

Intelligent Dependable Autonomous Systems

DFKI Research Fellow Talk

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smartFactory^{KL}

IFS Innovative
Fabriksysteme

WS **KL** Werkzeugmaschinen
und Steuerungen
TU KAISERSLAUTERN



Production
Level 4

Overview

- Motivation
- Trustworthy AI and Dependability
- Dependability of Dynamical Systems
- Production Scenarios
 - Production Level 4
 - Multi-Agent AI Systems and Behavior Decomposition
 - Collaborative Robotics
 - Fault-Tolerant Control
 - Hybrid Modelling
- Conclusion
- Future Research

Motivation: Development of Safety-Critical Applications



Source: Prevent

Automotive

complex
distributed
dynamical
collaborative
autonomous

Production



Source: <http://www.huffingtonpost.de>



Source: Boeing Dreamliner

Avionics

Medical
Technology



Source: Klinikum Jena

Trustworthy AI and Dependability

Ethical Principles

- (i) Respect for human autonomy
- (ii) Prevention of harm
- (iii) Fairness
- (iv) Explicability

Technical Requirements

- Technical robustness and safety
- Resilience to attack and security
- Fallback plan and general safety
- Accuracy
- Reliability and Reproducibility

Dependability

European Commission, Directorate-General for Communications Networks,
Content and Technology, *Ethics guidelines for trustworthy AI*, Publications
Office, 2019, <https://data.europa.eu/doi/10.2759/177365>

Dependability of Dynamical Systems - Linguistic

What is Dependability?

- [Carter, 1982]: A system is dependable if it is trustworthy enough that reliance can be placed on the service it delivers
- [Laprie, 1992]: Dependability is that property of a computing system which allows reliance to be justifiably placed on the service it delivers
- [Dubrova, 2013]: Dependability is the ability of a system to deliver its intended level of service to its users

Dynamical (autonomous mobile robot) Systems

- **[Rüdiger, 2007]: Dependability in general is the capability of a system to successfully and safely fulfil its mission.**

[Carter, 1982] Carter, W.: A time for reflection. In Proc. 12th Int. Symp. on Fault Tolerant Computing (FTCS-12). IEEE Computer Society Press Santa Monica

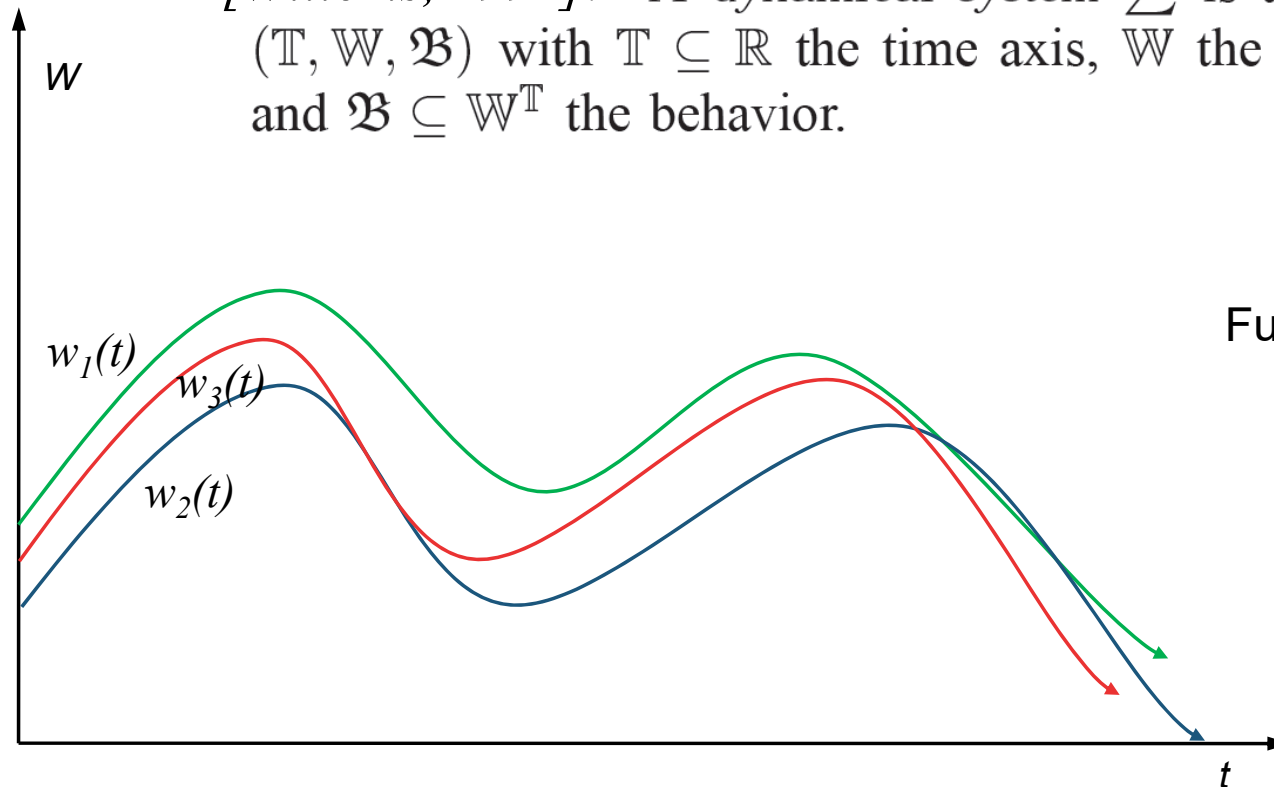
[Laprie, 1992] Laprie, J. C.: Dependability. Basic Concepts and Terminology. Ed. Springer Verlag.

[Dubrova, 2013] Dubrova, E.: Fault tolerant design: An introduction. Springer, DOI 10.1007/978-1-4614-2113-9

[Rüdiger, 2007] Rüdiger, J., Wagner, A., and Badreddin, E.: Behavior based definition of dependability for autonomous mobile systems. In European Control Conference 2007 (July 2-5), pp. 4146–4151.

Dynamical System – Behavior Description

[Willems, 1991]: A dynamical system Σ is a triple $\Sigma = (\mathbb{T}, \mathbb{W}, \mathfrak{B})$ with $\mathbb{T} \subseteq \mathbb{R}$ the time axis, \mathbb{W} the signal space, and $\mathfrak{B} \subseteq \mathbb{W}^{\mathbb{T}}$ the behavior.



