

# Getting started with Machine Learning

Date : 11th July 2021 | Speaker : Ayon Roy

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# Hello Buddy!

I am **Ayon Roy**

**B.Tech CSE ( 2017-2021 )**

Data Science Intern @ **Internshala**

( India's Leading Internship Provider Platform )

Brought **Kaggle Days Meetup** Community in India for the 1st time

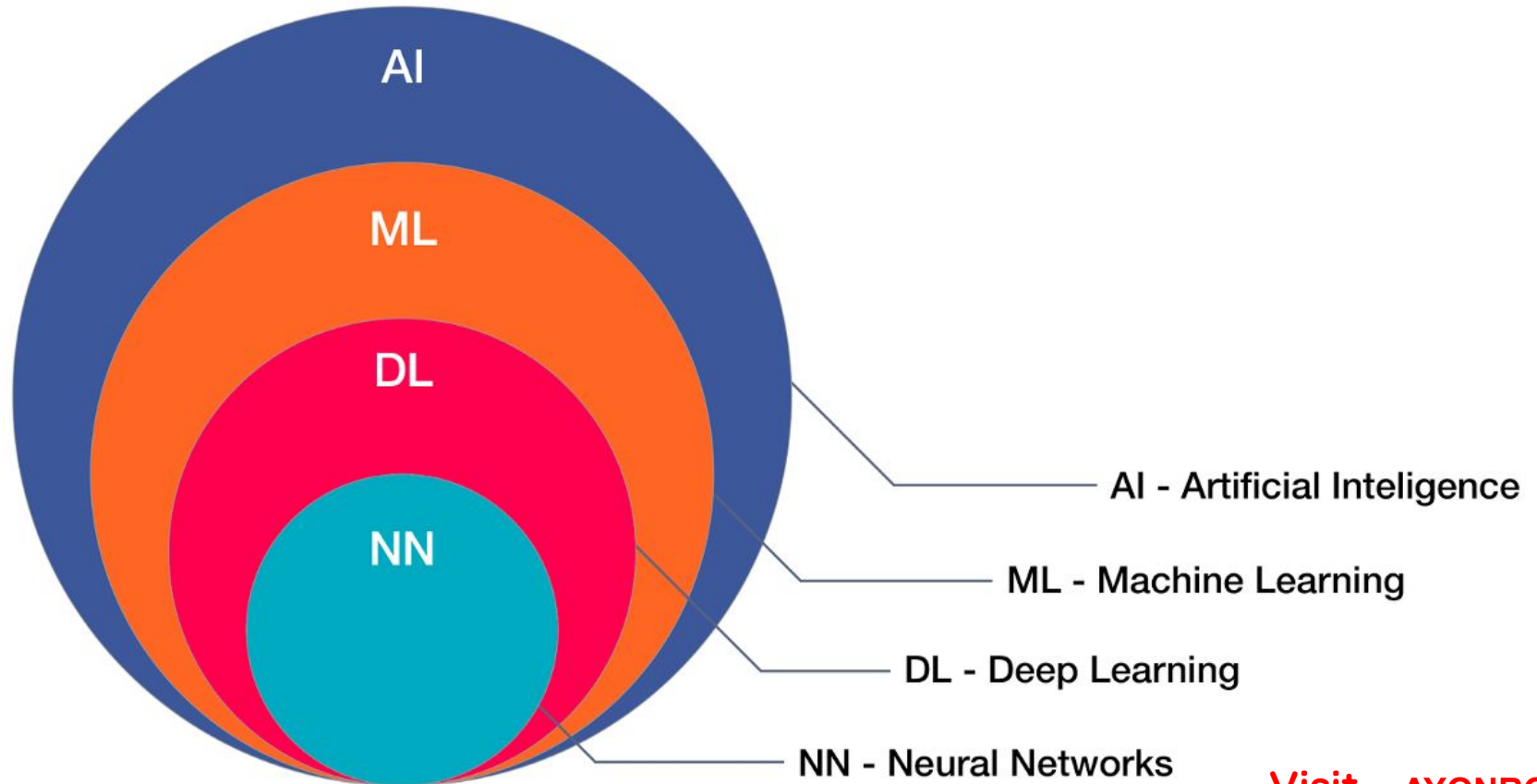
**If you haven't heard about me yet, you might have been living under the rocks. Wake up !!**

# Agenda

- What is Machine Learning ?
- How to start Machine Learning ?
- Initial steps in a Machine Learning Process
- A brief Intro to Data Pre-Processing, Exploratory Data Analysis, Data Visualization
- What's the current scenario & the scope of ML ?

