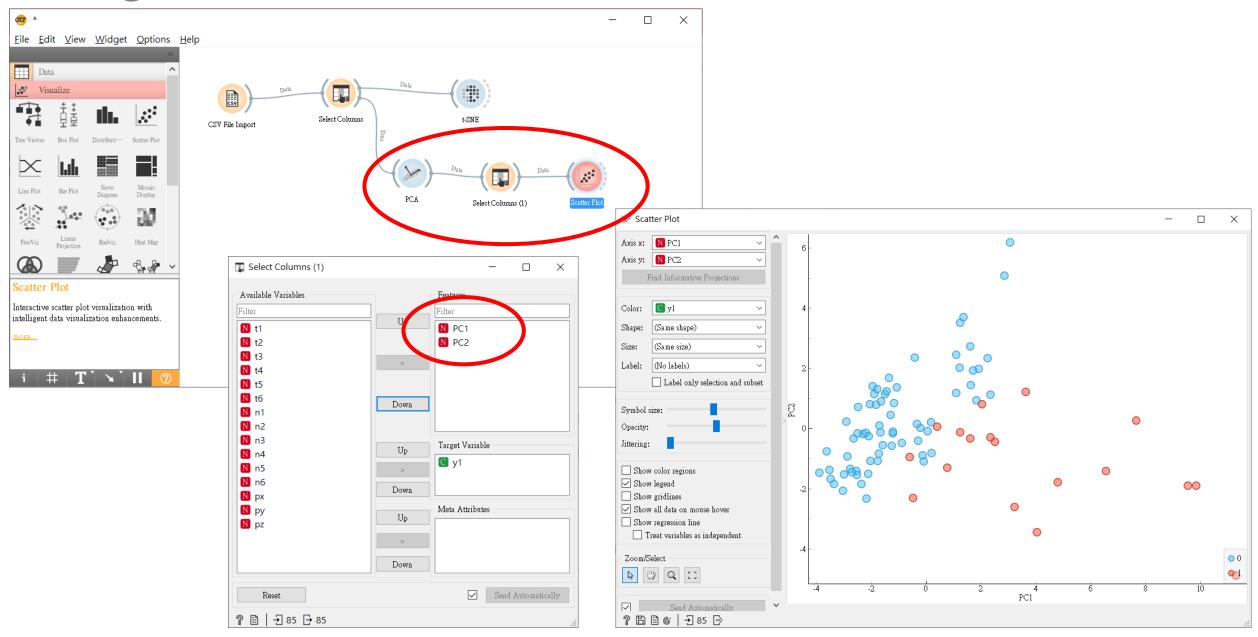
Interactively visual exploration – (2) Orange



