

Deep Reinforcement Learning-Based Control in Industrial Settings

Lleida-Tech

Agrobiotech Park, Gardeny Hill

October 25, 2024



NURIA NIEVAS VIÑALS

Industrial PhD Candidate

AI & BD Researcher

Applied Artificial Intelligence unit

Eurecat, Centre Tecnològic de Catalunya

nuria.nievas@eurecat.org





Introduction

Deep Reinforcement Learning-Based Control in Industrial Settings

Static environments



- Stability predictability
- Limited complexity
- Traditional control

Dynamic environments



- Technological advancements
- Increased data availability
- Market demands
- Dynamic control



Benefits of dynamic control in industry

- + Adaptability
- + Efficiency
- + Performance



Slower adoption in traditional industries

- × Integration complexity
- × Safety and reliability
- × Costs



Press Hardening Objectives

Deep Reinforcement Learning-Based Control in Industrial Settings

Hot Metal Forming - Press Hardening (Hot Stamping)

- Thermo-mechanical sheet deformation-based process
- Complex shapes and high strength
- Automotive industry



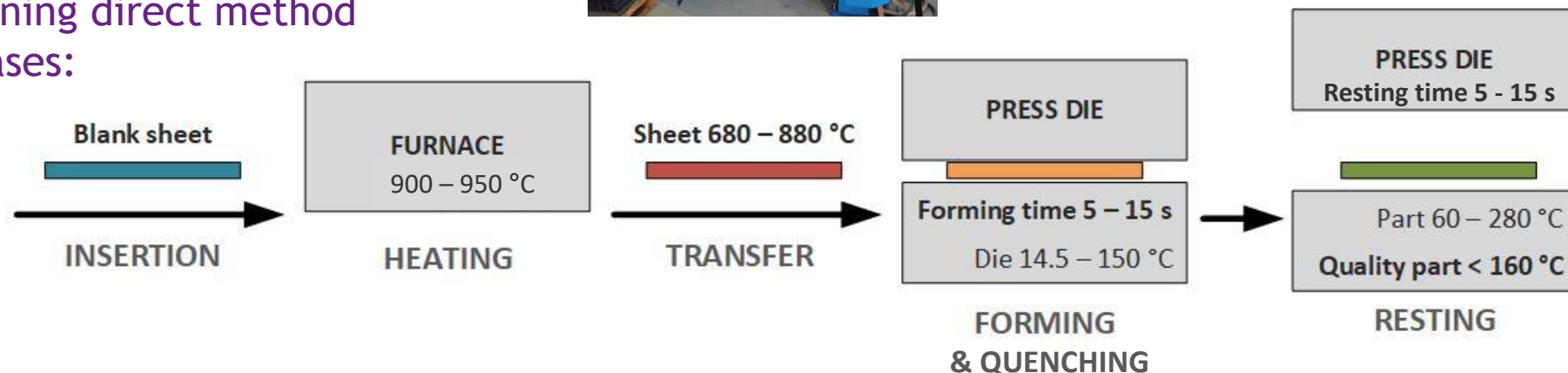
Control parameters

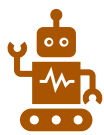
- Forming time: for each cycle
- Resting time: for each cycle

Objective of dynamic control

- Improve operational efficiency: Optimize the total time of a batch production
- Guarantee quality: parts $< 160^{\circ}\text{C}$

Press hardening direct method
process phases:

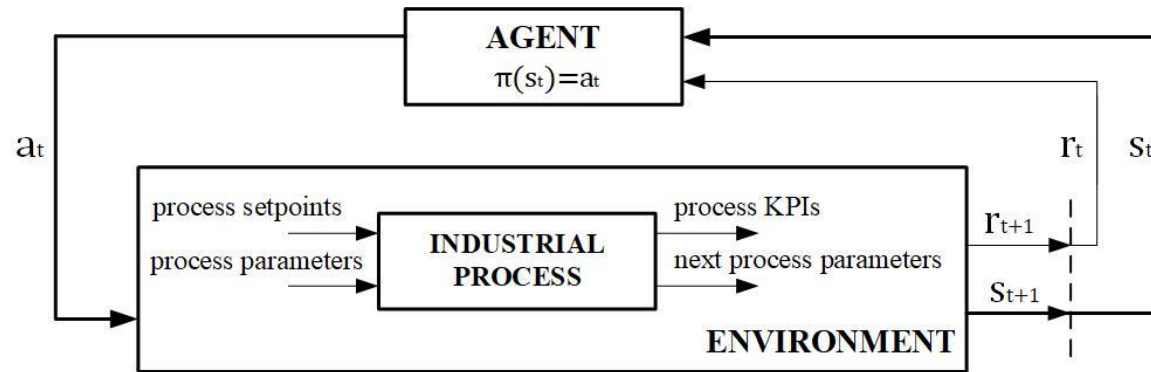




Reinforcement Learning Control

Deep Reinforcement Learning-Based Control in Industrial Settings

What is Reinforcement Learning?



