DYNAMIC ALGORITHM CONFIGURATION: Foundation of a New Meta-Algorithmic Framework

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ECAI2020















0 Silver

0 Bronze



Algorithm

Parameter

Instances

Objective

Algorithm Configuration



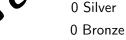
















Algorithm

Parameter

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Objective



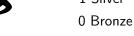
















Objective

Algorithm

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Instances

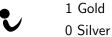
















0 Bronze

Algorithm

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Objective

Image sources given at the end of the presentation $% \left(1\right) =\left(1\right) \left(1\right) \left($









Parameter



1 Gold 1 Silver

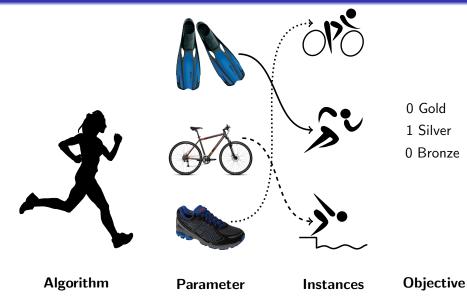


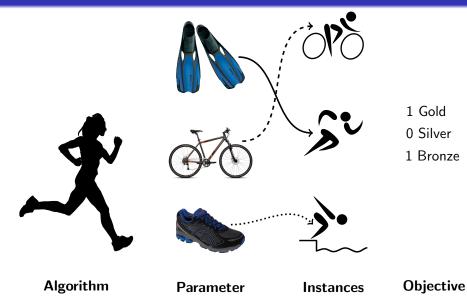


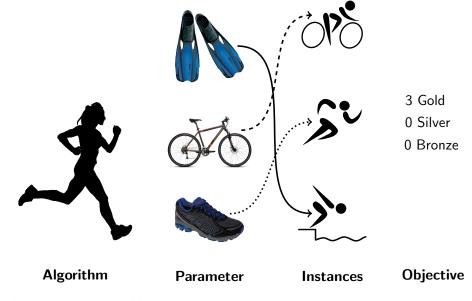
Instances



Per-Instance Algorithm Configuration



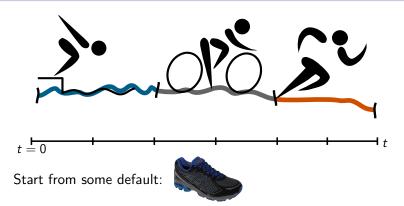


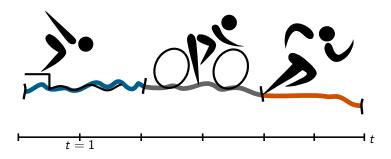




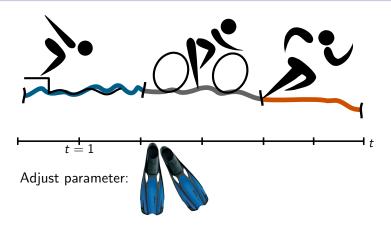
How can we solve such problem instances?

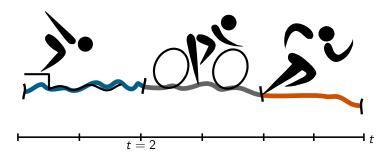
Dynamic Algorithm Configuration



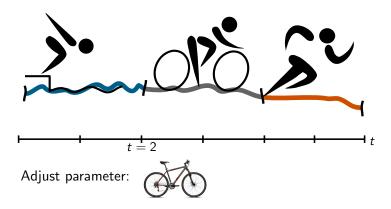


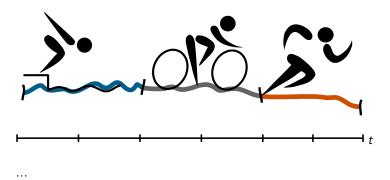
Observe state: Water

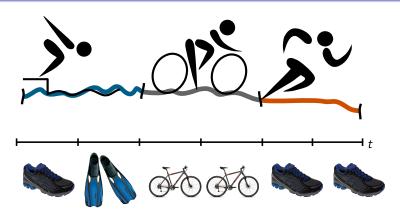




Observe state: Bike Trail







Dynamic Configuration as MDP

Formalize optimization as MDP $\mathcal{M} := (\mathcal{S}, \mathcal{A}, \mathcal{T}, \mathcal{R})$

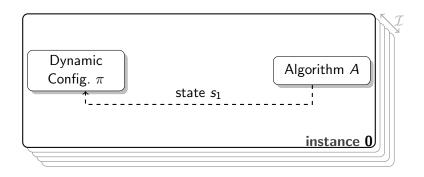
- lacksquare State Space ${\cal S}$
- lacksquare Action Space $\mathcal A$
- \blacksquare Transition Function \mathcal{T}
- Reward Function \mathcal{R}





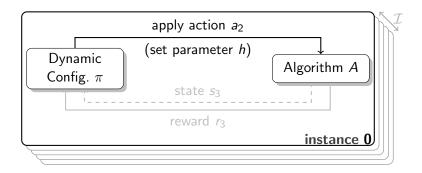
- Instances as context
- Contextual-MDP: $\mathcal{M}_{\mathcal{I}} := \{\mathcal{M}_i\}_{i \sim \mathcal{I}}$
- $M_i := (\mathcal{S}, \mathcal{A}, \mathcal{T}_i, \mathcal{R}_i)$





