

Week 8: Python for Machine Learning

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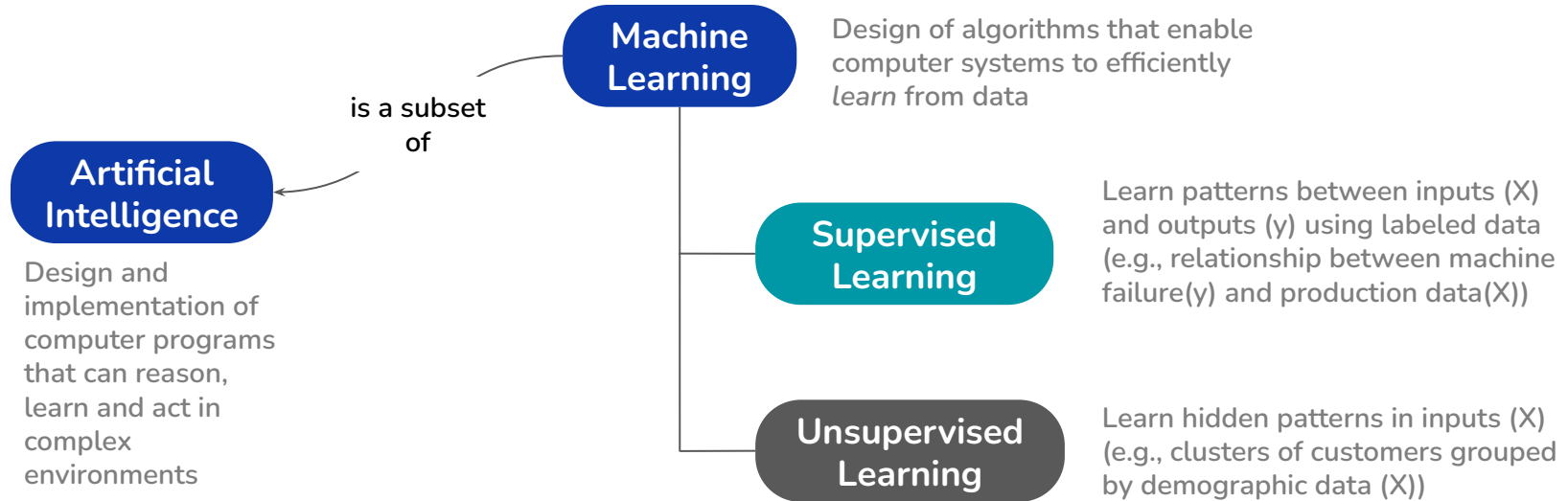
Agenda

In this session, we will discuss:

- Key Ideas of Machine Learning (ML)
- ML Workflow - An Overview
- ML for Regression and Classification Tasks
- Hyperparameter Tuning and Model Evaluation
- Serializing ML Models