Orange



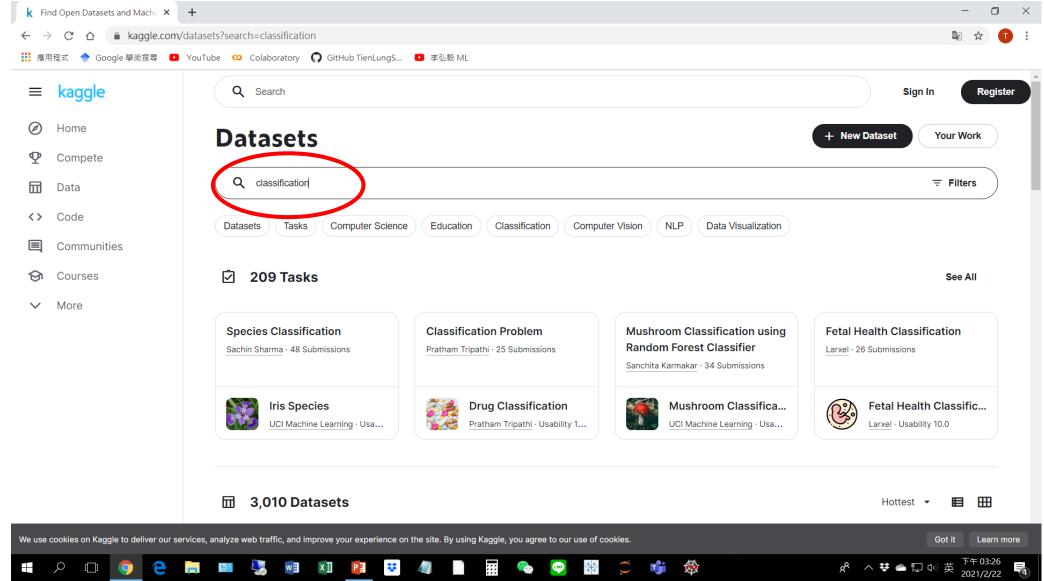
Orange



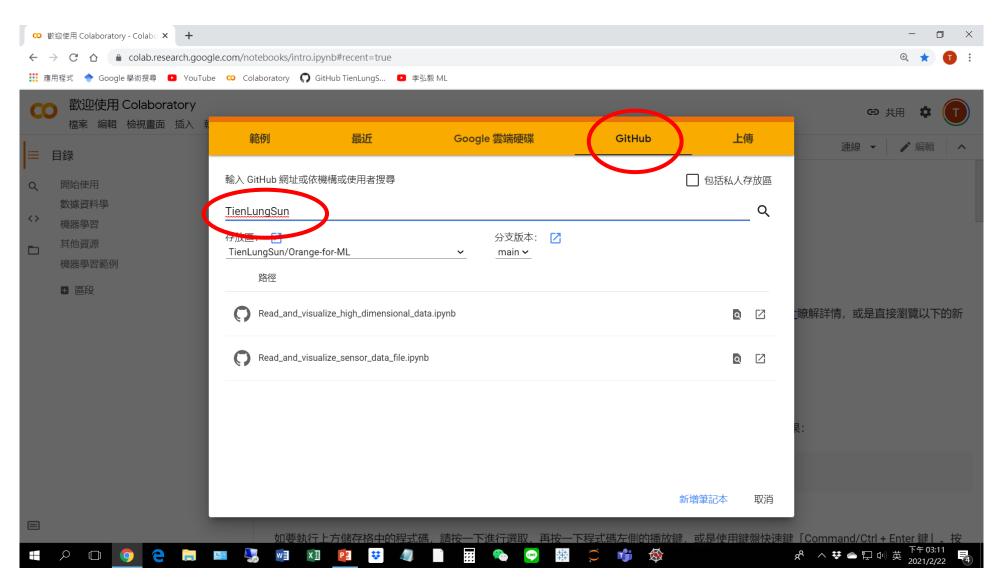
Practice – Orange

- 1. Download and install Orange
- 2. Visualize the 3M TUG data file
- 3. Search Kaggle to find a classification data file
- 4. Use Orange to visualize the high dimensional data

