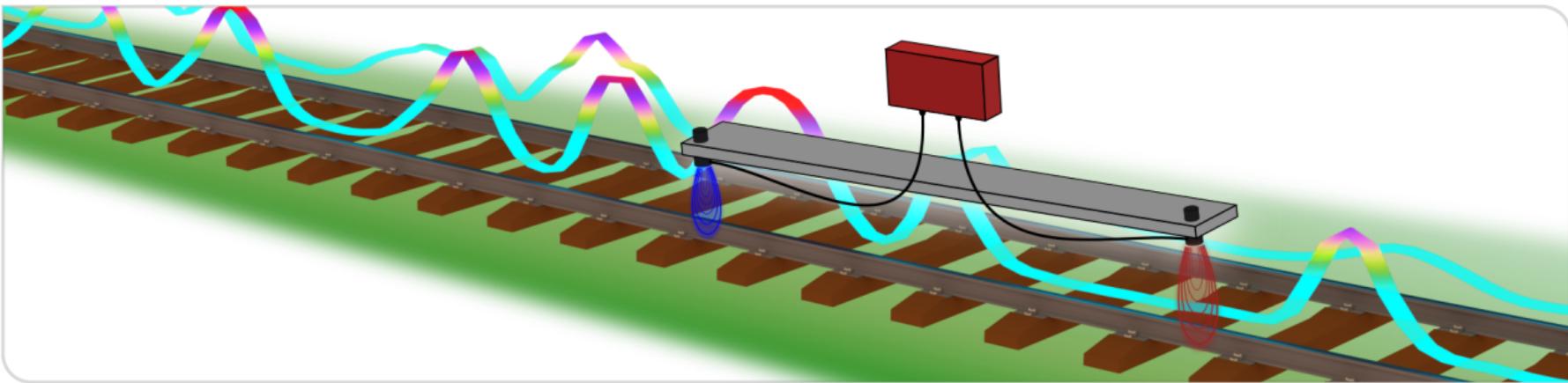


# Localization of Railway Vehicles using the Ferromagnetic Fingerprint of Rails

2nd iLoc Workshop, ITSC23

Bernd Kröper | 24.09.2023



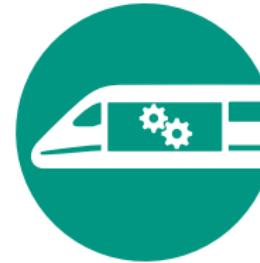
# Motivation

accurate and  
highly available



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pure onboard,  
stand-alone

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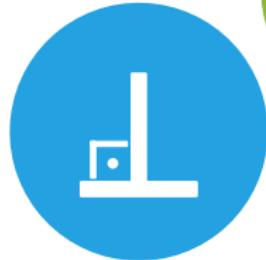
DIS: Difference  
Inductance Sensor

pure onboard,  
stand-alone

# Motivation

accurate and  
highly available

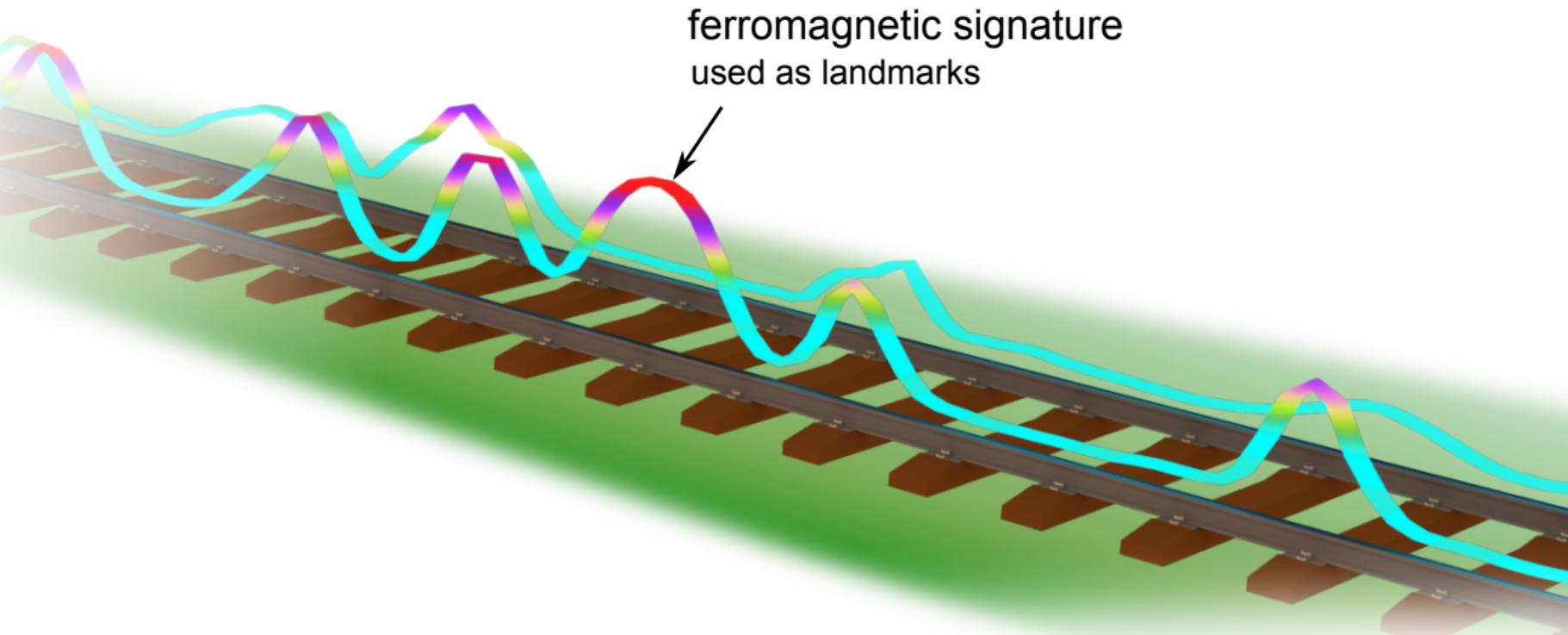
orthogonal  
measurement  
principle



DIS: Difference  
Inductance Sensor

pure onboard,  
stand-alone

# Difference Inductance Sensor (DIS)



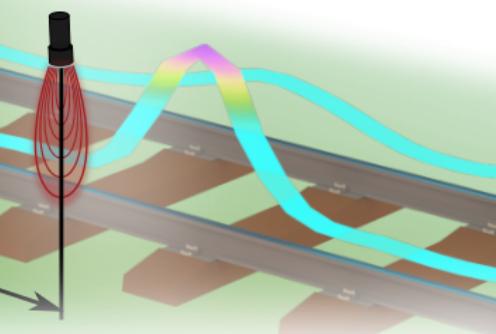
# Difference Inductance Sensor (DIS)



ferromagnetic signature  
used as landmarks

sensor head A & B  
pickup ferromagnetic  
signature below

$d$



# Difference Inductance Sensor (DIS)

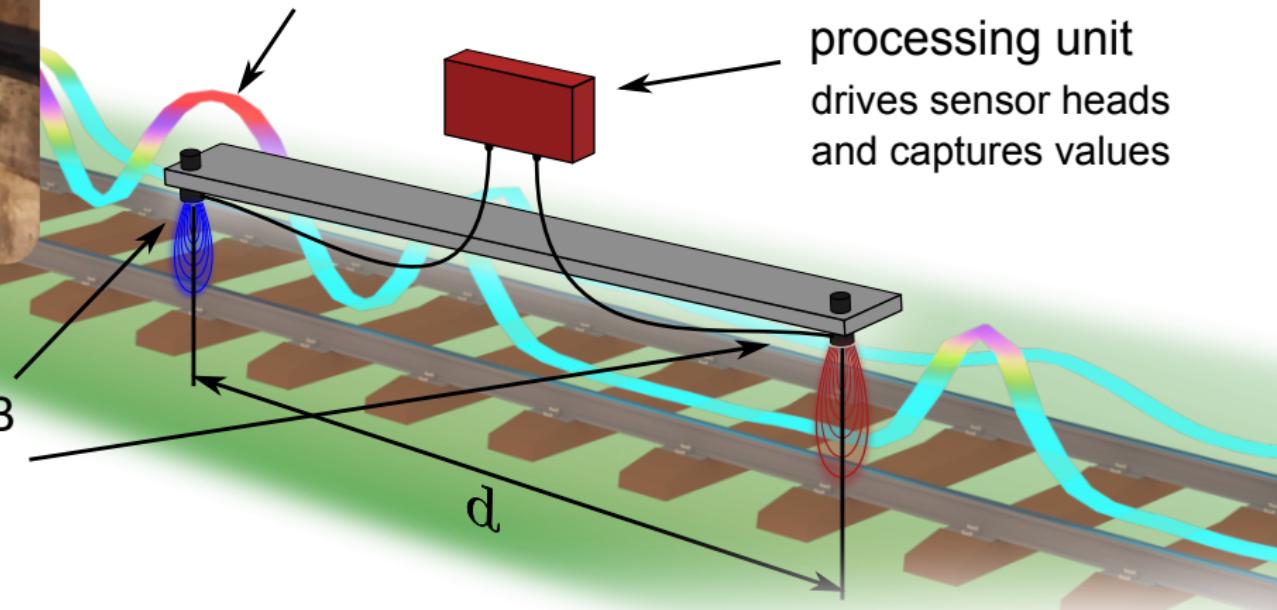


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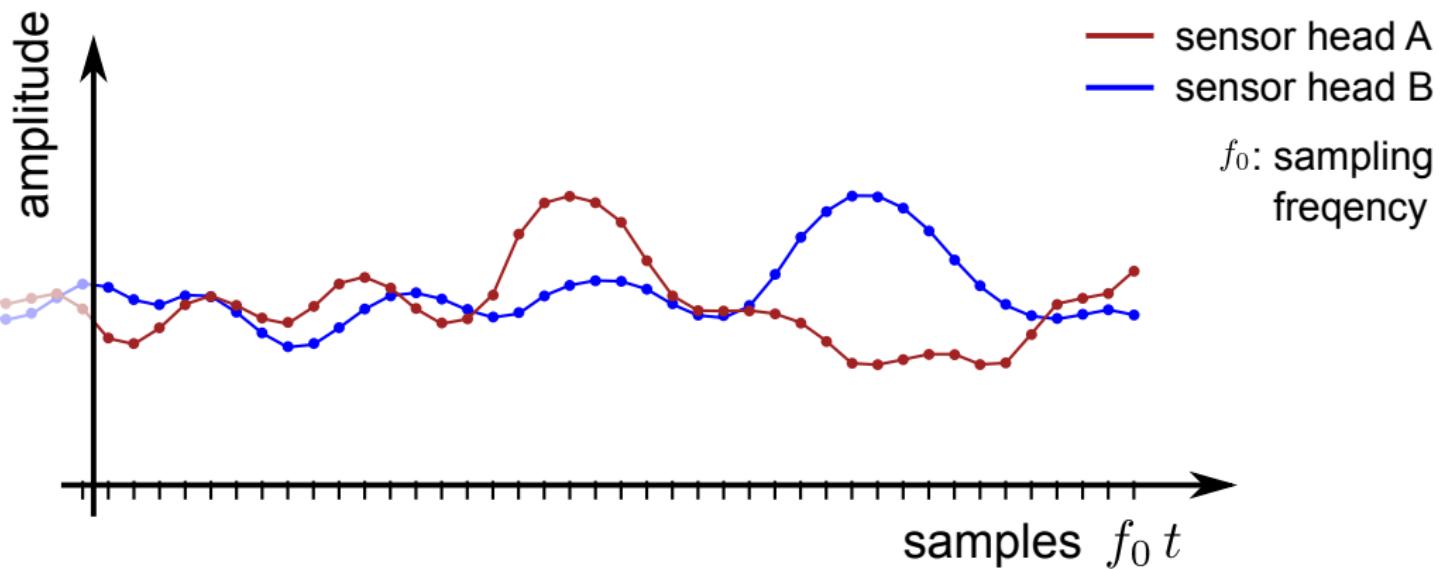
ferromagnetic signature  
used as landmarks