Functional representation
state space

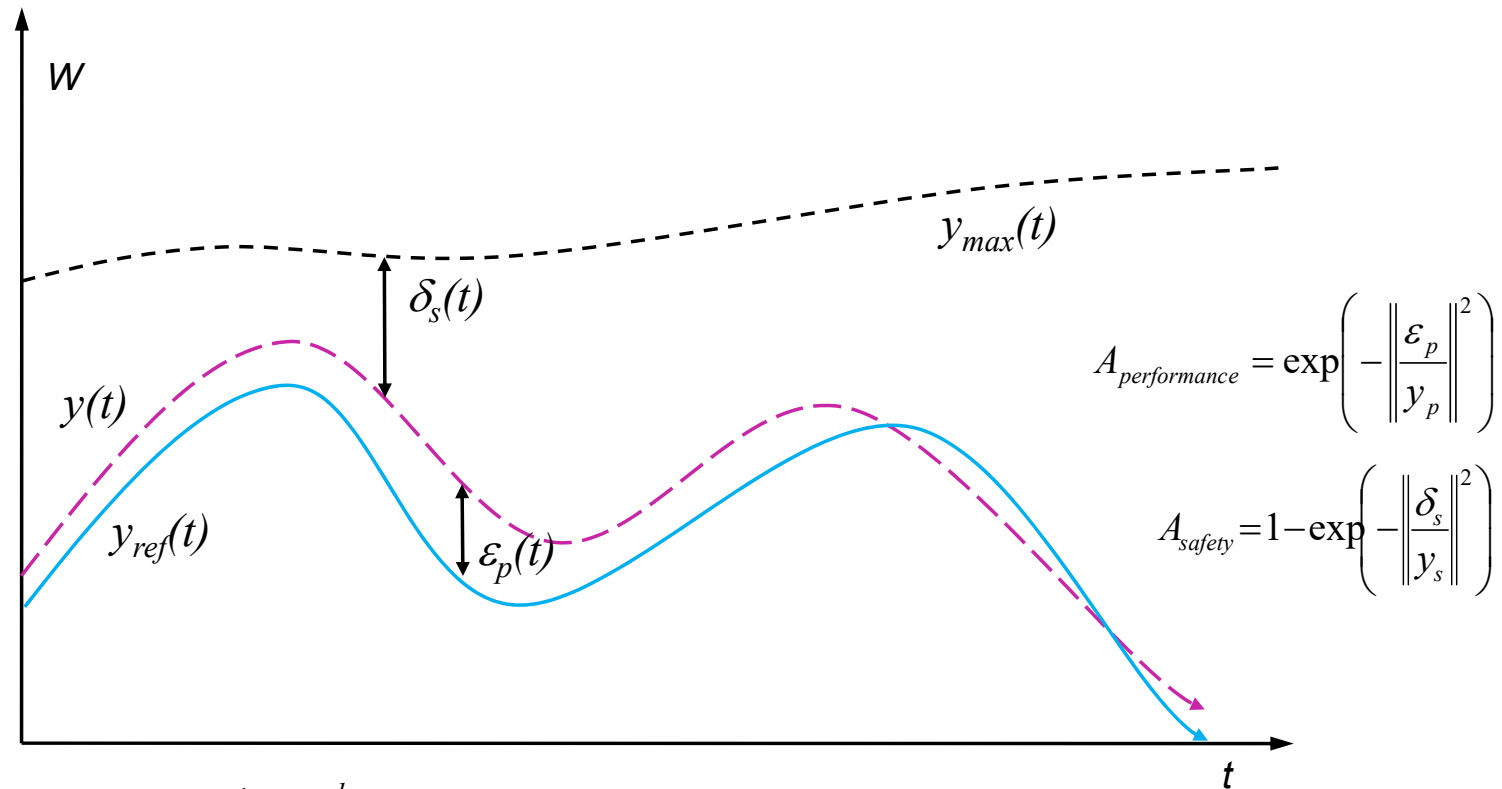
$$\dot{x}(t) = f(x(t), u(t))$$

$$y(t) = g(x(t), u(t))$$

$$x(t_0) = x_0$$

J.C. Willems: Paradigms and puzzles in the theory of dynamical systems. *Automatic Control, IEEE Transactions on*, 36(3):259-294, March 1991

Dependability of Dynamical Systems - Formal

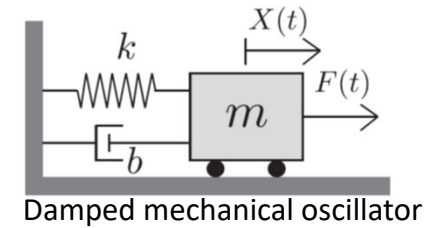
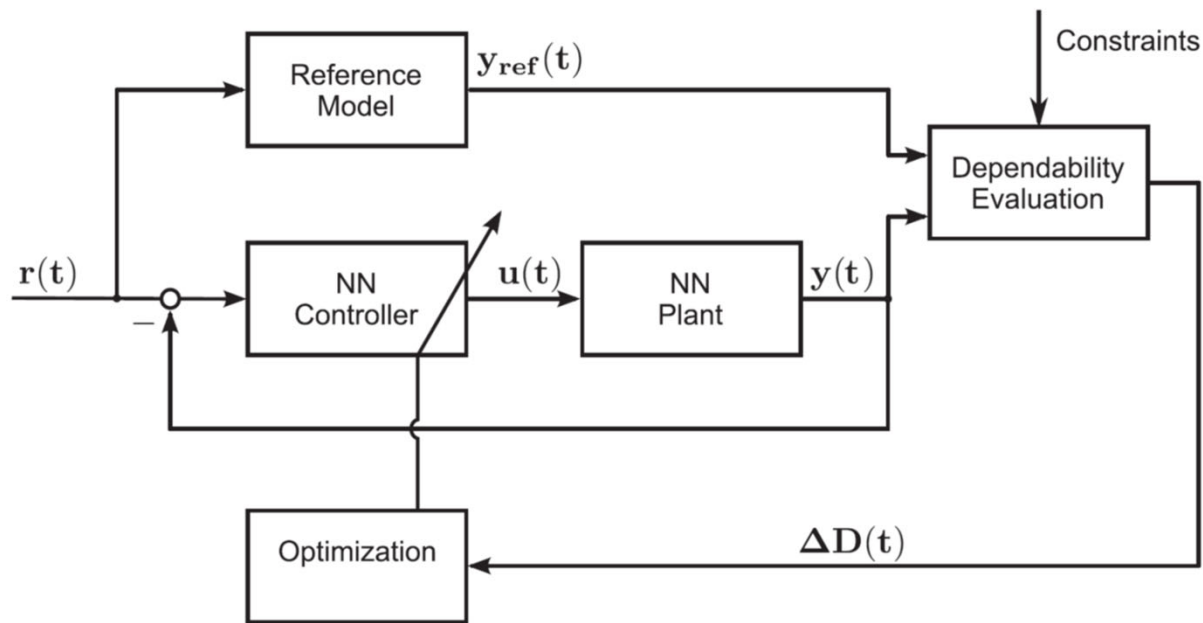


Dependability Measure

$$D(t) = 1 - \frac{1}{m} \int_0^t \sum_{j=1}^d a_j \left[1 - A_j(u(t), y_r(t), y(t), \Theta_j) \right] dt$$

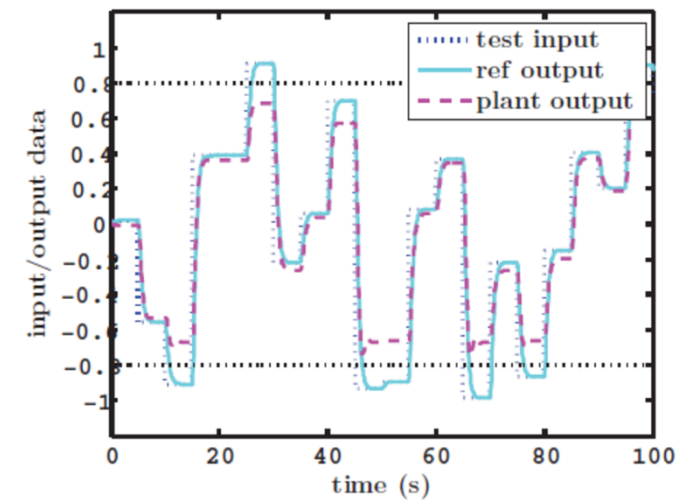
Wagner (2018): : Modeling, Analysis, and Design of Dependable Systems with Application to Robotics and Assistance Technology. Verlag Dr. Hut 2018, ISBN 978-3-8439-3559-3.