# What is Machine Learning ?











# Graphical Representation



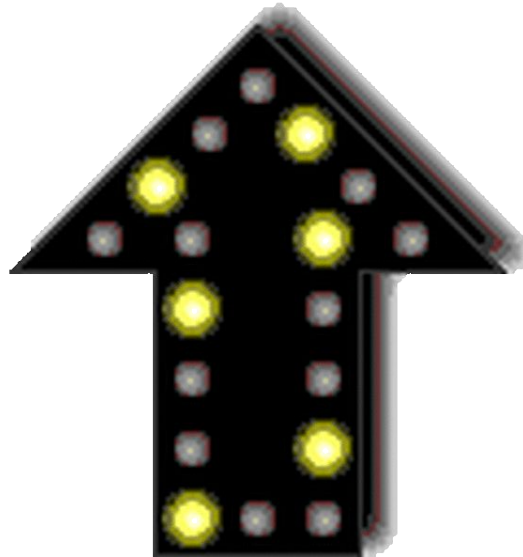
**A field of study that gives computers the capability to learn without being explicitly programmed.**

Machine learning is applied using Algorithms to process the data and get trained for delivering future predictions without human intervention. The inputs for Machine Learning is the set of instructions or data or observations.

# Applications of Machine Learning

APPLICATION	POTENTIAL ANNUAL VALUE BY 2026	KEY DRIVERS FOR ADOPTION
Robot-assisted surgery	 \$40B	Technological advances in robotic solutions for more types of surgery
Virtual nursing assistants	 20	Increasing pressure caused by medical labor shortage
Administrative workflow	 18	Easier integration with existing technology infrastructure
Fraud detection	 17	Need to address increasingly complex service and payment fraud attempts
Dosage error reduction	 16	Prevalence of medical errors, which leads to tangible penalties
Connected machines	 14	Proliferation of connected machines/devices
Clinical trial participation	 13	Patent cliff; plethora of data; outcomes-driven approach
Preliminary diagnosis	 5	Interoperability/data architecture to enhance accuracy
Automated image diagnosis	 3	Storage capacity; greater trust in AI technology
Cybersecurity	 2	Increase in breaches; pressure to protect health data

# How to start Machine Learning



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# Start with Maths for Machine Learning

But **why should I do Maths**  
first for Machine Learning ?

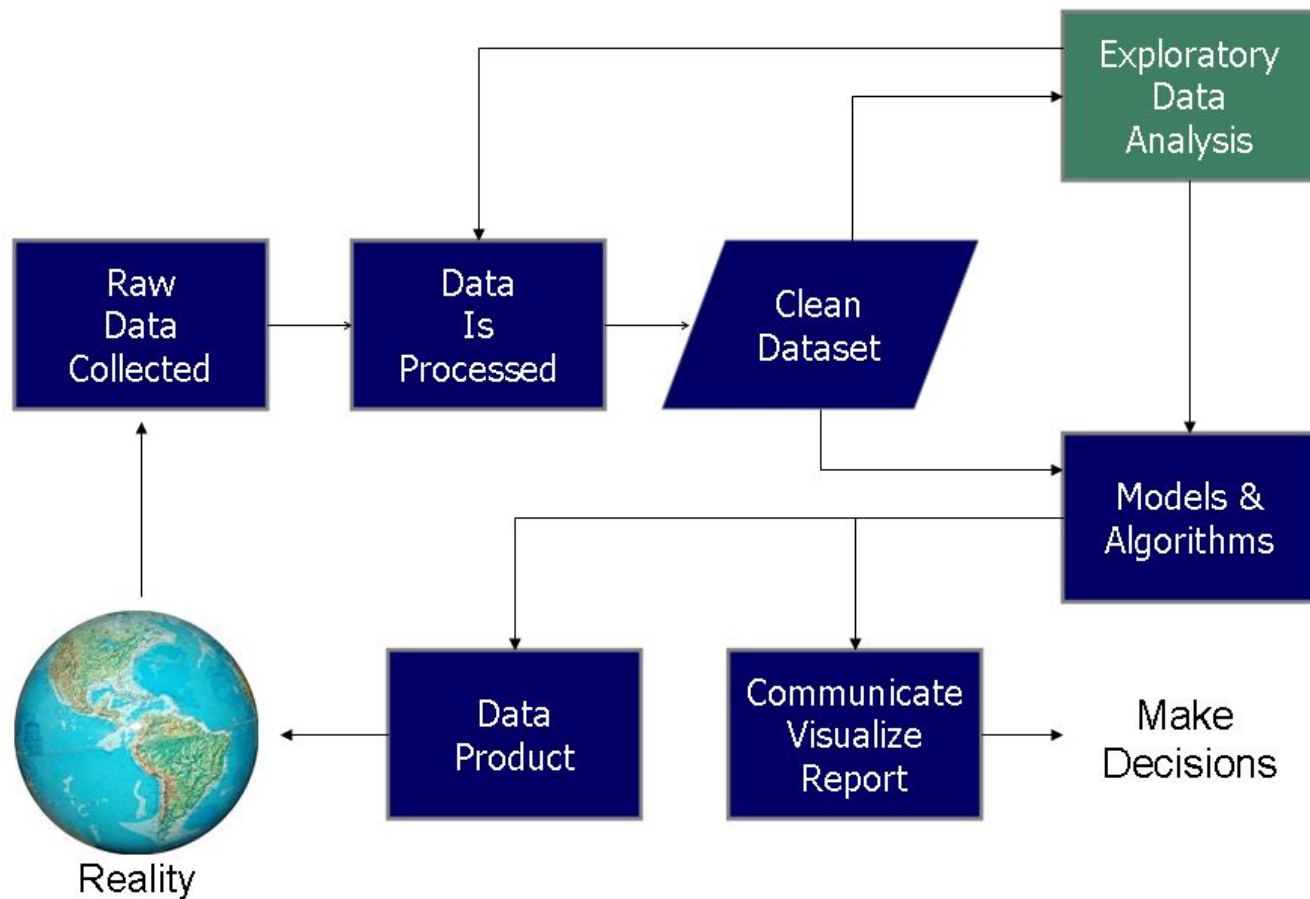
- Week 1 : Linear Algebra [B] <https://www.khanacademy.org/math/linear-algebra>
- Week 2 : Calculus [B] <https://www.youtube.com/playlist?list=PLZHQObOWTQDMsr9K-rj53DwVRMYO3t5Yr> or <https://www.mathsisfun.com/calculus/> ; want theoretical notes , find it at <https://the-learning-machine.com/article/machine-learning/calculus> .
- Week 3 : Probability [B] <https://www.edx.org/course/introduction-probability-science-mitx-6-041x-2>
- Week 4 : Statistics [B] <http://alex.smola.org/teaching/cmu2013-10-701/stats.html>
- Algorithms ( Only if you want to learn proper software development ) [ Highly optional ]  
This is an overview of what the students study as the subject Data Structures & Algorithm . So if you are fluent with this part , you can skip this !! <https://www.edx.org/course/algorithm-design-analysis-pennx-sd3x>