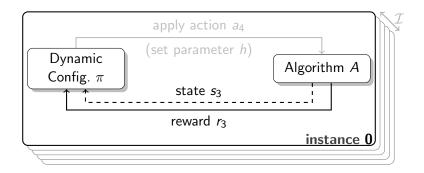






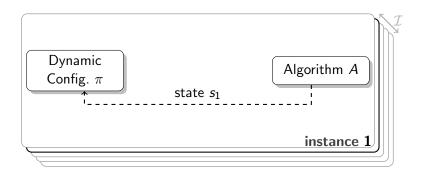






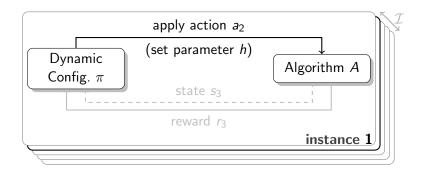






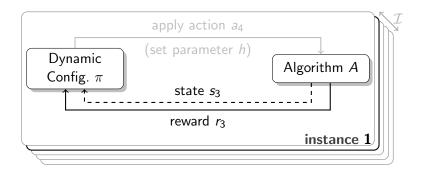






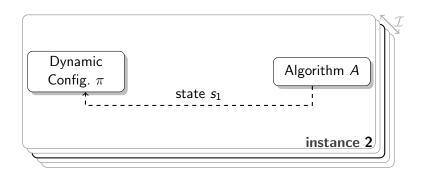






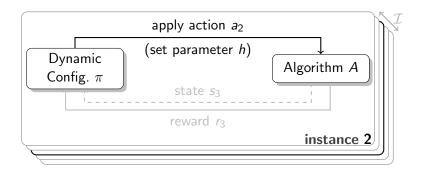






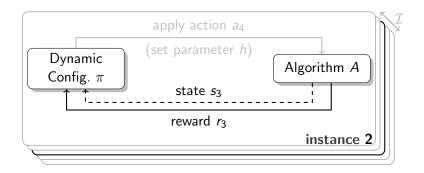
















We designed benchmarks to gain insights into the following settings:

- Effect of Short Effective Sequence Length
- Stochasticity of Reward Signal
- Homogeneity of Instances
- Generalization
- Scaling with the Number of Parameters
- Effect of Self-Paced Learning



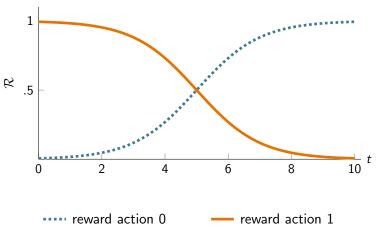


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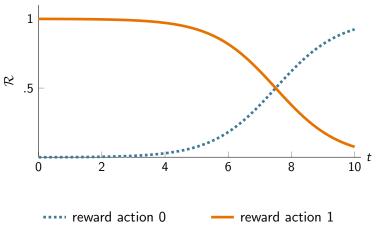






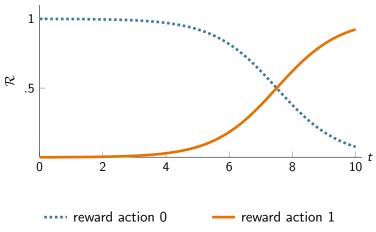


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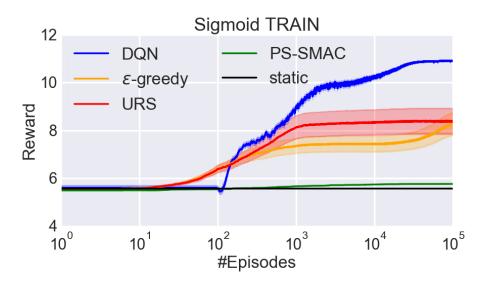
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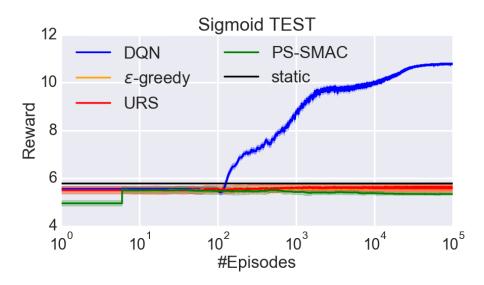


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Sigmoid: Instance sets



Sigmoid: Instance sets



Wrap-up

- We presented a new meta-algorithmic framework
- We demonstrated that the framework is a generalisation of prior frameworks
- Finally we demonstrated the effectiveness of the framework
 - Effect of Short Effective Sequence Length
 - Stochasticity of Reward Signal
 - Homogeneity of Instances
 - Generalization
 - Scaling with the Number of Parameters
 - Effect of Self-Paced Learning





Image Sources

- Female Runner (by algotruneman under CC0 1.0)
- Running Shoe (under CC BY-NC 4.0)
- Bike (under CC BY-NC 4.0)
- Flippers (under CC BY-NC 4.0)
- Cycling Pictogram (by Parutakupiu & Thadius856 under public domain)
- Athletics Pictogram (by Parutakupiu & Thadius856 under public domain)
- Swimming Pictogram (by Parutakupiu & Thadius856 under public domain)
- The above images were flipped/rotated or scaled and used to depict abstract representations of Algorithm Configuration