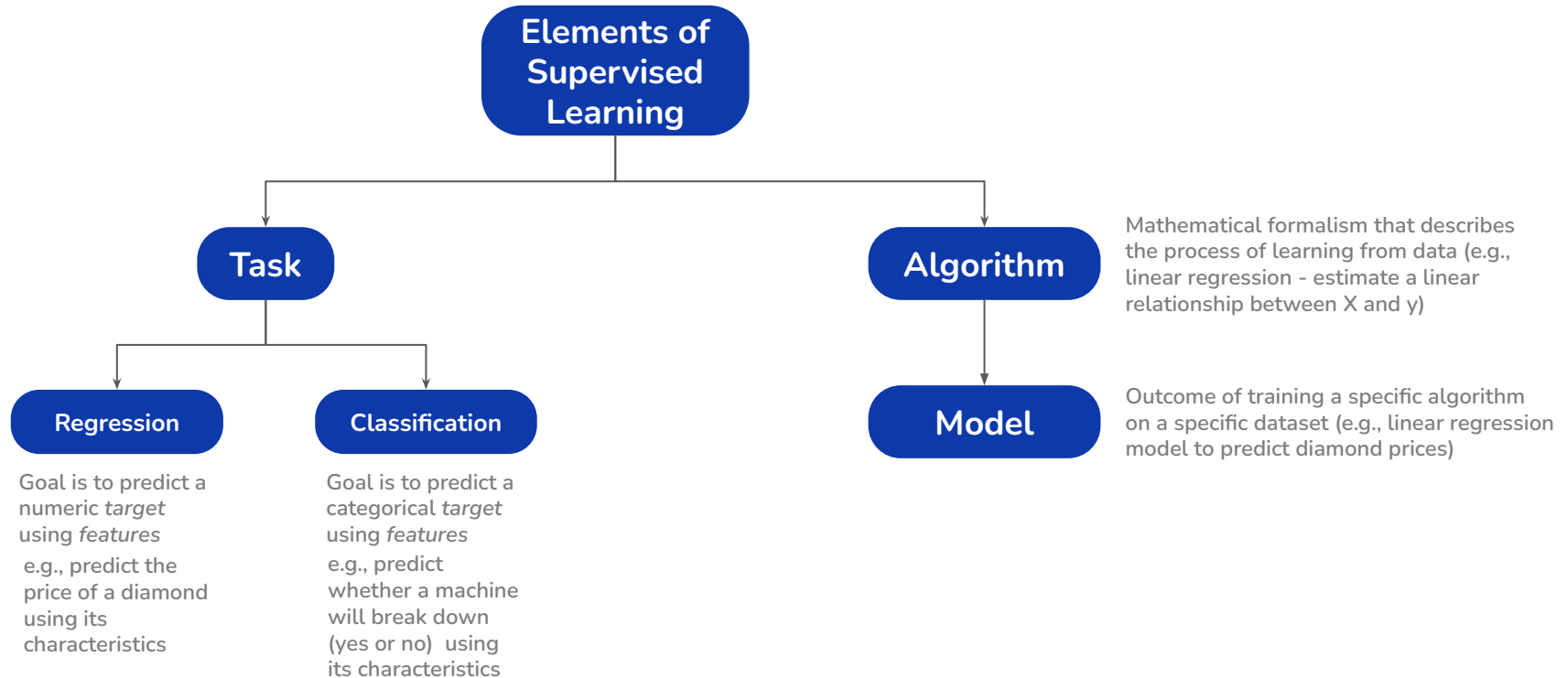
Machine Learning - An Introduction

Terminology



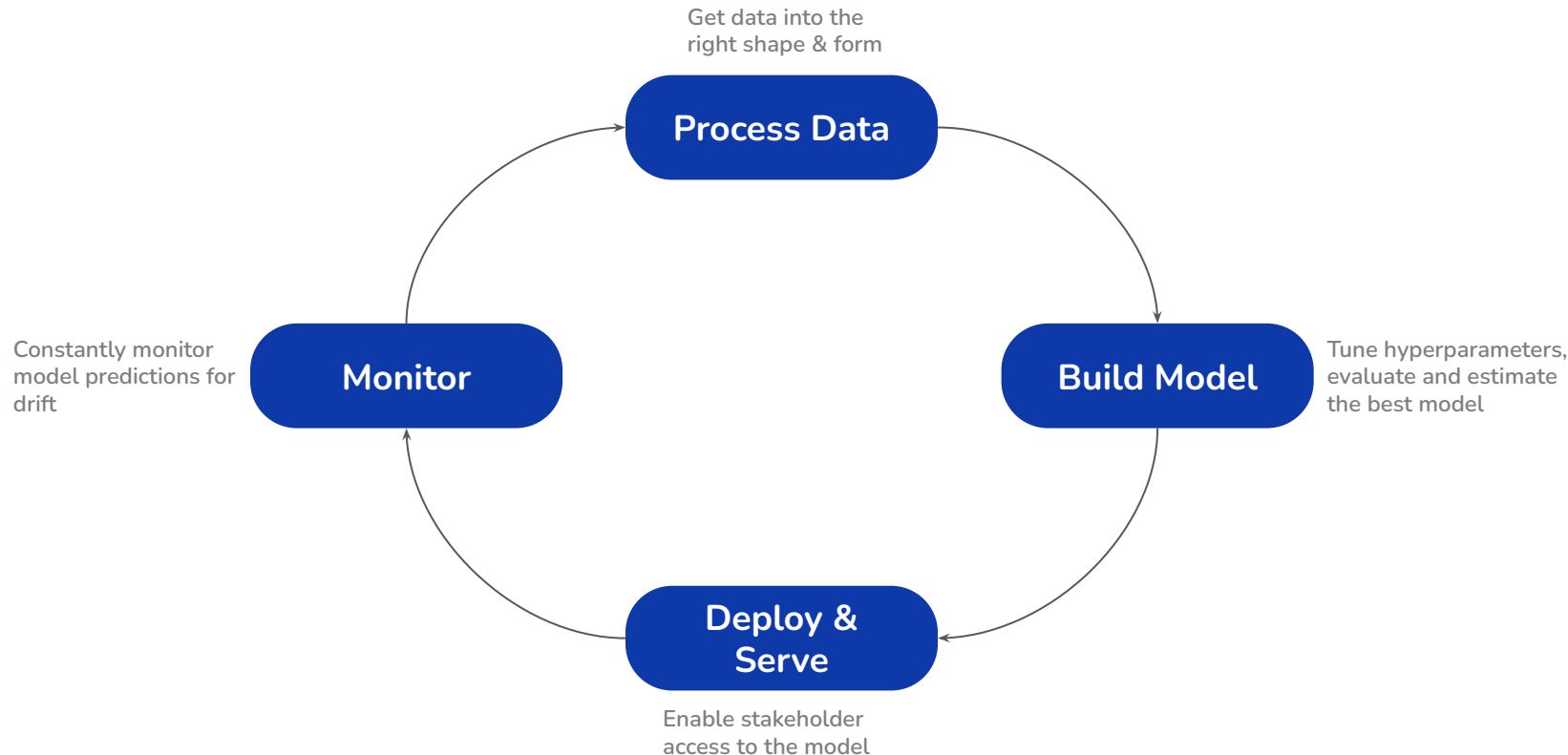
Machine Learning - An Introduction

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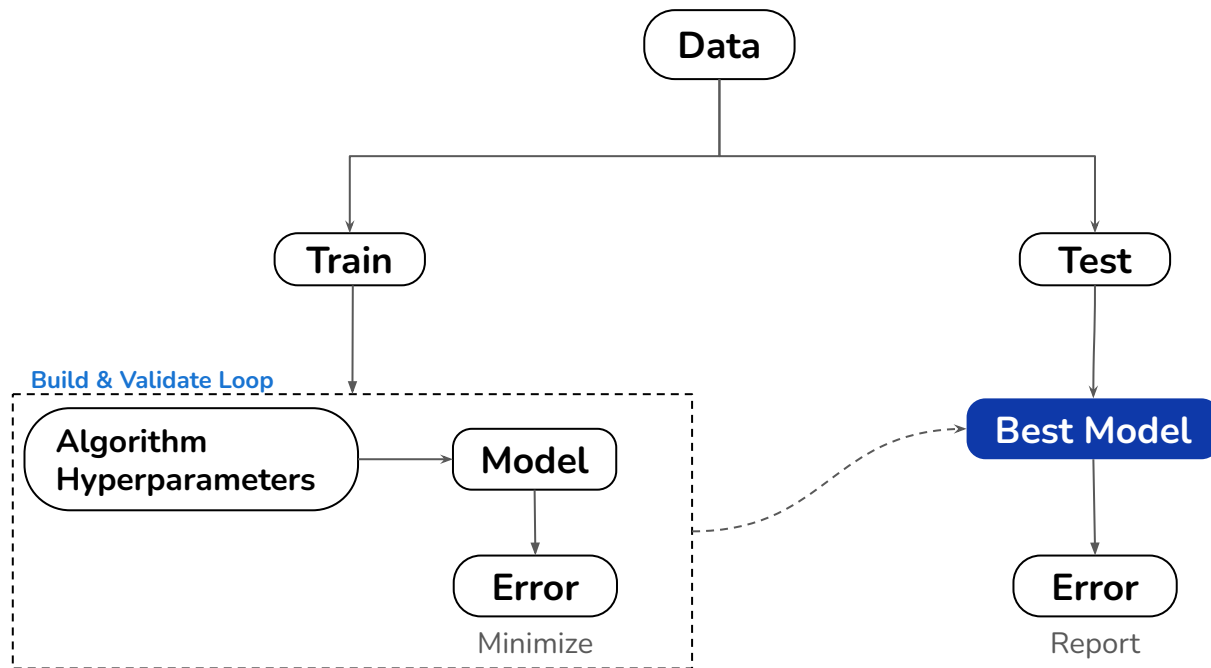
ML Workflow - An Overview



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Machine Learning - An Introduction

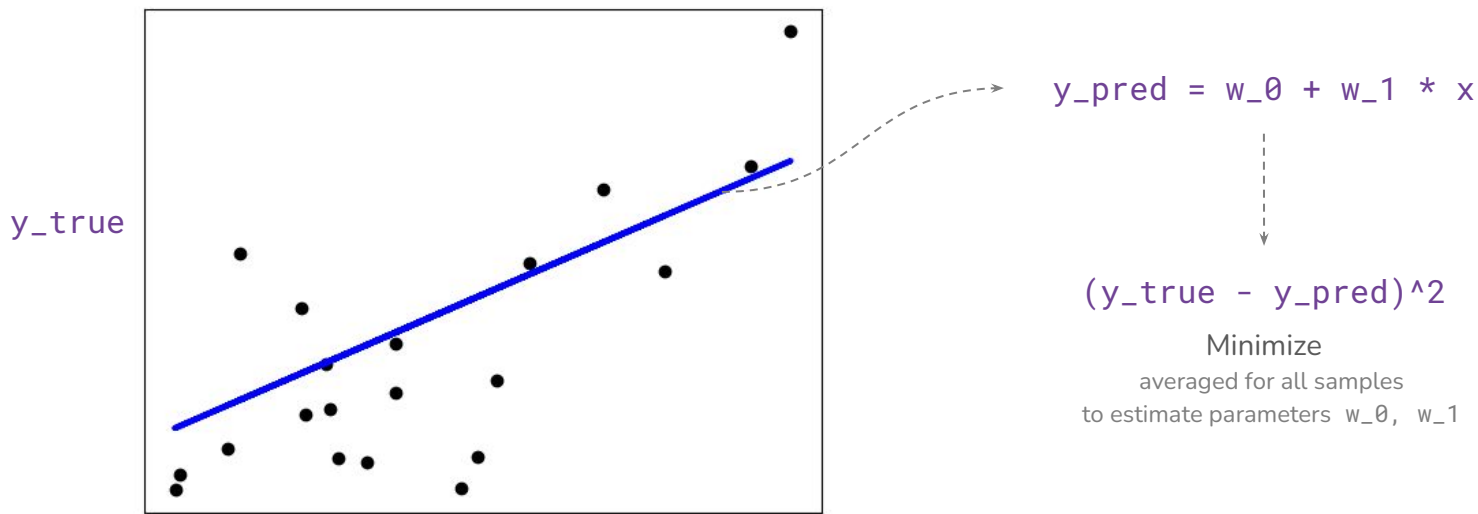
Supervised Learning Components for Efficient Learning



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Supervised Learning for Regression Tasks

Algorithm: Linear Regression estimates a linear model that is the best predictor of the target given the features in the training data. The best predictor is derived by minimizing the mean squared differences between the true targets and the targets predicted by the linear approximation.

