

# Modelling of braking curves

## White, grey and black box approaches

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November 14, 2017

# Section 1

## Motivation

## Henry

Job: ETCS Expert

Employer: Infrastructure manager

Challenges: Ensure safety, maintain or increase capacity



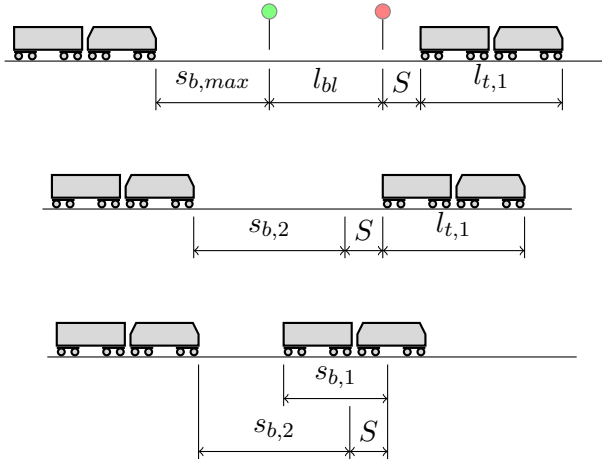
# User: Infrastructure Manager Executive

“I need to ensure that signals are practically never overrun while at the same time, the load on my network increases every year.”



# Efficient operation of trains on limited infrastructure

Appropriate braking curves and the moving block system help to reduce headway.



# Section 2

## Introduction to braking curves

# User: Infrastructure Manager Executive

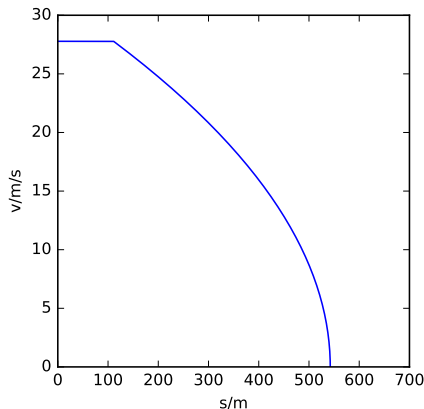
“With the moving block system, I can improve infrastructure utilisation - I only need to find the braking curves!”



# What is a braking curve?

ATP systems rely on braking curves to describe the train's braking capability.

- To supervise train velocity, ATP systems predict the future braking capability of the train

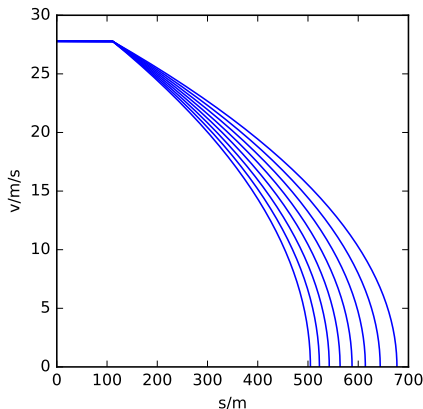




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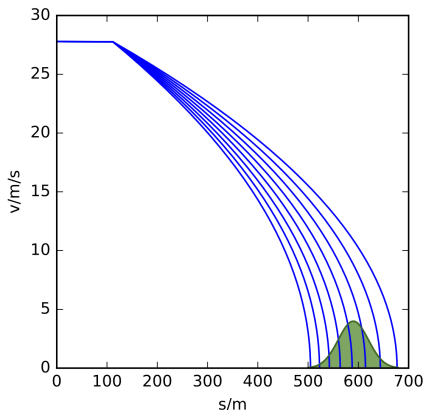
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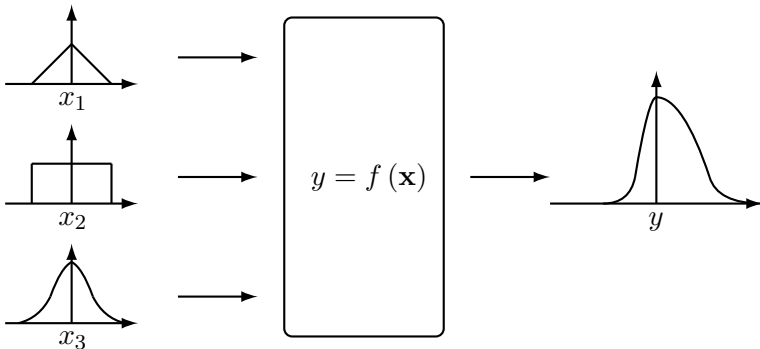
ATP systems rely on braking curves to describe the train's braking capability.

- To supervise train velocity, ATP systems predict the future braking capability of the train
- However, there is not *the* braking capability
- Braking curves exhibit a randomised behaviour



# How to obtain a braking curve?

To obtain a braking curve, the stochastic behaviour of the system needs to be analysed, typically by help of a Monte Carlo Simulation.



