



Industrial Examples of Graph Applications

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CINTI 2023 Conference

With direct contribution from:

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Personal Intro

@Bosch Budapest since 2015

- Bosch Scientist – Academic Relations team
- Automotive communications and security testing background
- Knowledge management – Product cybersecurity

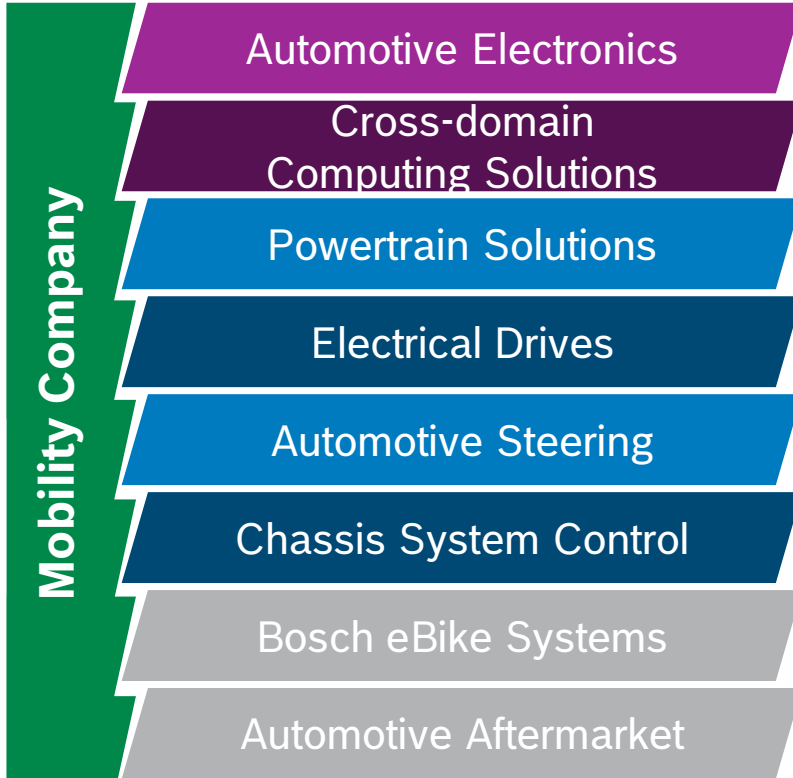
- PhD in Physics (experimental particle physics, ELTE, Budapest, Postdoc at CERN, Geneva) – photon structure measurements
- MBA in international management (University of Geneva)
- 20 years in industry:
 - Embedded systems, real-time control applications
 - HW-SW codesign & optimization
 - Functional safety and cybersecurity
 - Aerospace and automotive industries

