

Trust and safety in the context of complex AI systems: challenges related to verification and validation

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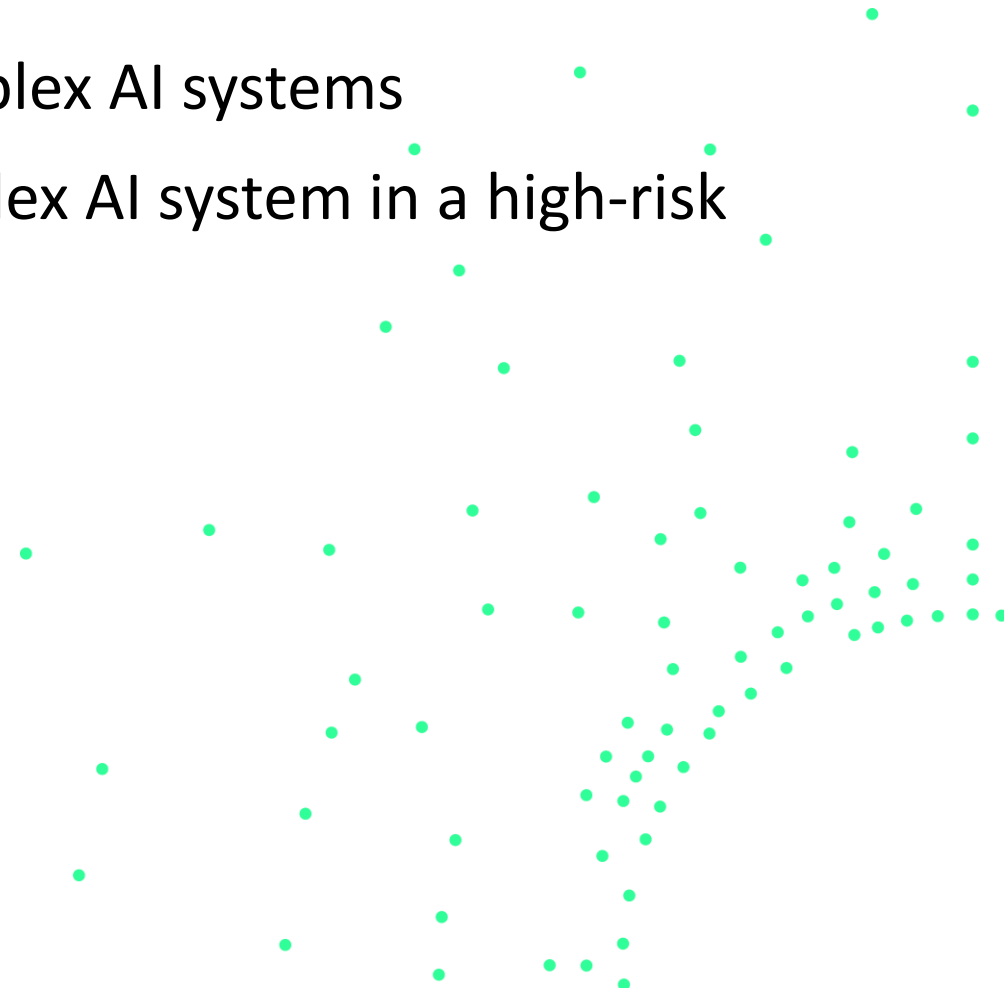
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Agenda



- Introduction
- Challenges for verification and validation in complex AI systems
- Lessons learned from implementation of a complex AI system in a high-risk environment
- Conclusion



Introduction



- Trust
- Safety
- Assumptions
- Domain related constraints
- Interoperability: data and functionalities

Trust in the system:

- ✓ Trust that the algorithm makes the correct decision under the correct assumption.
- ✓ Understand what happened when the algorithm failed.
- ✓ Predictions in line with expectations.
- ✓ Knowing that the algorithm performs badly on a specific set of data is very useful (thorough knowledge of limitations is very important).

Introduction



Safety:

- ✓ Safety of the system.
- ✓ Safety of the environment.
- ✓ Safety of the process.