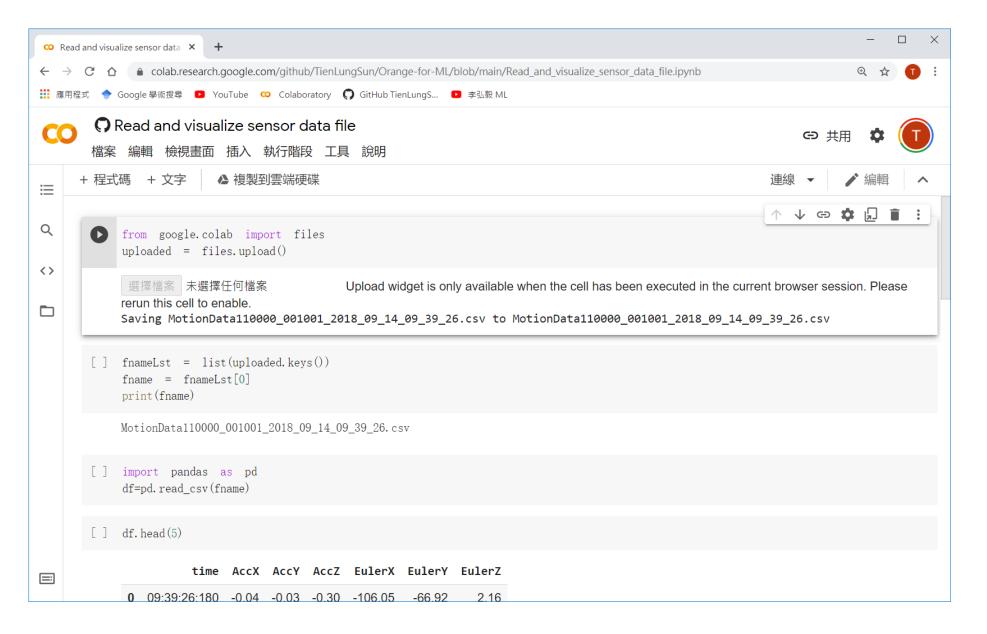
Kaggle



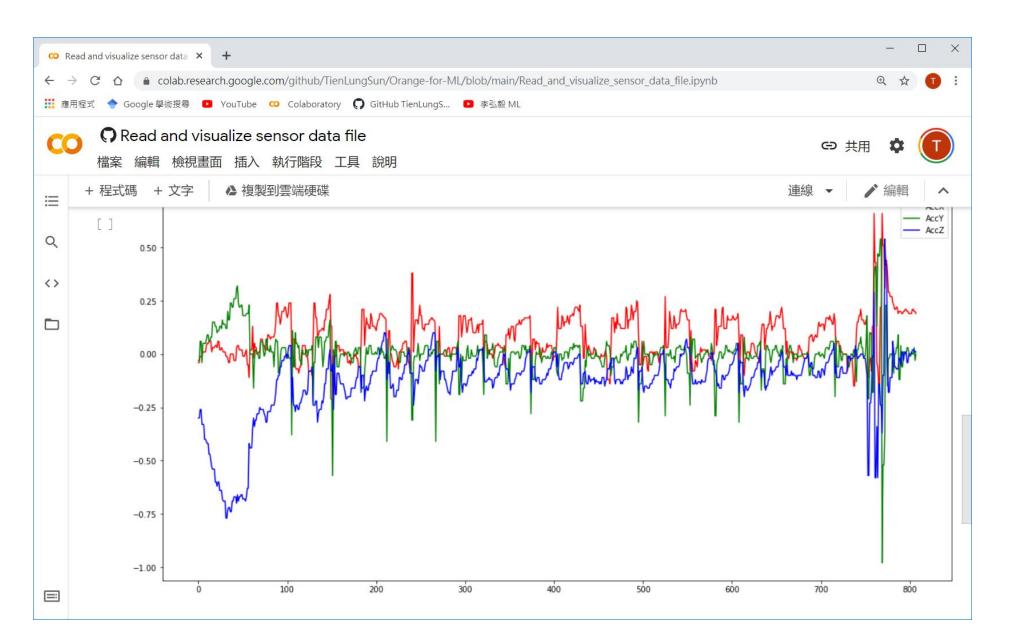
Data visualization – (3) Python coding



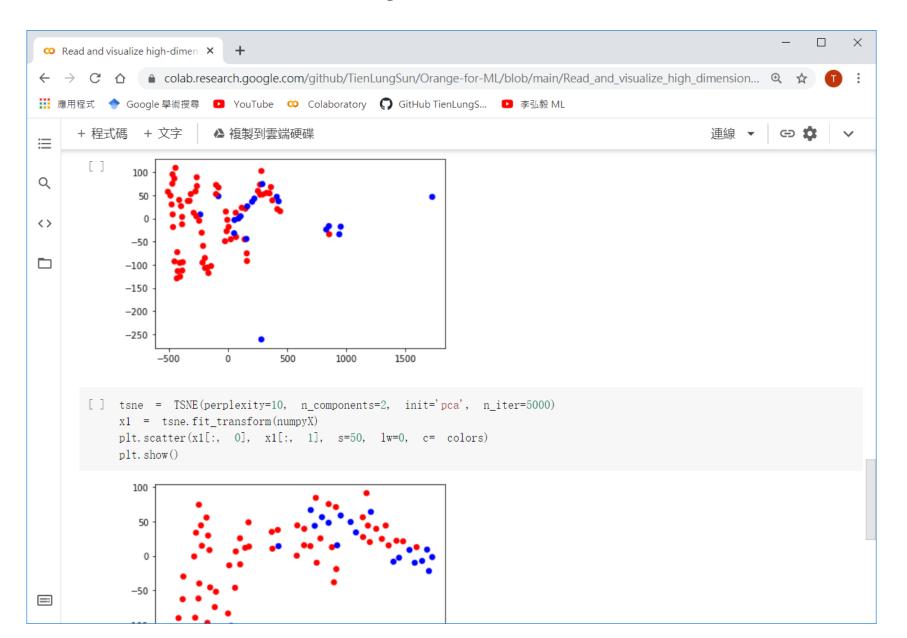
Data visualization with Python



Data visualization with Python



Data visualization with Python

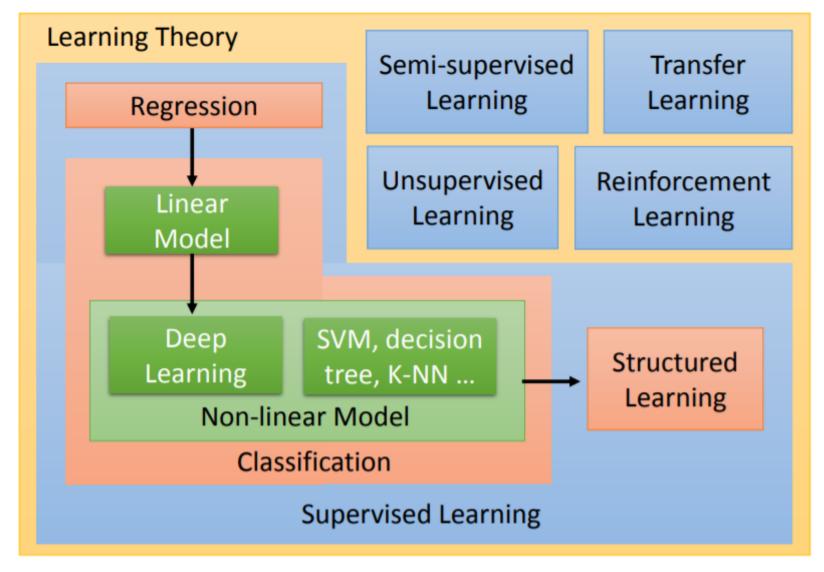


Practice – Python coding

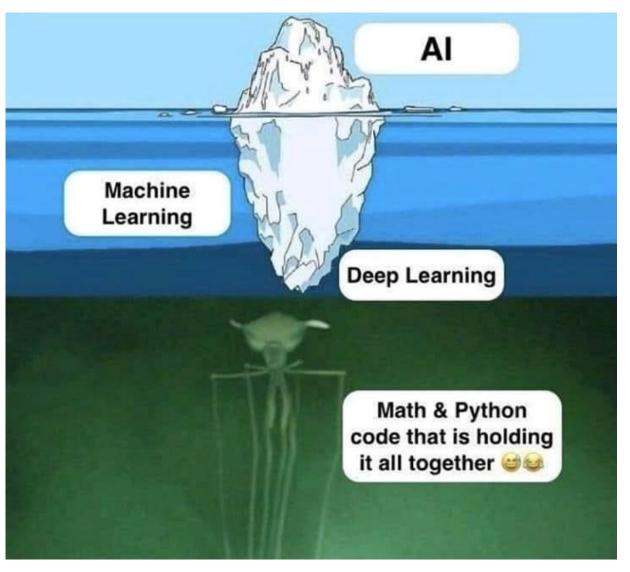
- 1. Log in to Colab
- 2. Run python code to visualize the motion sensor data file
- 3. Run python code to visualize the 3M TUG data file
- 4. Run python code to visualize the two data files you download from Kaggle