**Start** with Python  
&  
try to **implement** those  
Mathematical Concepts

**Start exploring Libraries  
& then start Machine  
Learning Courses**

- 
- Introduction to python for data science [B] <https://www.datacamp.com/courses/intro-to-python-for-data-science>
  - Want to dive deeper into Data Visualization & Pre-Processing ? Look into Data Visualization & Pre-Processing section in miscellaneous resources . [ Highly optional ]
  - Want to explore the field of Deep Learning ? See the Deep Learning Section in miscellaneous resources . [ Highly optional ]
  - Want to explore the field of Natural Language Processing [ NLP ] ? See the Natural language Processing Section in miscellaneous resources . [ Highly optional ]
  - See how ML codes are written and made to work at - > <https://github.com/maykulkarni/Machine-Learning-Notebooks> or <https://github.com/GokuMohandas/practicalAI/blob/master/README.md> . [ Highly optional ]
  - Find useful resources here at <https://github.com/ujjwalkarn/Machine-Learning-Tutorials/blob/master/README.md> . [ Highly optional ]

# Initial Steps in a Machine Learning Process





# What is Data Pre-Processing ?

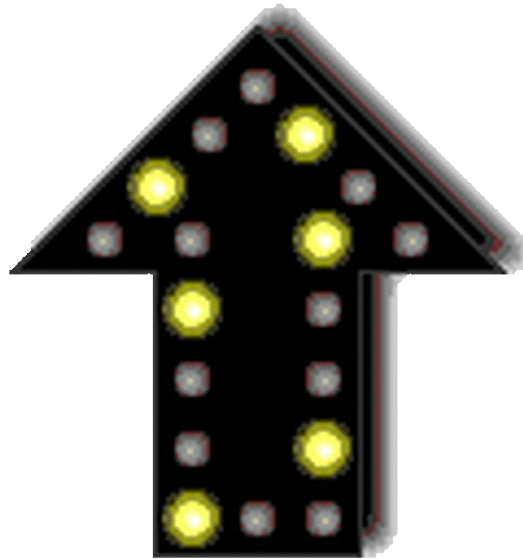
It is a technique that transforms raw data into an understandable format.

# Why do we need it ?

Raw data ( Real world data ) is always messy and that data cannot be sent through a model. That would cause certain errors.

So we need to preprocess data before sending through further analysis.