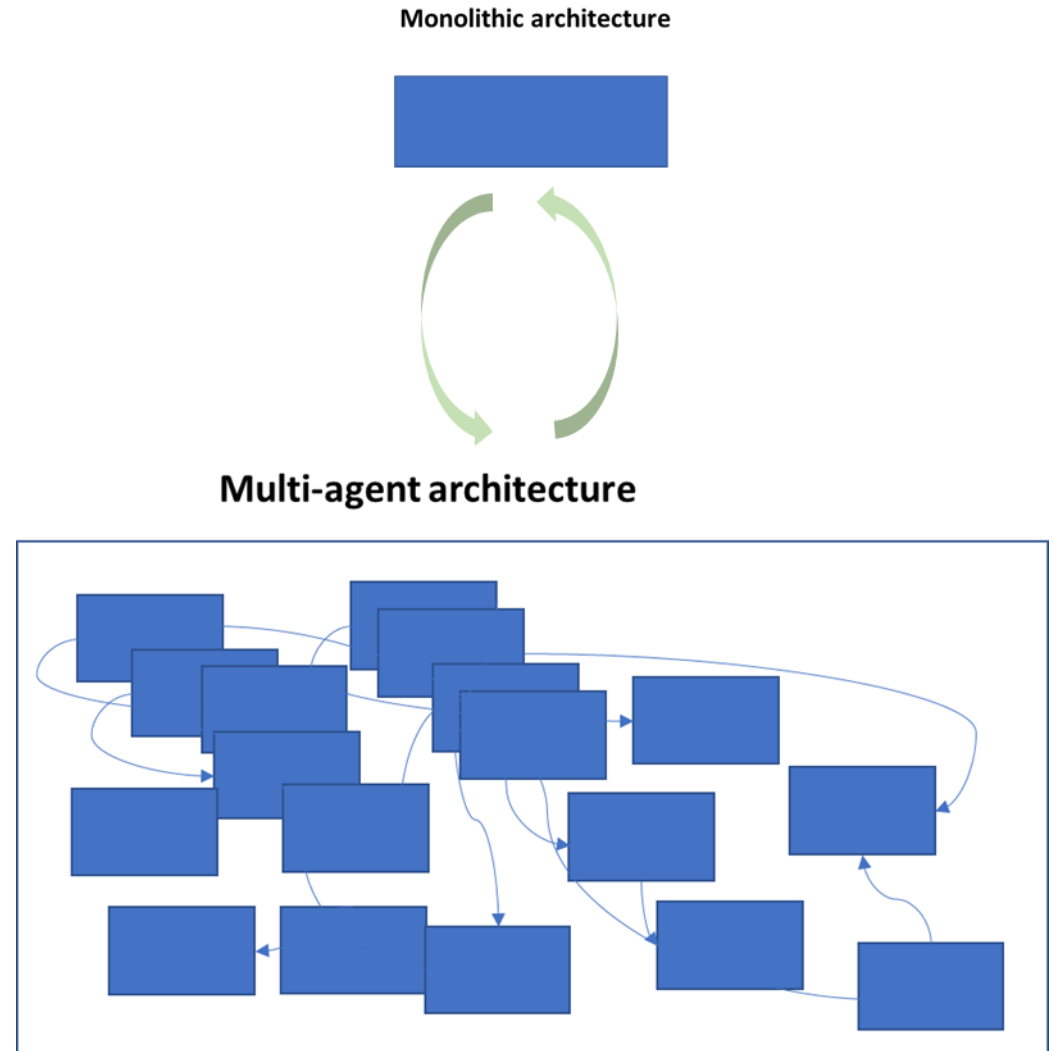
Introduction



- How much data is needed for the method to work.
- Specific method features (assumptions: example – population remains the same).
- React to abrupt changes in environment.
- Sensitive decision.
- Context is not fixed (different companies, sensors, etc.).
- Time and space sparsity of the data.
- Continuous real-time functioning without interruptions.