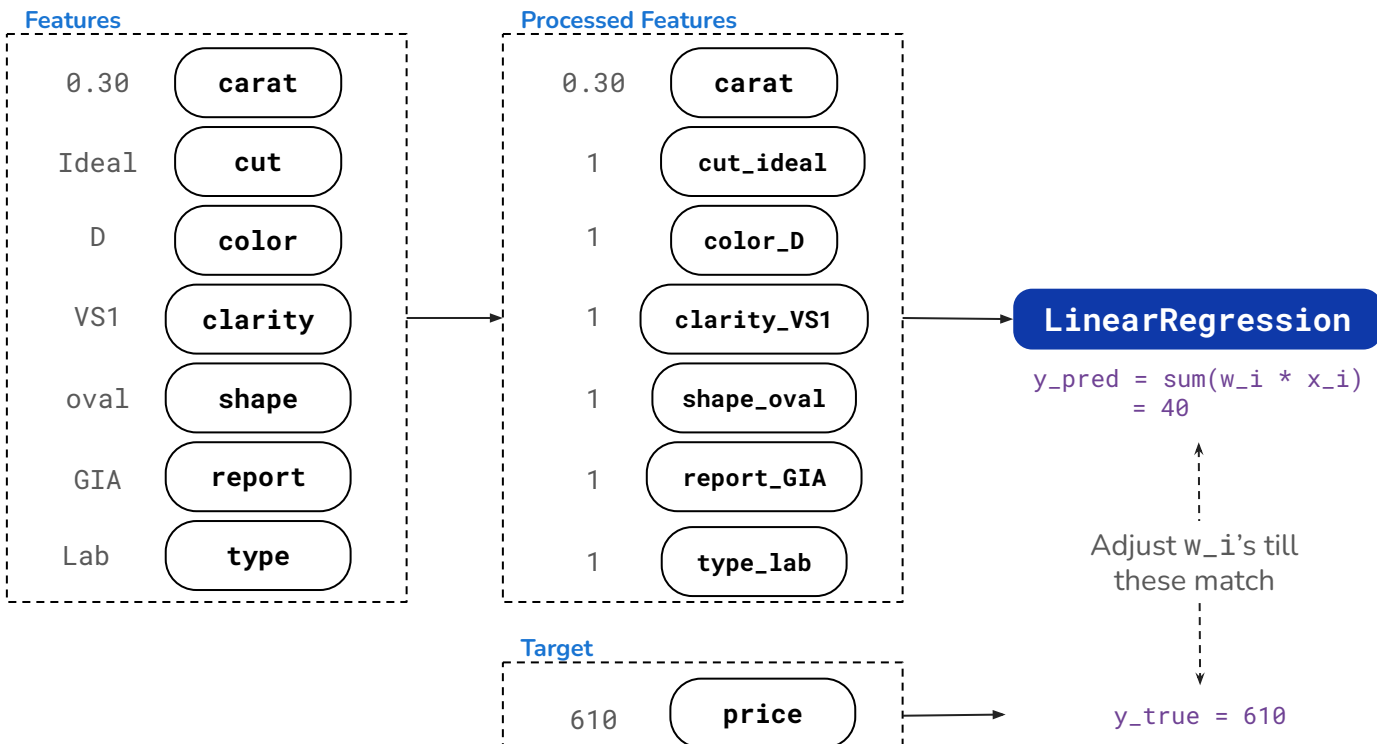


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Supervised Learning for Regression Tasks