processing unit  
drives sensor heads  
and captures values

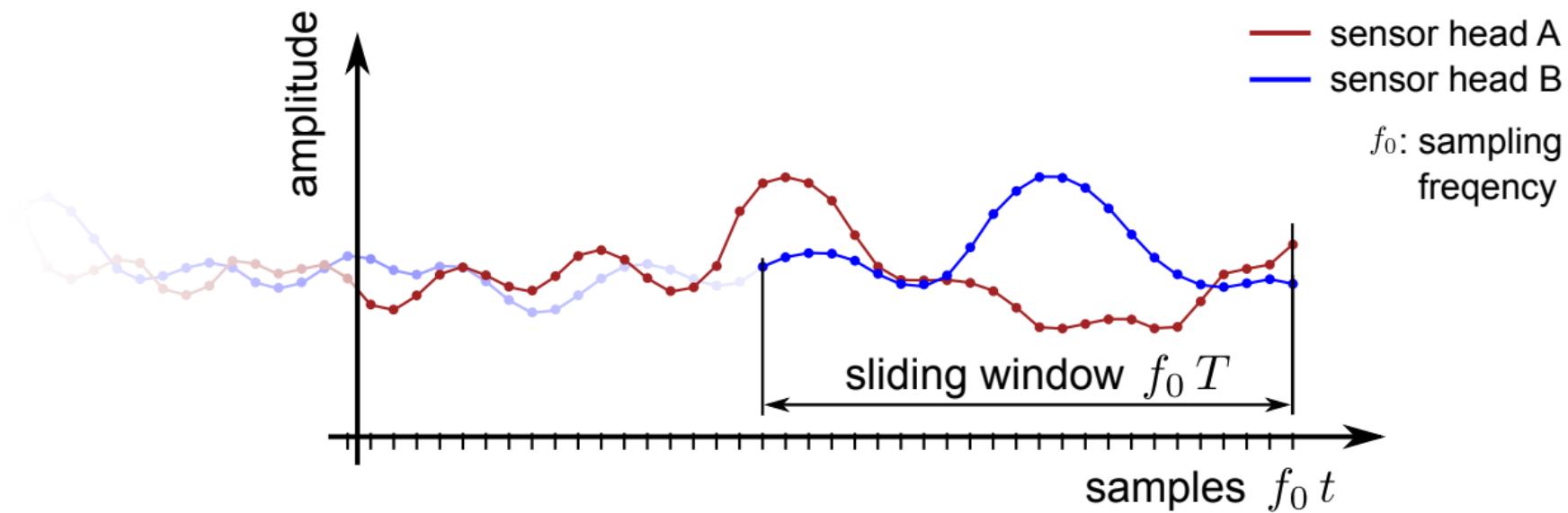
$d$



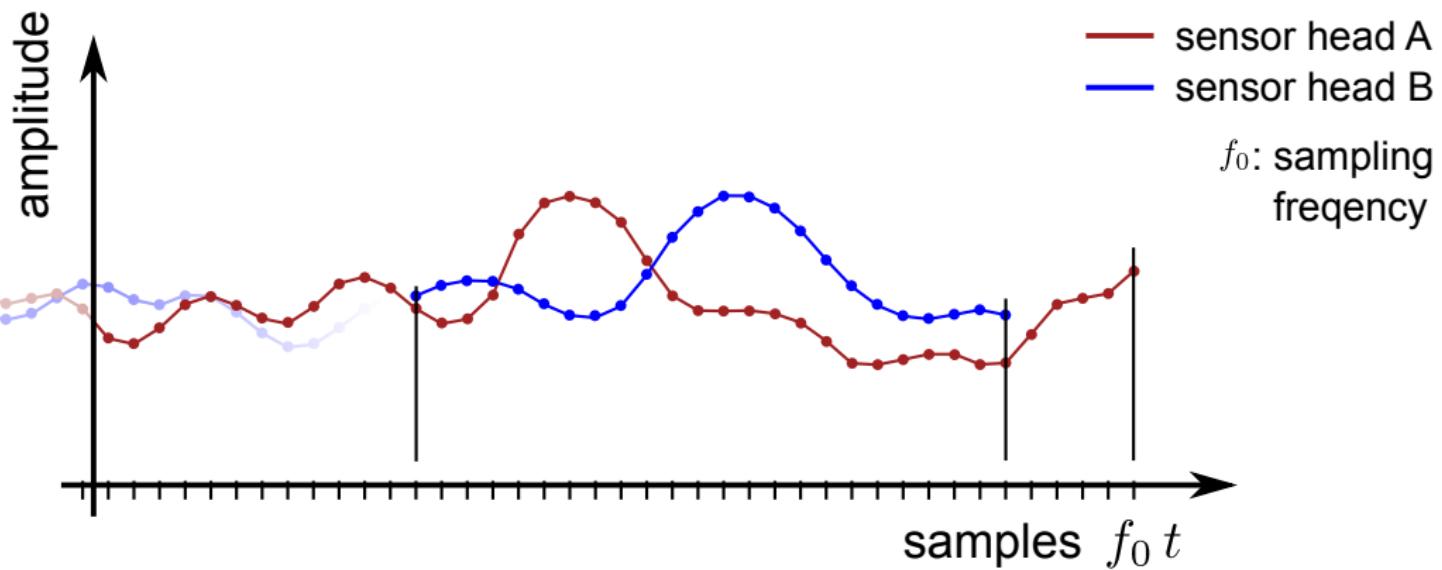
# Velocity Estimation



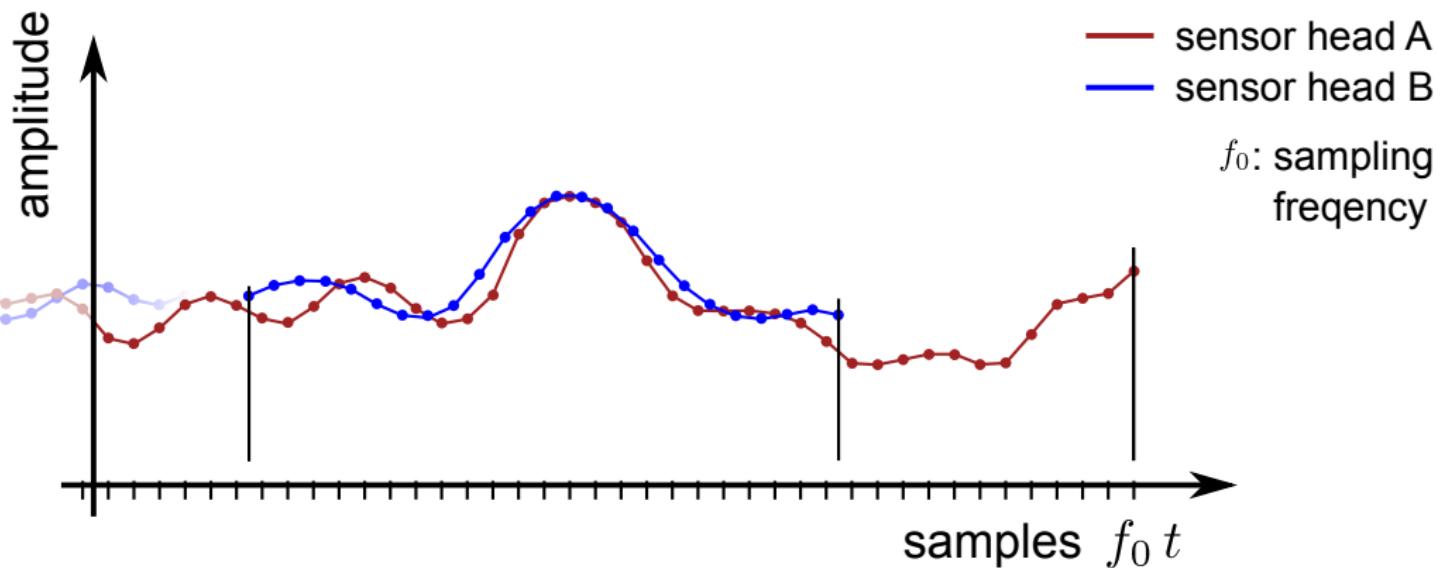
# Velocity Estimation



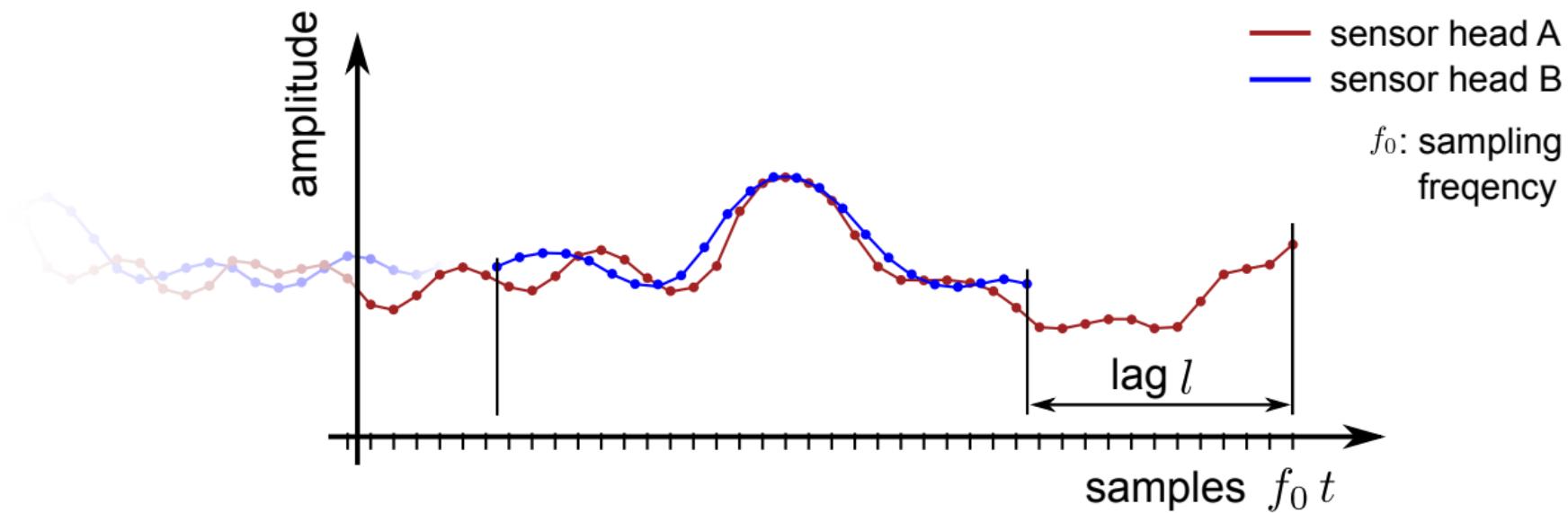
# Velocity Estimation



# Velocity Estimation

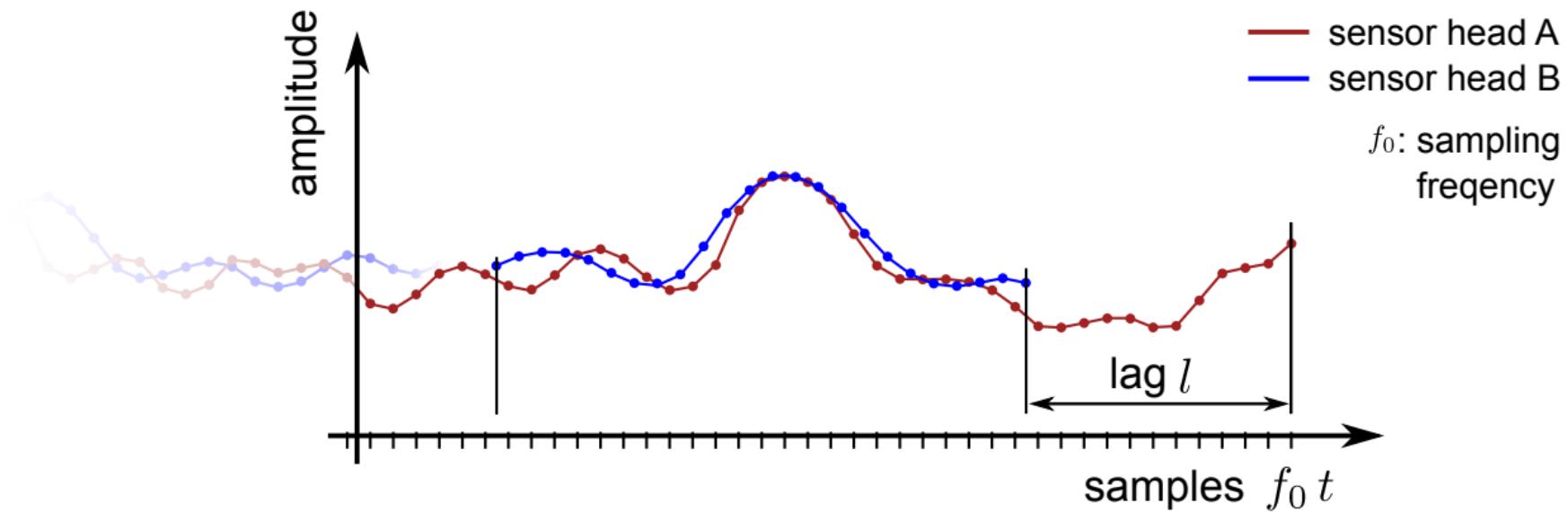


