# A Novel AutoML Solution with NePS

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Modality 1/2

### Motivation

Our main motivation was to create a versatile one click AutoML solution that can adapt to any tabular dataset using 8 different state-of the-art regression models.

NePS was chosen as an hyperparameter optimizer because it was the option which we had experience in exercises and it was transparent with the source code.

## **Used Methods**

Bayesian Hyperparameter Optimization

Successive Halving

IQR Variable Scale Outlier Detection

Standardization

# Week 1

Week 2

Week 3

Week 4

Week 5

Week 6

Week 7

Week 8

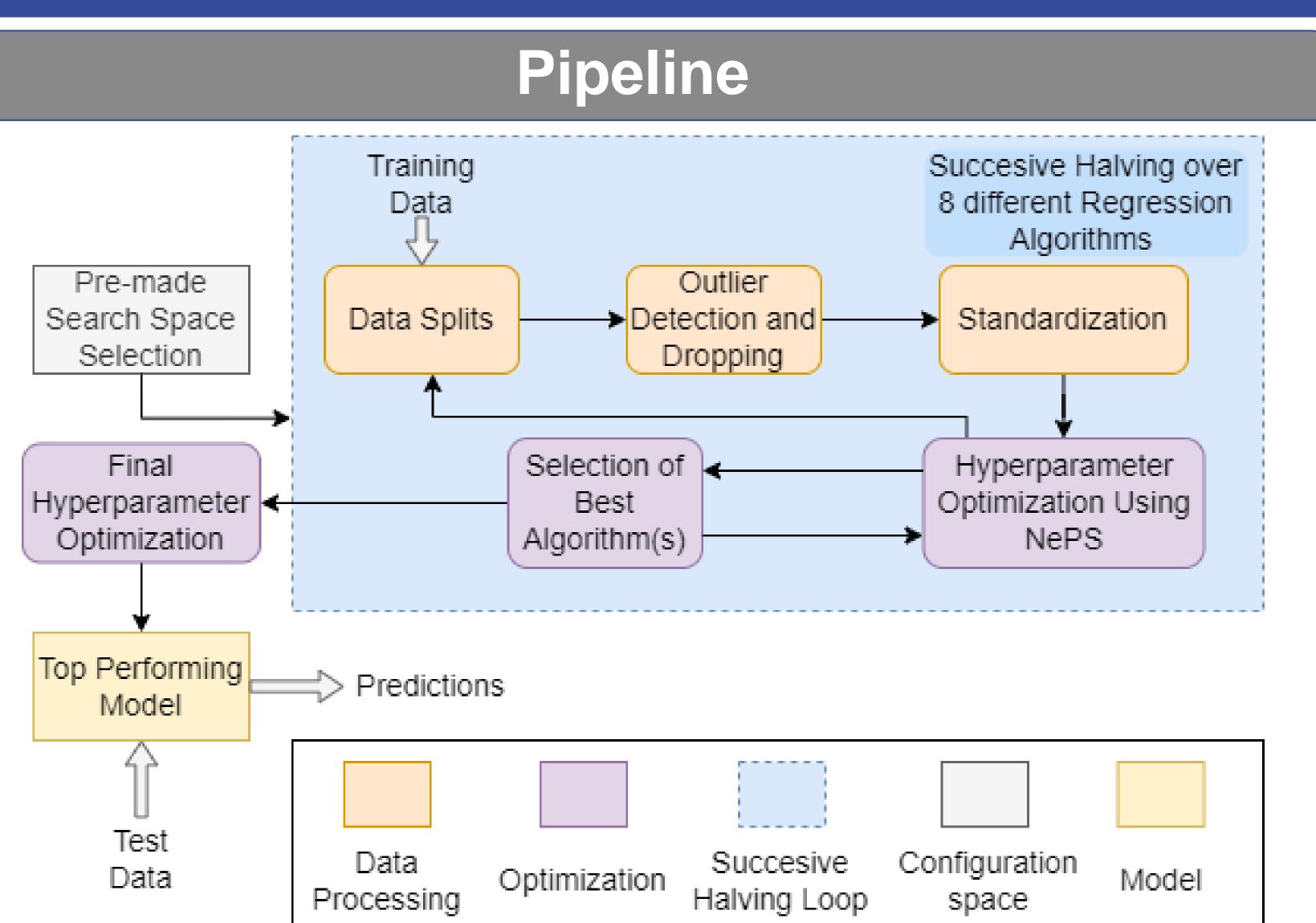
Week 9

Week 10

Bonus

Literature

# Our Approach



# Details

The pipeline consists of a main successive halving loop which works on 8 different regression algorithms such as:

- Sklearn and XGB Random Forest
- Gradient Boost
- Ada Boost
- MLP
- Bayesian Ridge
- Elastic Net
- Dart Boost

Optimization run on the given configuration space, the IQR scale and the validation split ratio. The resulting best model is then saved to be used for predictions.

#### Resources Used

For development:

- 1 Ryzen 9 5900HX (Laptop)

- Total compute estimate: 50 CPU-hours

### Workforce:

- 2 full week on average

# **Empirical Results**

# y\_prop Dataset

Validation 0.1229

Test 0.1135

> Chosen Algorithm: SKlearn Random Forest

Bike Sharing

Dataset

Test

0.9755

Validation

0.9423

For the Bike Sharing Dataset, our AutoML solution found the Gradient Boost as the most optimal algorithm.

# Brazilian Houses Dataset

Validation 0.9997

Test 0.997

Chosen Algorithm: SKlearn Random Forest

For the Brazilian Houses
Dataset, our AutoML
solution found the MLP
as the most optimal
algorithm.

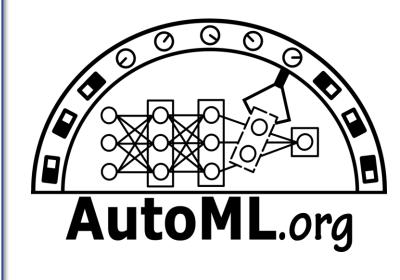
## Exam Dataset

Validation 0.9016

Test 0.8868

Chosen Algorithm:
Gradient Boost

For the Exam Dataset, our AutoML solution found the Dart Boost as the most optimal algorithm.



Number of queries

for test score

generation: 5



For the y\_prop dataset, our AutoML solution found the XGB Random Forest as the most optimal algorithm.

Chosen Algorithm:

XGB Random Forest