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	 DeviceProtection	TechSu
0	7590- VHVEG	Female	No	Yes	No	1	No	No phone service	DSL	No	 No	
1	5575- GNVDE	Male	No	No	No	34	Yes	No	DSL	Yes	 Yes	
2	3668- QPYBK	Male	No	No	No	2	Yes	No	DSL	Yes	 No	
3	7795- CFOCW	Male	No	No	No	45	No	No phone service	DSL	Yes	 Yes	
4	9237- HQITU	Female	No	No	No	2	Yes	No	Fiber optic	No	 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
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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