

## The Ex-Def Baseline

- Is just to cross-validate all learners and pick the best one (no pre-processors considered).
- Proposed (and used only) in the 2013 Auto-WEKA paper.
- Claimed to be outperformed but in fact (much) stronger than one would expect.
- We extend this baseline by also including pre-processors, meta (ensembles) learners, and some random parameter tuning.

## Naivity Assumption

### Formulation

Let  $\phi(D, c_1 \circ \dots \circ c_{k+1})$  be the prediction performance of the pipeline consisting of transformers  $c_1, \dots, c_k$  and estimator  $c_{k+1}$  on dataset  $D$ . *Naive* assumption:

$$c_i^* \in \arg \min_{c_i} \phi(D, c_1 \circ \dots \circ c_{k+1})$$

is *invariant* to the choices of  $c_1, \dots, c_{i-1}, c_{i+1}, \dots, c_{k+1}$ .

If we consider the search space an urn and denote as  $Y$  the event to observe an optimal pipeline in the urn, then

$$\begin{aligned} P(Y \mid c_1, \dots, c_{k+1}) &\propto P(c_1, \dots, c_{k+1} \mid Y)P(Y) \\ &= P(c_i \mid Y) \prod_{j=1, j \neq i}^{k+1} P(c_j \mid Y)P(Y) \end{aligned}$$

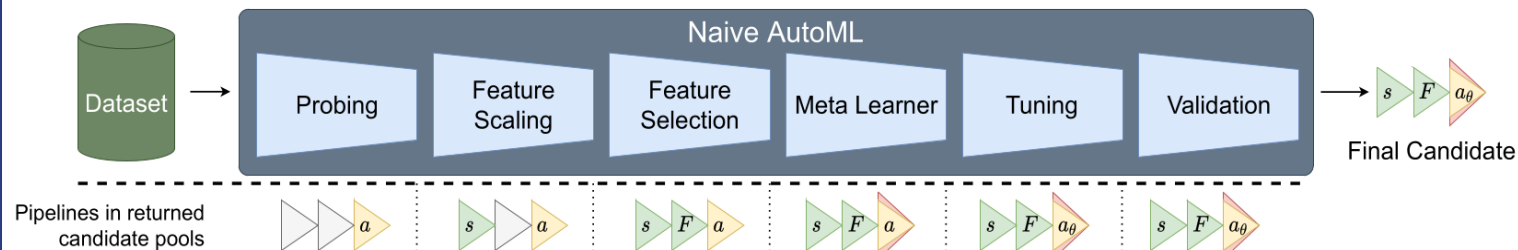
### Implication

The naivity assumption allows to optimize all parts in isolation. Or at least to have only a loose coupling.

## Naive AutoML

### Idea

Optimize different aspects of the ML pipeline in sequentially *isolated stages*.



### Results (ranks after 1h runtime, summarizing 67 datasets from AutoML Benchmark)