Dependability Optimization: Robot Control



$$y_p = 0.2 \quad y_s = 0.08, \quad y_{max} = 0.8, \quad y_{min} = -0.8,$$

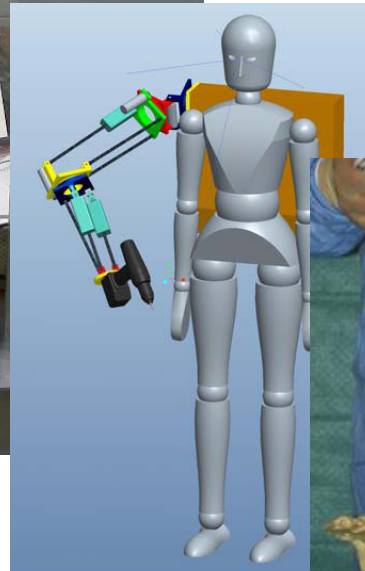
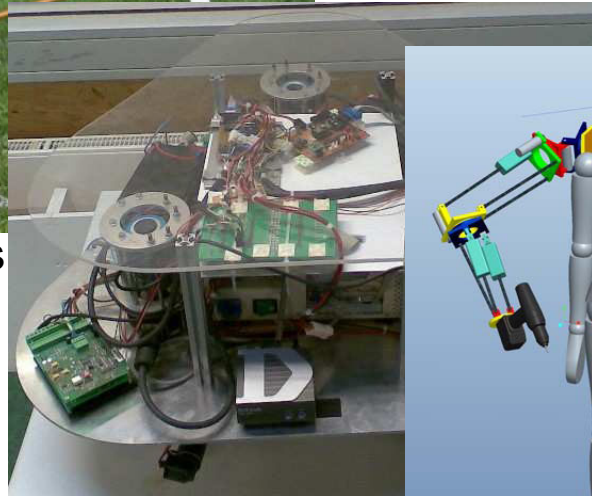
$$a_p = 0.4, \quad a_{s+} = a_{s-} = 0.3$$



Dependable Robotics Research Group (Univ. Heidelberg): Autonomous Mobile Robots and Assistance Systems



Mobile Robots



Bionic Exoskeletons

Rehabilitation
Systems



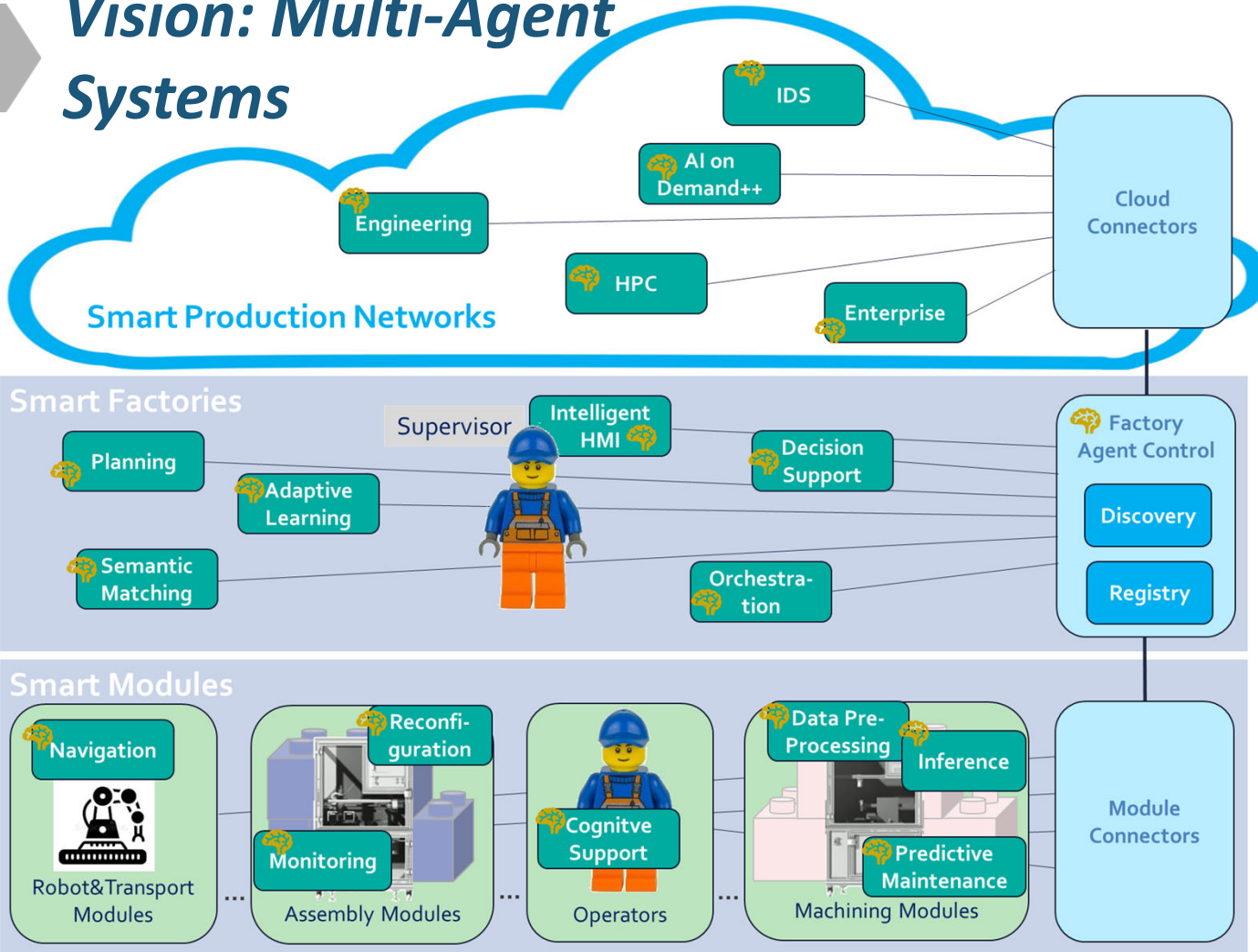
Surgical Robots

Production Level 4 (Autonomous Human-Centered Shared Production)



Source: SmartFactory-KL

Vision: Multi-Agent Systems



MAS4AI

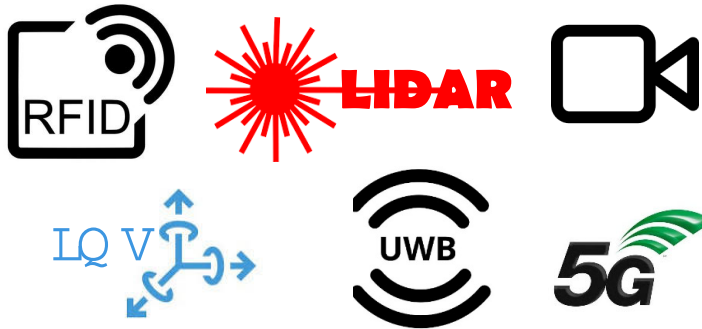
- **Multi-agent AI systems**

- Diverse AI technologies and combinations: knowledge-based, hierarchical planning and model-based machine learning
- AI components in different hierarchy layers
- Unified, reusable and scalable solutions

- **Autonomous modular production**

- Semantic Self-description capabilities of technical modules, optimization of configurations and processes
- Human assistance with awareness of human capabilities
- Highly reactive and safe solutions

