

AI and Control

Opportunities to tame usage of resources

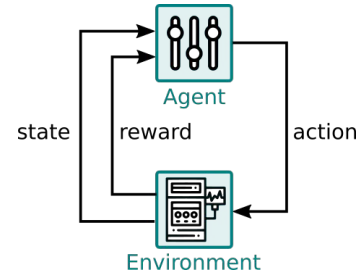
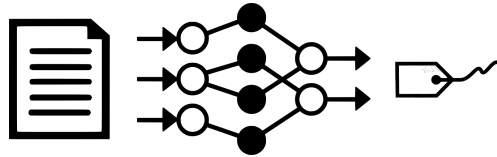
Sophie Cerf

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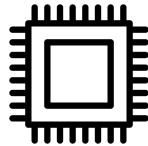


My vision of AI

- Used to make predictions & take actions

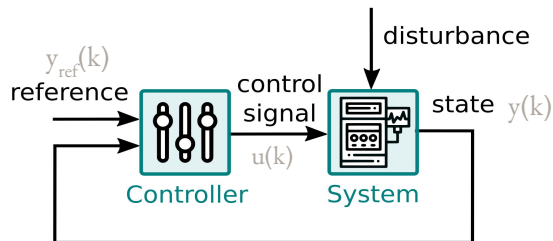
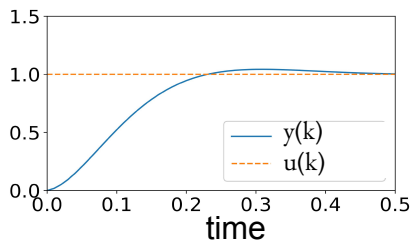


- Limits: data intensive, use of resources



Control Theory

*Field that study and control **dynamical** systems*

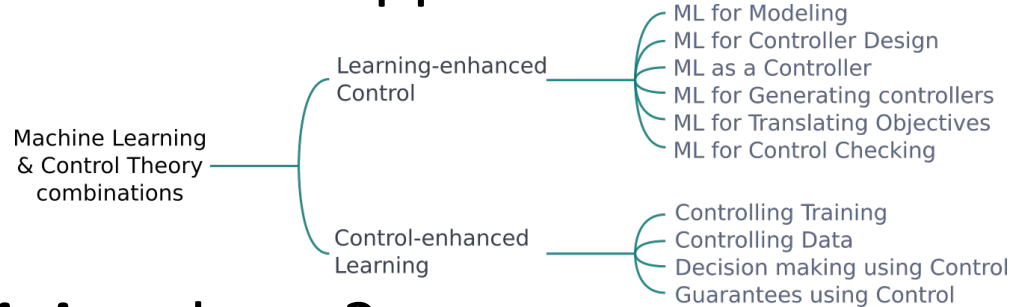


- System's model
 $y(k+1) = ay(k) + bu(k)$
- Controller
 $u(k) = K[y_{ref}(k) - y(k)]$

- Objectives among stabilisation, tracking, optimization, etc.
- Some interesting characteristics
 - the Feedback principle
 - Guaranteed behavior
 - Model-based approach

AI \Leftrightarrow Control

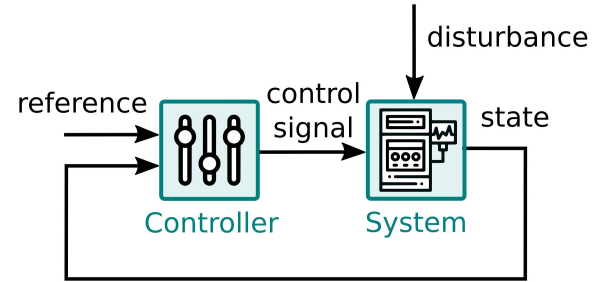
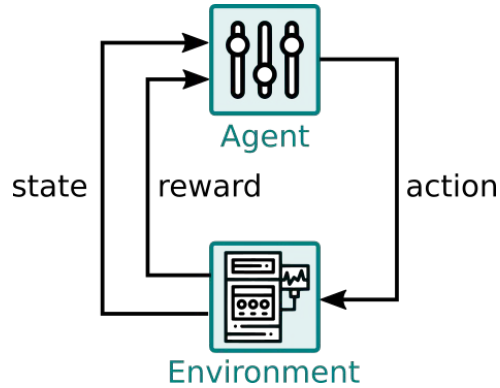
- Data vs. model-based approaches



- Why **combining** them ?