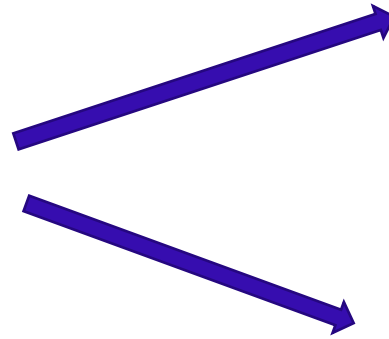
Challenges for verification and validation in complex AI systems

- Multiple providers
- Proprietary systems
- Black box behaviour
- Rigid versus adaptive behaviour
- Information on training data
- Online training capabilities
- Information on specific assumptions
- Interoperability: data and functionalities

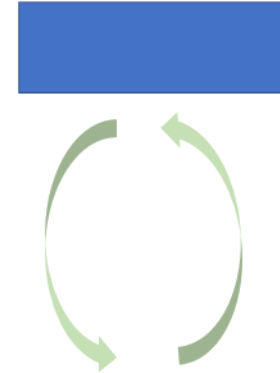


Lessons learned from implementation of a complex AI system in a high-risk environment

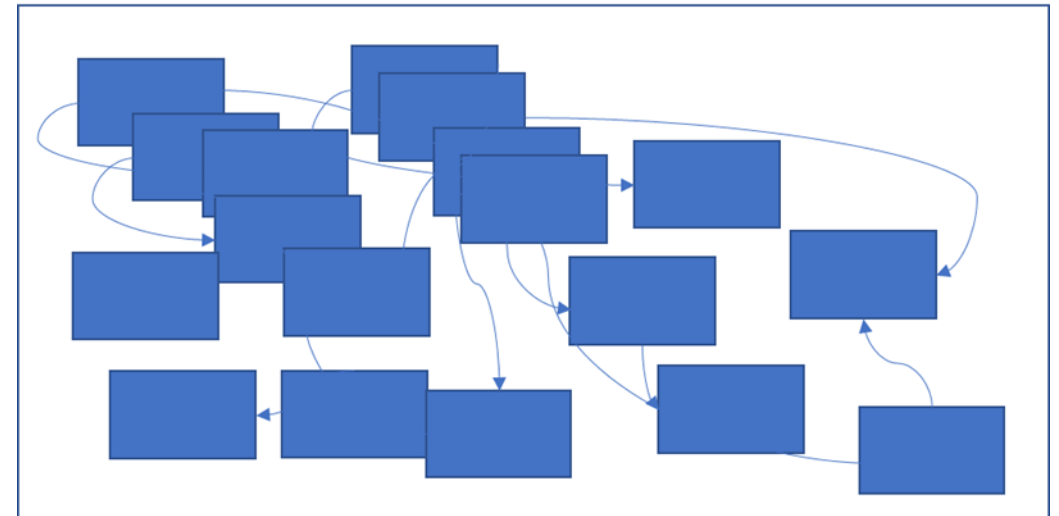
- ✓ Multiple providers
- ❖ Proprietary systems
- ❖ Black box behaviour
- ✓ Rigid versus adaptive behaviour
- ✓ Information on training data
- ✓ Online training capabilities
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- ✓ Interoperability: data and functionalities



Monolithic architecture



Multi-agent architecture



Lessons learned from implementation of a complex AI system in a high-risk environment



Drilling operations:

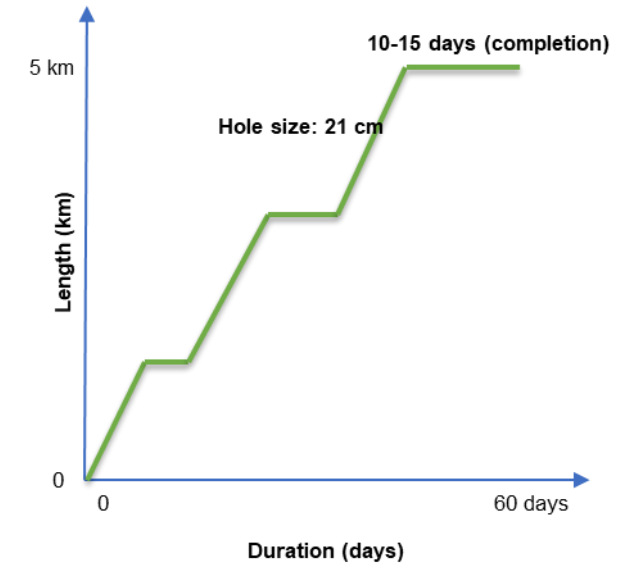
- Oil and gas
- Geothermal
- CO2 storage
- Hydrogen storage, production, etc.