- Continuous control
- Autonomous learning
- Long-term decision making
- Handle uncertainty

Challenges of Reinforcement Learning in the industry

- Slow training
- Exploration limited
- Safety constraints
- Cost of failures and non-optimal actions

Challenges of RL
in the Industry

Problem modeling

Environment for training

Training and validation

Deployment



Surrogate Model-Driven RL Training

Deep Reinforcement Learning-Based Control in Industrial Settings

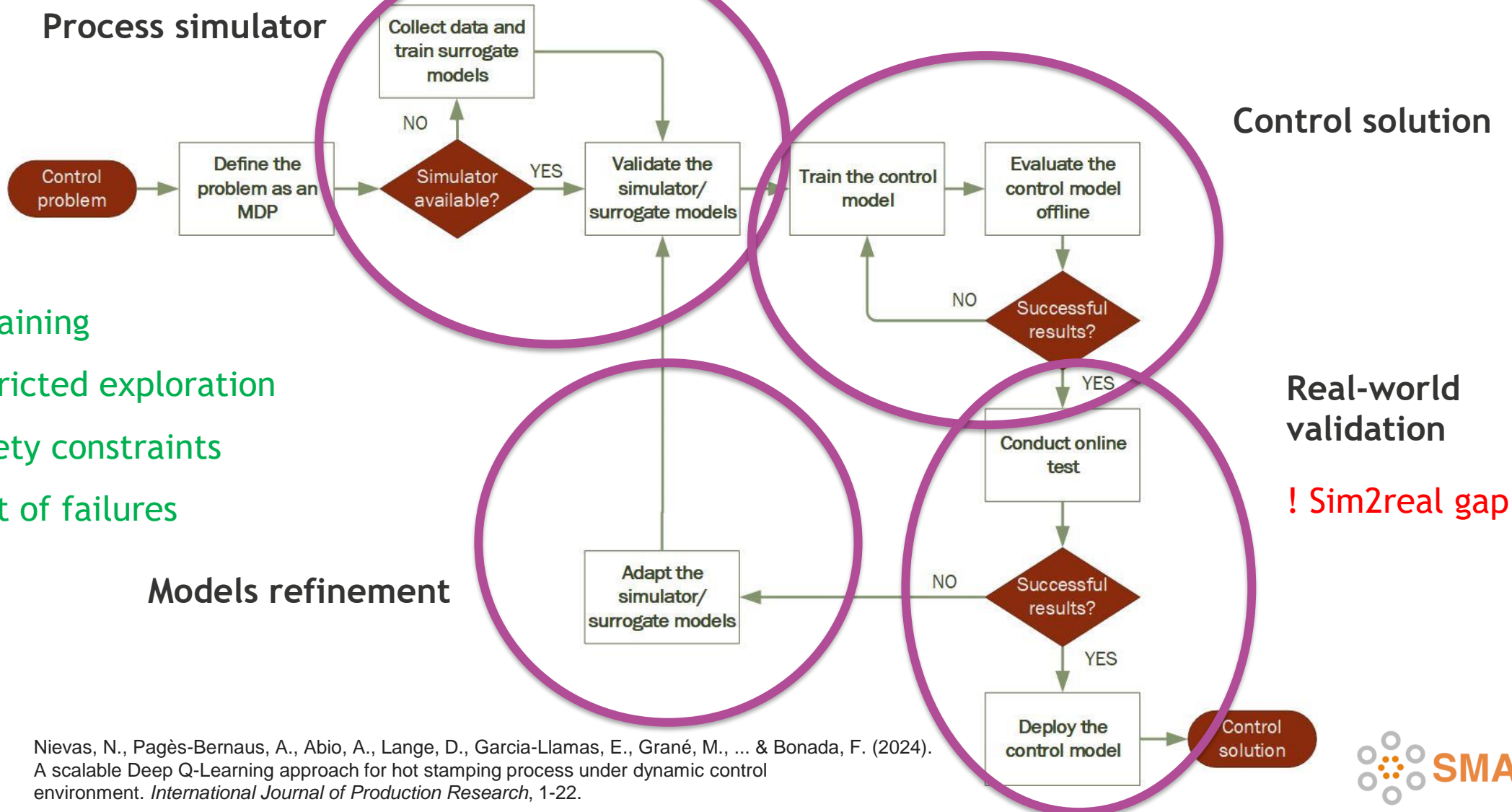


ARTIFICIAL INTELLIGENCE & OPTIMIZATION CONGRESS



Universitat de Lleida

Process simulator



- ✓ Fast training
- ✓ Unrestricted exploration
- ✓ No safety constraints
- ✓ No cost of failures