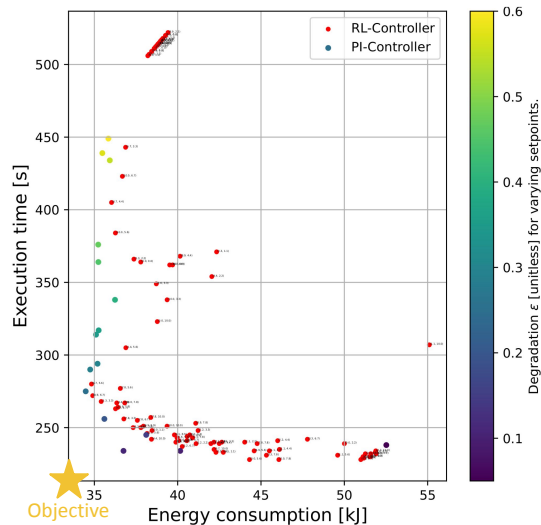
- Performance
- Safety
- Data efficiency
- Frugality

Using Control instead of AI



Using Control instead of AI

Similar performance, with significantly different inference cost

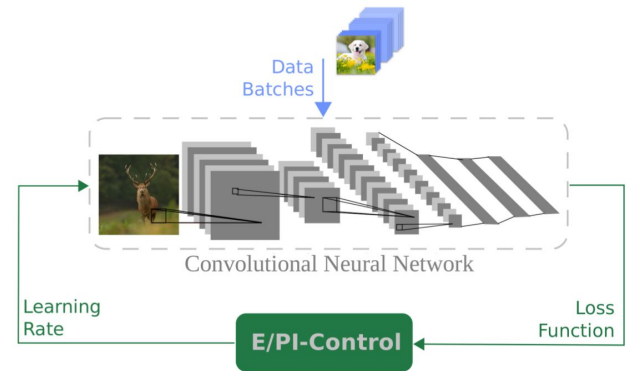
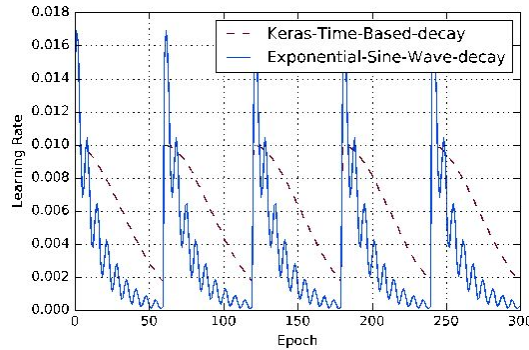


- Control: PI controller
 - 2 parameters
- $$\mathbf{u}(t_i) = (K_I \Delta t_i + K_P) \cdot e(t_i) - K_P \cdot e(t_{i-1}) + \mathbf{u}(t_{i-1})$$
- AI: Reinforcement learning PPO
 - 8100 parameters

Sophie Cerf, Raphaël Bleuse, Valentin Reis, Swann Perarnau, Eric Rutten. Sustaining Performance While Reducing Energy Consumption: A Control Theory Approach. EURO-PAR 2021 - 27th International European Conference on Parallel and Distributed Computing, Aug 2021, Lisbon, Portugal. pp.334-349, ([10.1007/978-3-030-85665-6_21](https://doi.org/10.1007/978-3-030-85665-6_21)). ([hal-03259316](https://hal.archives-ouvertes.fr/hal-03259316))
A. Raj, S. Perarnau and A. Gokhale, "A Reinforcement Learning Approach for Performance-aware Reduction in Power Consumption of Data Center Compute Nodes," 2023 IEEE International Conference on Cloud Engineering (IC2E), Boston, MA, USA, 2023, pp. 121-130, doi: [10.1109/IC2E59103.2023.00022](https://doi.org/10.1109/IC2E59103.2023.00022).