- Family with 3 teens

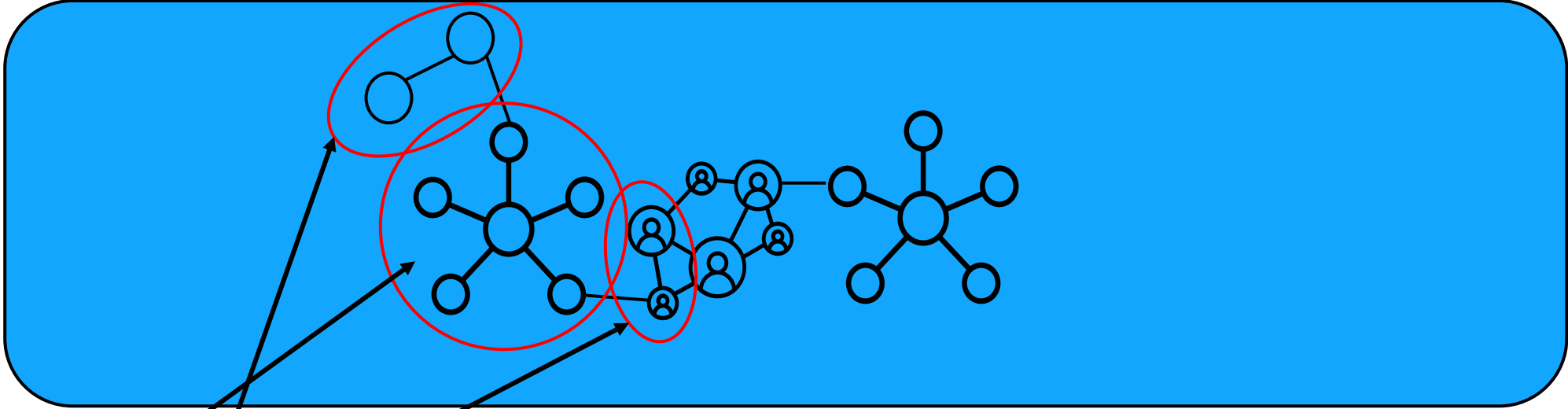


Knowledge graphs link engineering data across the organization

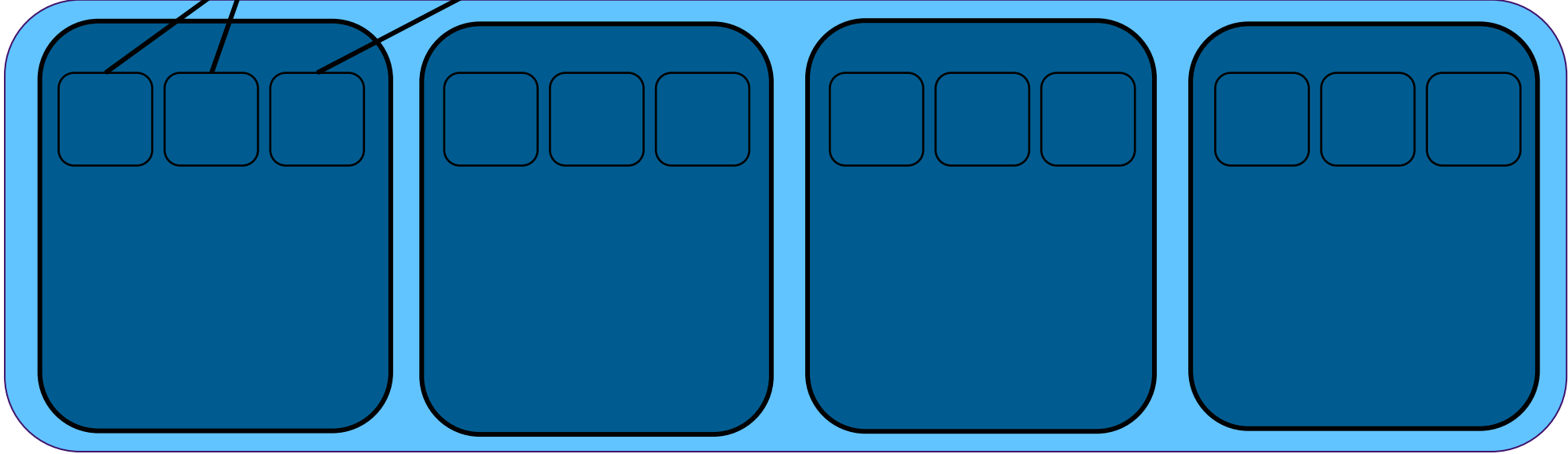
What we work on Divisions @ Budapest



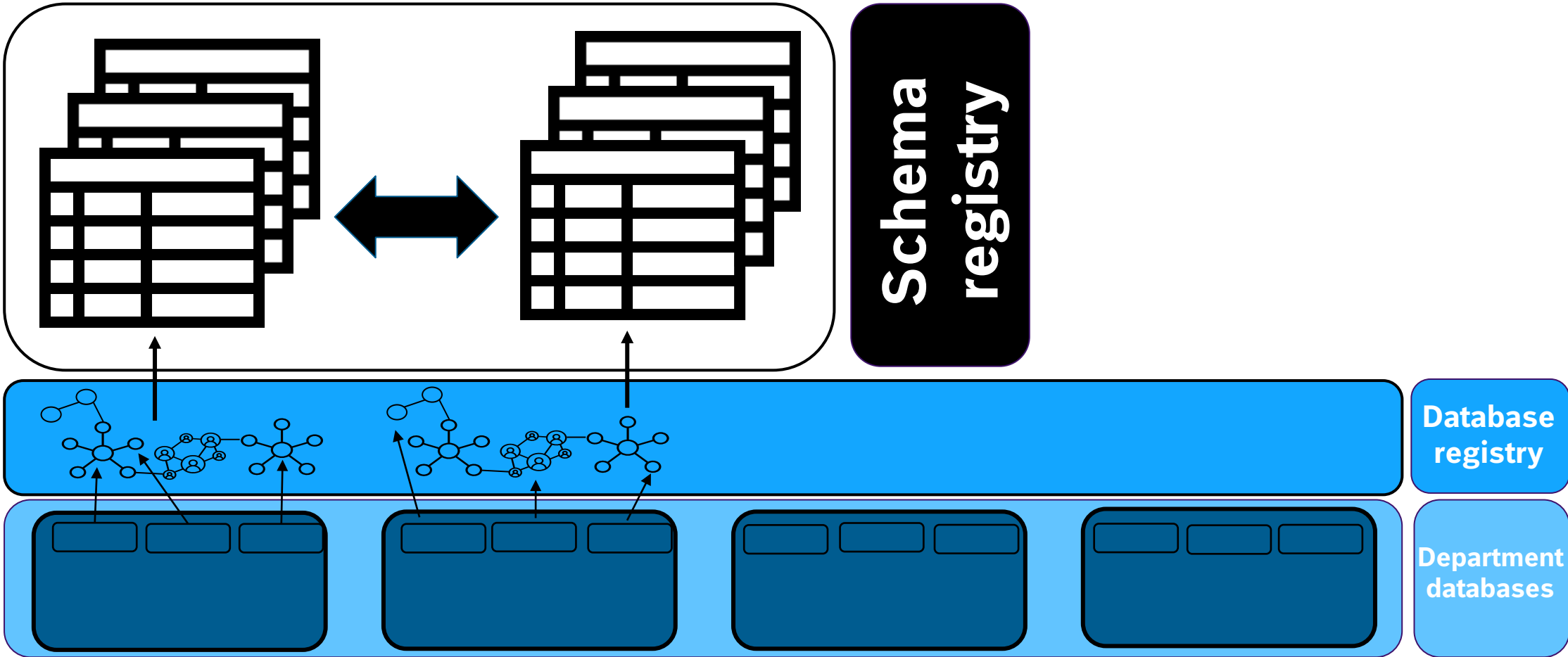
The presence of most Bosch Mobility Company divisions at ECB offers extraordinary synergy possibilities