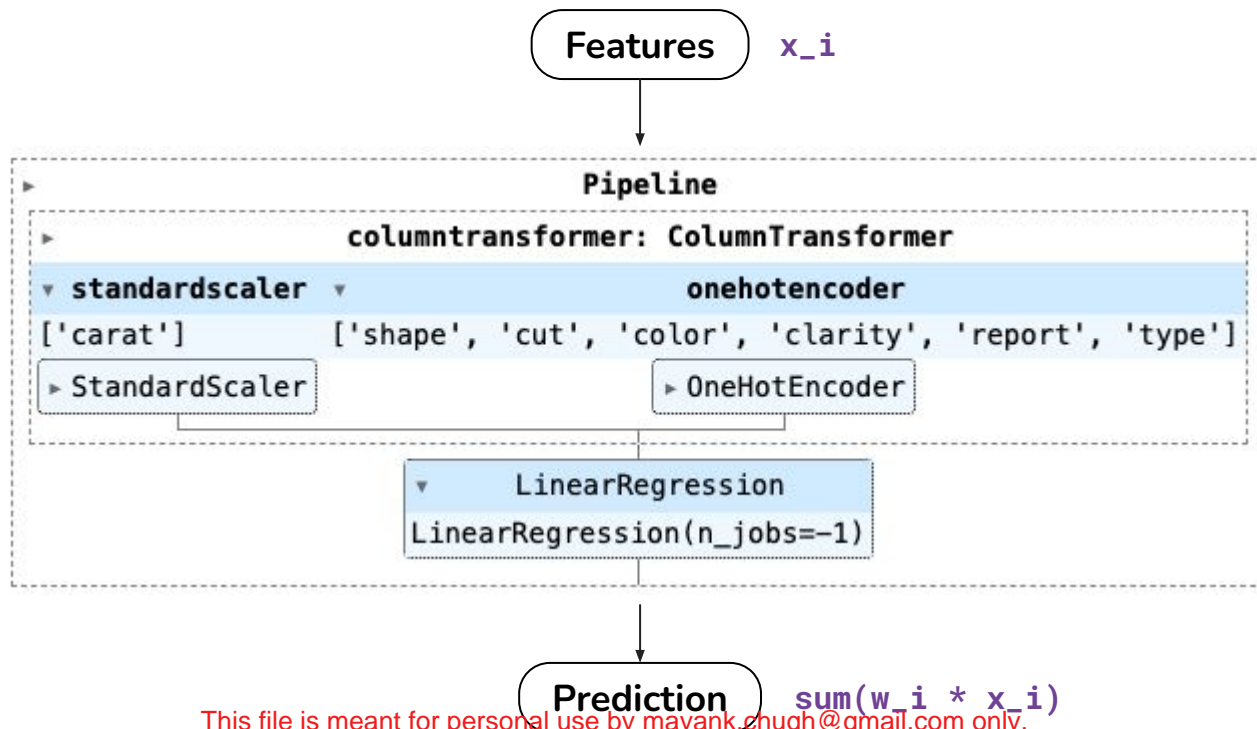
Example: Diamond Price Prediction **Model**



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Supervised Learning for Regression Tasks

Example: Diamond Price Prediction with scikit-learn

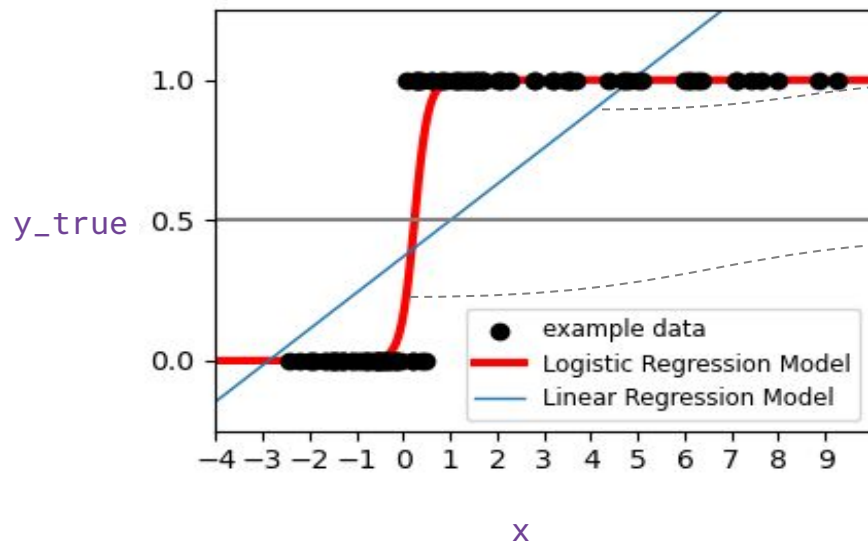


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[\[Notebook\]](#)

Supervised Learning for Classification Tasks

Algorithm: Logistic Regression squishes a linear model through a logistic function to estimate probabilities of the target classes. The best model minimizes the negative log probability predicted by the linear approximation + logistic squisher.