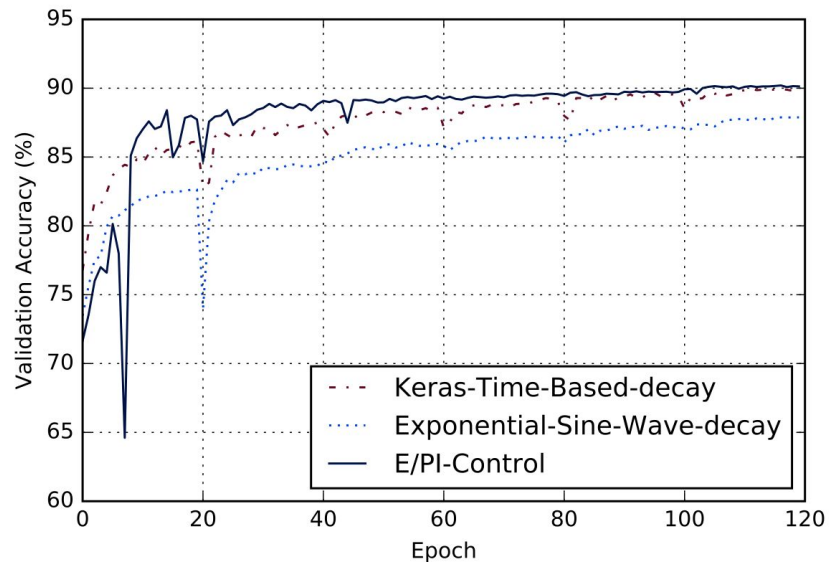
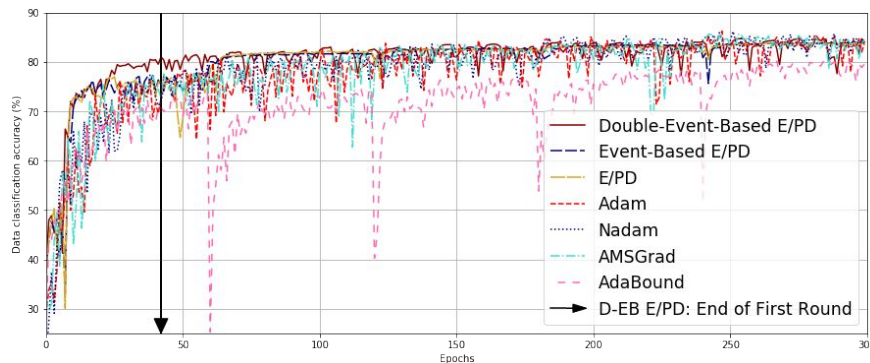
Using Feedback in Training

- Shift from *time-based* adaptation to *feedback*
 - Exploration vs. exploitation trade-off
 - **Training rate** evolution law



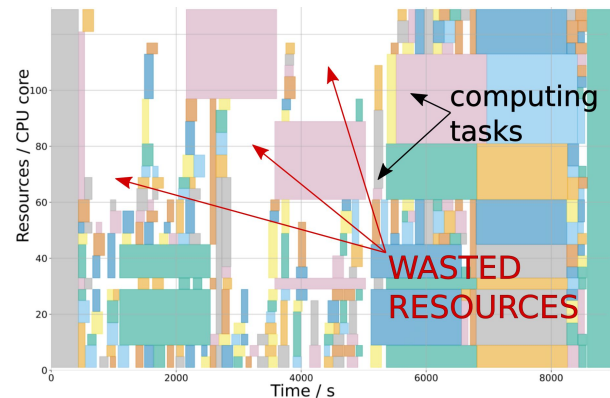
Using Feedback in Training

- Faster convergence
 - save up to 67% training time
- Less training
 - switch to the next batch based on the learning speed



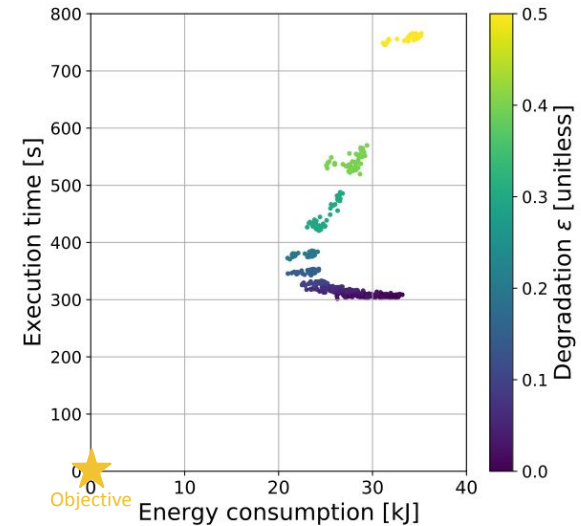
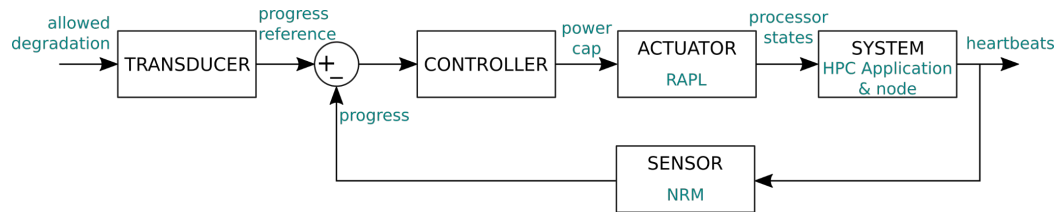
AI as High Performance Computing