Main goal of drilling operation is:

- To reach the desired depth as fast as possible.
- To avoid incidents.

Incidents cause delays in the process and can lead to damaging situations (hole collapse, drill-string getting mechanically stuck, etc.)

- Continuous estimation of downhole conditions.
- Proactive actions (weak symptoms)



Lessons learned from implementation of a complex AI system in a high-risk environment



- Context: **dynamic**
- Data: varying **quality and availability**
- Safety of the process: **highly relevant**
- Criticality of the process: **highly critical process**
- Level of human involvement in the decision making: **no human intervention**

Motivation&Objectives:

- ✓ **Increase safety**
- ✓ **Improve performance**
- ✓ **Achieve consistency in the process**

Lessons learned from implementation of a complex AI system in a high-risk environment



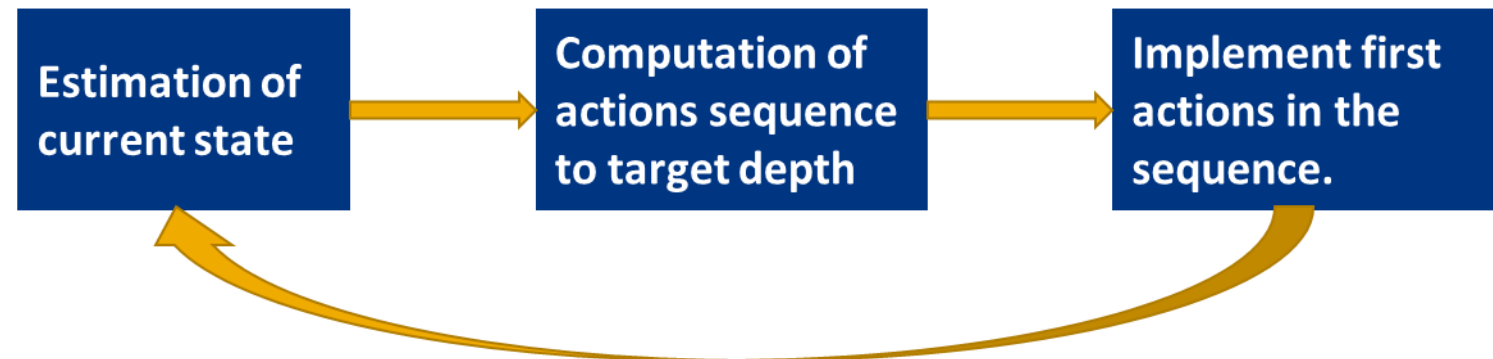
- Aim

To demonstrate autonomous drilling.