# Steps to be followed



# Read the data

```
# Read the data in the CSV file using pandas
df = pd.read_csv('../input/creditcard.csv')
df.head()
```

	Time	V1	V2	V3	V4	V5	V6	V7	V8	V9	...	V21	V22	V23	V24	
0	0.0	-1.359807	-0.072781	2.536347	1.378155	-0.338321	0.462388	0.239599	0.098698	0.363787	...	-0.018307	0.277838	-0.110474	0.066928	0.12
1	0.0	1.191857	0.268151	0.168480	0.448154	0.060018	-0.082361	-0.078803	0.085102	-0.255425	...	-0.225775	-0.638872	0.101288	-0.339846	0.16
2	1.0	-1.358354	-1.340163	1.773209	0.379780	-0.503198	1.800499	0.791461	0.247676	-1.514654	...	0.247998	0.771679	0.909412	-0.689281	-0.32
3	1.0	-0.966272	-0.185226	1.792993	-0.863291	-0.010309	1.247203	0.237609	0.377436	-1.387024	...	-0.108300	0.005274	-0.190321	-1.175575	0.64
4	2.0	-1.158233	0.877737	1.548718	0.403034	-0.407193	0.095921	0.592941	-0.270533	0.817739	...	-0.009431	0.798278	-0.137458	0.141267	-0.20

Fig 1 : Dataset

# Checking the Missing Values

```
# Looking at the ST_NUM column  
print df['ST_NUM']  
print df['ST_NUM'].isnull()
```

Out:

```
0    104.0  
1    197.0  
2      NaN  
3    201.0  
4    203.0  
5    207.0  
6      NaN  
7    213.0  
8    215.0
```

Out:

```
0    False  
1    False  
2     True  
3    False  
4    False  
5    False  
6     True  
7    False  
8    False
```

# Replacing the Missing Values

---

A very common way to replace missing values is using a median.

```
# Replace using median
median = df['NUM_BEDROOMS'].median()
df['NUM_BEDROOMS'].fillna(median, inplace=True)
```

# Standardizing the data

```
# Standardizing the features
df['Vamount'] =
StandardScaler().fit_transform(df['Amount'].values.reshape(-1,1))
df['Vtime'] =
StandardScaler().fit_transform(df['Time'].values.reshape(-1,1))

df = df.drop(['Time','Amount'], axis = 1)
df.head()
```

V22	V23	V24	V25	V26	V27	V28	Class	Vamount	Vtime
0.277838	-0.110474	0.066928	0.128539	-0.189115	0.133558	-0.021053	0	0.244964	-1.996583
-0.638672	0.101288	-0.339846	0.167170	0.125895	-0.008983	0.014724	0	-0.342475	-1.996583
0.771679	0.909412	-0.689281	-0.327642	-0.139097	-0.055353	-0.059752	0	1.160686	-1.996562
0.005274	-0.190321	-1.175575	0.647376	-0.221929	0.062723	0.061458	0	0.140534	-1.996562
0.798278	-0.137458	0.141267	-0.206010	0.502292	0.219422	0.215153	0	-0.073403	-1.996541

Fig 7 : Standardized dataset

# Exploratory Data Analysis





# What is Exploratory Data Analysis ?

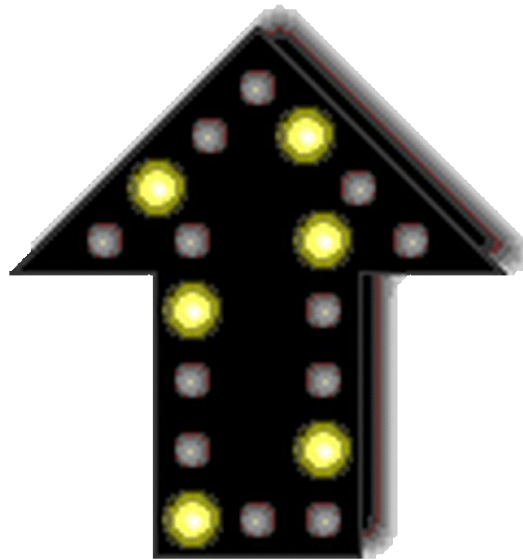
A critical process of performing initial investigations on data so as to discover patterns, to spot anomalies, to test hypothesis and to check assumptions with the help of summary statistics and graphical representations.