Information Structure



Sensor Data

Information



Actions

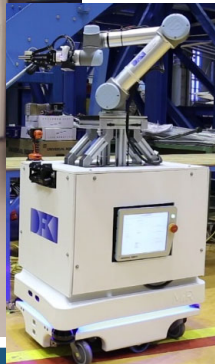
Knowledge



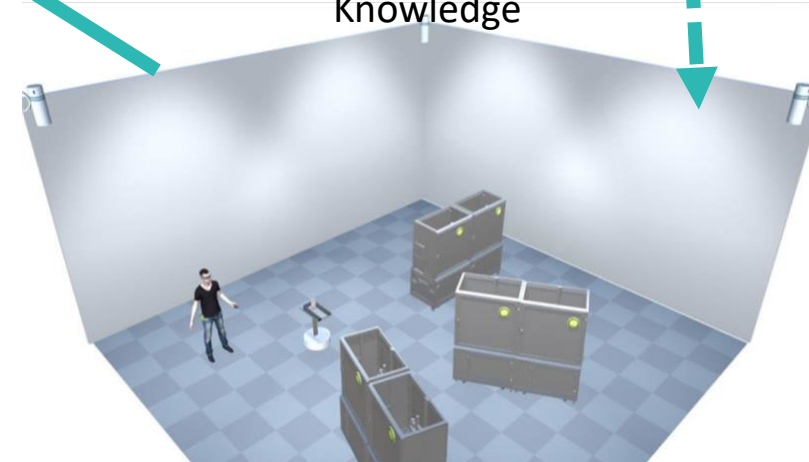
Smart Maintenance



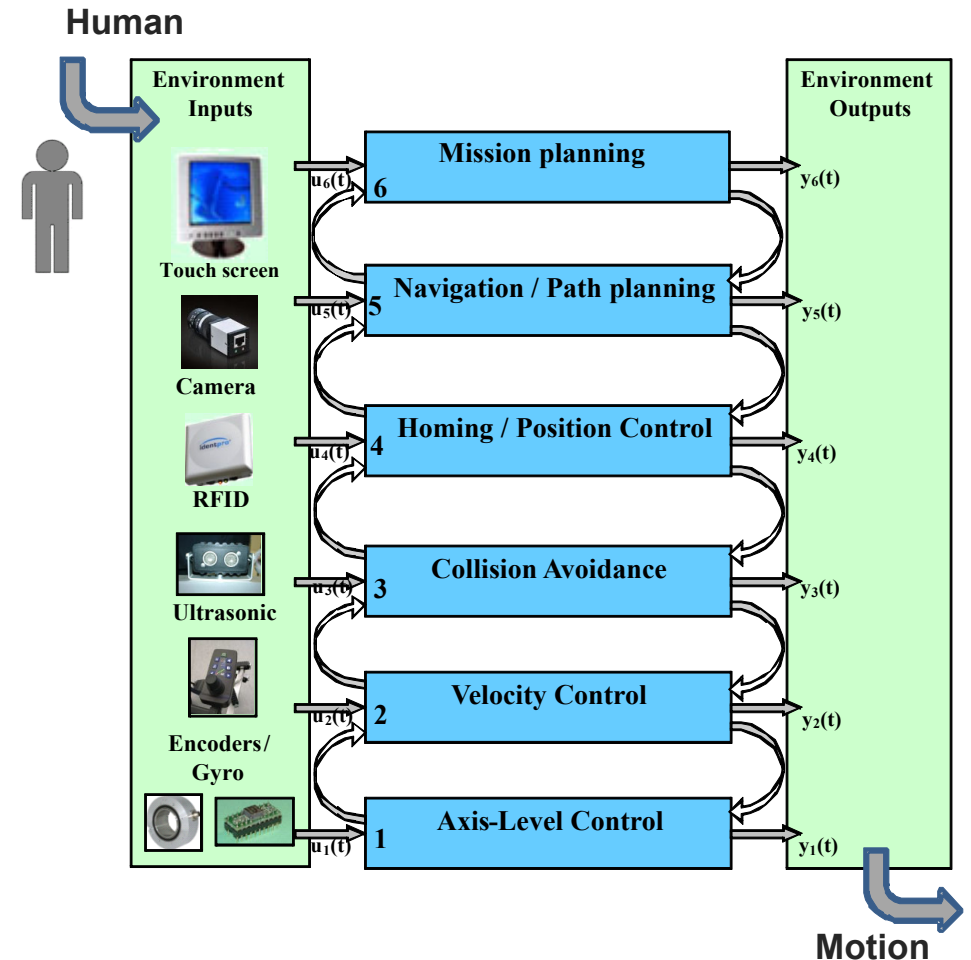
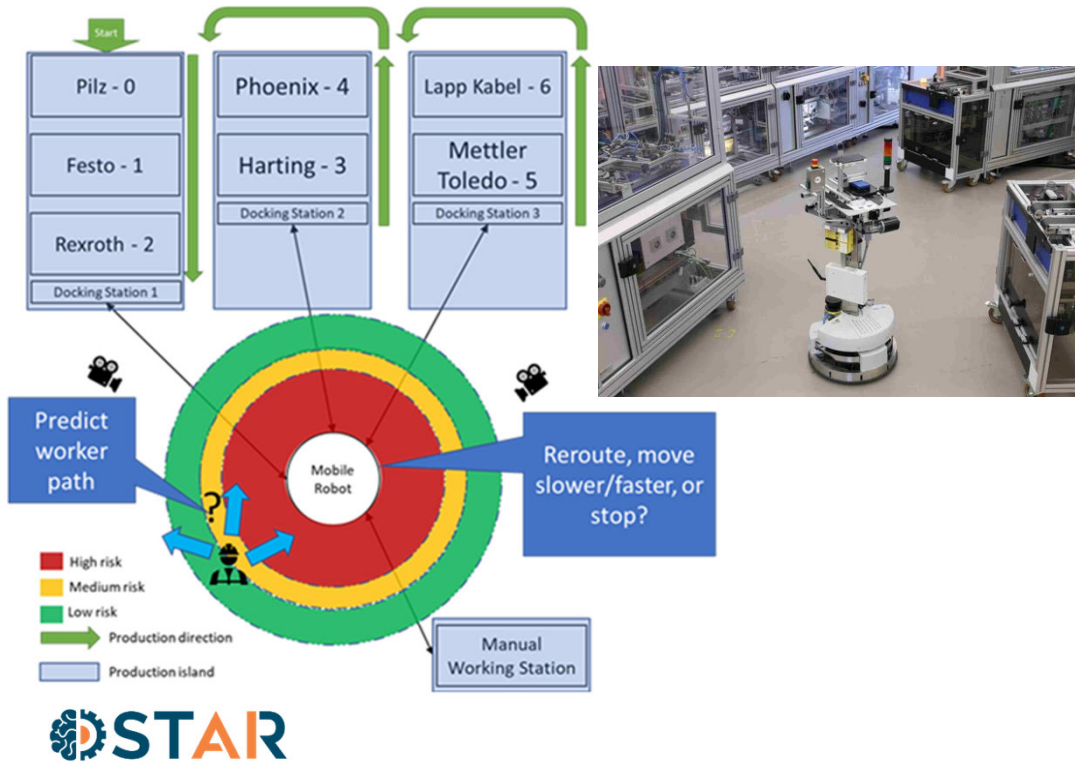
Smart Assembly



Robot Assistance

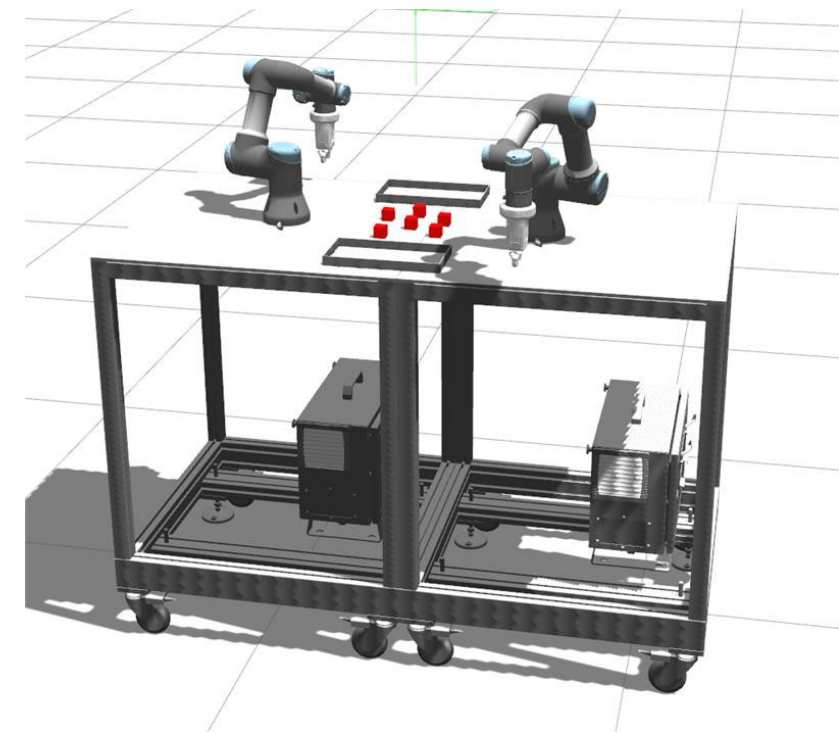
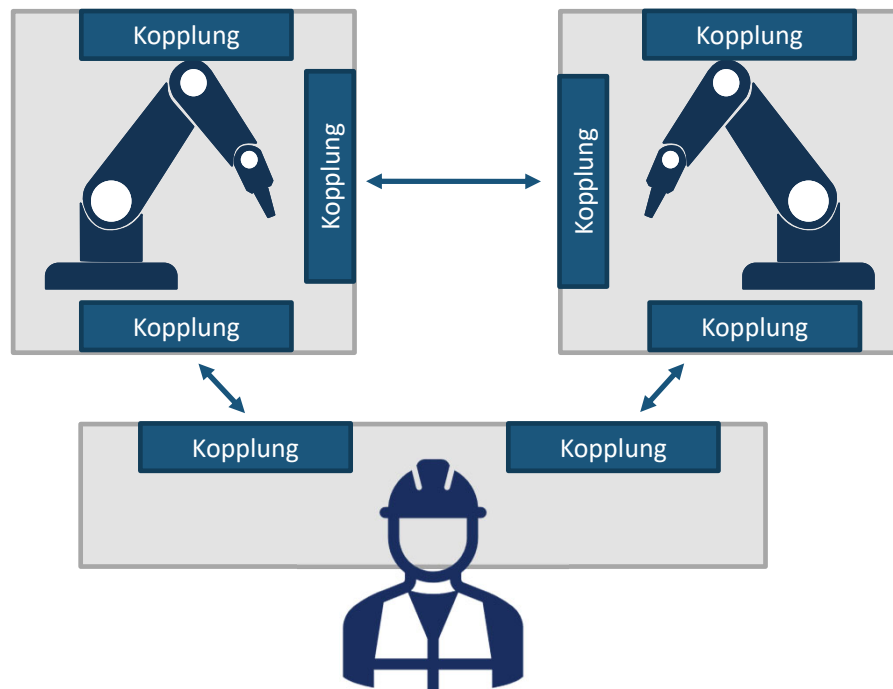


