

Siminole kick-off meeting – Task 4

Simulation-based discriminative learning

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Discriminative learning

- Simulate $D = \{\mathbf{x}_i, y_i\}_{i=1}^n$
 - $\mathbf{x} \in \mathbb{R}^d$: observation vector
 - $y \in \{1, \dots, K\}$: label
- Examples
 - \mathbf{x} : tank observables, y : number of muons
 - \mathbf{x} : images or films on the focal plane, y : shower / background
 - \mathbf{x} : ATLAS observables, y : Higgs / background

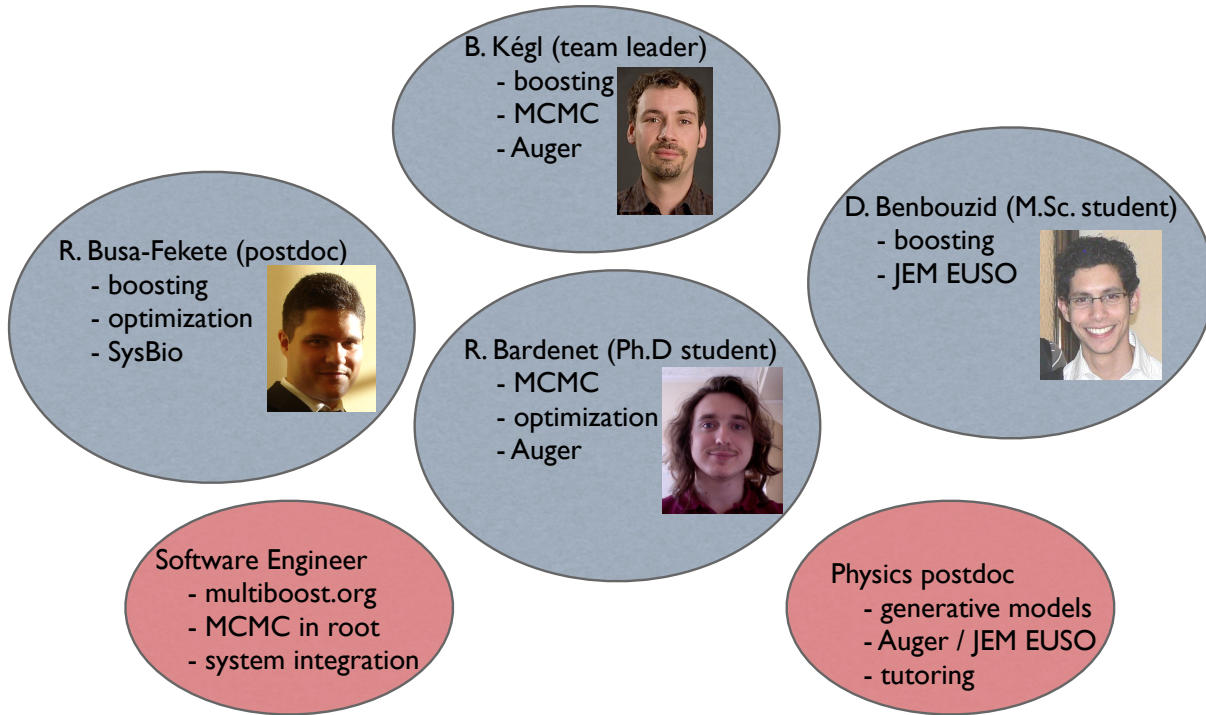
Discriminative learning

- Goal: learn $\hat{y} = f(\mathbf{x})$ based on $D = \{\mathbf{x}_i, y_i\}_{i=1}^n$
 - minimize $\mathbb{P}\{f(\mathbf{x}) \neq y | D\}$
 - off-line analysis (**reconstruction**), on-line **trigger**, **hypothesis testing**
- Why **AdaBoost**?
 - we have the **expertise**
 - **scales** well for large problems
 - **versatility** (multiclass, non-differentiable weak learners)

Discriminative learning

- Research goals
 - **large-scale** issues (speed, memory, automatic hyperparameter optimization)
 - **beyond classification** (ranking, regression, conditional random fields, reinforcement learning)
 - **unbalanced** data: **cascade** architecture, **Neyman-Pearson** learning
- Deliverable
 - modular AdaBoost **software** (`multiboost.org`)

The AppStat team at LAL



- + M. Sebag and C. Germain (TAO/LRI)
- Formal ties to the [TAO](#), [Auger](#) and [JEM-EUSO](#) teams