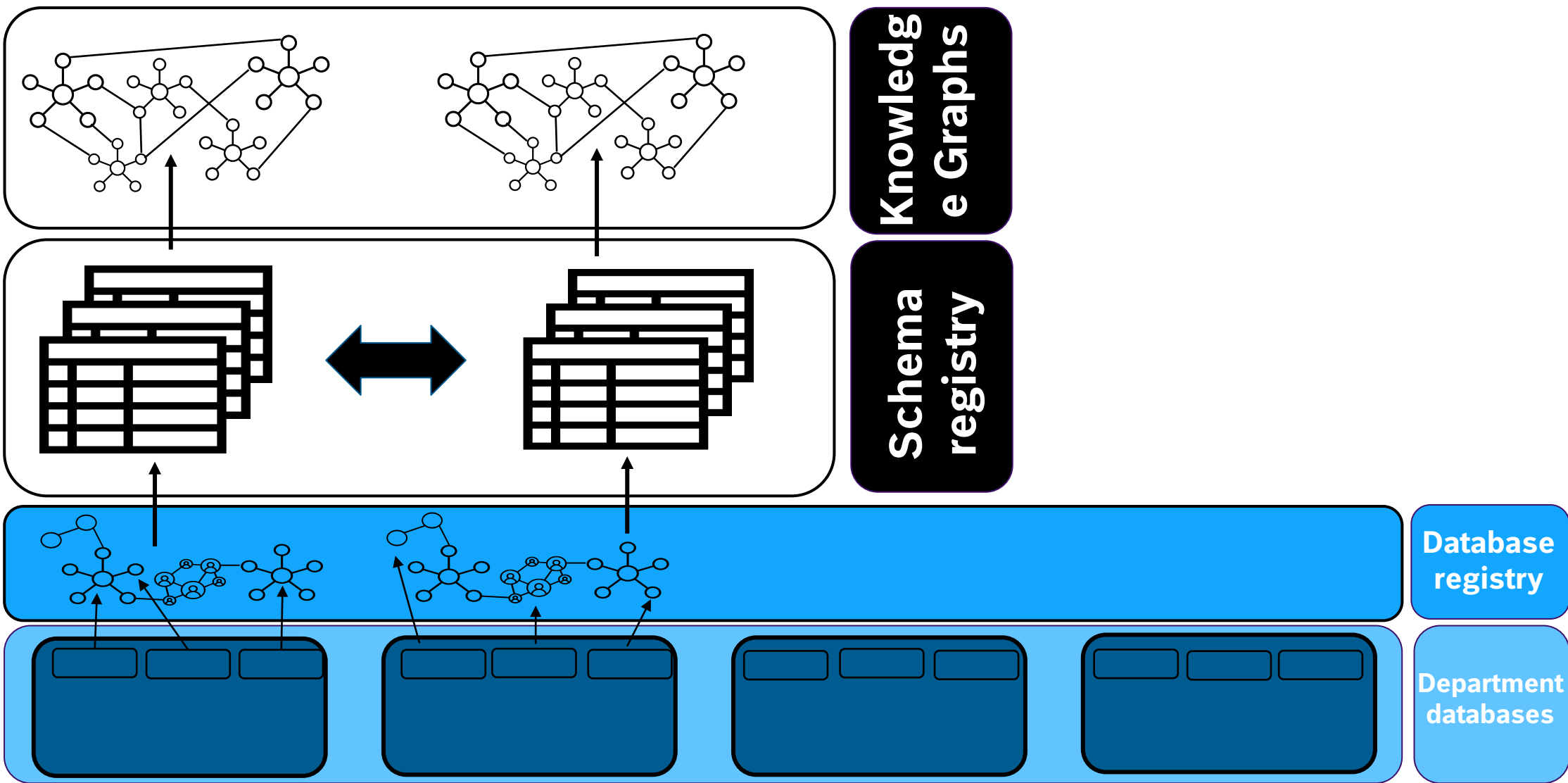


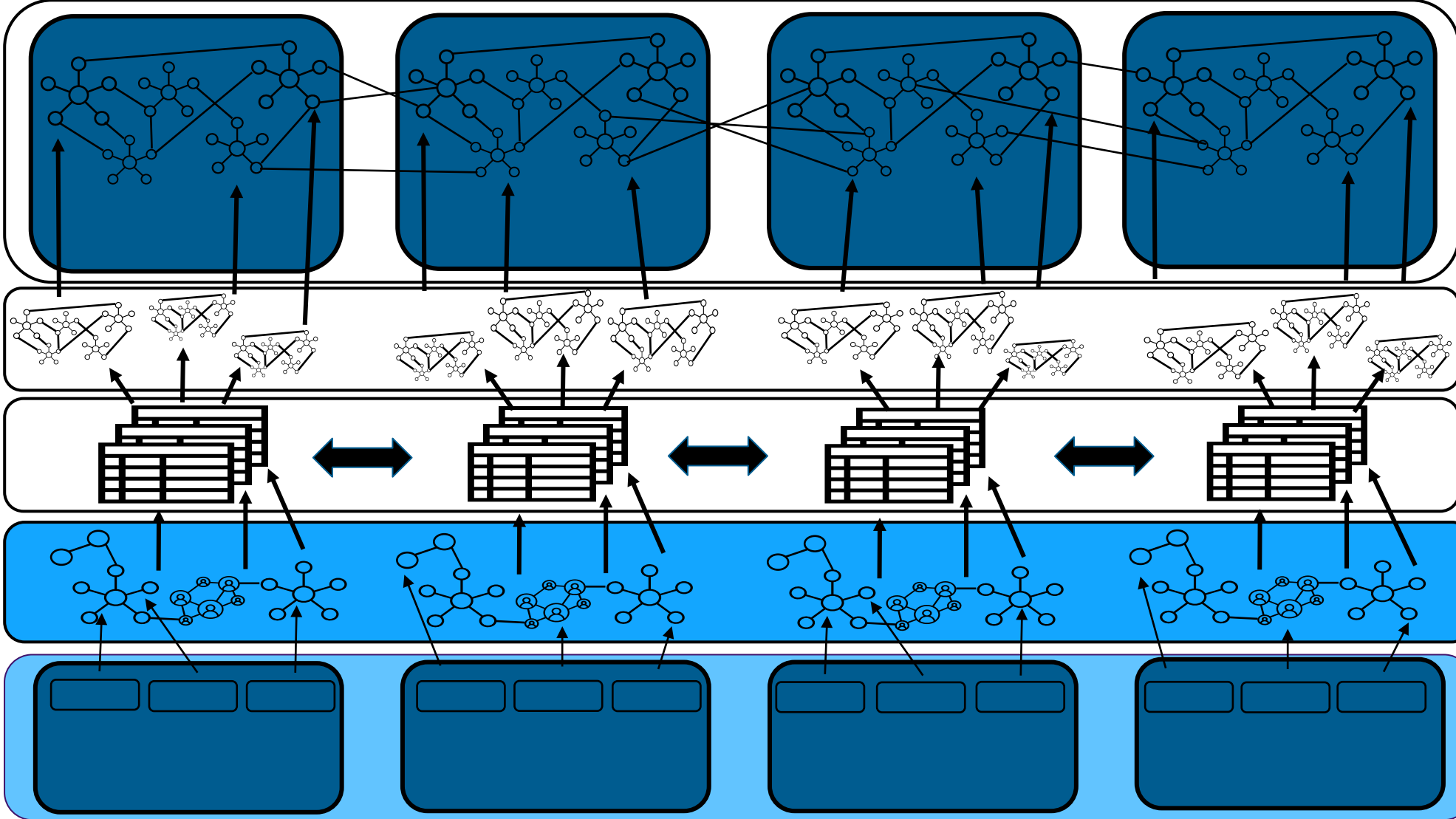
Database registry



Department databases







**ORG
Data
map**

**Knowledge
Graphs**

**Schema
registry**

**Database
registry**

**Department
databases**

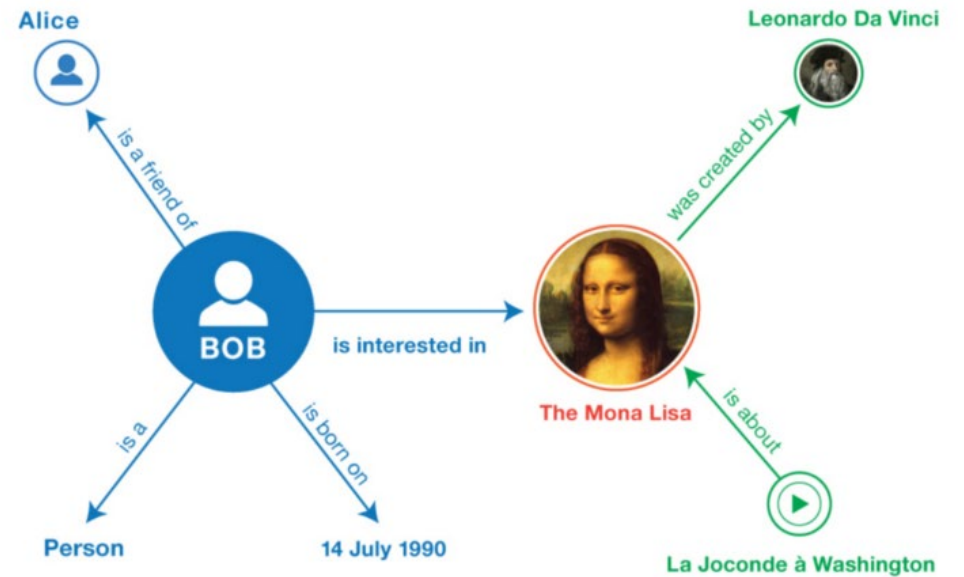
Knowledge Graphs in Engineering Terms

A **vocabulary** refers to a set of terms and concepts to describe data.

A **taxonomy** is a hierarchical framework of the terms.

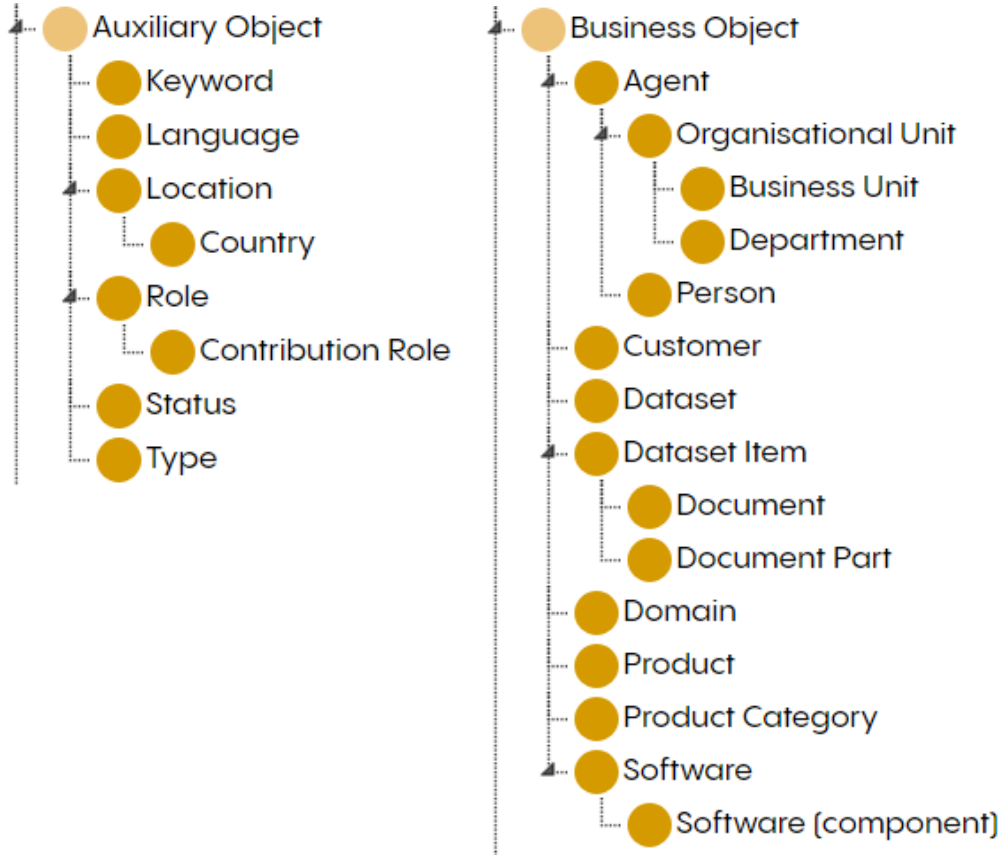
An **ontology** is the composition of properties and how they are related, by defining a set of terms and relational expressions that represent the entities.

A **triple** is a statement about semantic data in the form of subject-predicate-object expression.