Behavior Decomposition for Autonomous Mobile Robot



RNBC (Recursive Nested Behavior-based Control) Structure
 Badreddin1989; Bartolein2007; Wagner2010; Wagner2016

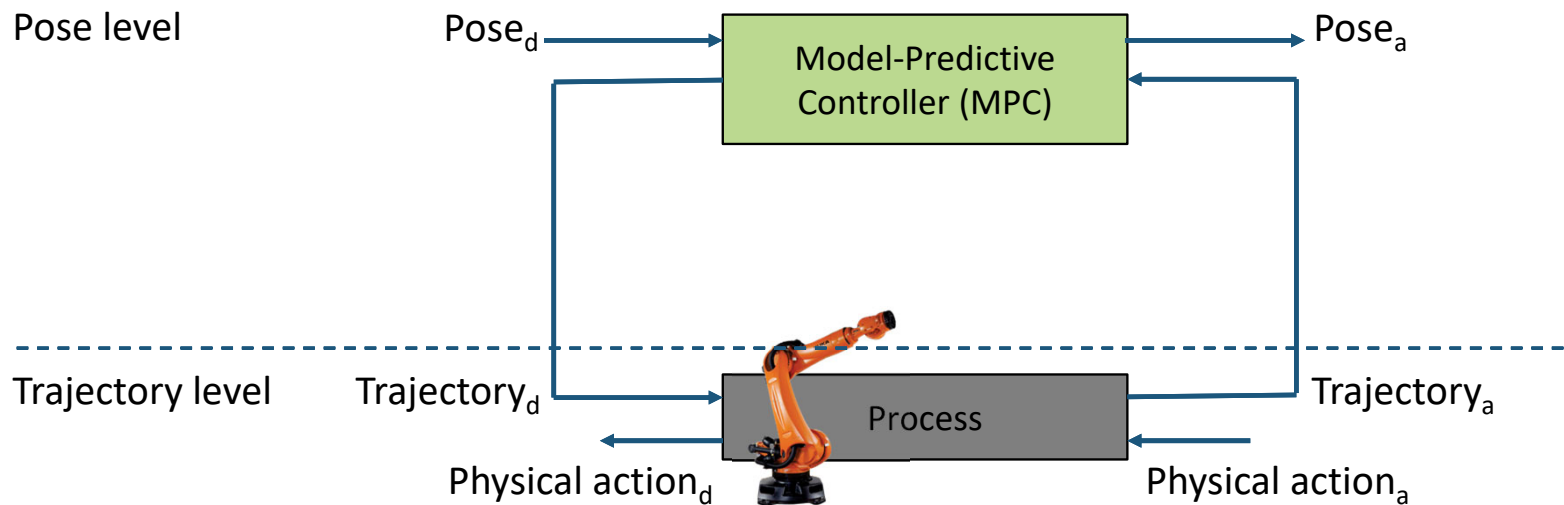
Collaborative Robots



Nigora Gafur et al., 2021

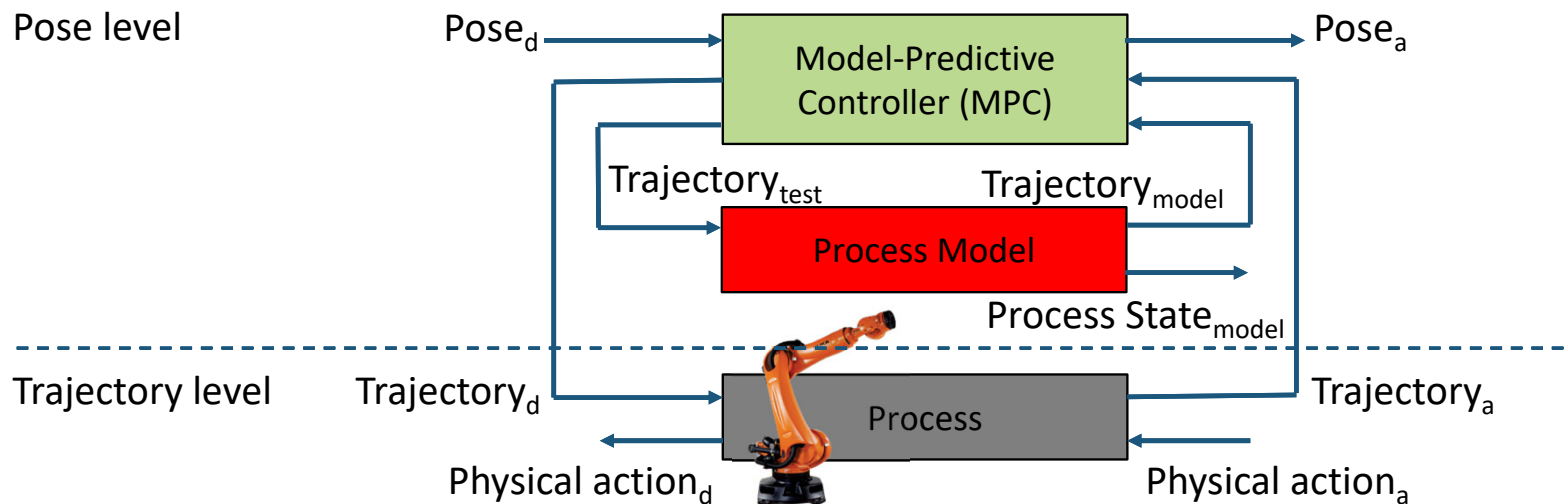
Planning und Model-Predictive Control

$$J_i(\mathbf{x}_i^k, \mathbf{u}_i^k) := (\mathbf{x}_i^{N_p} - \mathbf{x}_i^f)^T \mathbf{Q}_i^f (\mathbf{x}_i^{N_p} - \mathbf{x}_i^f) + \sum_{k=0}^{N_p-1} ((\mathbf{x}_i^k - \mathbf{x}_i^f)^T \mathbf{Q}_i^x (\mathbf{x}_i^k - \mathbf{x}_i^f) + \mathbf{u}_i^k{}^T \mathbf{R}_i^u \mathbf{u}_i^k + (\mathbf{u}_i^{k+1} - \mathbf{u}_i^k)^T \mathbf{R}_i^d (\mathbf{u}_i^{k+1} - \mathbf{u}_i^k))$$

