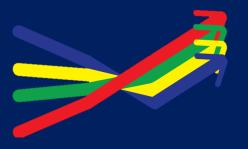
Object Oriented Programming for Machine Learning



Learn and transform, learn and predict

The different steps of the Machine Learning pipeline:

- Learn parameters from data
- Use those parameters to make transformations or predictions

Procedural Programming in ML

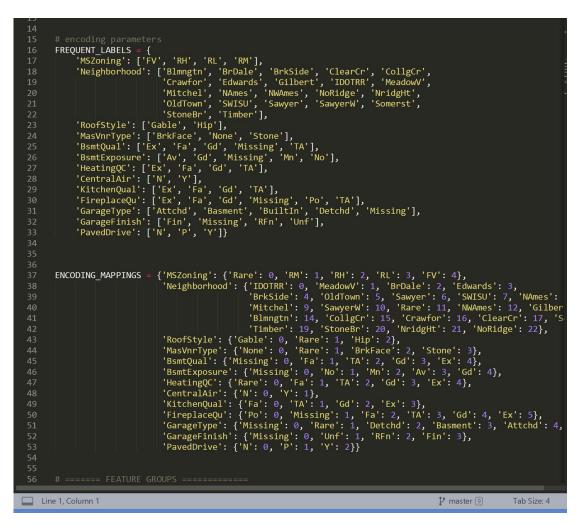
Code:

- Learn the parameters
- Make the transformations
- Make the predictions

Data:

- Store the parameters
- Mean values, regression coefficients, etc.

Procedural Programming: Hard-coded values



Straightforward



 Save multiple objects or data structures



```
OUTPUT_SCALER_PATH = 'scaler.pkl'
OUTPUT_MODEL_PATH = 'lasso_regression.pkl'
```





Object Oriented Programming - OOP

In Object-oriented programming (OOP) we write code in the form of "objects".

This "objects" can store **data** and can also store instructions or procedures (**code**) to modify that data, or do something else, like obtaining predictions.

- Data ⇒ attributes, properties
- Code or Instructions ⇒ methods (procedures)

OOP for Machine Learning

In Object-oriented programming (OOP) the "objects" can learn and store parameters

- Parameters get automatically refreshed every time model is re-trained
- No need of manual hard-coding
 - Methods:
 - Fit: learns parameters
 - Transform: transforms data with the learned parameters
 - Attributes: store the learn parameters

A Class

Creating a Class in Python

class MeanImputer:

pass

A Class - ___init___()

The properties or parameters that the class takes whenever it is initialized, are indicated in the __init__() method.

The first parameters will always be a variable called *self*.

We can give any number of parameters to __init__()

```
class MeanImputer:

    def __init__(self, variables):
        self.variables = variables
```

A Class - __init__()

The properties or parameters that the class takes whenever it is initialized, are indicated in the __init__() method.

The first parameters will always be a variable called self.

We can give any number of parameters to __init__()

```
class MeanImputer:
    def __init__(self, variables):
        self.variables = variables
```

```
>> my_imputer = MeanImputer(
>> variables = ['age', 'fare']
>> )
>> my_imputer.variables
['age', 'fare']
```

A Class - methods

Methods are functions defined inside a class and can only be called from an instance of that class.

The first parameters will always be a variable called self.

Our fit() methods learns parameters

```
class MeanImputer:
    def __init__(self, variables):
        self.variables = variables

def fit(self, X, y=None):
        self.imputer_dict_ =
        X[self.variables].mean().to_dict()

        return self
```

A Class - methods

Methods are functions defined inside a class and can only be called from an instance of that class.

The first parameters will always be a variable called self.

Our fit() methods learns parameters

```
>> my_imputer = MeanImputer(
>> variables = ['age', 'fare']
>> )

>> my_imputer.fit(my_data)
>> my_imputer.imputer_dict_

{'age': 39, 'fare': 100}
```

A Class - methods

Methods are functions defined inside a class and can only be called from an instance of that class.

The first parameters will always be a variable called self.

Our fit() methods learns parameters

Our transform() method transforms data

```
class MeanImputer:
    def __init__(self, variables):
         self.variables = variables
   def fit(self, X, y=None):
         self.imputer dict =
         X[self.variables].mean().to dict()
         return self
    def transform(self, X):
         for x in self.variables:
              X[x] = X[x].fillna(
                    self.imputer_dict_[x])
         return X
```