# Why use ML for marketing? Strategies and use cases

MACHINE LEARNING FOR MARKETING IN PYTHON

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#### Types of machine learning

- Supervised learning
  - Given X, can we predict Y?
  - Classification when Y is categorical (e.g. Churned/Not-churned, Yes/No, Fish/Dog/Cat).
  - Regression when Y is continuous (e.g. Purchases, Clicks, Time Spent on Website).
- Unsupervised learning
  - Given X, can we detect patterns and clusters that are homogenous?
- Reinforcement learning
  - Given a current state and a number potential actions, which path maximizes the reward?

#### Supervised learning data parts and steps

- 1. Define the target (dependent variable or Y) what do we want to predict?
  - Example 1 which customers will churn? [CLASSIFICATION]
  - Example 2 which customers will buy again? [CLASSIFICATION]
  - Example 3 how much will customers spend in the next 30 days? [REGRESSION]
- 2. Collect features (independent variables or X) which could have predictive power:
  - Example 1 Purchase patterns prior churning.
  - Example 2 Number of missed loan payments prior defaulting on a loan.

#### Supervised learning data format

- X by N+1 matrix:
  - X number of observations (customer, vendor, product)
  - N + 1 number of columns (N features + 1 target variable)

| Feature 1 | Feature 2 | ••• | Feature N | Target Y |
|-----------|-----------|-----|-----------|----------|
| 11        | 21        |     | N1        | Y1       |
| 12        | 22        |     | N2        | Y2       |

#### Unsupervised learning

Collect usage or purchase data and run model to identify homogenous groups i.e. clusters or segments of data:

- Example 1 customer segmentation by their product purchases.
- Example 2 product segmentation for bundling

#### Unsupervised learning data format

- X by N matrix:
  - X number of observations (customer, vendor, product)
  - N number of columns (N features)
- A list of independent variables (features) as separate columns for each observation

| Feature 1 | Feature 2 | ••• | Feature N |  |  |
|-----------|-----------|-----|-----------|--|--|
| 11        | 21        |     | N1        |  |  |
| 12        | 22        |     | N2        |  |  |

## Let's practice!

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# Preparation for modeling

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#### Data sample

telco\_raw.head()

|   | customerID     | gender | SeniorCitizen | Partner | Dependents | tenure | PhoneService | MultipleLines    | InternetService | OnlineSecurity | <br>DeviceProtection | TechSu |
|---|----------------|--------|---------------|---------|------------|--------|--------------|------------------|-----------------|----------------|----------------------|--------|
| 0 | 7590-<br>VHVEG | Female | No            | Yes     | No         | 1      | No           | No phone service | DSL             | No             | <br>No               |        |
| 1 | 5575-<br>GNVDE | Male   | No            | No      | No         | 34     | Yes          | No               | DSL             | Yes            | <br>Yes              |        |
| 2 | 3668-<br>QPYBK | Male   | No            | No      | No         | 2      | Yes          | No               | DSL             | Yes            | <br>No               |        |
| 3 | 7795-<br>CFOCW | Male   | No            | No      | No         | 45     | No           | No phone service | DSL             | Yes            | <br>Yes              |        |
| 4 | 9237-<br>HQITU | Female | No            | No      | No         | 2      | Yes          | No               | Fiber optic     | No             | <br>No               |        |

5 rows × 21 columns



#### Data types

telco\_raw.dtypes

| customerID      | object |
|-----------------|--------|
| gender          | object |
| SeniorCitizen   | object |
| Partner         | object |
| Dependents      | object |
| tenure          | int64  |
| PhoneService    | object |
| MultipleLines   | object |
| InternetService | object |
|                 |        |

| OnlineSecurity   | object  |
|------------------|---------|
| OnlineBackup     | object  |
| DeviceProtection | object  |
| TechSupport      | object  |
| StreamingTV      | object  |
| StreamingMovies  | object  |
| Contract         | object  |
| PaperlessBilling | object  |
| PaymentMethod    | object  |
| MonthlyCharges   | float64 |
| TotalCharges     | float64 |
| Churn            | object  |
|                  |         |

#### Separate categorical and numerical columns

Separate the identifier and target variable names as lists

```
custid = ['customerID']
target = ['Churn']
```

Separate categorical and numeric column names as lists

#### One-hot encoding

This is a typical categorical data type column

Color

Red

White

Blue

Red



#### One-hot encoding result

And this is how it looks when we transform it with one-hot encoding.

| Color |   | Red | White | Blue |
|-------|---|-----|-------|------|
| Red   | > | 1   | 0     | 0    |
| White | > | 0   | 1     | 0    |
| Blue  | > | 0   | 0     | 1    |
| Red   | > | 1   | 0     | 0    |

#### One-hot encoding categorical variables

One-hot encoding categorical variables

telco\_raw = pd.get\_dummies(data=telco\_raw, columns=categorical, drop\_first=True)



#### Scaling numerical features

```
# Import StandardScaler library
from sklearn.preprocessing import StandardScaler
# Initialize StandardScaler instance
scaler = StandardScaler()
# Fit the scaler to numerical columns
scaled_numerical = scaler.fit_transform(telco_raw[numerical])
# Build a DataFrame
scaled_numerical = pd.DataFrame(scaled_numerical, columns=numerical)
```

#### Bringing it all together

## Let's practice preprocessing data!

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## ML modeling steps

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#### Supervised learning steps

- 1. Split data to training and testing
- 2. **Initialize** the model
- 3. Fit the model on the training data
- 4. Predict values on the testing data
- 5. Measure model performance on testing data

#### Supervised learning with code

First, let's load the libraries.

```
from sklearn import tree
from sklearn.model_selection import train_test_split
from sklearn.metrics import accuracy_score
```



#### Supervised learning steps with code

```
# 1. Split data to training and testing
train_X, test_X, train_Y, test_Y = train_test_split(X, Y, test_size=0.25)
# 2. Initialize the model
mytree = tree.DecisionTreeClassifier()
# 3. Fit the model on the training data
treemodel = mytree.fit(train_X, train_Y)
# 4. Predict values on the testing data
pred_Y = treemodel.predict(test_X)
# 5. Measure model performance on testing data
accuracy_score(test_Y, pred_Y)
```

#### Unsupervised learning steps

- 1. **Initialize** the model
- 2. **Fit** the model
- 3. Assign cluster values
- 4. **Explore** results

#### Unsupervised learning with code

First, let's load the libraries.

```
from sklearn.cluster import KMeans
import pandas as pd
```



#### Unsupervised learning with code

```
1. Initialize the model
kmeans = KMeans(n_clusters=3)
# Fit the model
kmeans.fit(data)
# 3. Asign cluster values
data.assign(Cluster=kmeans.labels_)
# 4. Explore results
data.groupby('Cluster').mean()
```

# Let's go build some models!

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