Employ AI (ML/DL) to learn from high-dimensional data to assist human visual data exploration

How computer learns from data?



Math + Python coding



Why we need to study math + python coding together?

Theory is when you know everything, but nothing works. Practice is when you don't know anything, yet everything works. In Programming we combine theory and practice: nothing works, and we don't know why.

Run exemplar PyTorch code from GitHub

Starts to train your own NN for your thesis

Python development tools







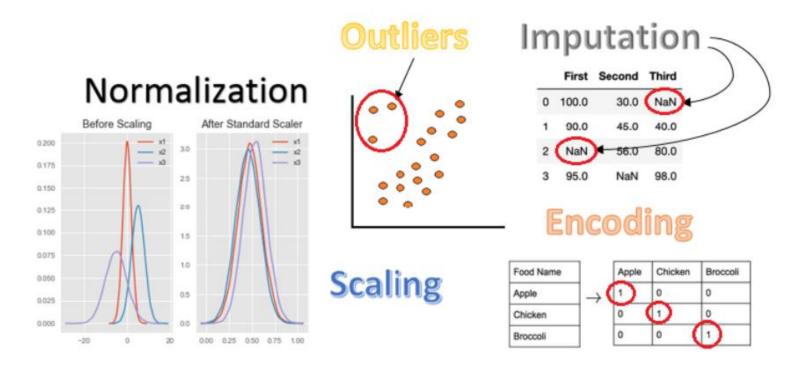


Before we start, two more issues not covered in this class but are important in AI development practice

Data pre-processing

Topics to be covered:

- 1. Standardization
- 2. Scaling with sparse data and outliers
- 3. Normalization
- 4. Categorical Encoding
- 5. Imputation



Data preprocessing in Python

Steps involved in data preprocessing:

- 1. Importing the required Libraries
- 2. Importing the data set
- 3. Handling the Missing Data.
- 4. Encoding Categorical Data.
- 5. Splitting the data set into test set and training set.
- 6. Feature Scaling.

https://aaaanchakure.medium.com/data-preprocessing-3cd01eefd438

Steps in Data Preprocessing in Machine Learning

- 1. Acquire the dataset
- 2. Import all the crucial libraries
- 3. Import the dataset
- 4. Identifying and handling the missing values
- 5. Encoding the categorical data
- 6. Splitting the dataset
- 7. Feature scaling

https://www.upgrad.com/blog/data-preprocessing-in-machine-learning/

Why Data Preprocessing?

Data in the real world is not clean

- incomplete: lacking attribute values, lacking certain attributes of interest, or containing only aggregate data, e.g., occupation="""
- noisy: containing errors or outliers, e.g., Salary="-10"
- inconsistent: containing discrepancies in codes or names, e.g., Age="42" Birthday="03/07/1997"
- No quality data, no quality mining results!
- Quality decisions must be based on quality data
 - Duplicates or missing data may cause incorrect or misleading analyses.

Data Preprocessing: Major Tasks

Data cleaning

 Fill in missing values, smooth noisy data, identify or remove outliers, and resolve inconsistencies

Data integration

Integration of multiple databases, data cubes, or files

Data transformation

Normalization and aggregation

Data reduction

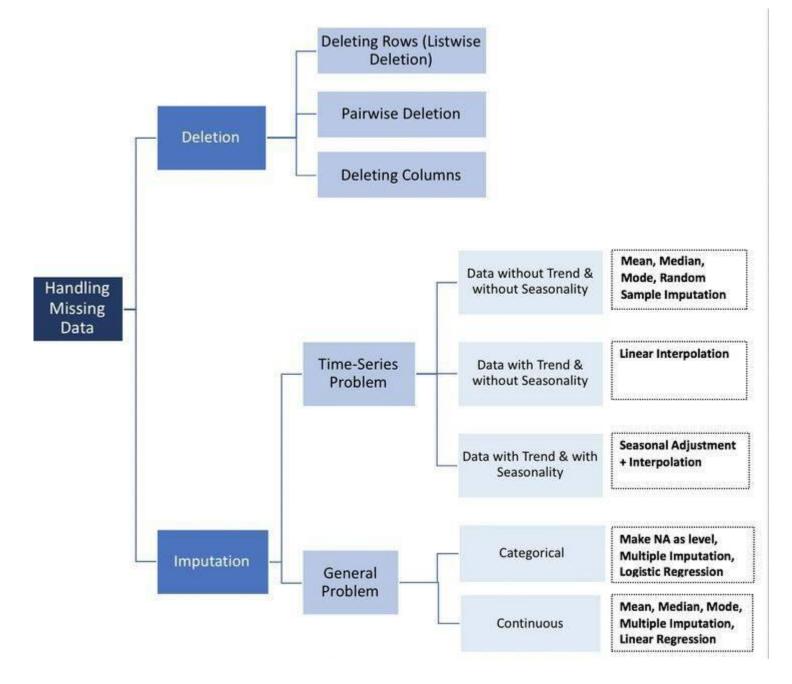
 Obtains reduced representation in volume but produces the same or similar analytical results

Data discretization

 Part of data reduction but with particular importance, especially for numerical data

Data Cleaning Tasks

- Fill in missing values
- Identify outliers and smooth out noisy data
- Correct inconsistent data
- Resolve redundancy caused by data integration