$$s = \text{sum}(w_i * x_i)$$

$$h(s) = 1/(1+e^{-s})$$

$$-\log(h(s)) + C * \text{sum}(|w|)$$

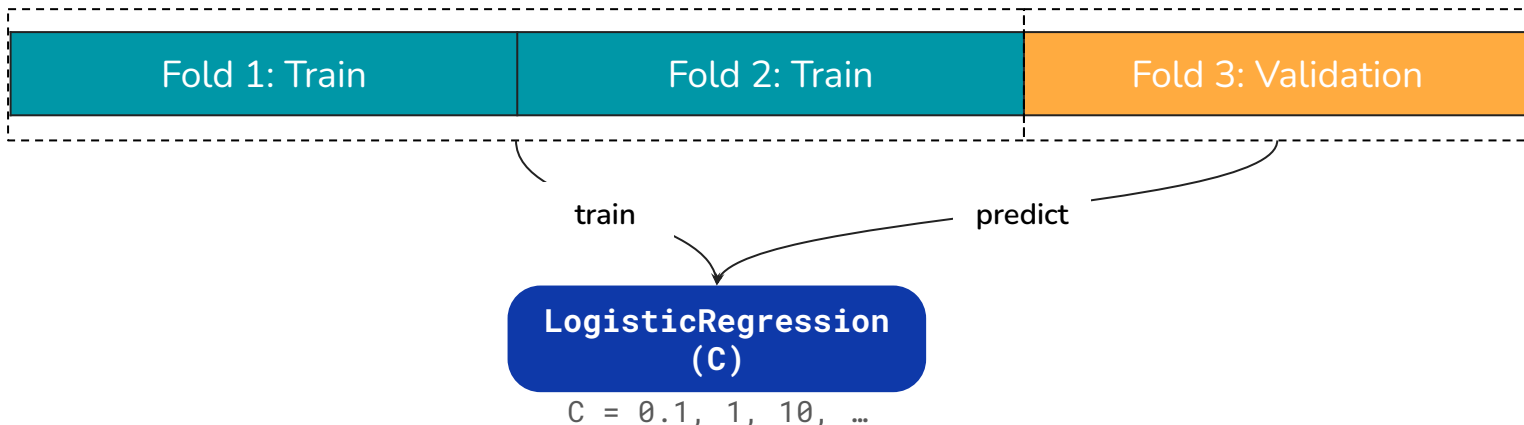
Minimize

averaged for all samples & a value of hyperparameter C
to estimate parameters w_0, w_1

Supervised Learning for Classification Tasks

Algorithm: To estimate the best value of the hyperparameter C , we divide training data into folds and estimate models with different values of C on the training folds. The performance of the model is then measured on the validation fold. The best value of C corresponds to the model with highest validation accuracy. This is called **hyperparameter tuning**.

