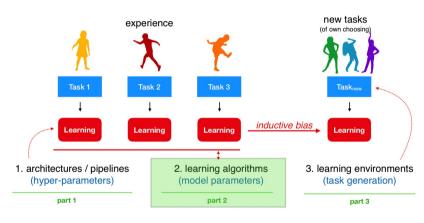
AutoML: Meta-Learning

Learning model priors

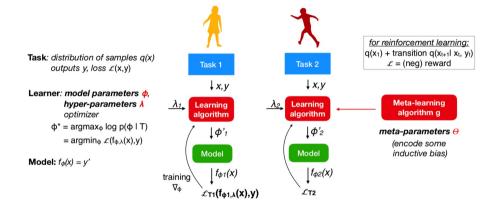
Bernd Bischl Frank Hutter Lars Kotthoff Marius Lindauer <u>Joaquin Vanschoren</u>

What can we learn to learn?

3 pillars

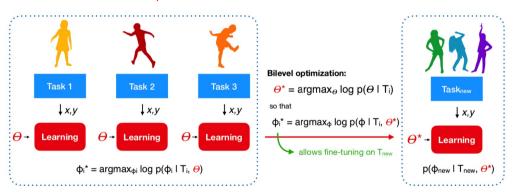


Terminology (reminder)



Strategy 1: bilevel optimization

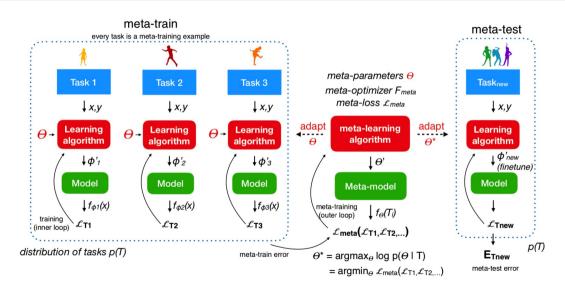
parameterize some aspect of the learner that we want to learn as meta-parameters θ meta-learn θ across tasks



 $\Theta(Prior)$, could encode an initialization ϕ , the hyperparameters λ , the optimizer,...

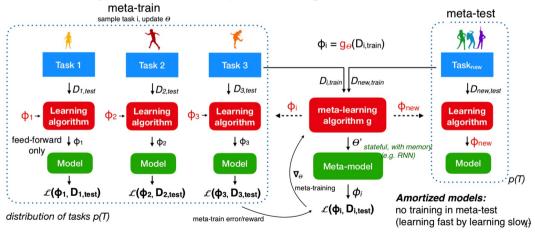
Learned θ^* should learn T_{new} from small amount of data, yet generalize to a large number of tasks

Meta-learning with bilevel optimization

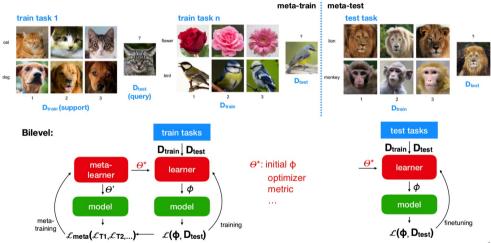


Strategy 2: black-box models

black box meta-model g_{θ} predicts ϕ given D_{train} (theta is hidden) hypernetwork where input embedding learned across tasks

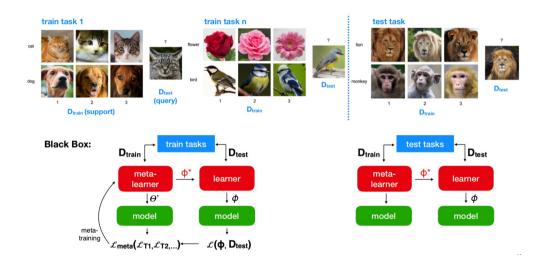


Example: few-shot classification

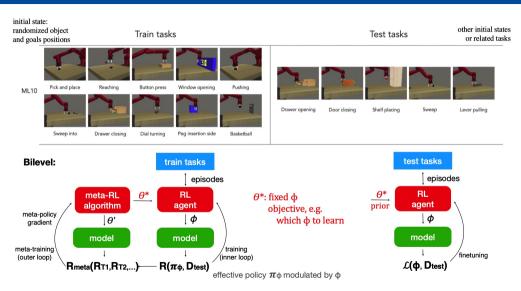


4

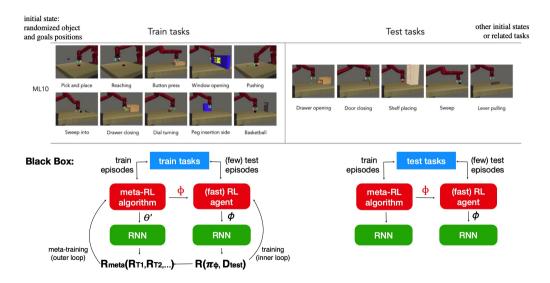
Example: few-shot classification



Example: meta-reinforcement learning

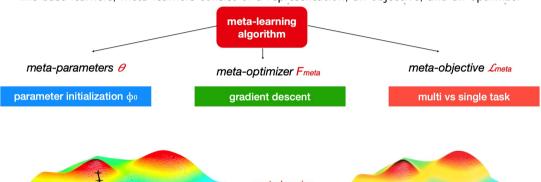


Example: meta-reinforcement learning



Taxonomy of meta-learning methods

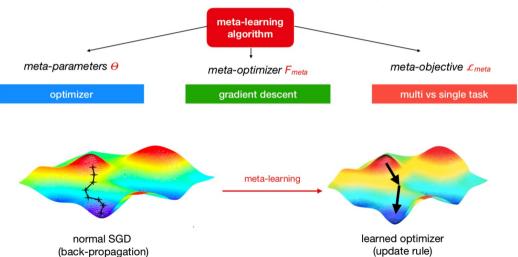
like base-learners, meta-learners consist of a representation, an objective, and an optimizer





Taxonomy of meta-learning methods

like base-learners, meta-learners consist of a representation, an objective, and an optimizer



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