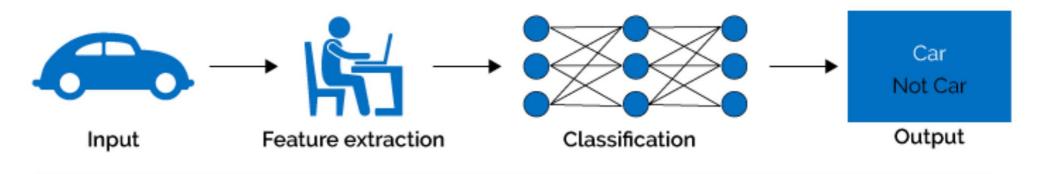


Data Transformation

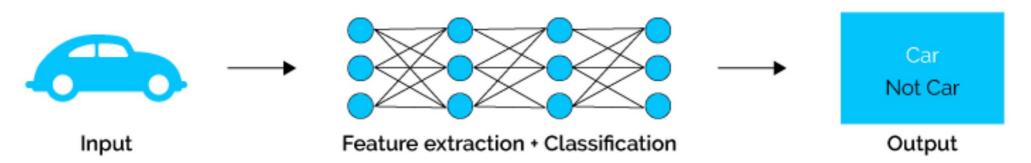
- Smoothing: remove noise from data
- Aggregation: summarization, data cube construction
- Generalization: concept hierarchy climbing
- Normalization: scaled to fall within a small, specified range
 - min-max normalization
 - z-score normalization
 - normalization by decimal scaling
- Attribute/feature construction
 - New attributes constructed from the given ones

Feature engineering

Machine Learning



Deep Learning



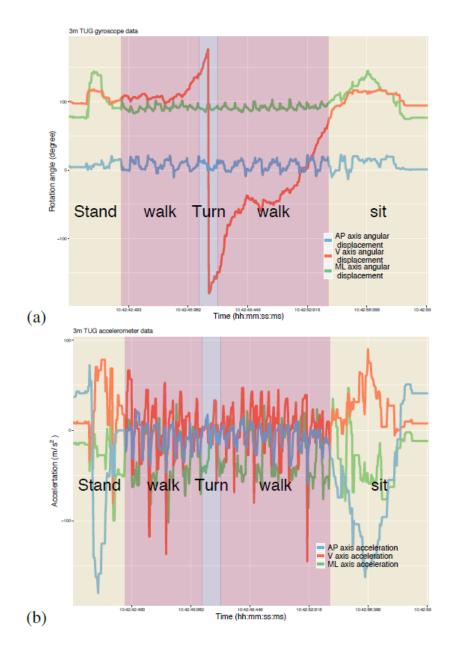


TABLE II SUMMARY OF FEATURES

TUG sensor-based features		
Feature name	Direction	Unit
Sit to stand		
Angle range	ML	deg
Acceleration range	V, AP	m/s^2
Completion time		S
Walking	-	
Acceleration CV	V, AP, ML	
Acceleration range	V, AP, ML	m/s^2
Acceleration median	V, AP, ML	m/s^2
Acceleration RMS	V, AP, ML	m/s^2
Angular velocity CV	V, AP, ML	
Angular velocity range	V, AP, ML	deg/s
Angular velocity median	V, AP, ML	deg/s
Angular velocity RMS	V, AP, ML	deg/s
Speed		m/s
Turning		·
Acceleration CV	V, AP, ML	
Acceleration range	V, AP, ML	m/s^2
Acceleration median	V, AP, ML	m/s^2
Acceleration RMS	V, AP, ML	m/s^2
Angular velocity CV	V, AP, ML	
	1	

Hsu, Y. C., Zhao, Y., Huang, K. H., Wu, Y. T., Cabrera, J., Sun, T. L., & Tsui, K. L. (2020). A novel approach for fall risk prediction using the inertial sensor data from the timed-up-and-go test in a community setting. IEEE Sensors Journal, 20(16), 9339-9350.

Feature engineering

- > Supervised Feature Selection Methods
 - > Wrapper Feature Selection Methods
 - > Filter Feature Selection Methods
- > Embedded or Intrinsic Feature Selection Methods
- > Feature Selection with Statistical Measures
 - > Univariate Feature Selection
- > Feature Selection Strategies
 - > Selection Method
 - > Transform Variables
- > Which Feature Selection Method is the Best?
- > Feature Selection Implementations
 - > Feature Selection For Regression models
 - Classification Feature Selection