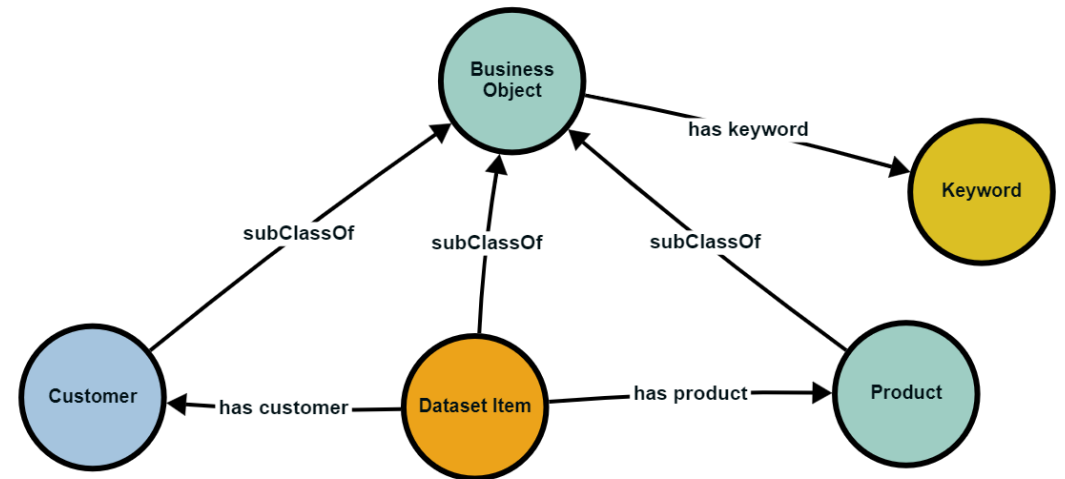


Knowledge Graphs in Engineering

Core data ontology



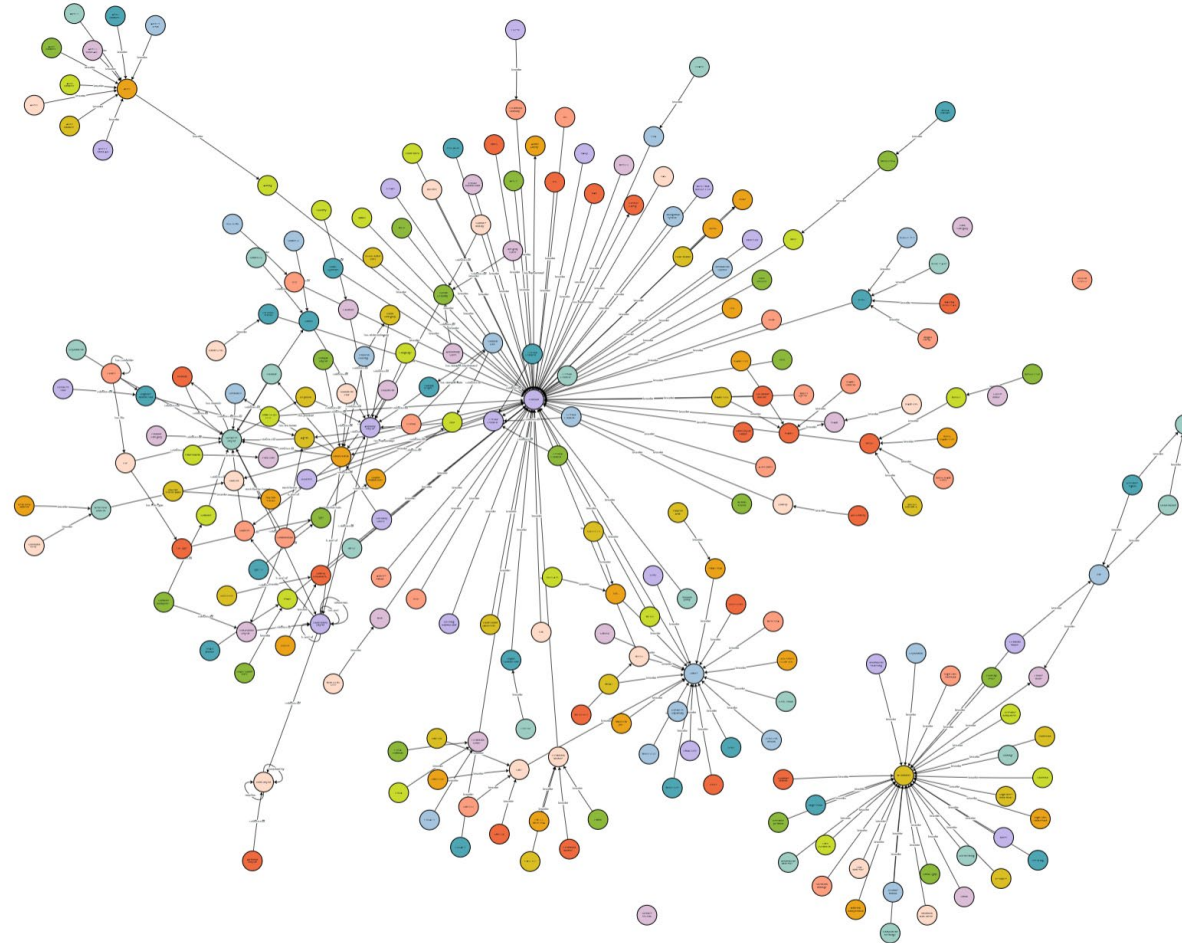
Our **core data ontology** is an abstract description of data assets and related meta-data objects.

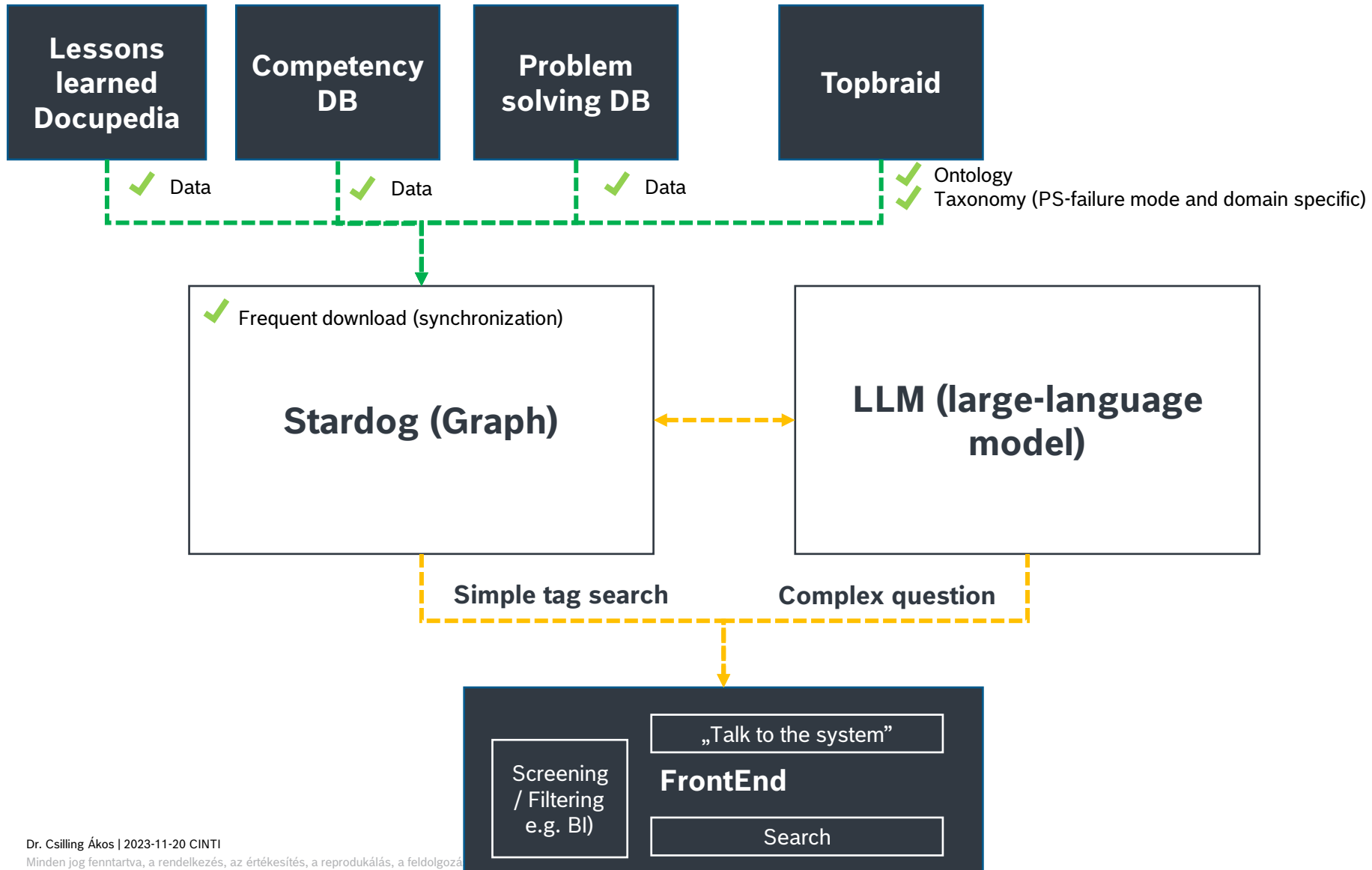


- Implemented in current ontology
- Imported from outer ontology

Knowledge Graphs in Engineering

KG in Stardog Explorer





Knowledge Graphs in Engineering

Key Take-Aways

Why Knowledge Graphs?

