

## AutoML Lecture: Notation Cheat Sheet

Symbol	Meaning
<b>Machine Learning</b>	
$\mathcal{D}$	Dataset
$\mathcal{D}_{\text{train}}$	Training dataset
$\mathcal{D}_{\text{val}}$	Validation dataset
$\mathcal{D}_{\text{test}}$	Test dataset
$\mathbf{D}$	Space of datasets
$\mathbf{x}$	Feature vector
$y$	Label
$(\mathbf{x}^{(i)}, y^{(i)})$	$i$ -th observation
$L(y, f(\mathbf{x}))$	(empirical) loss
$\mathcal{R}$	risk
$\mathcal{R}_{\text{emp}}$	empirical risk
$f(\mathbf{x})$	continuous prediction function
$\mathcal{H}$	hypothesis space where $f$ is from
$\hat{f}$	estimated prediction function
<b>Hyperparameter Optimization</b>	
$\lambda$	Hyperparameter configuration
$\lambda_i$	Value of $i$ -th hyperparameter
$\lambda_{\text{def}}$	Default hyperparameter configuration
$\hat{\lambda}$	finally returned hyperparameter configuration
$\lambda^*$	Optimal hyperparameter configuration
$\Lambda$	Space of possible hyperparameter configurations
$\mathcal{A}$	Algorithm (e.g. SVM, RF, DNN)
$\mathbf{A}$	Distribution or set of algorithms
$c(\lambda)$	Target cost function (e.g., empirical risk, validation loss, runtime)
$\hat{c}(\lambda)$	Surrogate (probabilistic) model of target function
$\mathcal{D}_{\text{HPO}} = \langle \lambda^{(t)}, c(\lambda^{(t)}) \rangle_{t=1}^T$	All observations collected for BO / HPO
<b>Gaussian Processes and Bayesian Optimization</b>	
$\mathcal{G}$	Gaussian process
$t$	BO loop counter
$T$	BO loop counter max, the counter runs from 1 to this value
$u$	Acquisition Function, no args
$\phi$	Standard Normal PDF
$\Phi$	Standard Normal CDF
$\mu$	Mean
$\sigma$	Standard Deviation
$\sigma^2$	Variance
$\nu$	Noise
$\mathbb{R}$	Real numbers set
$\mathbb{E}$	Expected value
$\kappa$	kernel
$c$	Constraint function
$\mathcal{N}$	Normal distribution

Symbol	Meaning
<b>Algorithm Selection</b>	
$\mathbf{x}_{\text{meta}}$	Vector of (meta-) features
$\mathcal{X}_{\text{meta}}$	Space of (meta-)features
$\mathcal{P}$	Portfolio (i.e., discrete set) of algorithms or hyperparameter configurations
$\mathcal{S}$	Schedule of algorithms or hyperparameter configurations
<b>Meta-Learning</b>	
$\theta$	Weights (a.k.a. parameters) of ML model (e.g., DNN)
$\phi$	Weights of meta-model
$\mathcal{D}_{\text{meta}}$	Meta-dataset
<b>Reinforcement Learning</b>	
$\pi$	Reinforcement learning policy
$\Pi$	Space of policies
$a$	action in RL-setting
$s$	state in RL-setting
$\mathcal{S}$	Space of states
$r$	Reward in RL-setting
$\mathcal{R}$	Random variable or function of reward
<b>Algorithm Configuration</b>	
$\kappa$	Cutoff (often runtime) of an algorithm run
$i$	a single instances (a.k.a. problem, dataset, task)
$\mathcal{I}$	Distribution over instances (a.k.a. problems, datasets, tasks)