Static vs. Dynamic Control

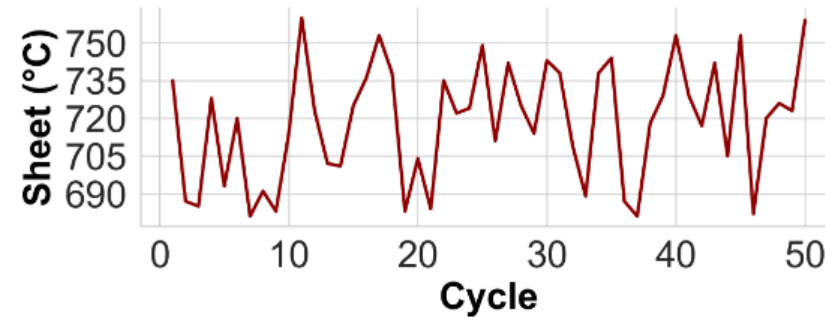
Deep Reinforcement Learning-Based Control in Industrial Settings

Results of training with a surrogate model of the hot stamping process

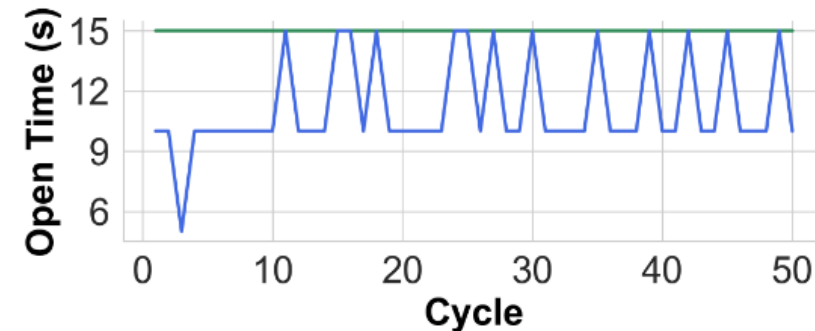
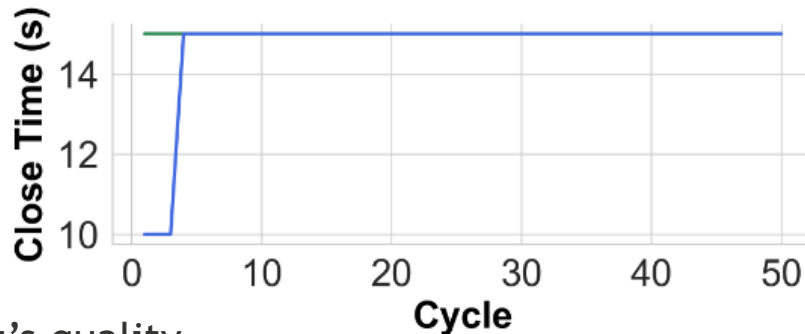
- Dynamics of the environment

Production of a batch of 50 parts comparing static versus dynamic approach

— Sheet (°C)
— Static
— Dynamic
- - - 160 (°C)

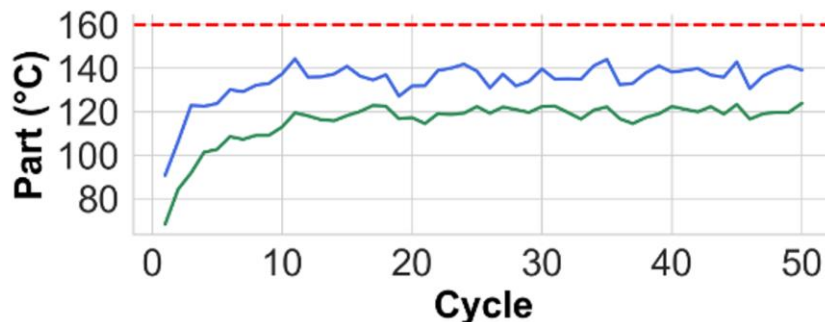


- Controlled variables: forming and resting times



The dynamic approach reduces times

- Final part's quality



The dynamic approach maintains quality

Thank you!



NURIA NIEVAS VIÑALS
Industrial PhD Candidate
AI & BD Researcher
Applied Artificial Intelligence unit
Eurecat, Centre Tecnològic de Catalunya
nuria.nievas@eurecat.org