- **Flexible data model**, easier to integrate **new data sources**
- **Dynamic** way to **connect data points**, revealing **intricate relationships**
- **Intuitive** and **efficient exploration** of information

Why Knowledge Graphs with LLM?

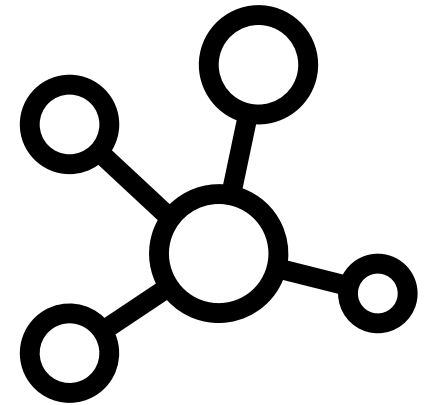
- Enabling more **complex, contextual questions** to retrieve information

Current Status:

- Mapping our data asset ✓
- Competence build up ✓

Next steps:

- Integrating more data assets
- Roll out first use-case (LL and Problem solving)



Behavior Prediction with Graph Neural Networks

Engineering Center Budapest

Our Focus Areas in Budapest



Automation

Environment Sensing

Radar, camera and ultrasonic sensors.



Perception, localization

Computer Vision with AI, e.g.:
Road Sign Recognition



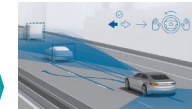
Prediction, planning

Predictive driver assistance functions, e.g.:

Automated Emergency Braking



Motion Control Advanced Driver Assistance



Traffic Jam Assist



Adaptive Cruise Control



Power Steering



ESP

Prototype vehicle built in ECB for testing of >L3 autonomous functions



We develop elements of key importance for the Bosch Automated Driving System.
Our goal: Deliver turnkey Automated Driving solutions to our customers.

Behavior Prediction with Graph Neural Networks

Intro to L4/RB-Stack

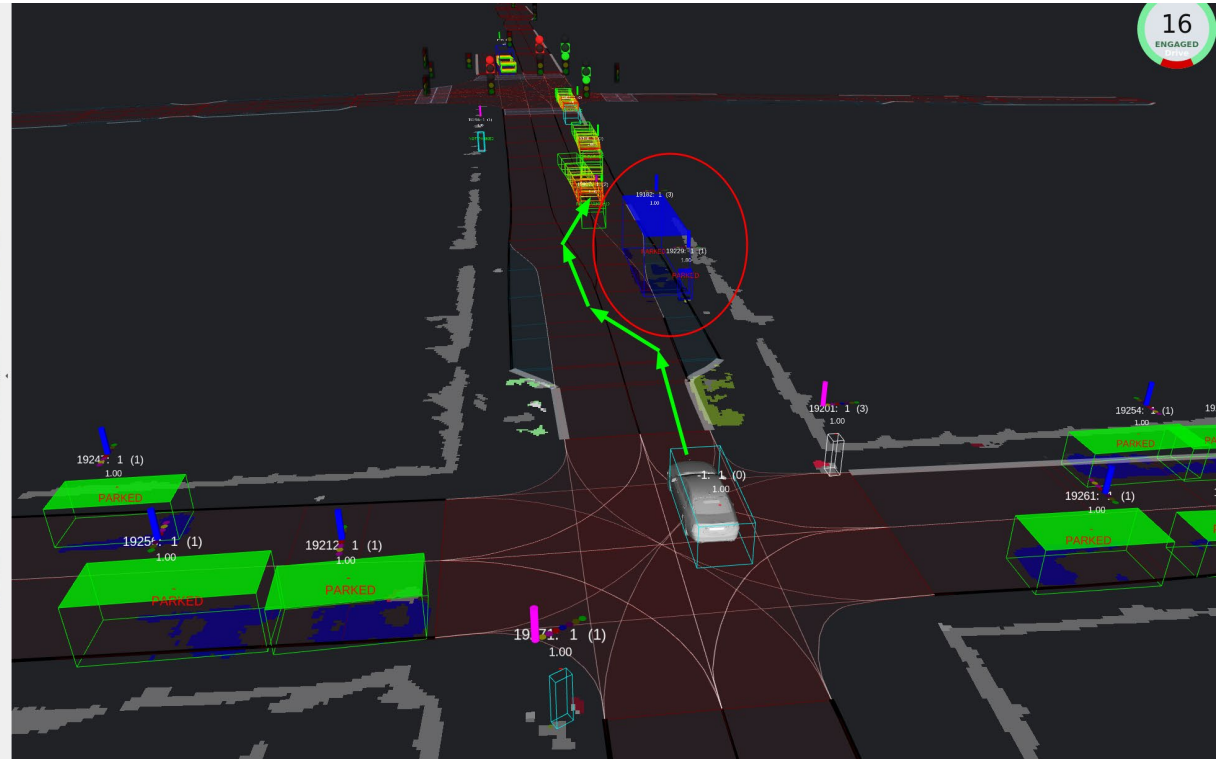
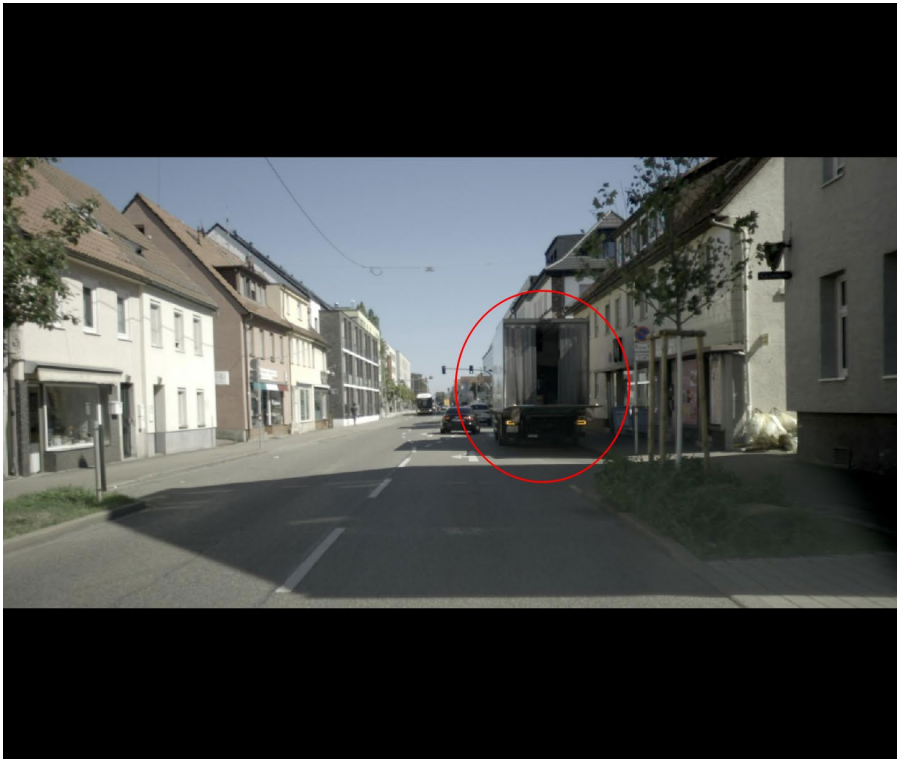
- Goal is SAE L4 Automation
- Sensor set
- Planning Machine Learning (PLNML): scope is end-to-end machine learning solutions in prediction and planning
- Planning component in the stack



Behavior Prediction with Graph Neural Networks

Problem Definition (example for parked car classification)

- Aim in L4 is full autonomy, ego vehicle cannot be stuck
- Parked car classification is a surprisingly hard problem



Dr. Csilling Akos | 2023-11-20 CIN11

Minden jog fenntartva, az értékesítés, az értelmezés, a reprodukálás, a feldolgozás, a továbbadás, és a jogvédelmi bejelentések joga is.

Behavior Prediction with Graph Neural Networks

Problem Definition (example for parked car classification)

- Parked car classification is a surprisingly hard problem



Behavior Prediction with Graph Neural Networks

Problem Definition

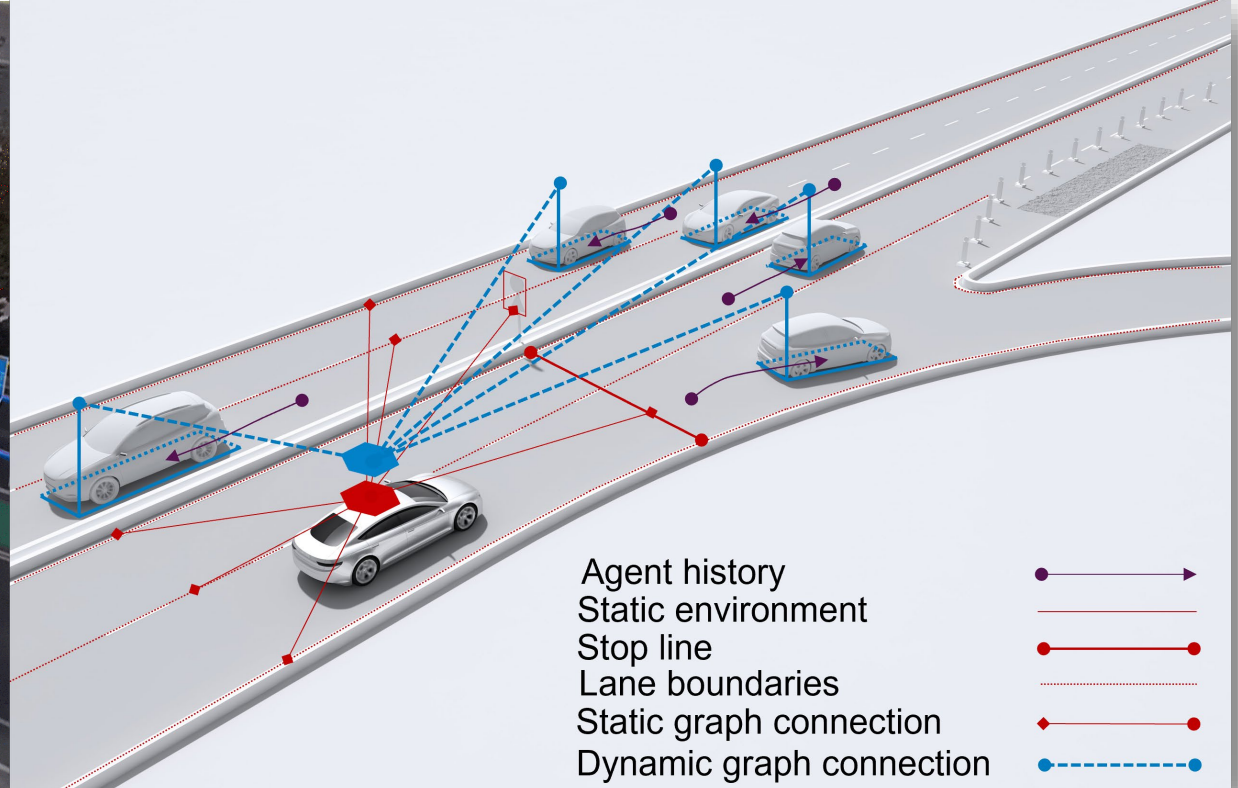
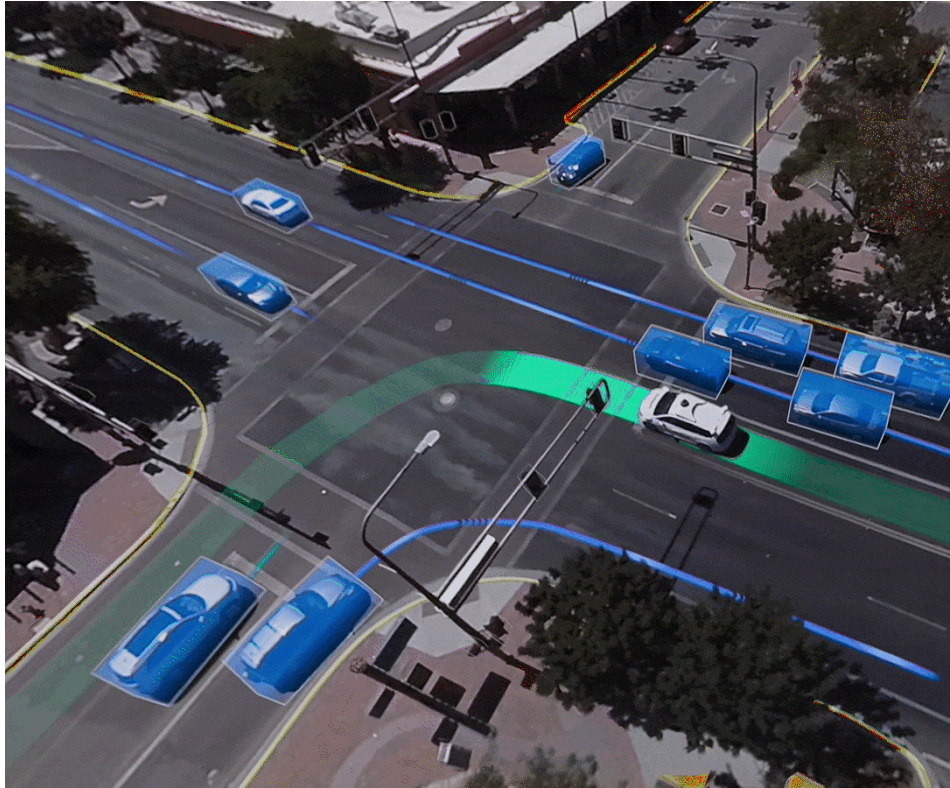
- System limitations for parked cars with Logistic Regression (LR)
 - General performance was not satisfactory for new behavior: nudging
 - No generalization to unseen data (e.g. different routes in Germany)
 - Typical problem clusters (static agents, no state change possible)
- Other system limitations (trajectory, ego lane change success, etc.)
- Proposed solution: Multi-task GNN
 - Modeling of agent-agent, and agent-environment relations and interactions
 - Serving not only parked car classification, but other prediction/planning tasks



