# Welcome to ML | Day 06

## How We Understand the Data? ¶

### **Import Library and Dataset**

[1]:	impo	rt pandas <mark>a</mark>	s pd										
[2]:	df =	pd.read_cs	v('train	.csv')									
[3]:	df												
[3]:		Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Ε
	0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.2500	NaN	
	1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	female	38.0	1	0	PC 17599	71.2833	C85	
	2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.9250	NaN	
	3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.1000	C123	
	4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.0500	NaN	
		•••											
	886	887	0	2	Montvila, Rev. Juozas	male	27.0	0	0	211536	13.0000	NaN	
	887	888	1	1	Graham, Miss. Margaret Edith	female	19.0	0	0	112053	30.0000	B42	
	888	889	0	3	Johnston, Miss. Catherine Helen "Carrie"	female	NaN	1	2	W./C. 6607	23.4500	NaN	
	889	890	1	1	Behr, Mr. Karl Howell	male	26.0	0	0	111369	30.0000	C148	
	890	891	0	3	Dooley, Mr. Patrick	male	32.0	0	0	370376	7.7500	NaN	

#### Data Size:

In [5]: df.shape

Out[5]: (891, 12)

In [7]: df.head()

Out[7]: Pa:

	Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Emb
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.2500	NaN	
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	female	38.0	1	0	PC 17599	71.2833	C85	
2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.9250	NaN	
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.1000	C123	
4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.0500	NaN	
4												•

#### **Data Visualization:**

In [10]: df.sample(5)

[10]:		Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Emba
	461	462	0	3	Morley, Mr. William	male	34.0	0	0	364506	8.0500	NaN	
	826	827	0	3	Lam, Mr. Len	male	NaN	0	0	1601	56.4958	NaN	
	848	849	0	2	Harper, Rev. John	male	28.0	0	1	248727	33.0000	NaN	
	874	875	1	2	Abelson, Mrs. Samuel (Hannah Wizosky)	female	28.0	1	0	P/PP 3381	24.0000	NaN	
	678	679	0	3	Goodwin, Mrs. Frederick (Augusta Tyler)	female	43.0	1	6	CA 2144	46.9000	NaN	

#### **Column Data Varieties:**

```
In [12]: df.info()
```

<class 'pandas.core.frame.DataFrame'> RangeIndex: 891 entries, 0 to 890 Data columns (total 12 columns):

#	Column	Non-Null Count	Dtype
0	PassengerId	891 non-null	int64
1	Survived	891 non-null	int64
2	Pclass	891 non-null	int64
3	Name	891 non-null	object
4	Sex	891 non-null	object
5	Age	714 non-null	float64
6	SibSp	891 non-null	int64
7	Parch	891 non-null	int64
8	Ticket	891 non-null	object
9	Fare	891 non-null	float64
10	Cabin	204 non-null	object
11	Embarked	889 non-null	object
dtype	es: float64(2	), int64(5), obj	ect(5)

memory usage: 83.7+ KB

#### **Existence of Missing Values:**

```
In [14]: df.isnull().sum()
```

Out[14]: PassengerId 0 Survived 0 Pclass 0

> Name 0 Sex 0 177 Age SibSp 0 Parch 0 Ticket 0 Fare 0

> Cabin 687 Embarked 2

dtype: int64

#### **Mathematical Representation:**

In [15]: df.describe()

[].		()							
Out[15]:		Passengerld	Survived	Pclass	Age	SibSp	Parch	Fare	
	count	891.000000	891.000000	891.000000	714.000000	891.000000	891.000000	891.000000	
	mean	446.000000	0.383838	2.308642	29.699118	0.523008	0.381594	32.204208	
	std	257.353842	0.486592	0.836071	14.526497	1.102743	0.806057	49.693429	
	min	1.000000	0.000000	1.000000	0.420000	0.000000	0.000000	0.000000	
	25%	223.500000	0.000000	2.000000	20.125000	0.000000	0.000000	7.910400	
	50%	446.000000	0.000000	3.000000	28.000000	0.000000	0.000000	14.454200	
	75%	668.500000	1.000000	3.000000	38.000000	1.000000	0.000000	31.000000	
	max	891.000000	1.000000	3.000000	80.000000	8.000000	6.000000	512.329200	

### **Correlation Examination:**

In [16]: df.corr(numeric\_only = True)

Out[16]:

	Passengerld	Survived	Pclass	Age	SibSp	Parch	Fare
Passengerld	1.000000	-0.005007	-0.035144	0.036847	-0.057527	-0.001652	0.012658
Survived	-0.005007	1.000000	-0.338481	-0.077221	-0.035322	0.081629	0.257307
Pclass	-0.035144	-0.338481	1.000000	-0.369226	0.083081	0.018443	-0.549500
Age	0.036847	-0.077221	-0.369226	1.000000	-0.308247	-0.189119	0.096067
SibSp	-0.057527	-0.035322	0.083081	-0.308247	1.000000	0.414838	0.159651
Parch	-0.001652	0.081629	0.018443	-0.189119	0.414838	1.000000	0.216225
Fare	0.012658	0.257307	-0.549500	0.096067	0.159651	0.216225	1.000000

In [ ]: