



autoBAHN4mining

(semi-)autonomous mining trains pilot project@Salzwelten-Hallein, Salzburg

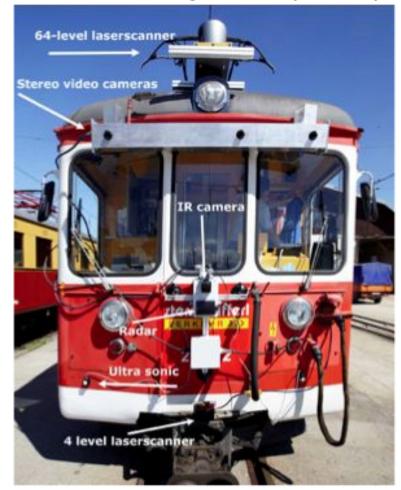
Felix Hörbinger, Wolfgang Pree CS@PLUS

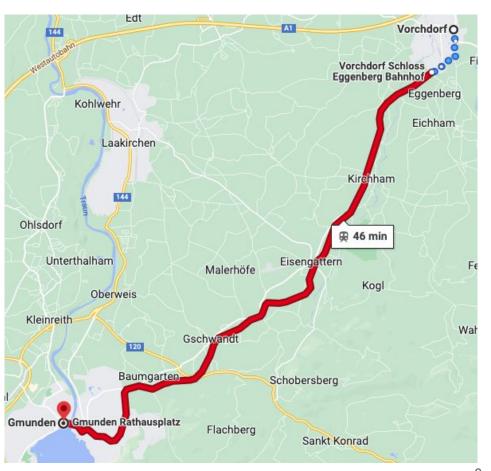


benefits for mines

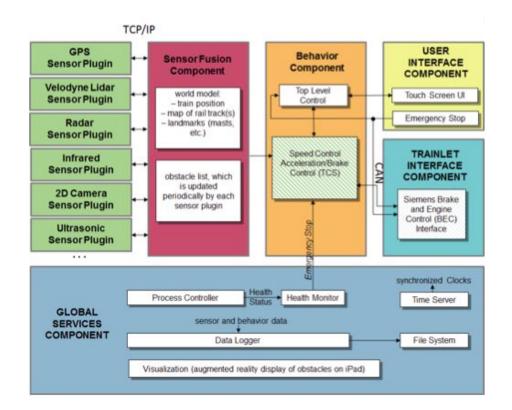
- improves safety
- reduces train driver costs by a factor of 20 or more (one human operator for 20-30 trains)
 - + counteracts potential/actual train driver shortages
- reduces infrastructure costs (maintenance of signals, etc.)
- can increase train utilization, potentially towards 24/7
- straight-forward and thus inexpensive to upgrade existing trains

how it all began ... (2008)





'conventional' implementation (NO deep learning)







mining trains—THE low-hanging fruit of autonomous driving

- rails define path where a train can move
- constant weather/climate—in all underground mines, all over the world
- train speed typically < 20 km/h
 - => requires obstacle detection only up to 50-70 meters

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- simple maps
- train position with accuracy of 1-2 meters sufficient
- liability can be significantly reduced by putting a human in the loop
- low probability of encountering an obstacle
 one human operator might monitor/control 20-30 trains, maybe many more
- controlled environment (only employees are in the mines)

self-driving mining train

:: self-driving truck/car

2 cameras + 1 basic lidar

6 high-end lidars + camera(s) + radar + (GPS + IMU)

+ sensors at rear





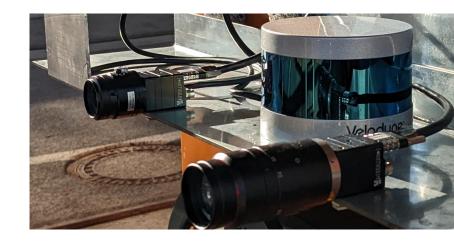
autoBAHN4mining@Salzwelten Hallein



state-of-the-art, off-the-shelve hardware

- Ximea high-resolution cameras
- Velodyne lidar (100 m range Puck)
- nVidia GPU (Jetson Nano Kit)





lean, cutting-edge software







- deep learning for sensor data processing
 - obstacle recognition
 - rail detection
- plug-in sensor fusion framework
 - core data structure: a list of objects classified according to 'obstacleness', forming the basis of a world model
 - allows any degree of sensor redundance
 - implemented with imperative programming language Java
 - => can be formally verified

behavior component

ML-based image processing

ML-based point cloud processing

world model:

obstacles, classified in dangerous and unproblematic ones

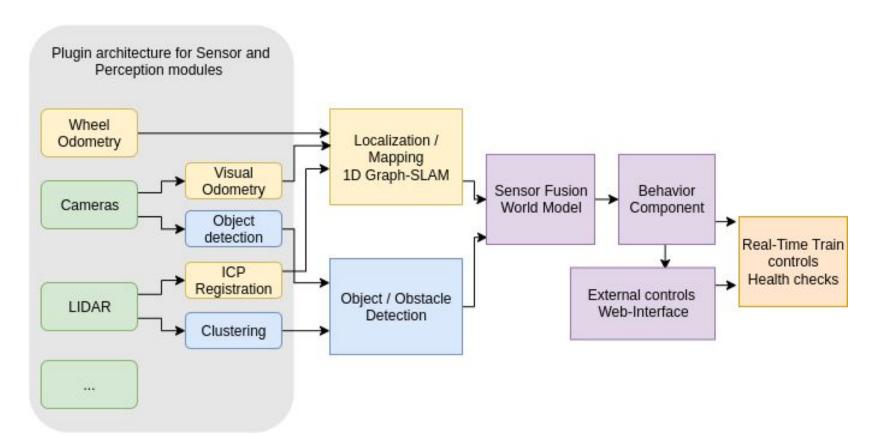
landmarks

train position on map

obstacles

speed control/braking/emergency braking

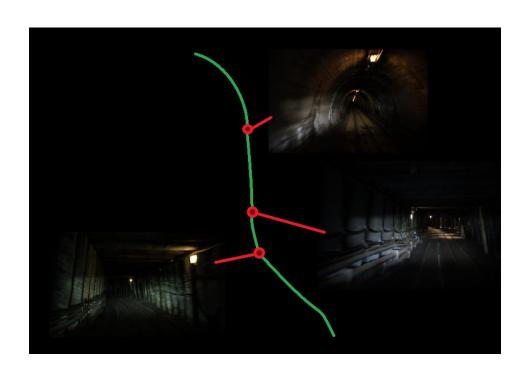
a closer look at the autoBAHN4mining architecture



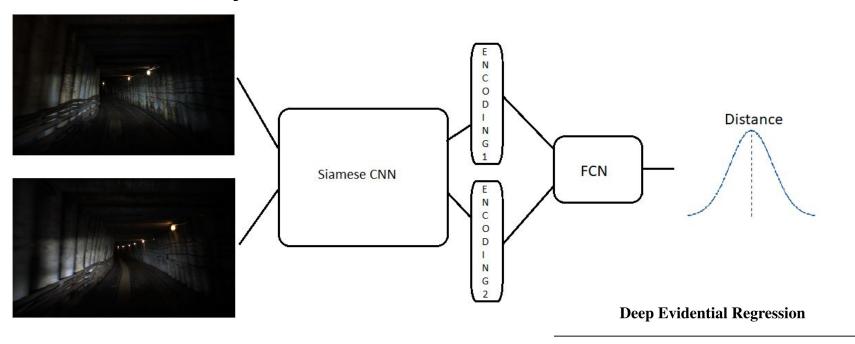
Mapping

Problems:

- no GPS/GNSS signals
- tracks can be slippery odometer not accurate
- compass disturbed by mineral deposits and the train itself
- tunnels very uniform—visual or LIDAR odometry hard

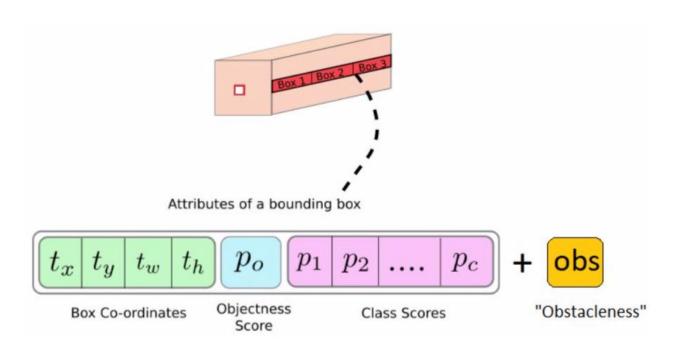


Visual Odometry

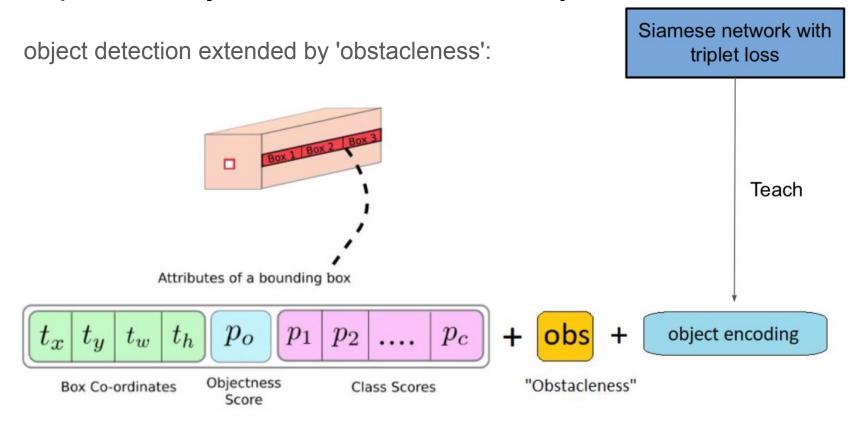


improved object detection

object detection extended by 'obstacleness':

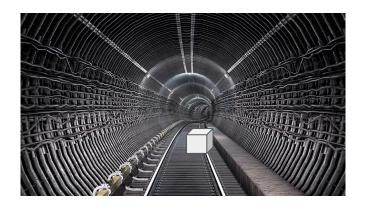


improved object detection + similarity metric



additional autoBAHN4mining tools/apps

virtual reality simulation (Unity 3)



remote monitoring & control by a human (Dart/Flutter app)



some future research tasks

- test the generalization to new object classes on the tracks
- adversarial attacks on "obstacleness" for better understanding of network data augmentation with Generative Adversarial Networks (GANs)
- comprehensive tests based on the simulator

Thanks for your attention!