# Why do we need it ?

1. Detection of mistakes & missing data
2. Checking of assumptions
3. Preliminary selection of appropriate models
4. Determining relationships among the explanatory variables

**With EDA, we can make sense of the data we have and then figure out what questions we want to ask and how to frame them**

# Major Steps to be followed



# Import the Libraries

```
# Importing required libraries.  
import pandas as pd  
import numpy as np  
import seaborn as sns #visualisation  
import matplotlib.pyplot as plt #visualisation  
%matplotlib inline  
sns.set(color_codes=True)
```

# Check the type of Data

```
# Checking the data type  
df.dtypes
```

```
Make          object  
Model         object  
Year          int64  
Engine Fuel Type  object  
Engine HP     float64  
Engine Cylinders float64  
Transmission Type object  
Driven_wheels  object  
Number of Doors float64  
Market Category object  
Vehicle Size  object  
Vehicle Style  object  
highway MPG   int64  
city mpg      int64  
Popularity    int64  
MSRP          int64  
dtype: object
```

# Dropping Irrelevant Columns

```
# Dropping irrelevant columns
df = df.drop(['Engine Fuel Type', 'Market Category', 'Vehicle Style',
             'Popularity', 'Number of Doors', 'Vehicle Size'], axis=1)
df.head(5)
```

	Make	Model	Year	Engine HP	Engine Cylinders	Transmission Type	Driven_wheels	highway MPG	city mpg	MSRP
0	BMW	1 Series M	2011	335.0	6.0	MANUAL	rear wheel drive	26	19	46135
1	BMW	1 Series	2011	300.0	6.0	MANUAL	rear wheel drive	28	19	40650
2	BMW	1 Series	2011	300.0	6.0	MANUAL	rear wheel drive	28	20	36350
3	BMW	1 Series	2011	230.0	6.0	MANUAL	rear wheel drive	28	18	29450
4	BMW	1 Series	2011	230.0	6.0	MANUAL	rear wheel drive	28	18	34500

Dropping irrelevant columns.

# Renaming the Columns

```
# Renaming the column names
df = df.rename(columns={"Engine HP": "HP", "Engine Cylinders":
"Cylinders", "Transmission Type": "Transmission", "Driven_Wheels":
"Drive Mode", "highway MPG": "MPG-H", "city mpg": "MPG-C", "MSRP":
"Price" })
df.head(5)
```

	Make	Model	Year	HP	Cylinders	Transmission	Drive Mode	MPG-H	MPG-C	Price
0	BMW	1 Series M	2011	335.0	6.0	MANUAL	rear wheel drive	26	19	46135
1	BMW	1 Series	2011	300.0	6.0	MANUAL	rear wheel drive	28	19	40650
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4	BMW	1 Series	2011	230.0	6.0	MANUAL	rear wheel drive	28	18	34500

Renaming the column name.

# Removing the Duplicates

```
# Dropping the duplicates  
df = df.drop_duplicates()  
df.head(5)
```

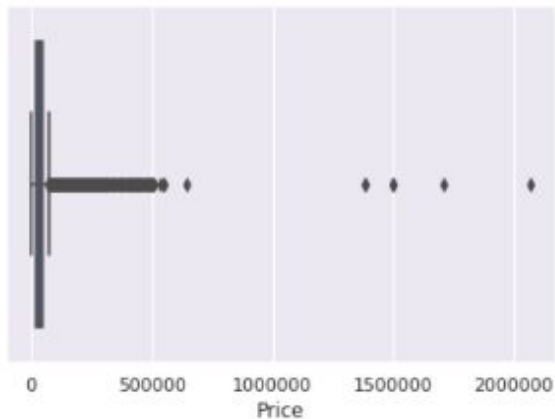
	Make	Model	Year	HP	Cylinders	Transmission	Drive Mode	MPG-H	MPG-C	Price
0	BMW	1 Series M	2011	335.0	6.0	MANUAL	rear wheel drive	26	19	46135
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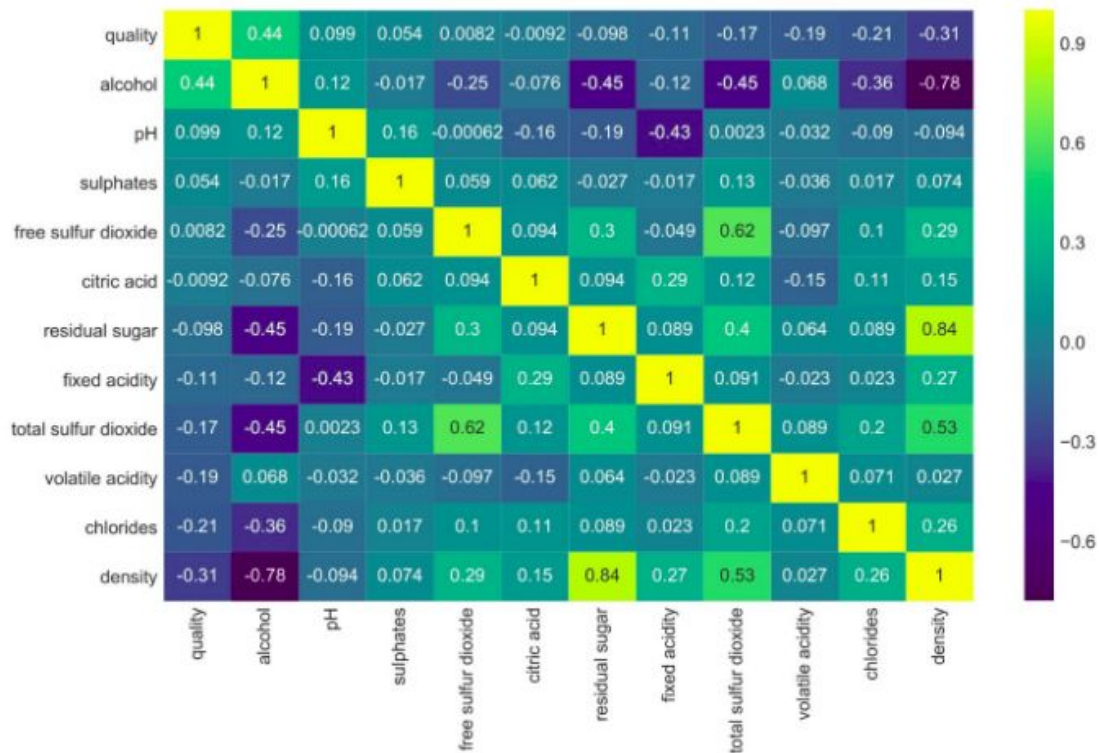
# Detecting the Outliers

```
sns.boxplot(x=df['Price'])
```