Training Data for 3-Fold Cross Validation

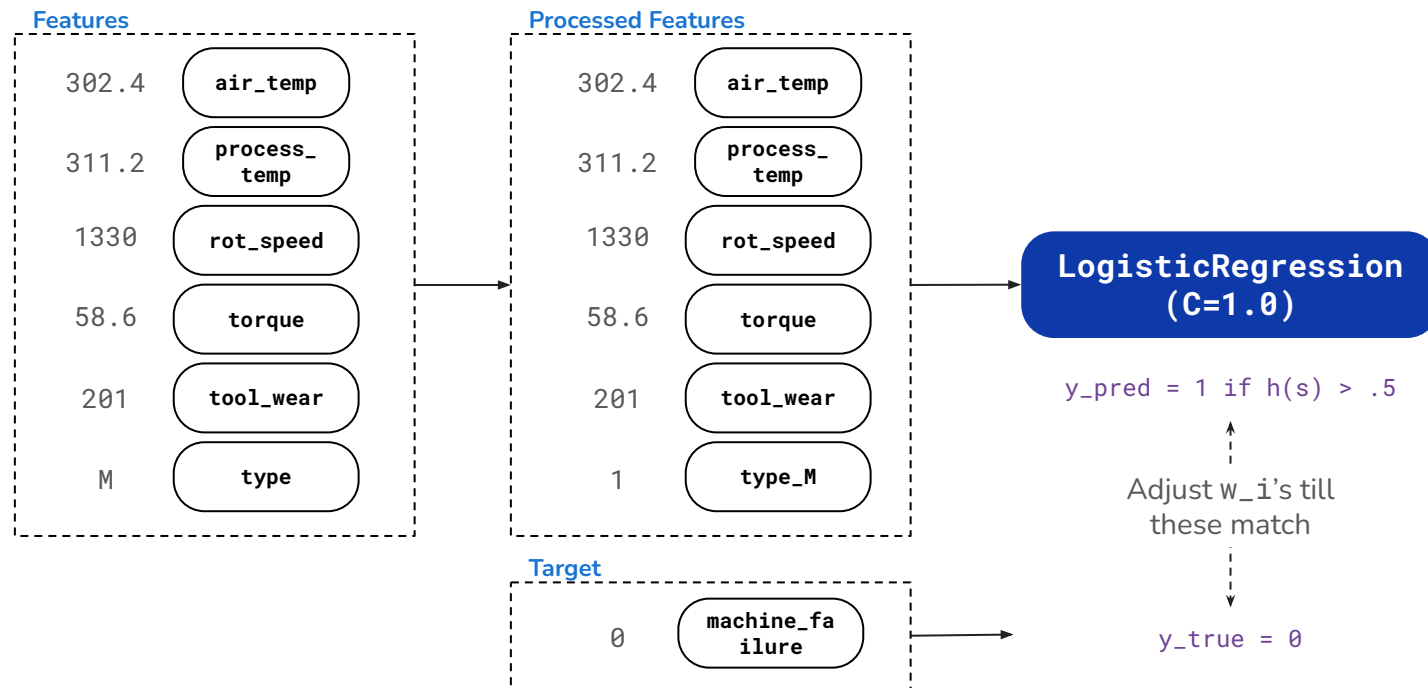


Shuffle folds and estimate best C

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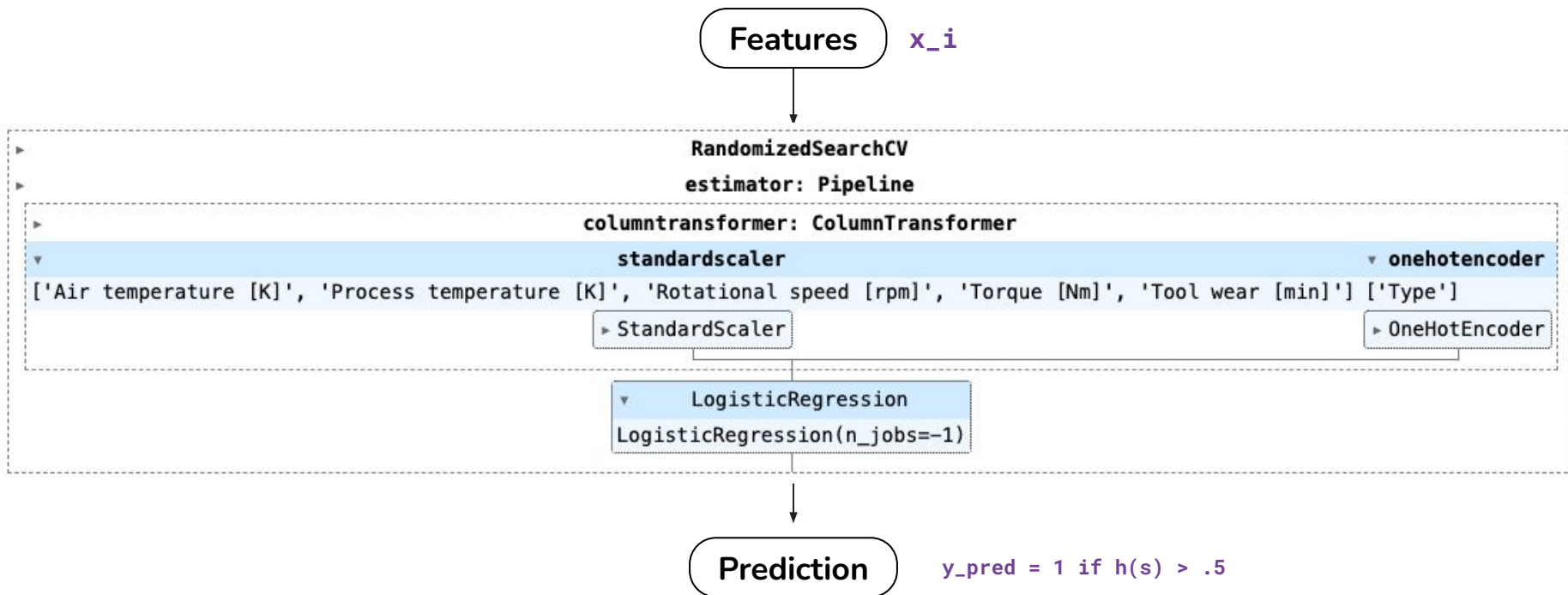
Supervised Learning for Classification Tasks