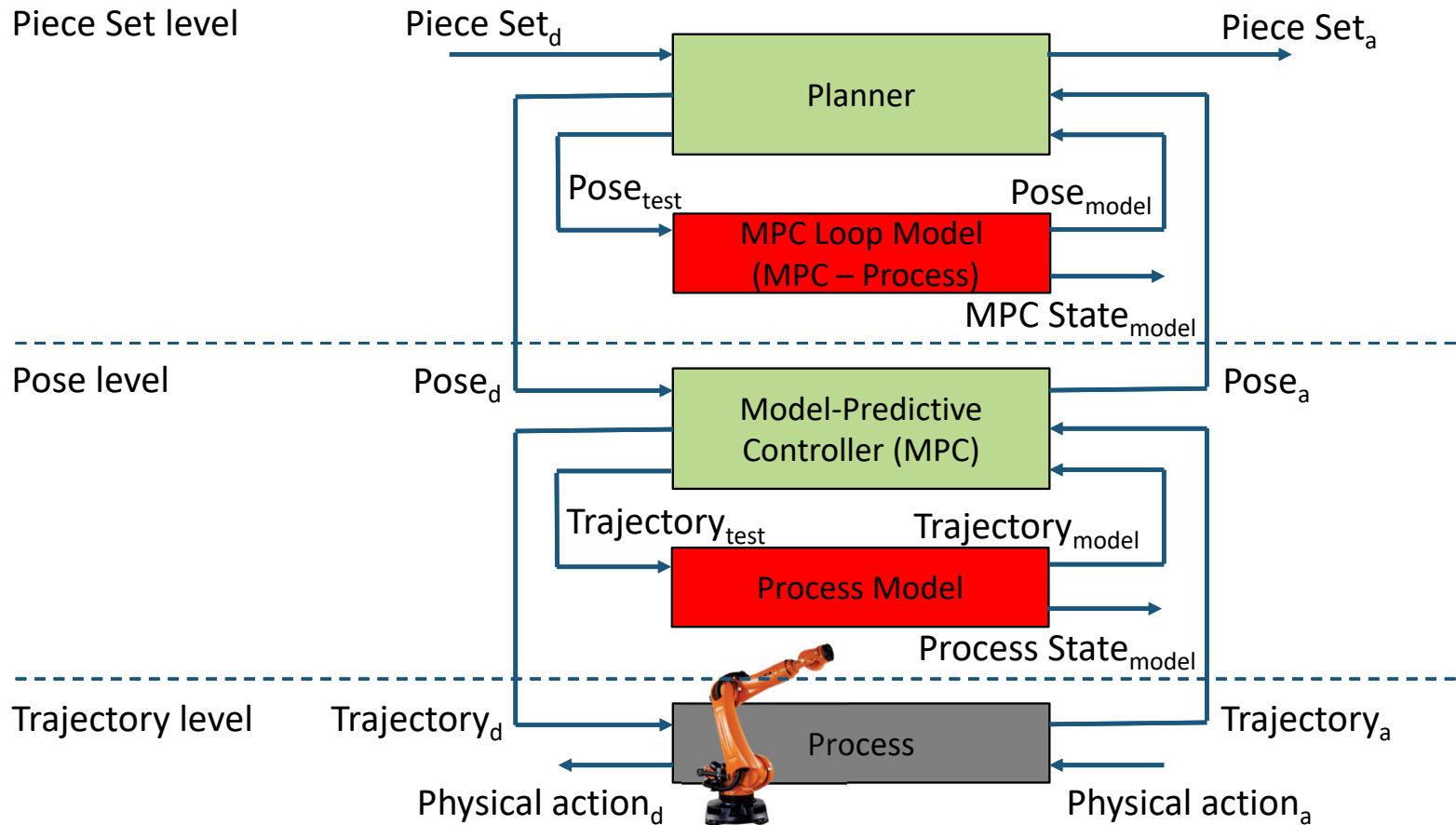


Planning und Model-Predictive Control

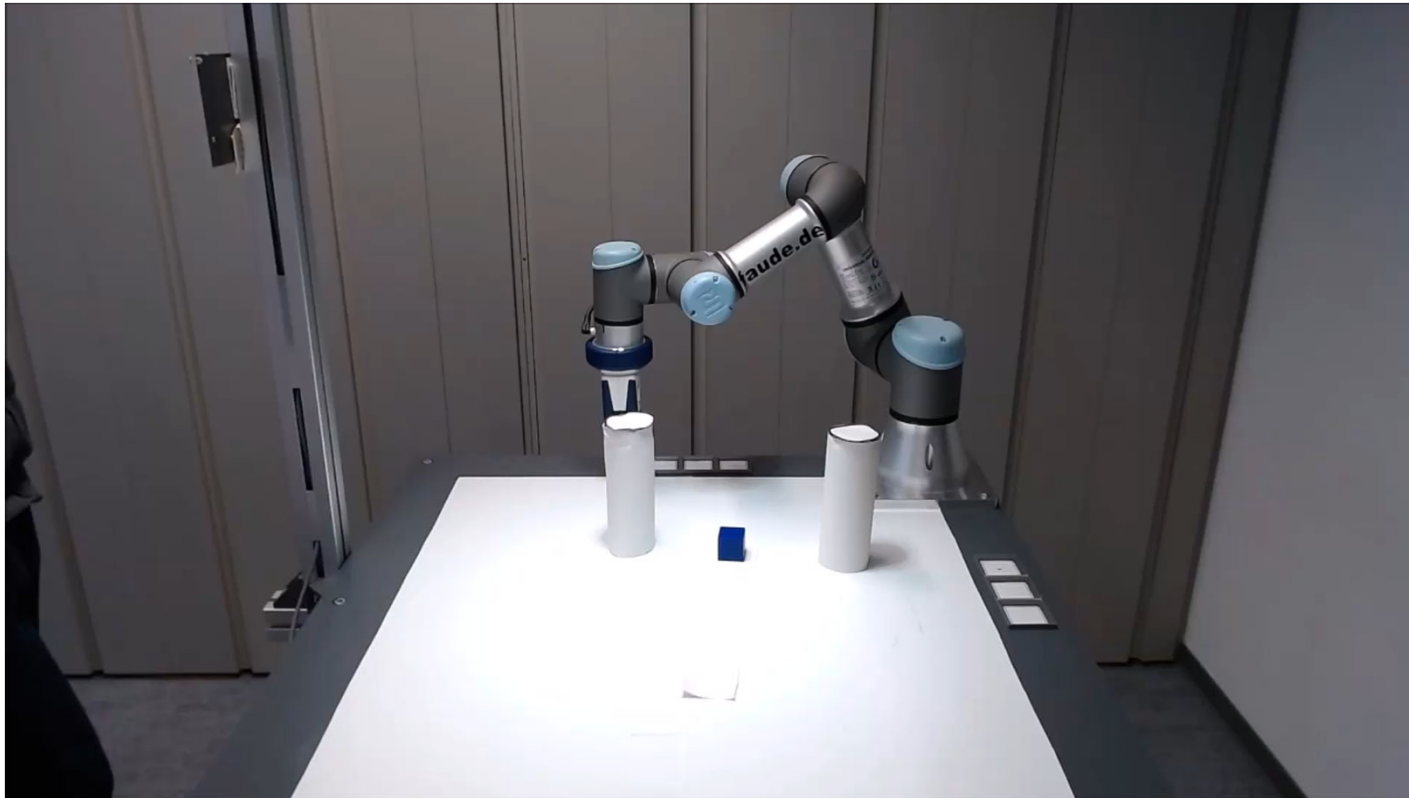
$$J_i(\mathbf{x}_i^k, \mathbf{u}_i^k) := (\mathbf{x}_i^{N_p} - \mathbf{x}_i^f)^T \mathbf{Q}_i^f (\mathbf{x}_i^{N_p} - \mathbf{x}_i^f) + \sum_{k=0}^{N_p-1} ((\mathbf{x}_i^k - \mathbf{x}_i^f)^T \mathbf{Q}_i^x (\mathbf{x}_i^k - \mathbf{x}_i^f) + \mathbf{u}_i^k{}^T \mathbf{R}_i^u \mathbf{u}_i^k + (\mathbf{u}_i^{k+1} - \mathbf{u}_i^k)^T \mathbf{R}_i^d (\mathbf{u}_i^{k+1} - \mathbf{u}_i^k))$$