<matplotlib.axes.\_subplots.AxesSubplot at 0x7f69f68edc18>



# Correlation Matrix etc.



# What's **Data Visualization** ?

Data visualization is the graphical representation of information and data.

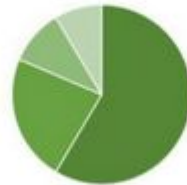
By using visual elements like charts, graphs, and maps, data visualization tools provide an accessible way to see and understand trends, outliers, and patterns in data.

# Difft. Types of Data Visualization methods

Charts ->



Line

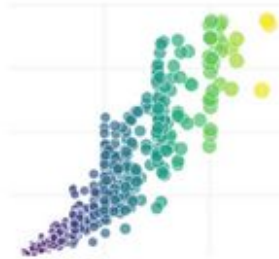


Pie



Bar

Plots ->



Bubble



Scatter

# Diff. Types of Data Visualization methods

Maps ->

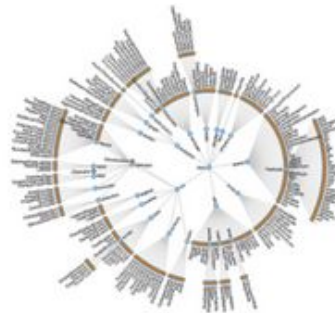


Heat

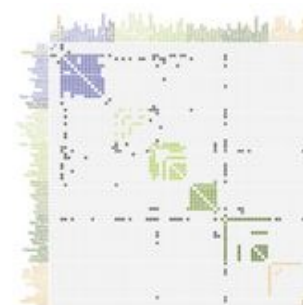


Dot distribution

Trees & Matrix ->



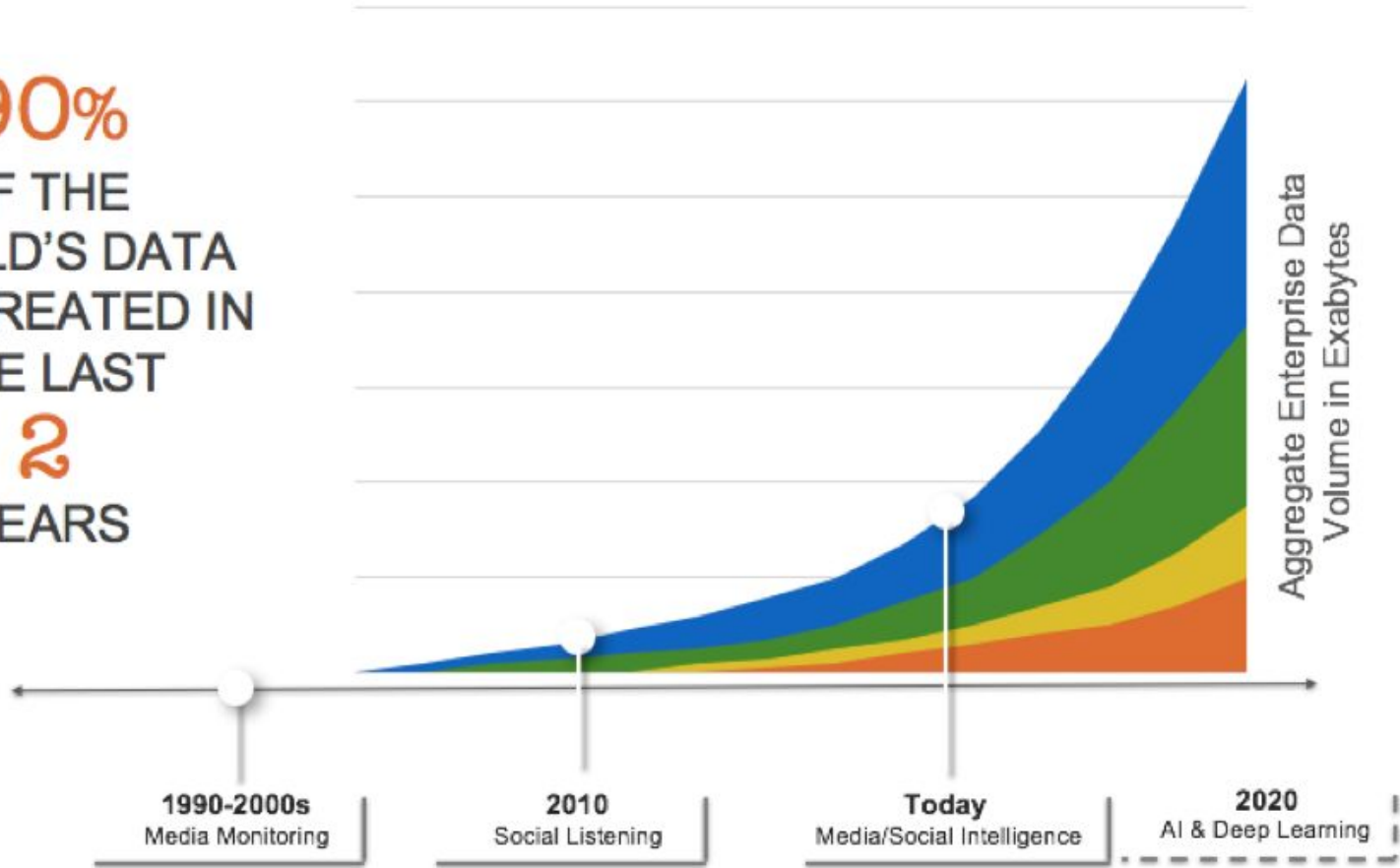
Tree



Matrix

# What's the Current Scenario ?

**90%**  
OF THE  
WORLD'S DATA  
WAS CREATED IN  
THE LAST  
**2**  
YEARS



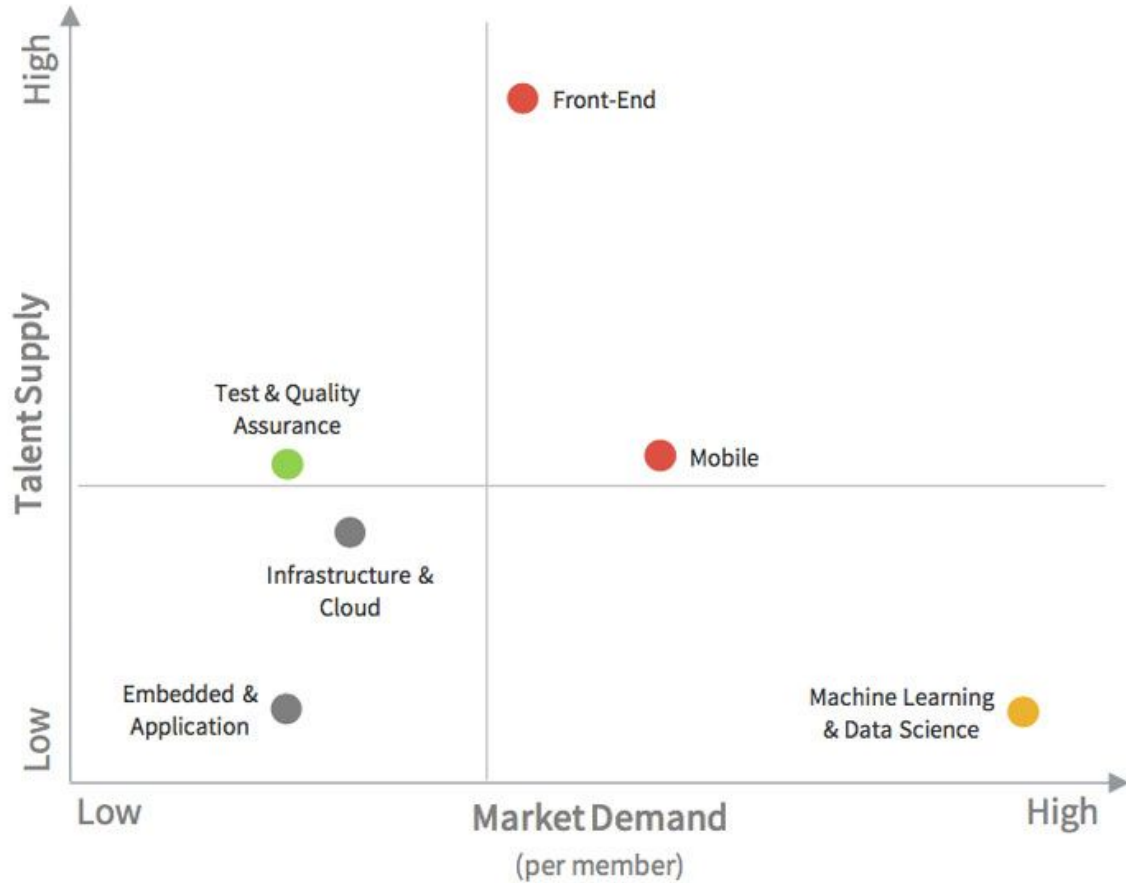
# But why Machine Learning now ?