# Velocity Estimation



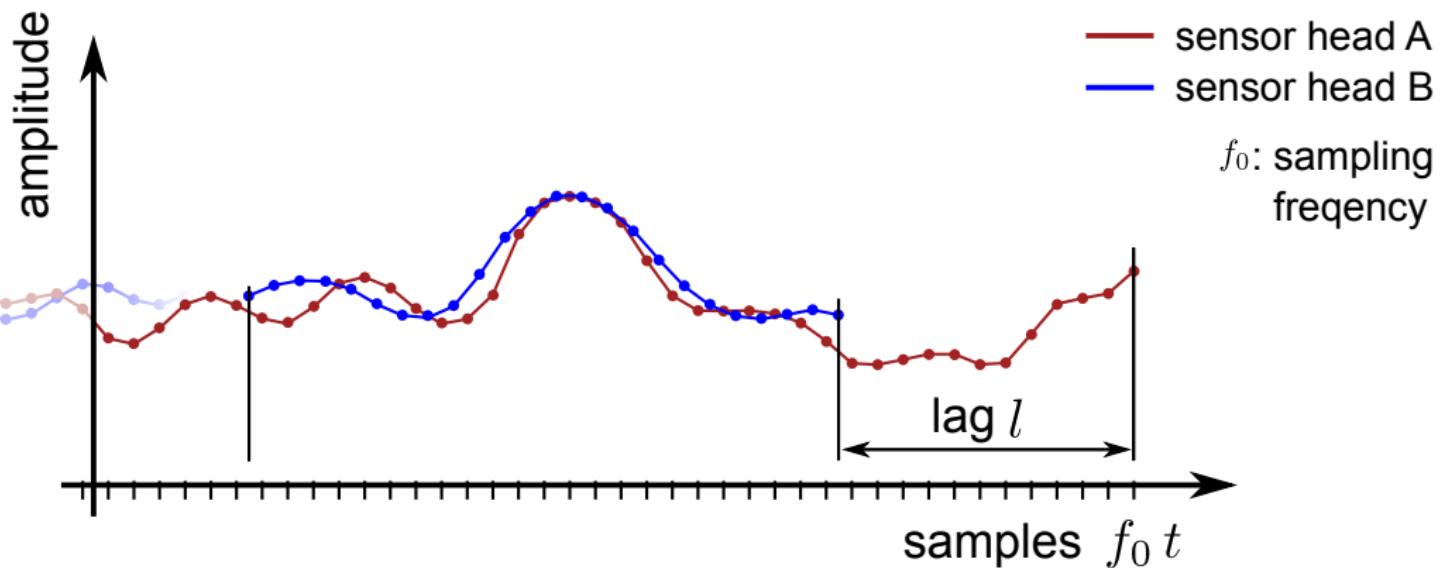
- $l$  gives current velocity, found by maximizing cross-correlation

# Velocity Estimation

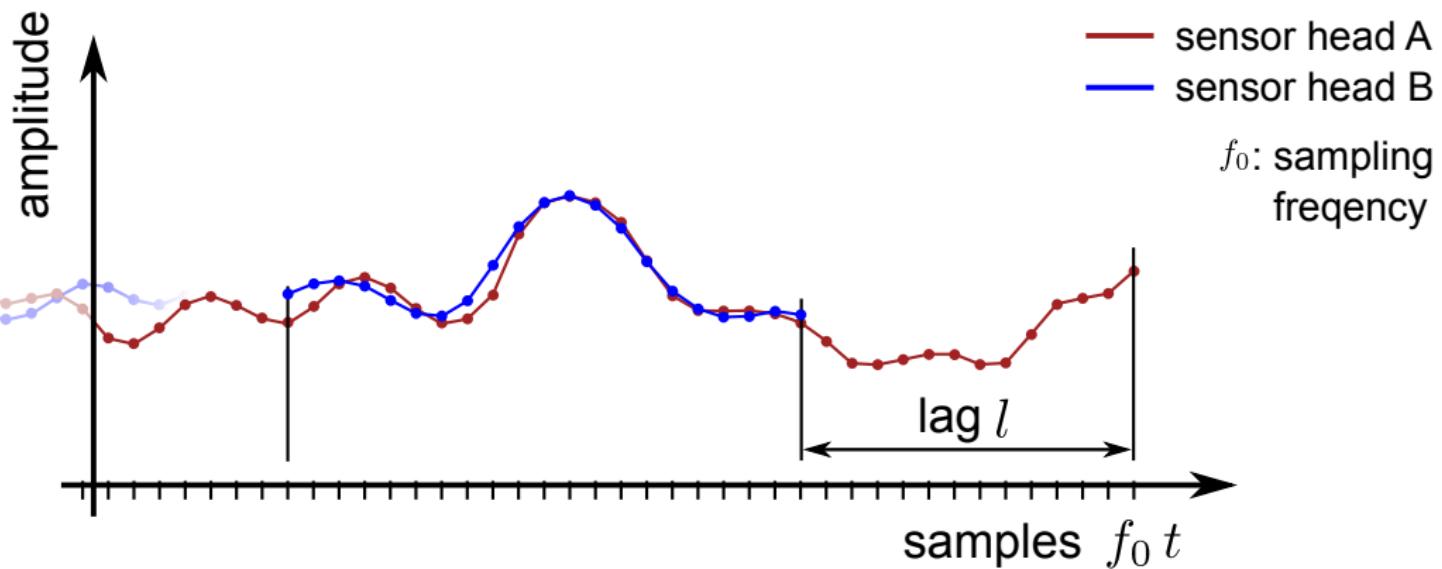


- $l$  gives current velocity, found by maximizing cross-correlation
- good solution for constant velocity

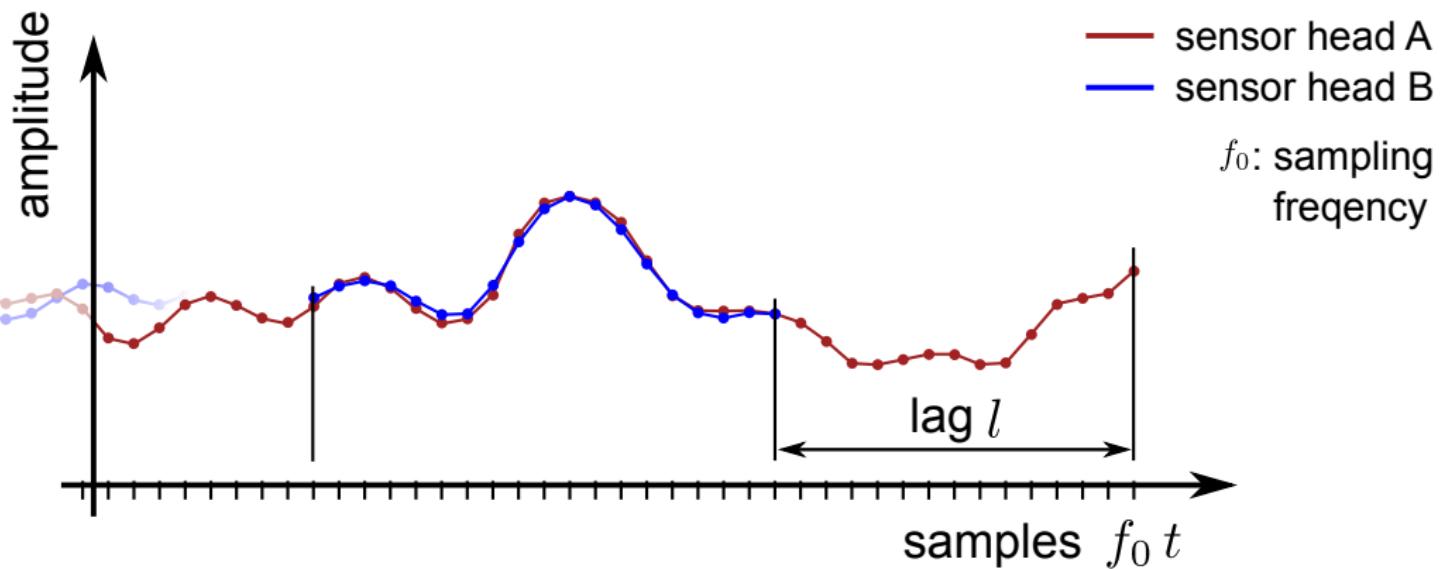
# Acceleration Estimation



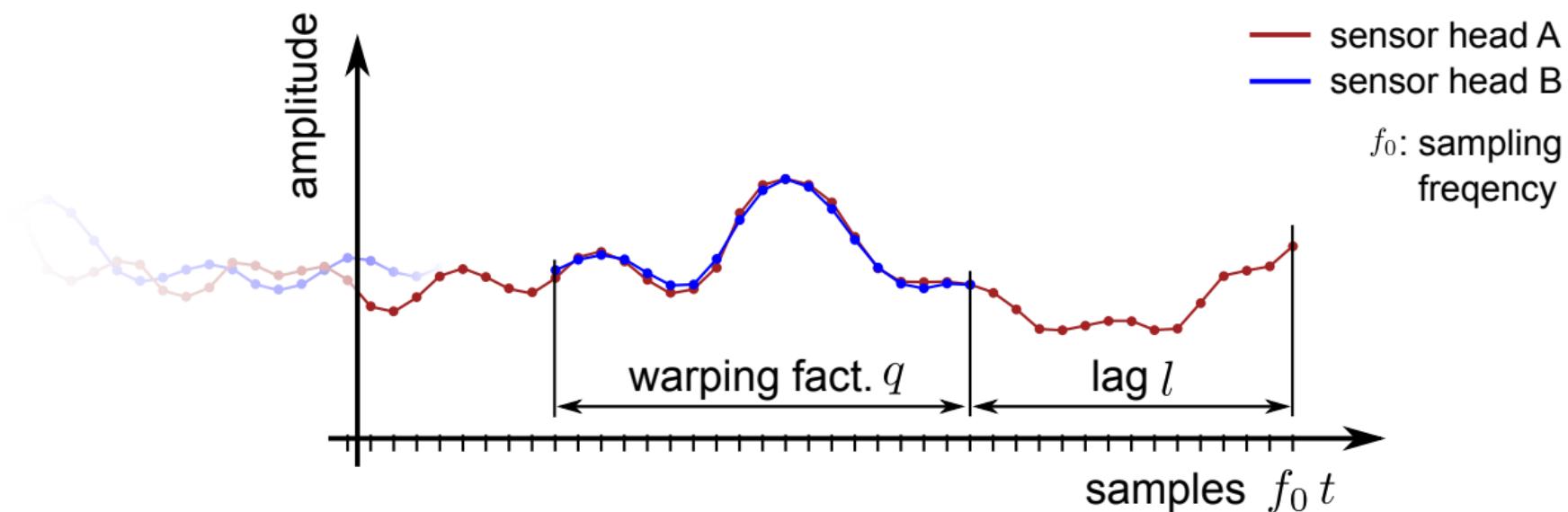
# Acceleration Estimation



# Acceleration Estimation

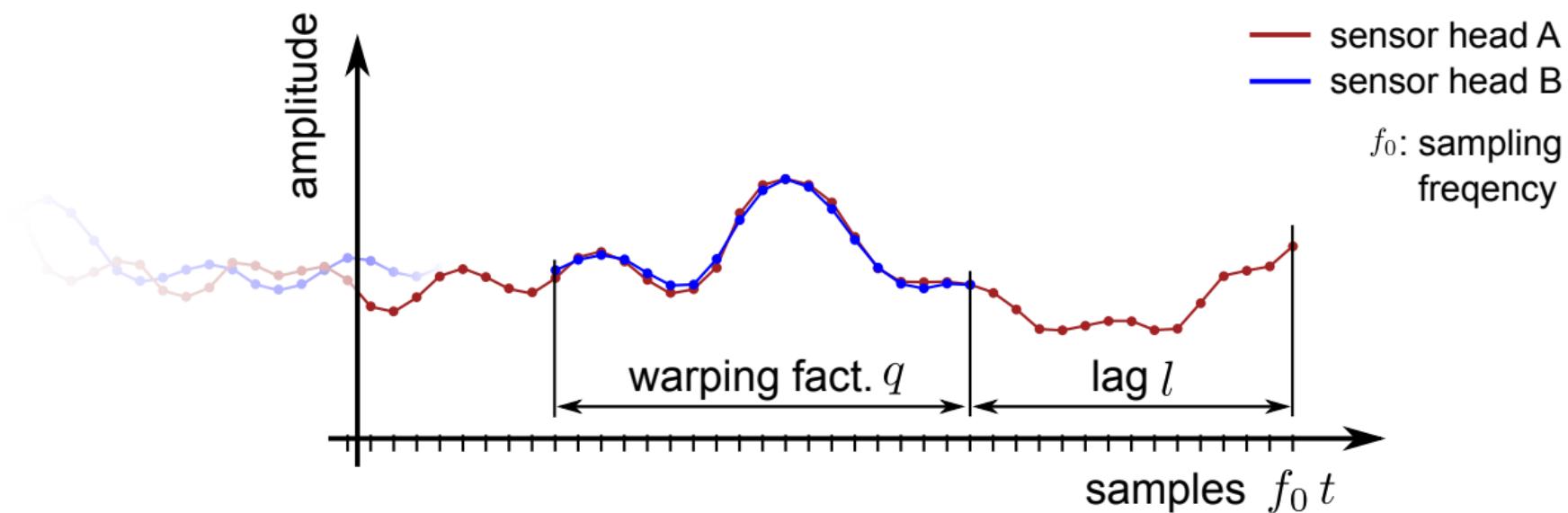


# Acceleration Estimation



- $q$  gives current acceleration, found by maximizing cross-correlation

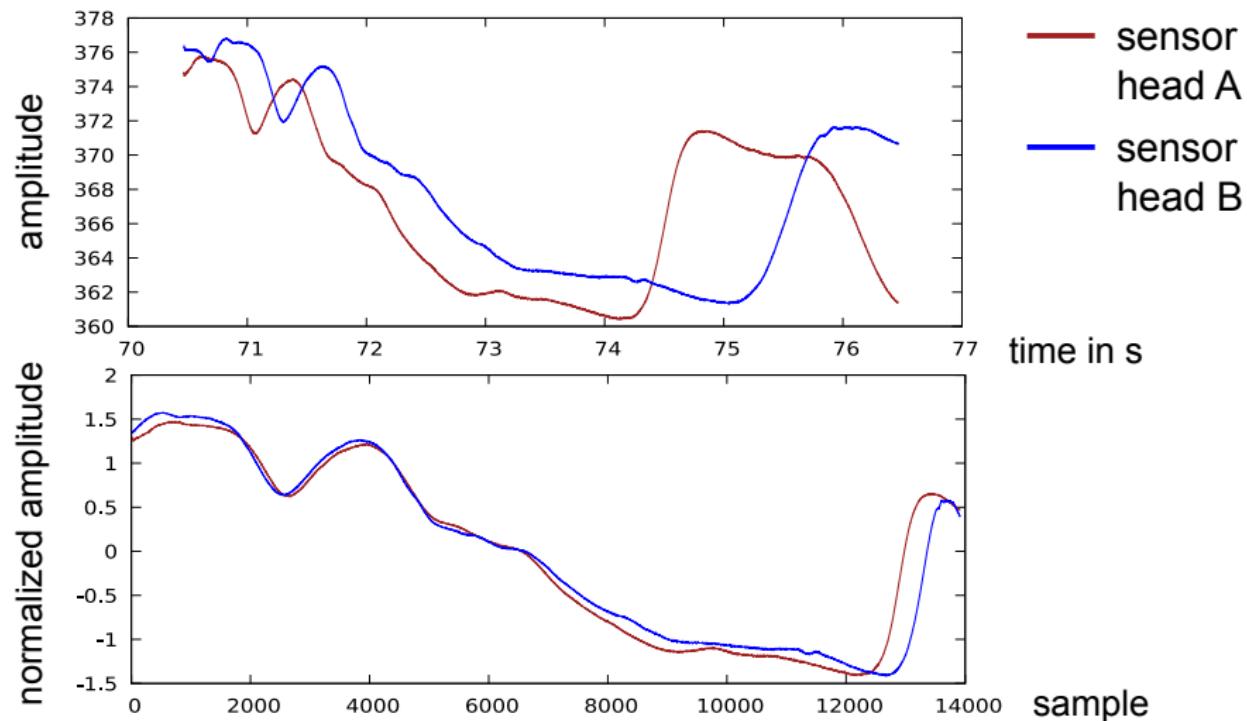
# Acceleration Estimation



- $q$  gives current acceleration, found by maximizing cross-correlation
- finite state machine to handle standstill

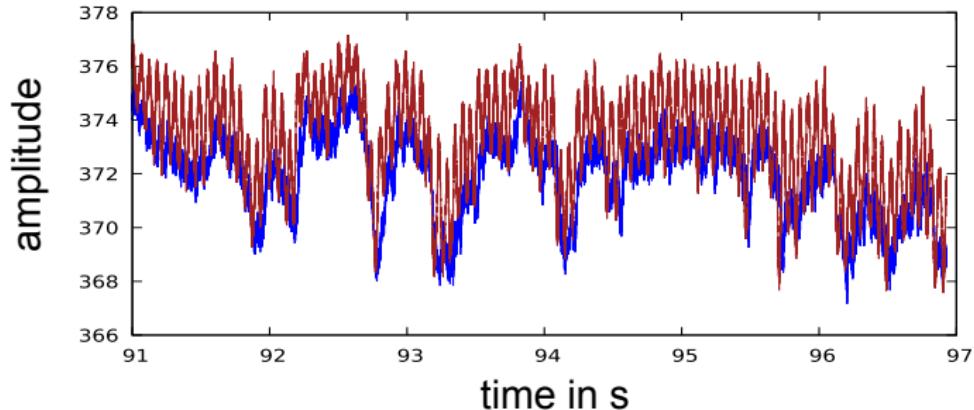
# Velocity and Acceleration Estimation

- sample of measured ferromagnetic signals
- signals after normalization, shifting and warping



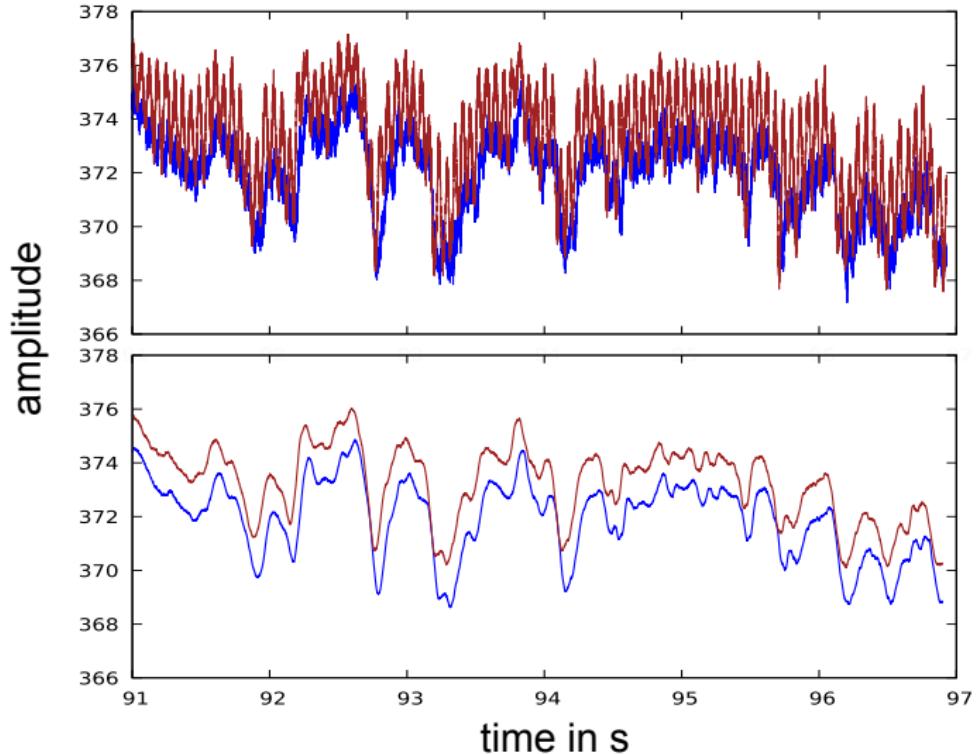
# Noise reduction

- raw sensor data with noise  
e.g. quantization, electrical components with 16.7 Hz



# Noise reduction

- raw sensor data with noise  
e.g. quantization, electrical components with 16.7 Hz
- significant reduction using a lowpass filter (moving average)

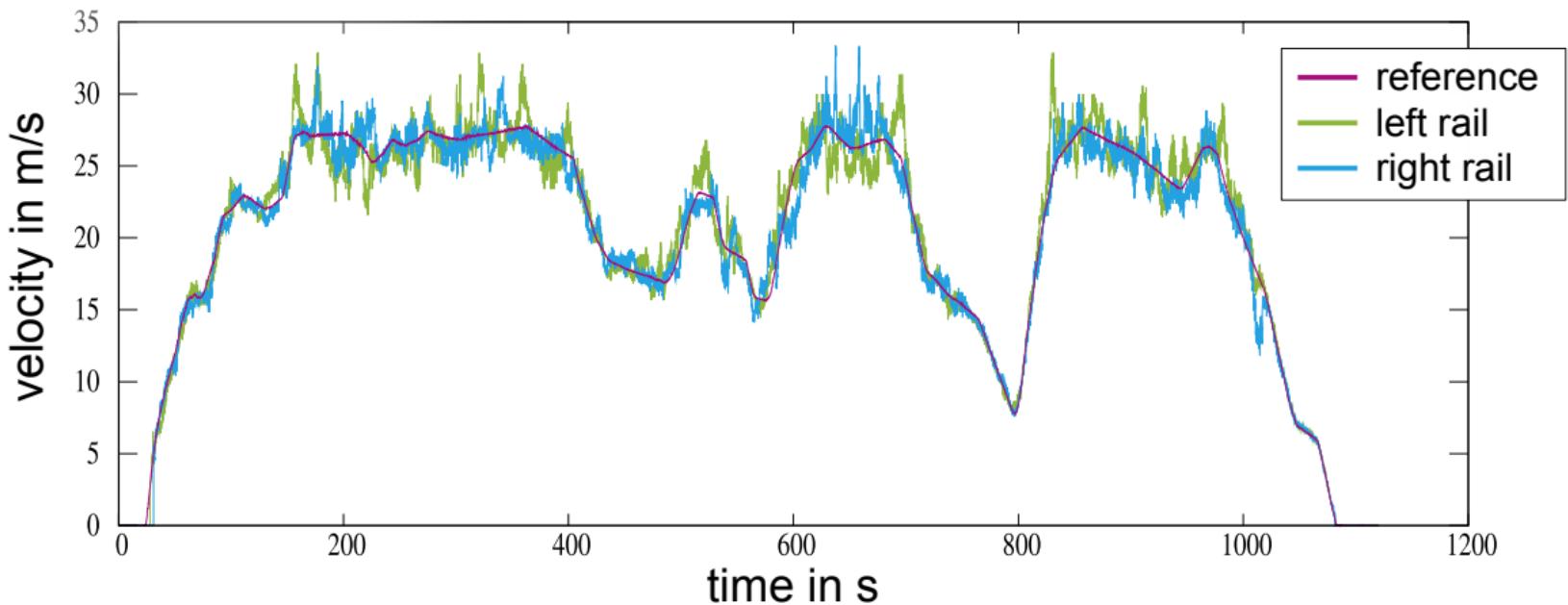


# Evaluation Setup

- mounted on railway vehicle
  - two sensors over both rails
  - distance to rail = 100 mm
  - $d = 553$  mm
  - sampling frequency  $f_0 = 30$  kHz
- reference sensor:  
GNSS+INS+Odometer
- 130 km in operating traffic
- velocity up to 100 km/h



# Evaluation of Velocity Estimation

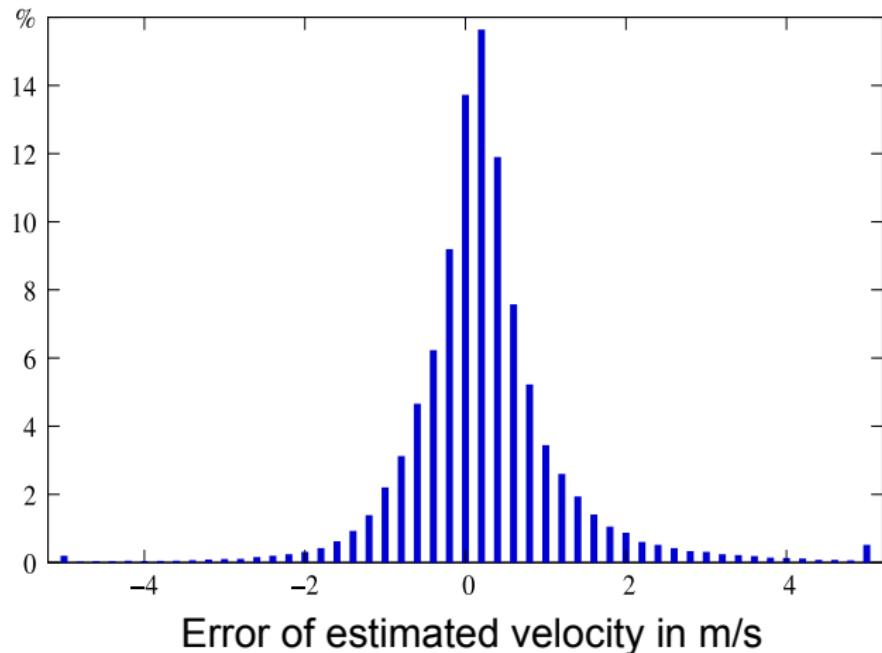


- above 0.1 m/s almost always a velocity was esitimated

# Evaluation of Velocity Estimation

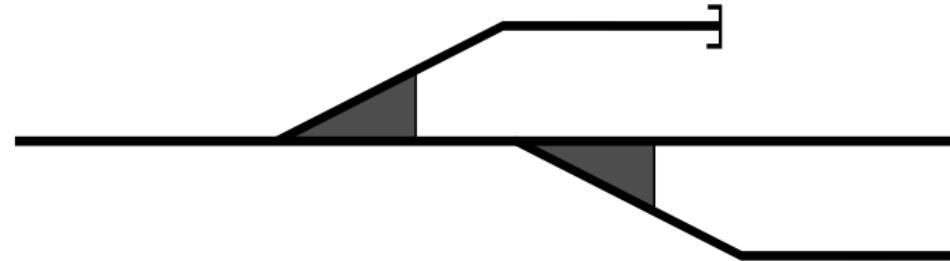
summary of errors over 130 km:

- 0.2 %: poor signal quality
- 1.0 %: no sufficient movement
- 98.8 %: accurate estimation
  - 80 %: error < 1 m/s
  - 93 %: error < 2 m/s
  - 97 %: error < 3 m/s



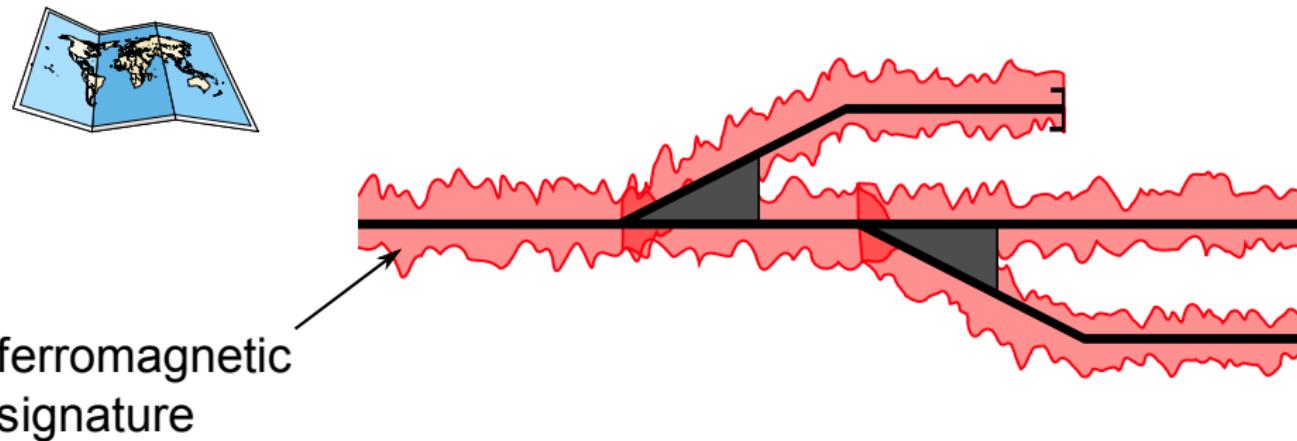
# Ferromagnetic Map

- Localization using ferromagnetic signatures is based on a map



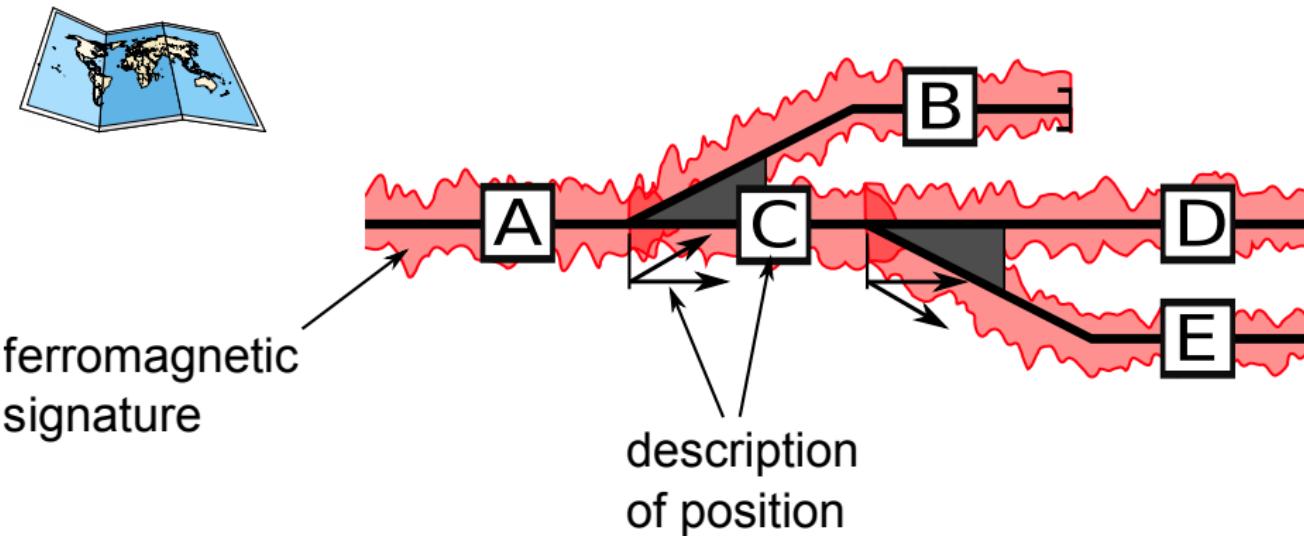
# Ferromagnetic Map

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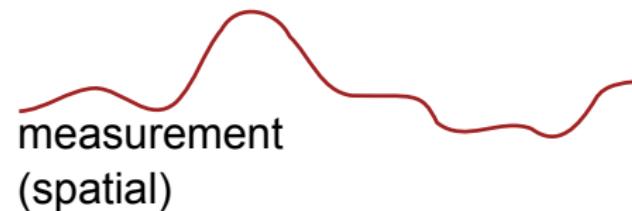
# Ferromagnetic Map

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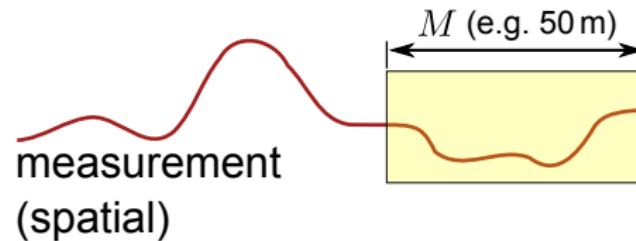
# Localization

- convert measured signals from time domain into spatial domain



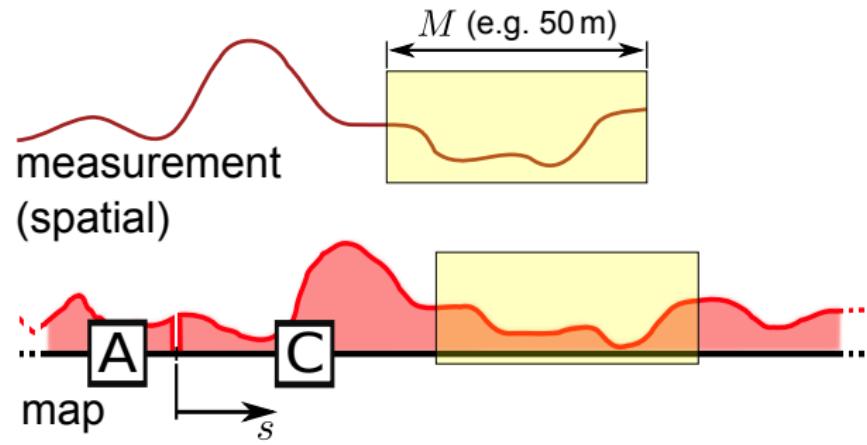
# Localization

- convert measured signals from time domain into spatial domain
- take most recent values of the signal as template



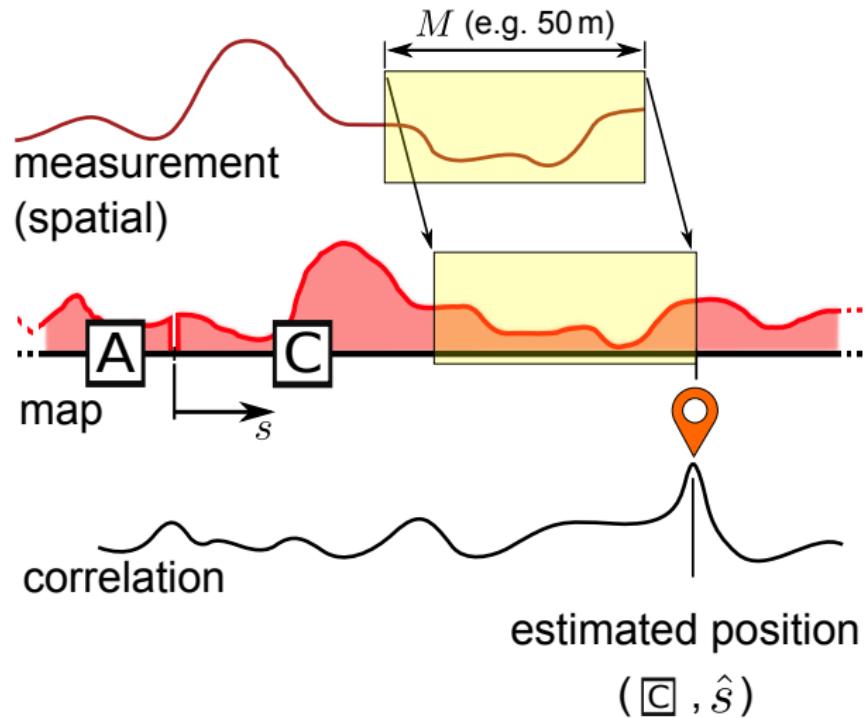
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- convert measured signals from time domain into spatial domain
- take most recent values of the signal as template
- search for template in map



# Localization

- convert measured signals from time domain into spatial domain
- take most recent values of the signal as template
- search for template in map
- take the position with the highest similarity

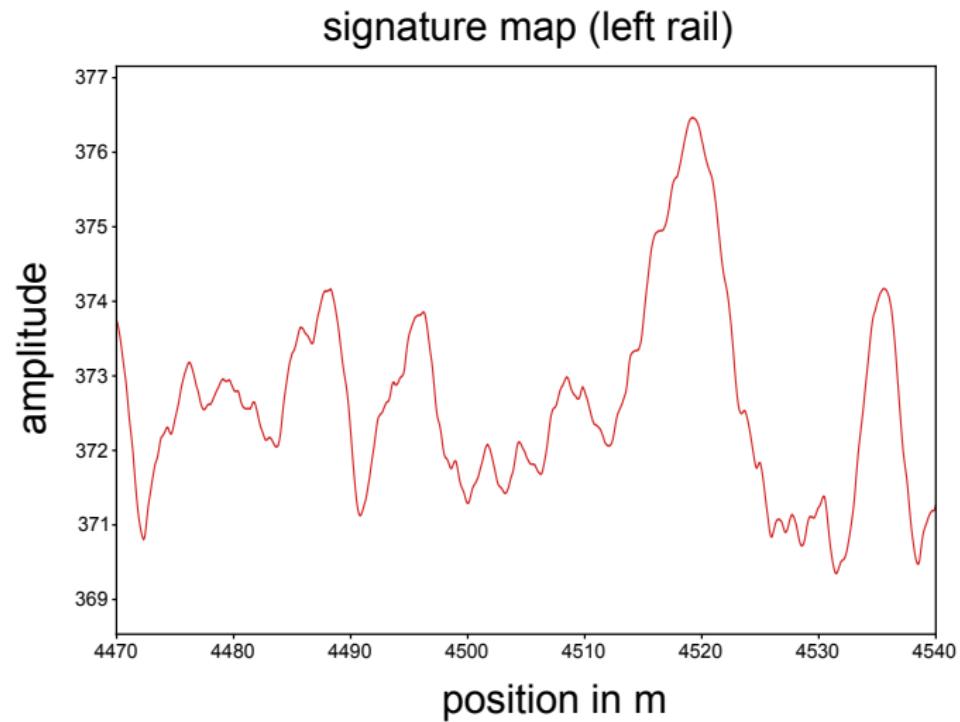


# Evaluation of Position Estimation

- evaluation on section of 12 km that was passed four times

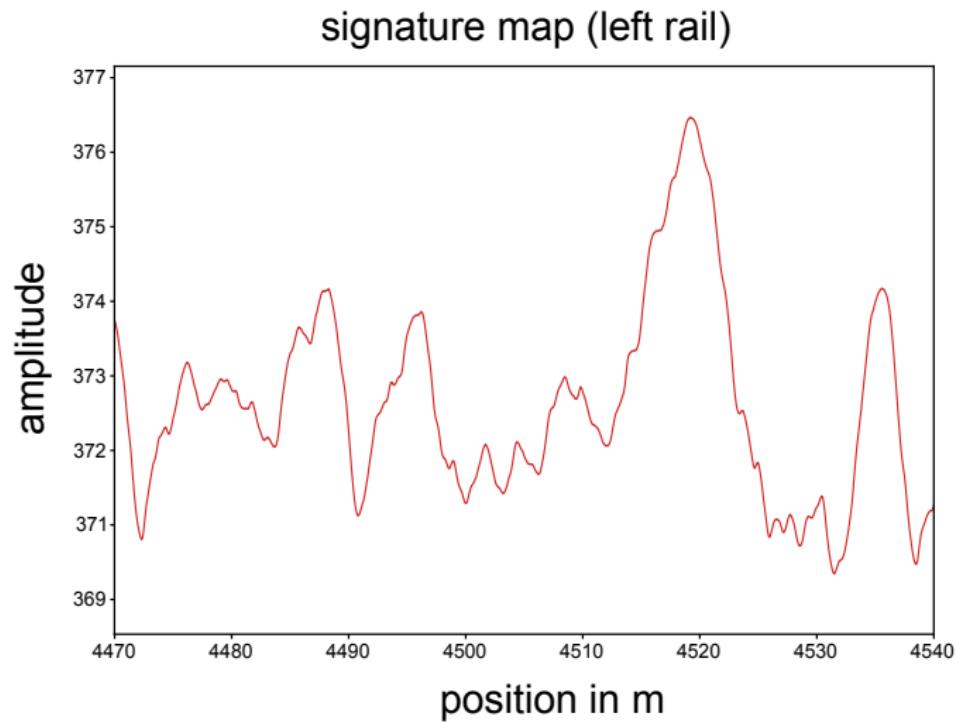
# Evaluation of Position Estimation

- evaluation on section of 12 km that was passed four times
- one drive used for mapping (sampled at 10 cm)



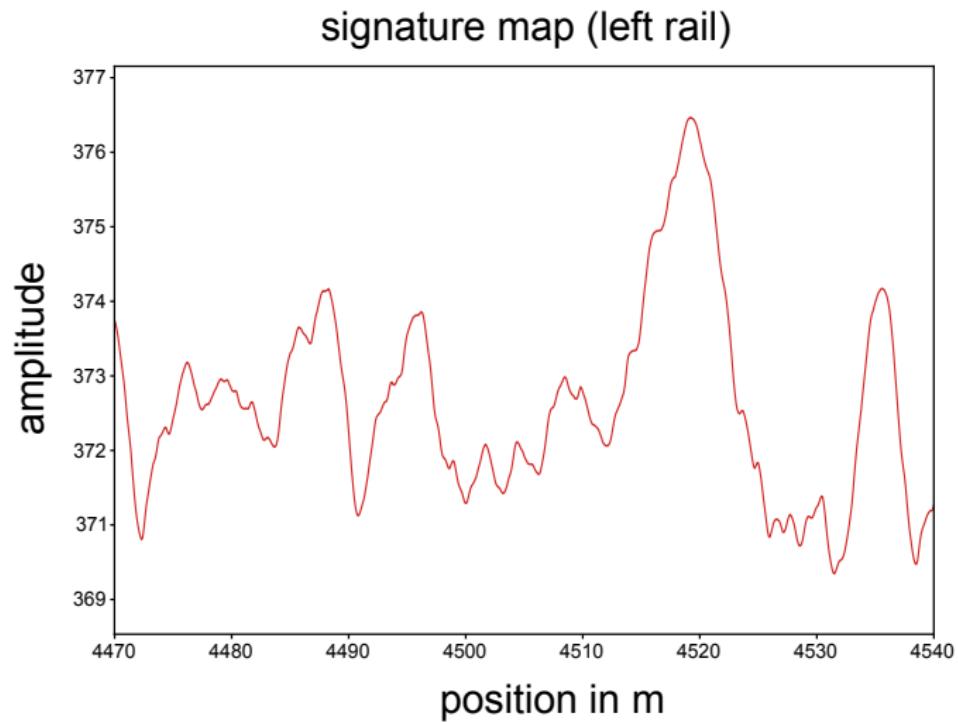
# Evaluation of Position Estimation

- evaluation on section of 12 km that was passed four times
- one drive used for mapping (sampled at 10 cm)
- remaining three drives used for localization  
(template length  $M = 50$  m)



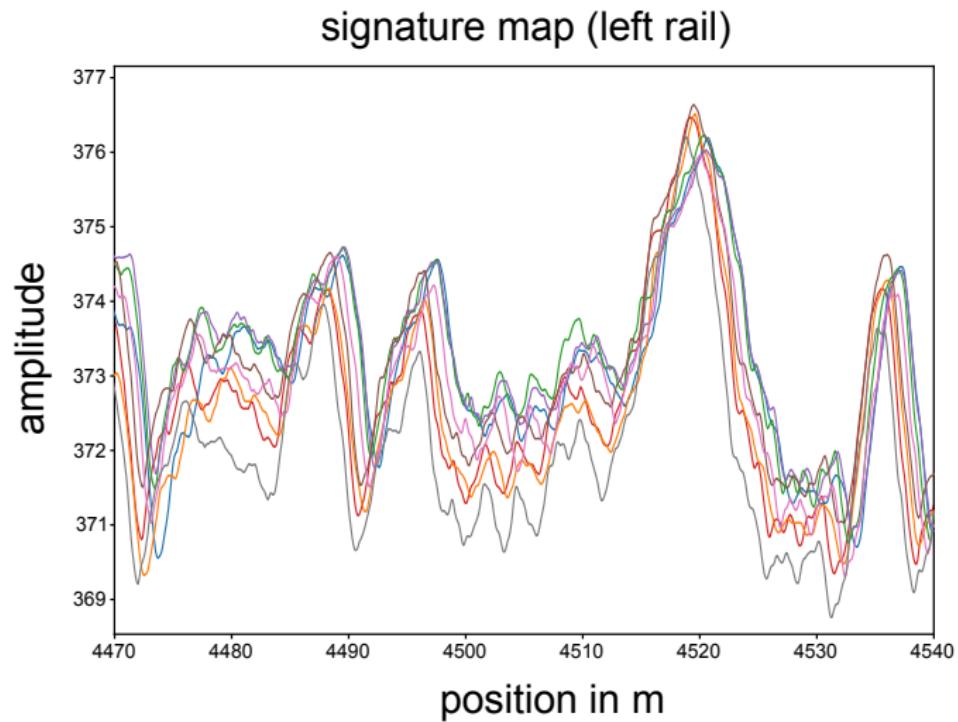
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- evaluation on section of 12 km that was passed four times
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- separately for left and right rail



# Evaluation of Position Estimation

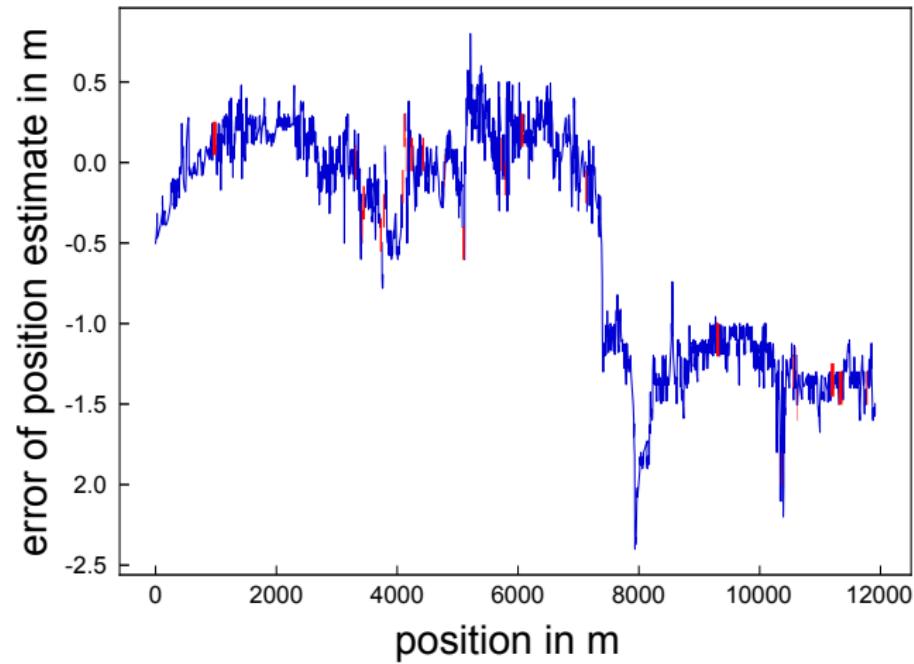
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# Evaluation of Position Estimation

For each localization drive:

- localization error estimated at 2500 equally distributed points



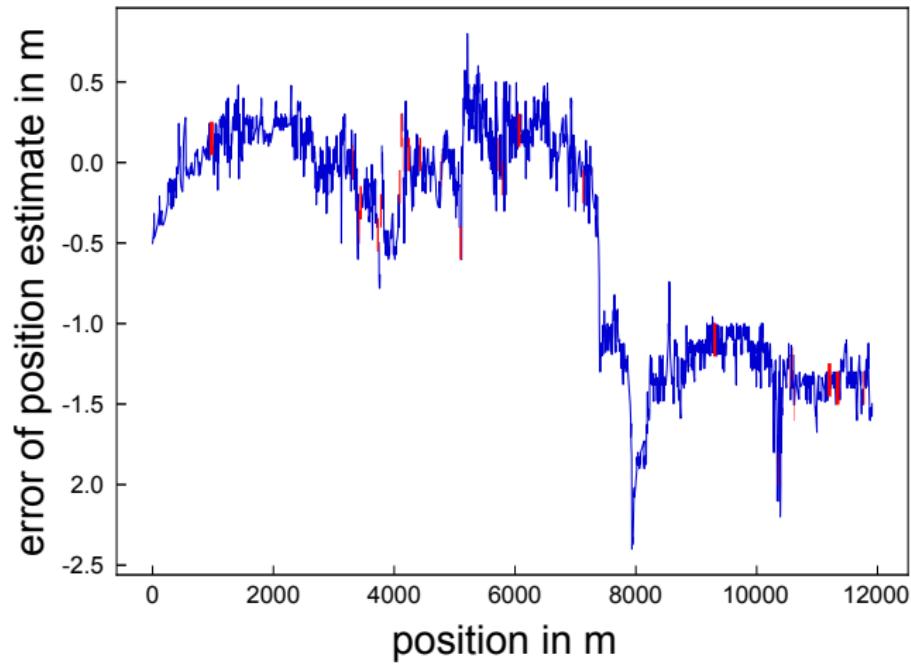
# Evaluation of Position Estimation

For each localization drive:

- localization error estimated at 2500 equally distributed points

Two types of error:

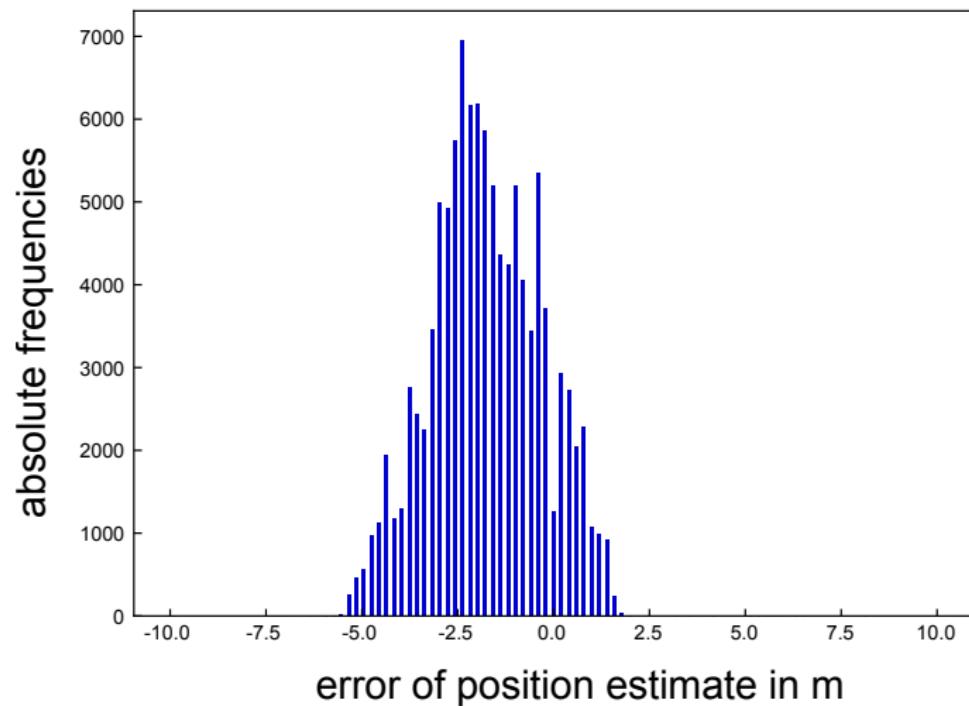
- small (< 10 m): synchronization with reference, imprecise mapping
- large: pattern mismatch, can be detected



# Evaluation of Position Estimation

histogram of errors of all experiments

- almost Gaussian
- mean of  $-2.2\text{ m}$  due to synchronization with reference
- $95\% < 10\text{ m}$



# Conclusion



localization without any changes  
to existing infrastructure

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method allows highly available and accurate  
velocity ( $97\% < 3 \text{ m/s}$ ) and position ( $95\% < 10 \text{ m}$ )

# Conclusion



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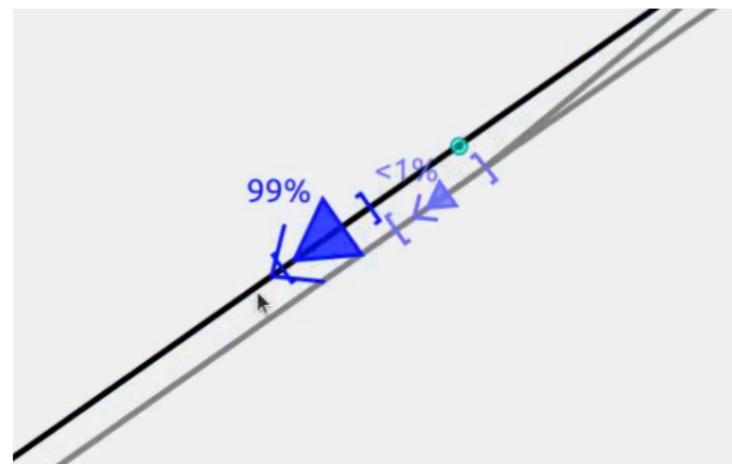
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high potential to improve existing systems

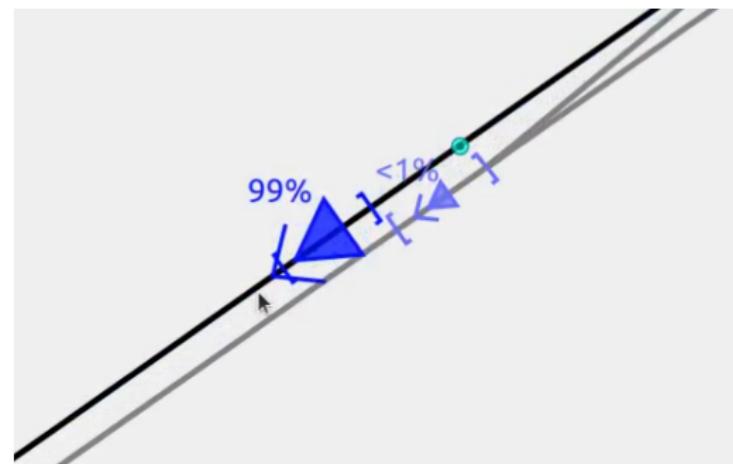
# Work in progress

- combine velocity and position estimation



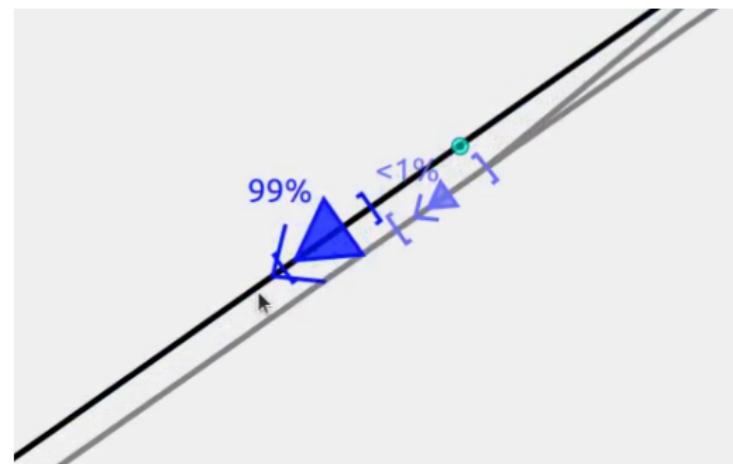
# Work in progress

- combine velocity and position estimation
- combine both rails



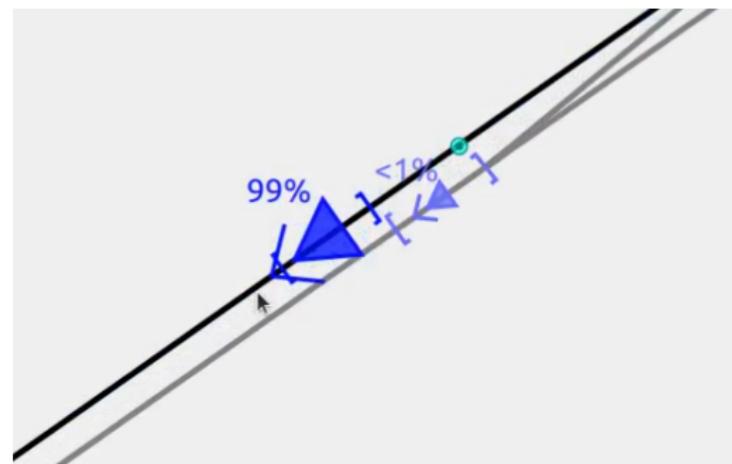
# Work in progress

- combine velocity and position estimation
- combine both rails
- online processing

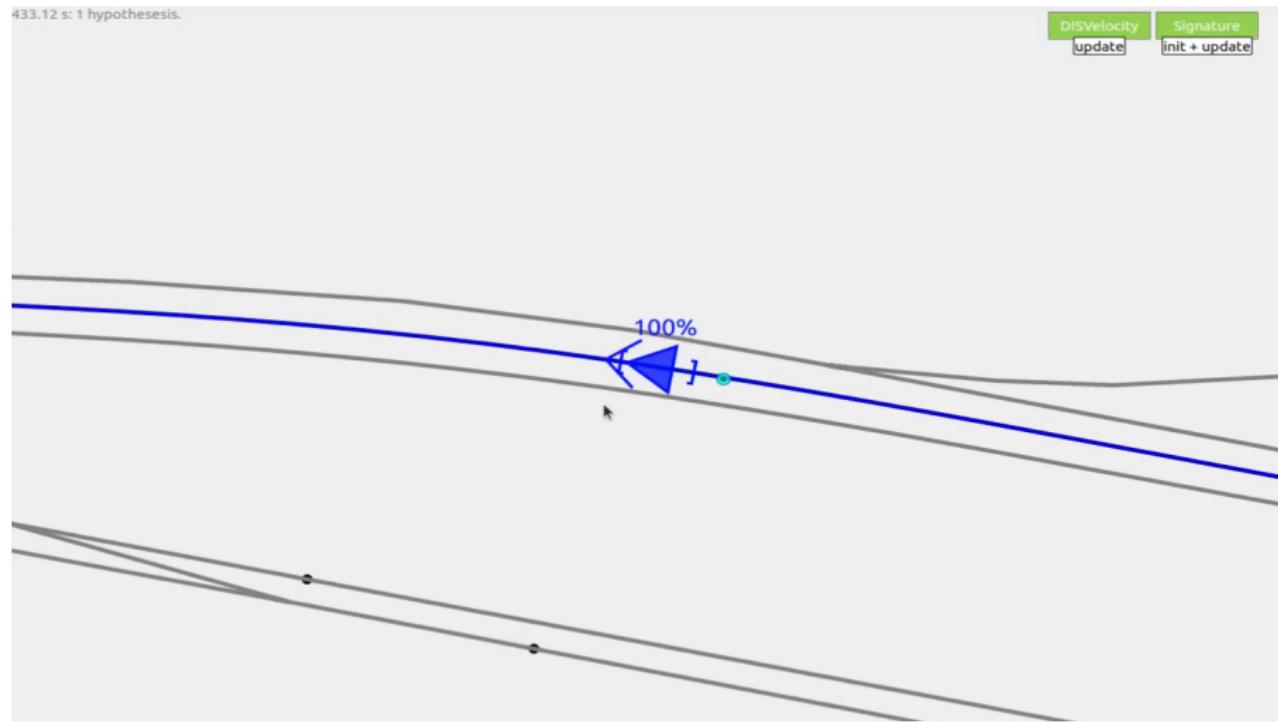


# Work in progress

- combine velocity and position estimation
- combine both rails
- online processing
- long term stability of signatures



# Demo



# Thank you