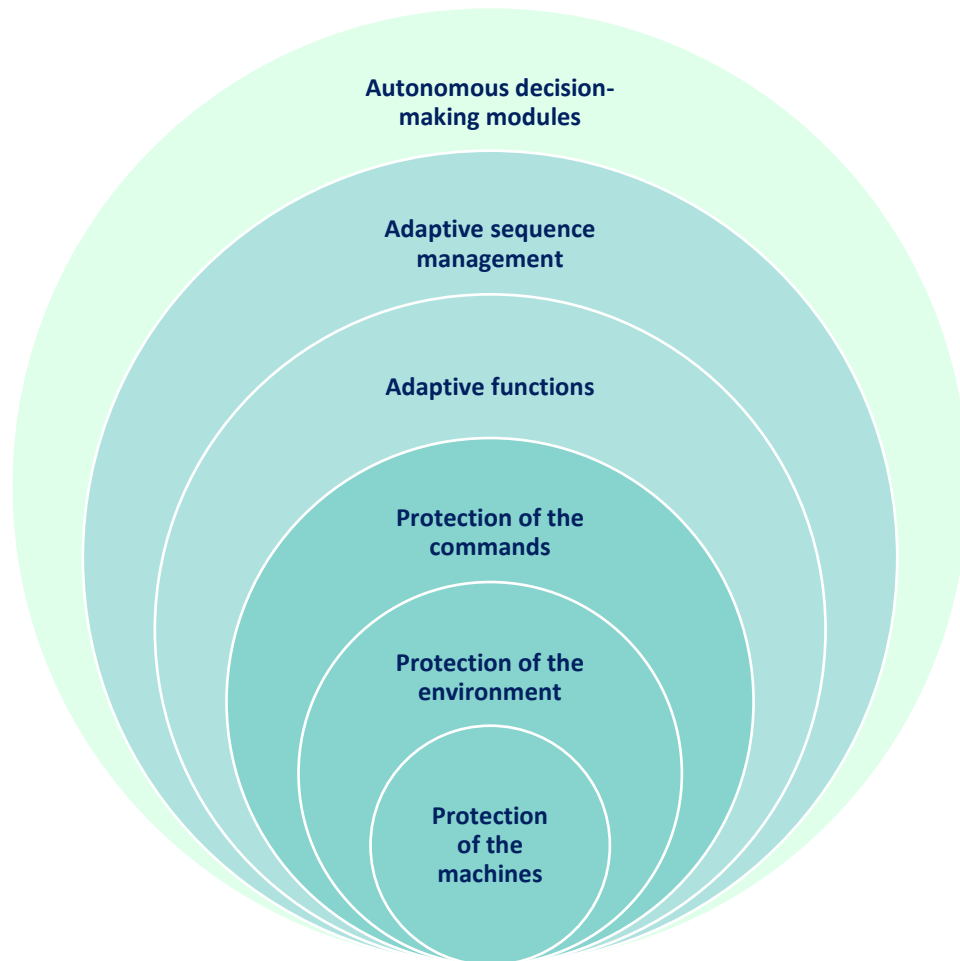
- How

1. Autonomous decision-making on the fly, handling of unexpected situations.
2. Executing these decisions automatically.

- Strategy



Lessons learned from implementation of a complex AI system in a high-risk environment