Example: Machine Failure Prediction **Model**



Supervised Learning for Classification Tasks

Example: Machine Failure Prediction with scikit-learn

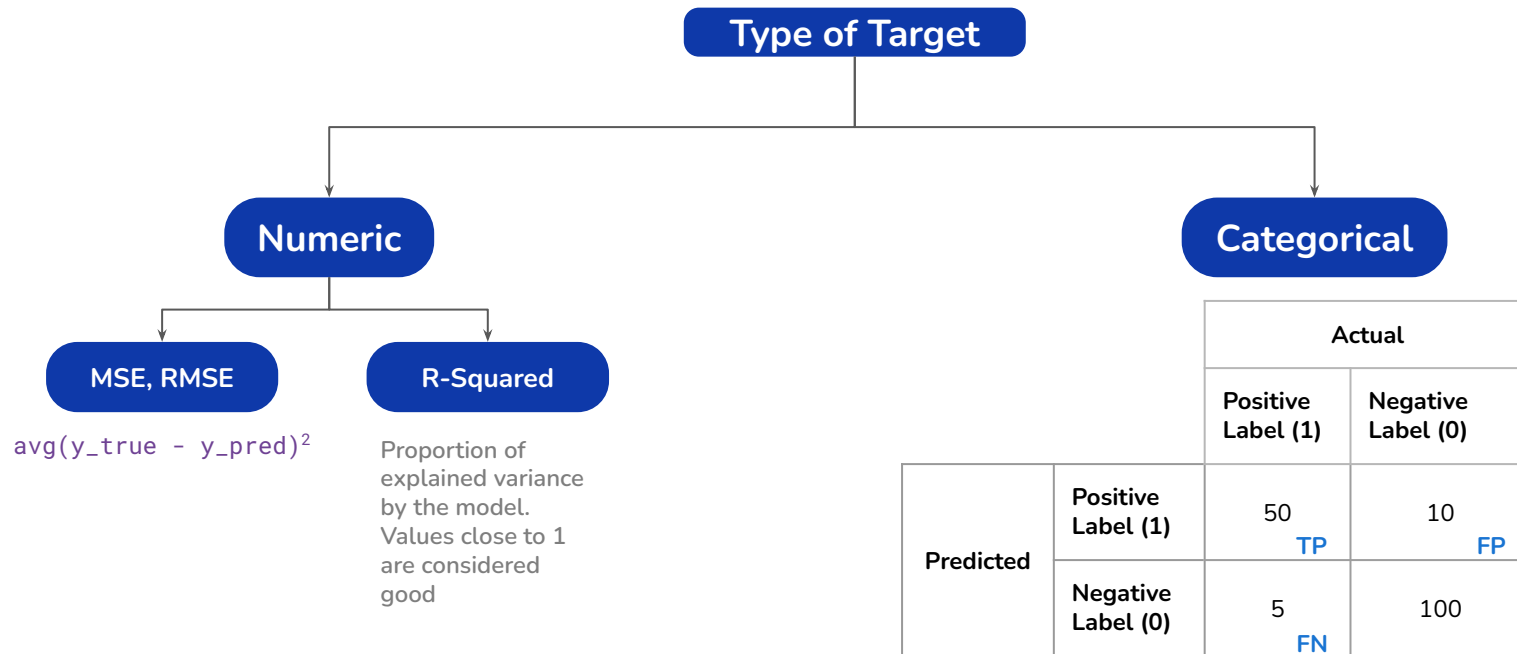


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Model Evaluation

Model evaluation is conducted by comparing predictions and targets on train, validation and test data



$$\text{Micro-precision} = \text{TP} / (\text{TP} + \text{FP}) = 50 / (50 + 10) = 0.833$$

$$\text{Micro-recall} = \text{TP} / (\text{TP} + \text{FN}) = 50 / (50 + 5) = 0.909$$

$$\text{Micro-F1 score} = 2 * (\text{Micro-precision} * \text{Micro-recall}) / (\text{Micro-precision} + \text{Micro-recall}) = 2 * (0.833 * 0.909) / (0.833 + 0.909) = 0.870$$

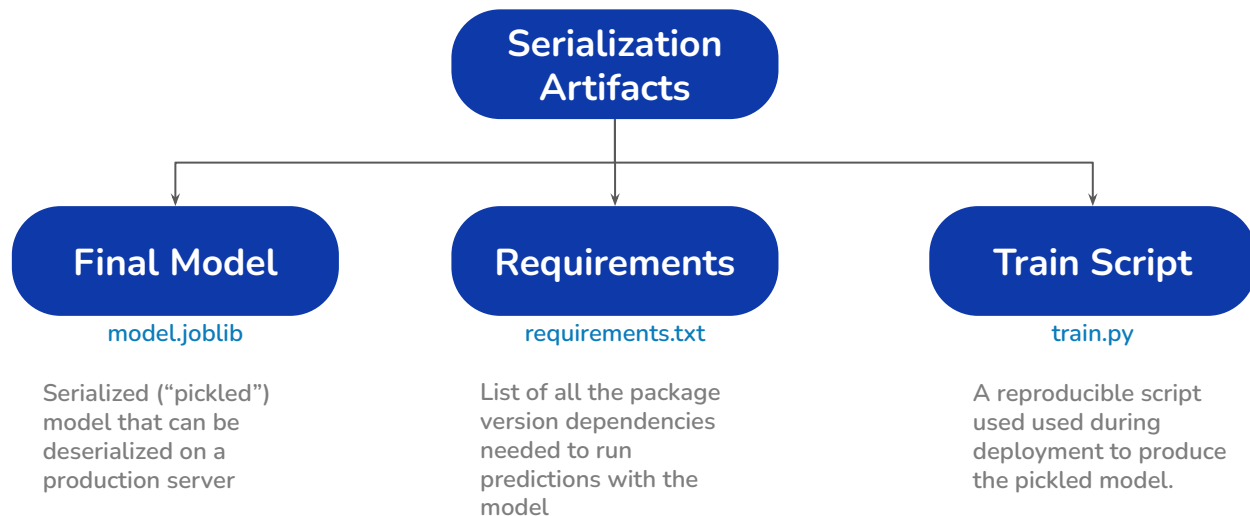
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Model Serialization

Model serialization is the process of creating persistent, deployable artifacts after model training is complete. Three artifacts are created during serialization - a byte-stream conversion (e.g., binary file) of the model Python object, a requirements file and a training script.



Summary

A paradigm of machine learning from input-output exemplars

Supervised Learning

is used to solve

Regression & Classification Tasks

Objectives: predict a numeric target (regression) or a categorical target (classification)

by building

Models

using

Data

&

Algorithms

Train, Validation and Test

Mathematical formalism detailing the process of learning from data

serialized to

Deployable Artifacts

Pickled model pipeline, a requirements file and a training script

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