- Tasks in shared computing centers
 - scheduling
- Management of **unused** resources
 - inject small tasks to fill the cluster *without impacting the main tasks*
- Open direction
 - federated learning



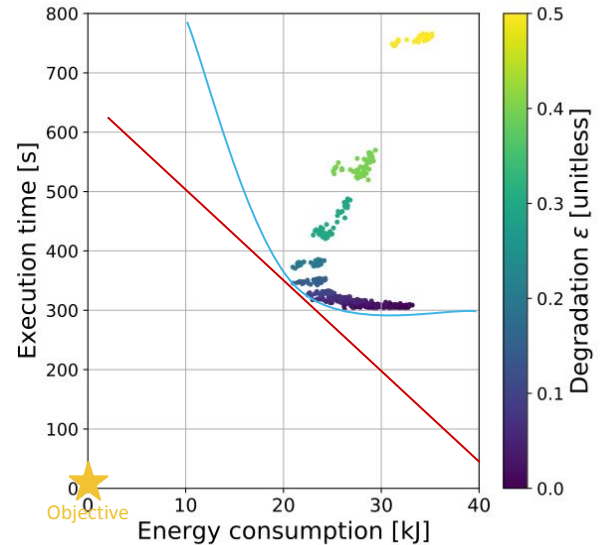
From Efficiency to Sufficiency

- Reduce CPU **energy** usage in memory-intensive phases
- Acceptable **performance degradation** as a design objective



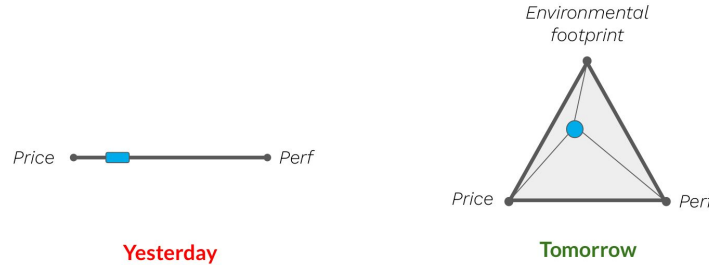
AI shrinkability

- Can algorithms handle low resource conditions ?
- Which tunable configurations at runtime ?
 - Software: models
 - Hardware: architectures

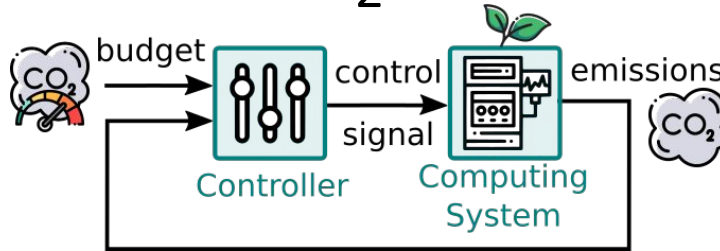


AI within limits

- Environmental footprint as a design objective



- Means to ensure a CO₂ budget



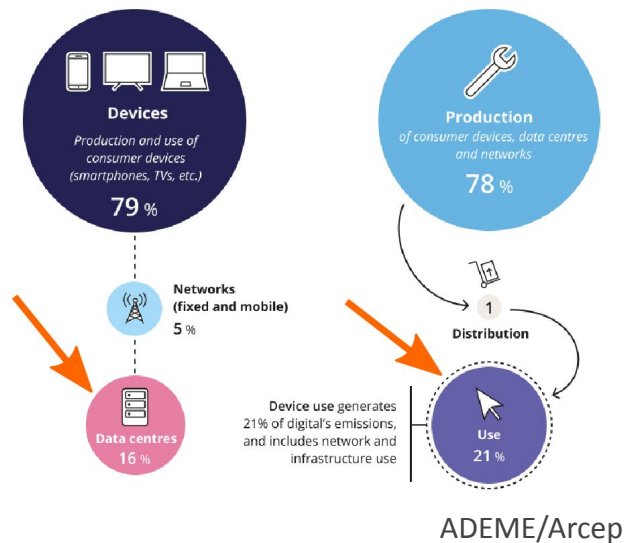
Conclusion

- **AI and Control:** Opportunities to tame usage of resources
- **Limitations**
 - focus on usage phase, datacenter, climate change
 - Rebound effect

Devices and their production account for the overwhelming majority of the digital carbon footprint

Breakdown of the digital carbon footprint in 2020 by ICT component (%)

Breakdown of the digital carbon footprint in 2020 by life cycle stage (%)



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Thank You

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