Behavior Prediction with Graph Neural Networks

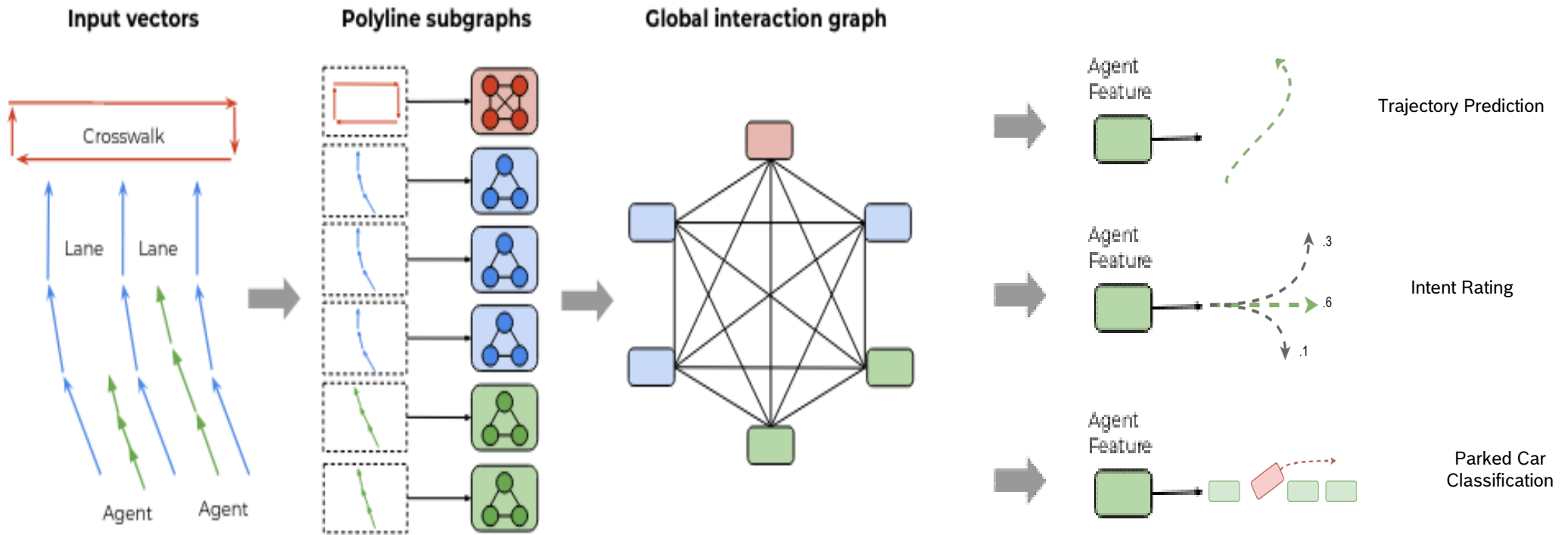
Solution: Multi-task GNN

- Proposal: vectornet working on graph representation
- Input of the model is a fused, vectorized environment and agents



Behavior Prediction with Graph Neural Networks

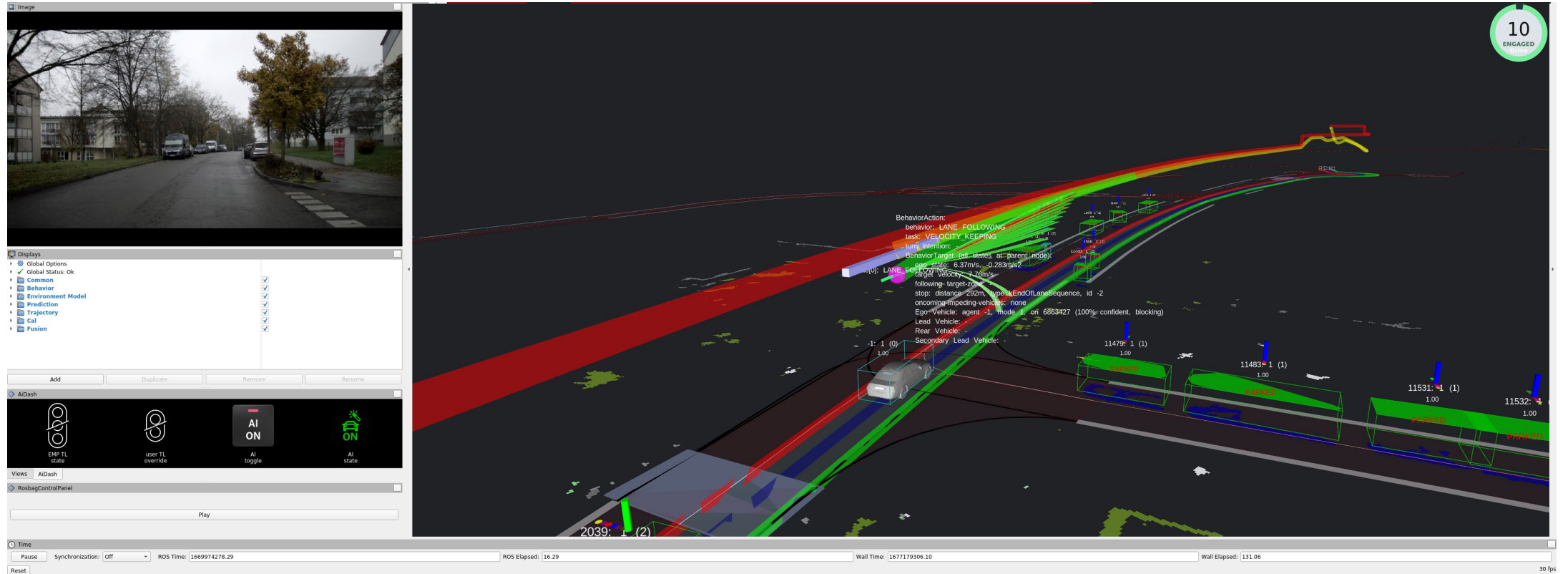
Solution: Multi-task GNN



Behavior Prediction with Graph Neural Networks

Comparison between LR and GNN

- Example of a static agent at an intersection

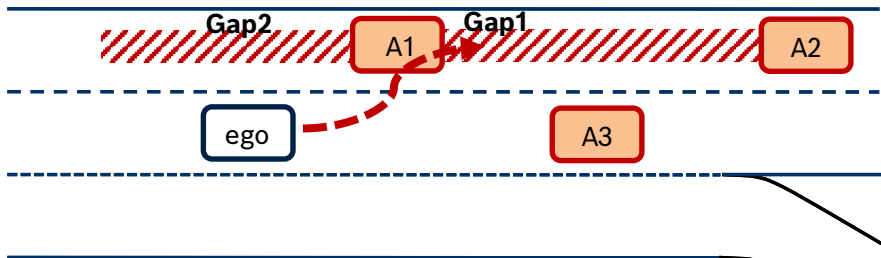
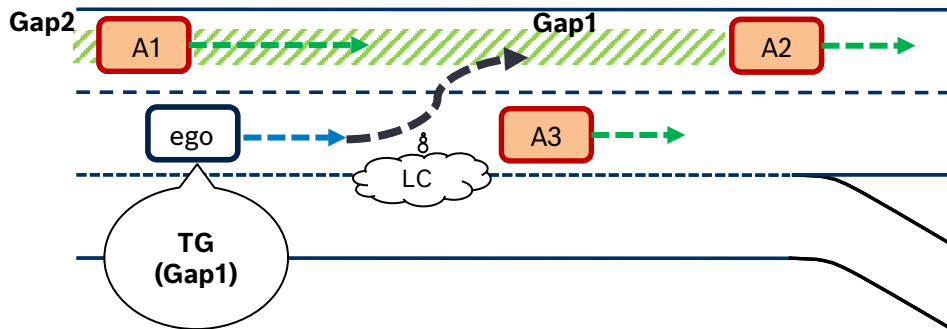


Behavior Prediction with Graph Neural Networks

Outlook to other Planning tasks

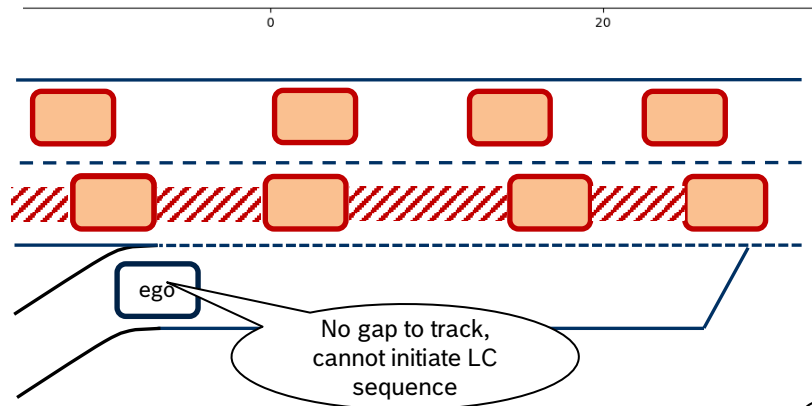
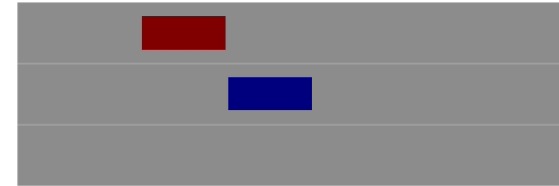
Ego Lane Change Success

- Veto the possibly unsuccessful lane changes
- Rating the track gaps (Gap1, Gap2)



Hybrid Trajectory Prediction

- Take into account interactions for a more assertive lane change
- Useful in merges and dense traffic



Evaluating Cybersecurity with Attack Graphs

Security of Automotive Products

Cybersecurity is Mandatory to Ensure Road Safety

Originally, cars were designed as closed systems. Cyber-security was not an issue.