Planning und Model-Predictive Control

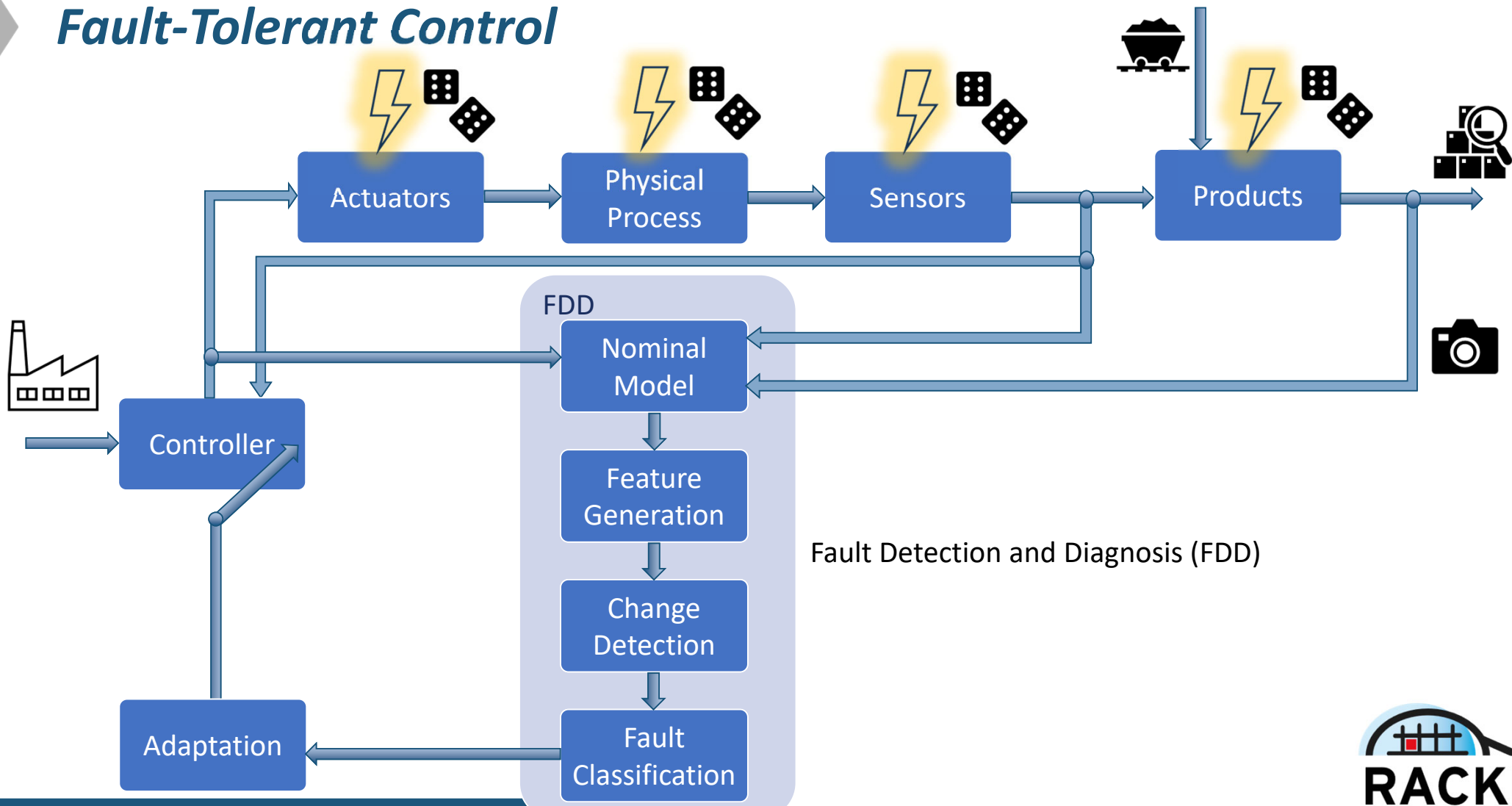


Robot Pick-and-Place with MPC and Collision Avoidance



Nigora Gafur et al., 2022

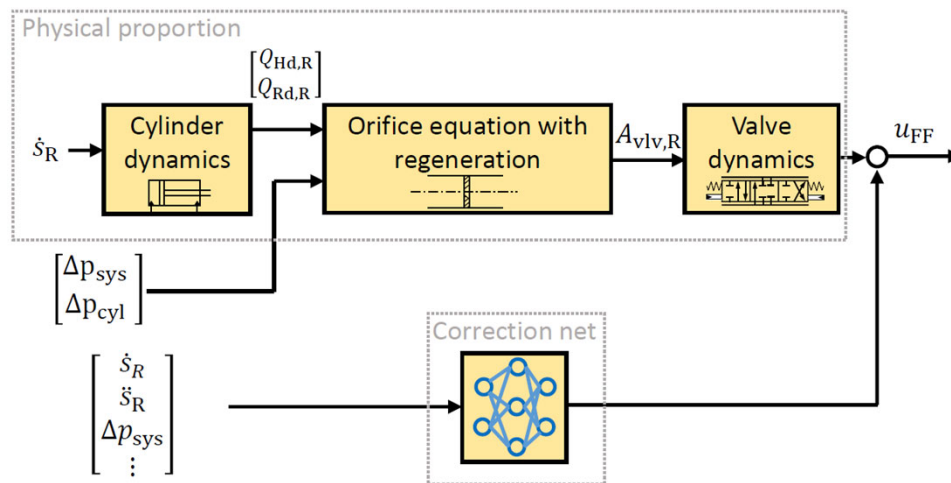
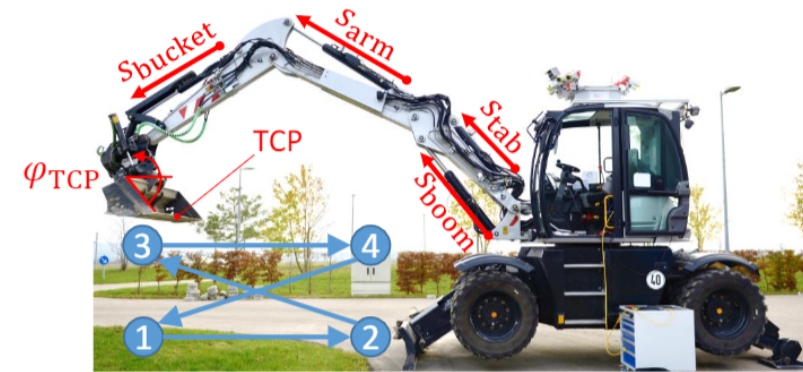
Fault-Tolerant Control



Fault Detection and Diagnosis (FDD)

Hybrid Data-Driven Modelling for Inverse Control of Hydraulic Excavators

- Development of Hybrid models based on expert knowledge and data
- Low effort system modelling
- Simplified general expert model available
- Only small set of experimental data available



4-Point-Cycle, without payload

	Boom Cylinder		Arm Cylinder	
	RMSE	R2	RMSE	R2
White Box	54.97	0.951	91.85	0.955
Hybrid 1	21.38	0.993	31.39	0.994
Hybrid 2	22.06	0.992	31.72	0.994
Black Box	21.27	0.993	31.21	0.994

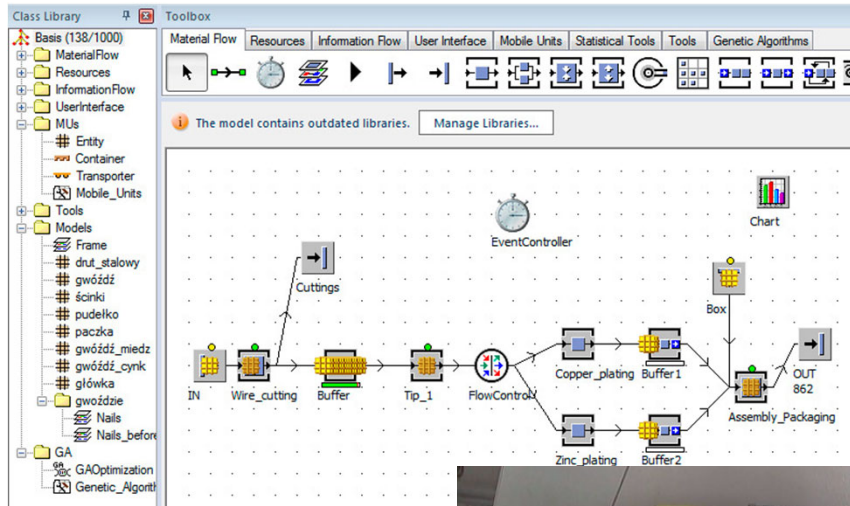
4-Point-Cycle, with additional 1000 kg

	Boom Cylinder		Arm Cylinder	
	RMSE	R2	RMSE	R2
White Box	102.58	0.951	127.38	0.975
Hybrid 1	31.11	0.994	37.35	0.995
Hybrid 2	33.26	0.994	36.72	0.995
Black Box	51.06	0.988	37.14	0.995

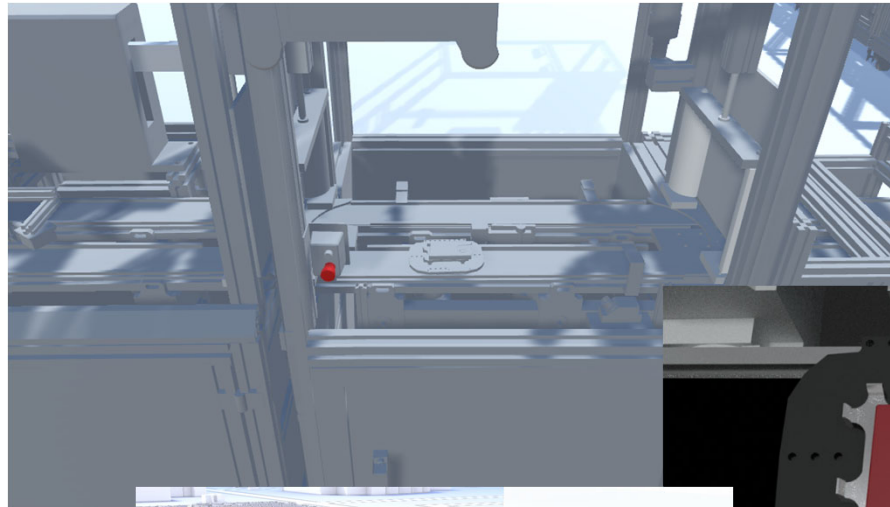
Conclusion

- Design of dependable production systems can be facilitated by design principles and a specific methodology
 - Formal specification of dependability properties and dependability measure
 - Explicit Description of *desired* and *not desired* system behavior
 - Information Structure is important factor for system design
 - Appropriate system decomposition for managing complexity
 - Identification of critical system variables and dependencies is challenging
- Role of Artificial Intelligence
 - Information extraction, knowledge representation, and decision making
 - Improving model accuracy and reducing effort by combining knowledge and data
 - Prediction of system behavior for optimal planning and control
 - Modelling of human behavior for advanced assistance

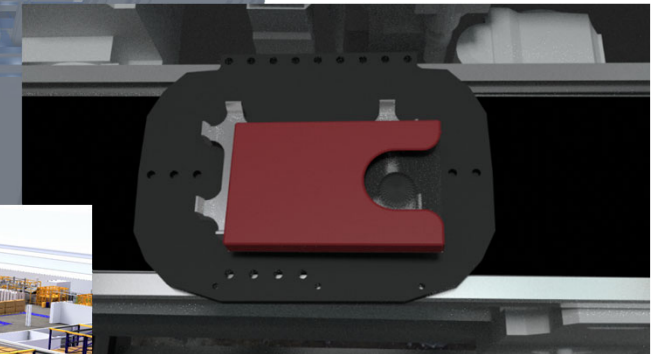
Future AI-Driven Dependability Research



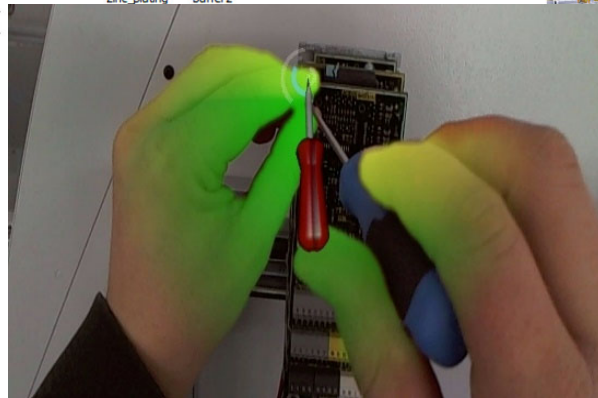
Production processes



Machine Behavior



Product Quality



Worker assistance



Tracking/Visualization

Thank You for Attention!



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References

- European Commission, Directorate-General for Communications Networks, Content and Technology, *Ethics guidelines for trustworthy AI*, Publications Office, 2019, <https://data.europa.eu/doi/10.2759/177365>
- Carter, 1982: Carter, W.: A time for reflection. In Proc. 12th Int. Symp. on Fault Tolerant Computing (FTCS-12). IEEE Computer Society Press Santa Monica
- Laprie, 1992: Laprie, J. C.: Dependability. Basic Concepts and Terminology. Ed. Springer Verlag.
- Dubrova, 2013: Dubrova, E.: Fault tolerant design: An introduction. Springer, DOI 10.1007/978-1-4614-2113-9
- Rüdiger, 2007]:Rüdiger, J., Wagner, A., and Badreddin, E.: Behavior based definition of dependability for autonomous mobile systems. In European Control Conference 2007 (July 2-5), pp. 4146–4151.
- Jan C. Willems, Paradigms and Puzzles in the Theory of Dynamical Systems, *IEEE Transactions on Automatic Control*, Vol. 36 No. 3, March 1991.
- Wagner 2018: Modeling, Analysis, and Design of Dependable Systems with Application to Robotics and Assistance Technology. Verlag Dr. Hut 2018, ISBN 978-3-8439-3559-3.
- E. Badreddin, “Recursive Control Structure for Mobile Robots”, International Conf. on Intelligent Autonomous Systems 2 (IAS.2), Amsterdam, pp. 11-14, Dec. 1989.
- C. Bartolein, A. Wagner, M. Jipp, E. Badreddin: Multilevel Intention Estimation for Wheelchair Control, in Proc. of the European Control Conference 2007, Kos, Greece, July 2-5, 2007.
- Wagner A, Jipp M., Kandil A., Eck C, Badreddin E. Generic system architecture for dependable interactive systems: A flying robot example. *IEEE International Conference on System, Man and Cybernetics*: 2010:2129-2136, 2010.
- Wagner 2016: Wagner, A., Bartolein, C., Badreddin, E., Multi-Level Human-Machine-Interaction Monitoring and System Reconfiguration. IFToMM/IEEE/euRobotics 25th International Conference on Robotics in Alpe-Adria-Danube Region – RAAD, Belgrade, 2016
- Gafur 2021: Gafur, N., Yfantis, V., Ruskowski, M.: Optimal scheduling and non-cooperative distributed model predictive control for multiple robotic manipulators. *IEEE/RSJ International Conference on Intelligent Robots and Systems (IROS)*, pp. 390-397, 2021. doi: 10.1109/IROS51168.2021.9636118.
- Gafur 2022: N. Gafur, L. Weber, V. Yfantis, A. Wagner, M. Ruskowski. Dynamic path planning and reactive scheduling for a robotic manipulator using nonlinear model predictive control. Submitted to Mediterranean Conference on Control and Automation (MED) 2022.
- Weigand et al. 2021: Jonas Weigand, Julian Raible, Nico Zantopp, Ozan Demir, Adrian Trachte, Achim Wagner, and Martin Ruskowski. Hybrid Data-Driven Modelling for Inverse Control of Hydraulic Excavators. In 2021 IEEE/RSJ International Conference on Intelligent Robots and Systems (IROS), pages 2127–2134, 2021