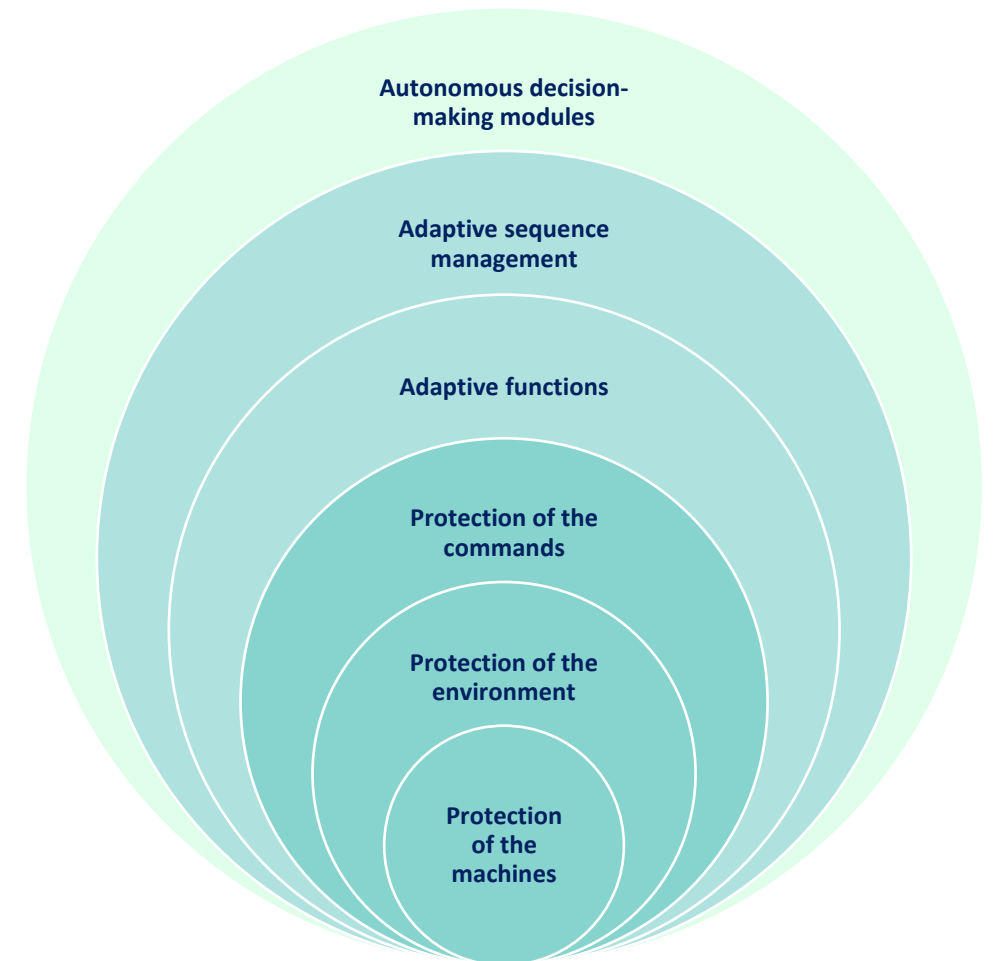


1. *On Transitions Functions Model for Decision-Making in Offshore Operations*, R. Mihai, B. Daireaoux, A. Ambrus, E. Cayeux and L. Carlsen, NORCE, 2022 IEEE 17th International Conference on Control & Automation (ICCA), 2022, pp. 309-314, doi: 10.1109/ICCA54724.2022.9831911
2. *Autonomous Decision-Making While Drilling*, E. Cayeux, B. Daireaoux, A. Ambrus, R. Mihai, and L. Carlsen, NORCE, *Energies*, vol. 14, no. 4, p. 969, Feb. 2021, doi: 10.3390/en14040969.
3. *Demonstration Of Autonomous Drilling On A Full-scale Test Rig*, R. Mihai, E. Cayeux, B. Daireaoux, L. Carlsen, A. Ambrus, P. Simensen (NORCE), M. Welmer (SEKAL), M. Jackson (NOV), SPE Annual Technical Conference and Exhibition, Houston, USA, 2022, <https://doi.org/10.2118/210229-MS>.

Autonomous drilling: protection features



- **Adaptive functions**
- **Protection functions:**
 - **Safe operating envelopes (SOE):**
 - Acceptable setpoints limits to respect operational constraints
 - **Fault detection Isolation and Recovery (FDIR):**
 - Quick events are quite common
 - Fast detection and reaction is often essential to avoid an escalating problem.
- **Safe mode management: transition from autonomous to manual drilling** (to address decreased situational awareness).



Addressing low situational awareness from a technological perspective – autonomous setting



- **What:** process state which is stable for a certain amount of time.
 - Give the user sufficient time to get into the current situation
- **When:** transition from autonomous to manual control.
 - Because a recovery procedure has not succeeded.
 - Because of an internal problem which impeded continuing in autonomous mode.
- **Why:** safe transition to manual control
 - Because situational awareness of user is reduced while running autonomous processes.
 - It might be a different user than the one who engaged the autonomous mode.
- **How:** Estimate which set of actions that can bring back the process in a stable condition
 - This set of action is context dependent.

Conclusion



- Some relevant aspects for verification and validation of complex AI systems were illustrated (also from an implementation of such a system).
- Situational awareness and an example how to cope with reduced situational awareness from a technological perspective.
- Information on specific assumptions and training data relevant to be shared for a smooth interplay in complex AI system.
- Interoperability of data and functionalities is crucial for complex AI systems (multiple providers perspective).

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Thank you!