- Many **different!** electronic control units (ECUs) on an inhomogeneous network.
- More and more features implemented in SW, including safety-critical drive-by-wire functions.
- External **connectivity** (Infotainment, SW update, emergency calls, v2x smart services).
- **Remote access**
 - Lock / unlock
 - Start / stop
 - User devices
- Cybersecurity management and evaluation is mandatory in most markets, ISO 21434 is the standard method.



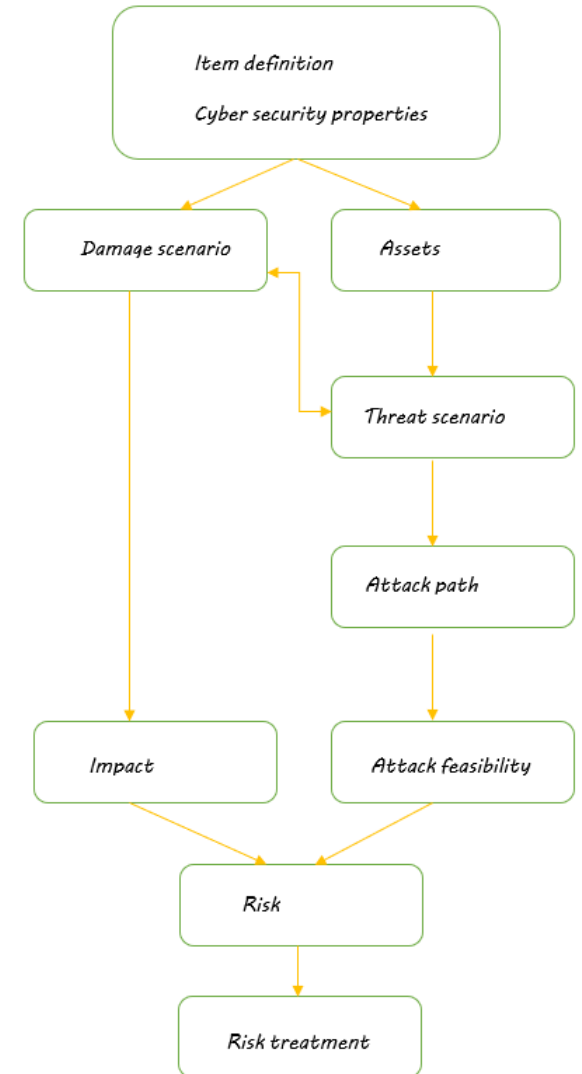
Autonomous driving (Illustrative). Photo: Shutterstock

Evaluating Cybersecurity with Attack Graphs

Threat Analysis and Risk Assessment

- ISO/SAE 21434 is the cybersecurity standard for road vehicles
- It mandates full lifecycle cybersecurity management, including the **Threat analysis and risk assessment**
- Systematic process to identify and treat risk
- Cybersecurity attacks can be combined into a “kill chain”
- Manually enumerating all possible attack paths in a complex system is practically impossible

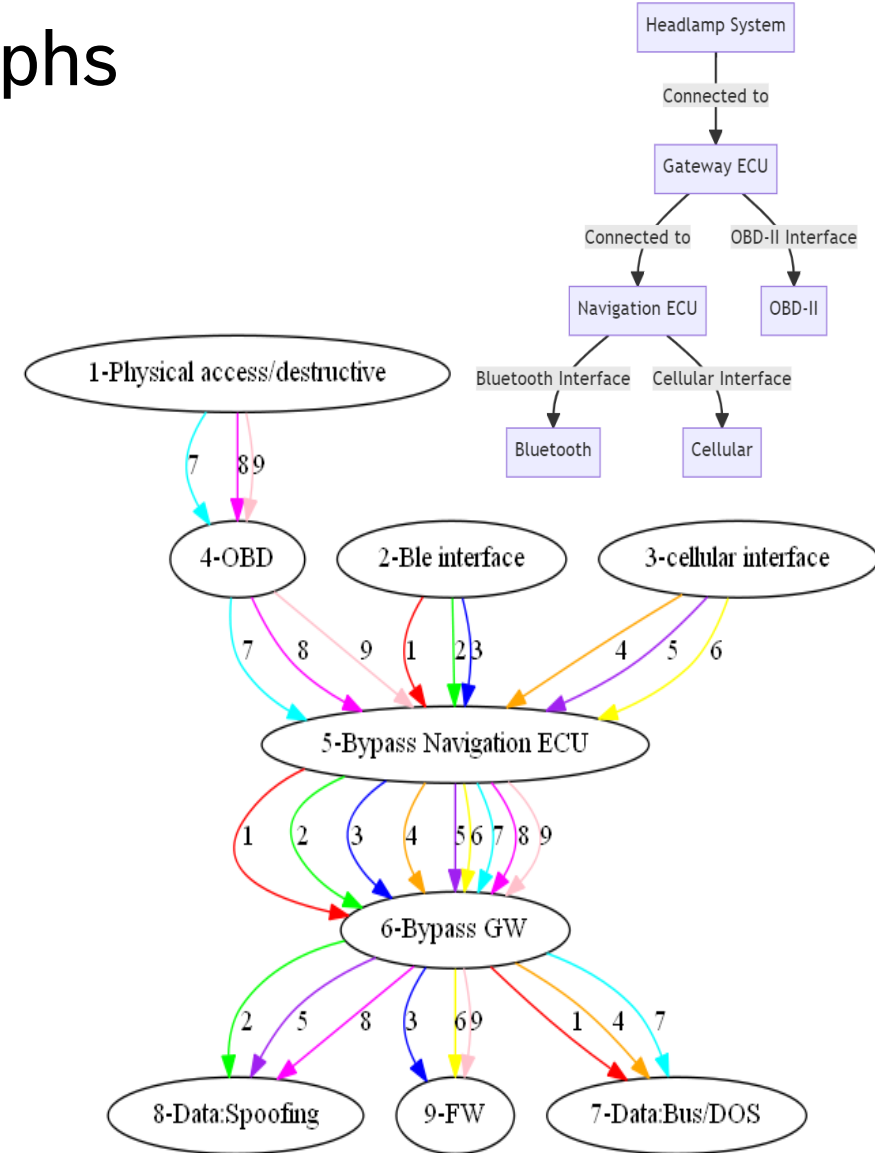
We need automation in the threat analysis



Evaluating Cybersecurity with Attack Graphs

Attack Graphs for TARA

- Understand the system composition
- Construct asset inventory
- Define external and internal interfaces
- Evaluate each attack step separately, in immediate context
- **Construct attack graph** and generate all possible paths
- Rank each path according to combined feasibility (likelihood)
- Consider security measures to reduce feasibility
- Add in damage scenarios to evaluate impact
- Risk is a combination of the impact with the feasibility



Industrial Examples of Graph Applications

Conclusions

- Graph computation methods are well suited to many industrial applications
- The basic concepts are available and accessible
- The challenge is to reformulate industrial questions into graph language and find the right model for the specific application