1. The sharp decrease in costs associated with data storage and processing.
2. The advent of the Internet economy and the explosion in mobile apps.
3. The abundance of open-source tools.
4. The development of a wealth of innovative ML and DL algorithms.
5. Availability of GPUs etc.



# The Scope of Machine Learning

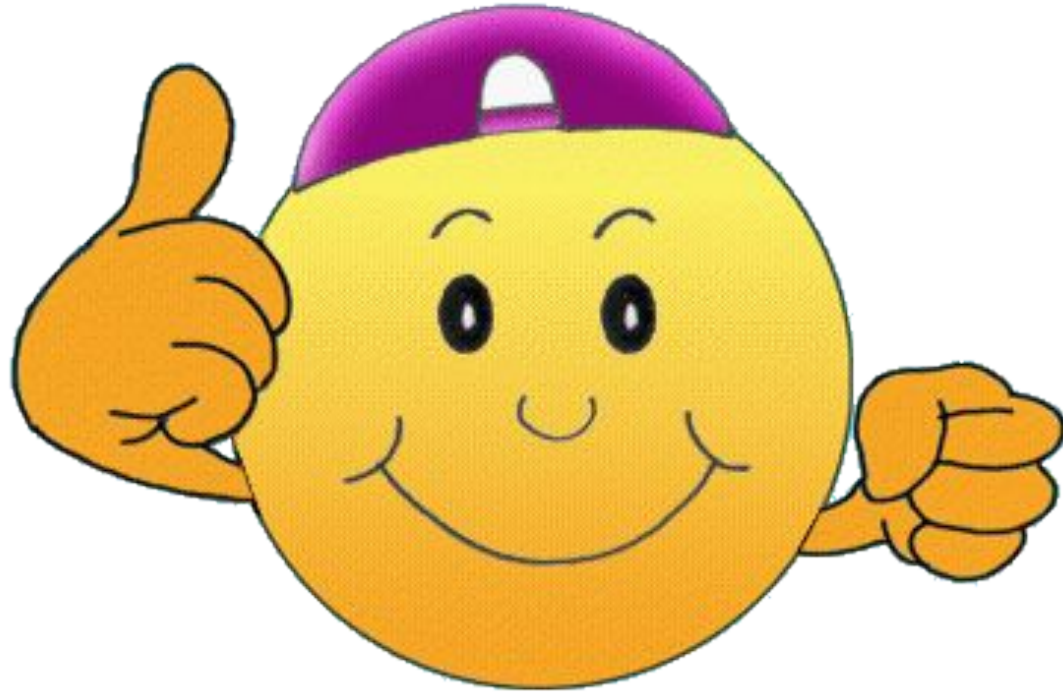
# Supply & Demand by Specialty



# Get the resources at

1. <https://github.com/aayoonn/100DaysOfMLCode>
2. <https://blog.ayonroy.ml/2020/12/01/personalized-guide-by-ayon-roy>

**GO FOR IT !**



**GOOD LUCK !**

Let me answer your Questions now.

Finally, it's your time to speak.



# Danke Schoen

Questions ? Any Feedbacks ? Did you like the talk?  
Tell me about it.

If you think I can help you,  
connect with me via

**Email** : aayoonn@gmail.com

**LinkedIn** : <https://www.linkedin.com/in/aayoonn/>

**Website** : <https://AYONROY.ML/>