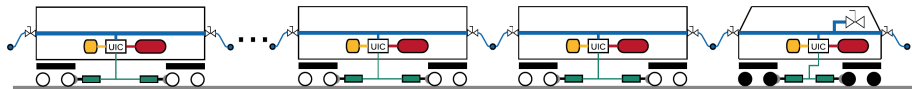
# Section 3

## White box approach

# Physical Modelling of the braking system

Which parameters can be identified and which effect do they have on the braking distance?

- Brake pipe: propagation velocity, flow resistances, train length



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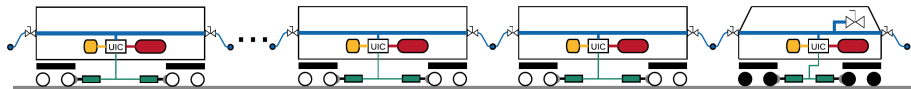
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- Distributor valve: Filling time, brake cylinder pressure



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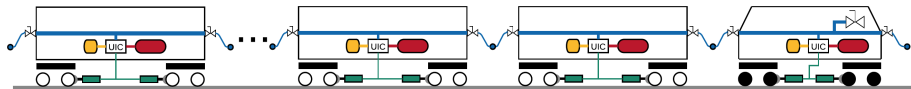
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- Wheel/rail contact: rail surface, contaminants, slip, ...

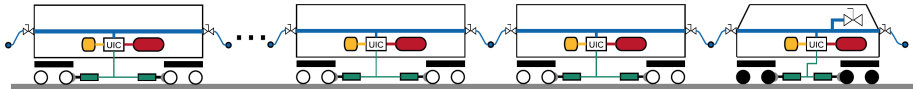




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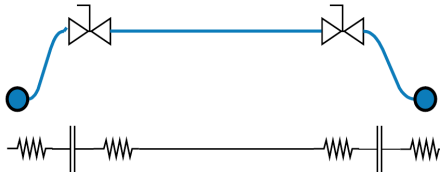


- Also discrete failure events need to be considered

# Brake pipe parameters

Brake pipe parameters determine the distribution of the brake command along the train.

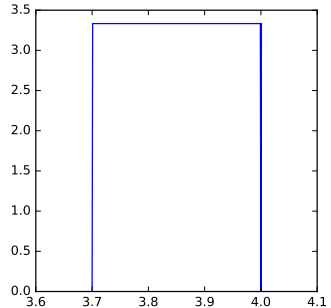
- Propagation velocity:
  - Required:  $c \geq 250 \frac{\text{m}}{\text{s}}$
  - May be considered lower limit
- Flow resistances:
  - Flow resistance in the individual wagons determine filtering behaviour of BP
- Train length:
  - Non-random input parameter
- Wagon position:
  - Distribution of braked mass in train and effective filling time influence overall braking distance



# Distributor valve parameters

Distributor valve parameters determine the effectivity of the brake command.

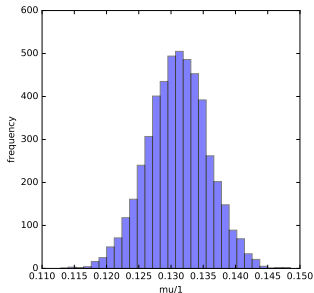
- Filling time  $t_f$ :
  - Brake modes P/R:  $(4 \pm 1)$  s
  - Brake modes G:  $(24 \pm 6)$  s
  - Uniform distribution (conservative)
- Brake cylinder pressure  $p_C$ :
  - Required:  $p_C = (3.8^{+0.2}_{-0.1})$  bar
  - Uniform distribution (conservative)



# Braking force generation parameters

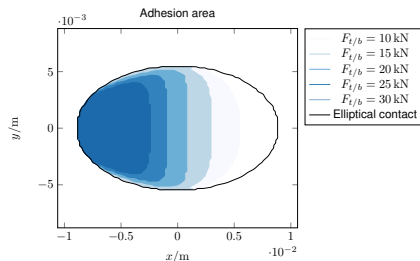
Parameters of the braking force generation subsystem determine the propagation of braking effort between  $p_C$  and wheel/disc.

- Efficiency
  - Typical dynamic efficiency:  
 $\eta \in [0.75, 95]$
  - Depending on maintenance state
  - Assumed uniform distribution
- Brake radius
  - Systematic variation with pad wear, not relevant for block brakes
- Pad/block friction coefficient  $\mu_B$ 
  - Mean friction coefficient depending on  $v_0$
  - Stochastic variation of instantaneous coefficient
  - Normal distribution appropriate



# Wheel-rail surface parameters

- Rail surface:
  - According to Hertzian theory
  - Non-Hertzian contacts due to hunting
- Contaminants:
  - Empirical estimation due to network
  - Mostly dry braking curves simulated
- Slip:
  - Curving motion, hunting impose 3D-slip on contact patch
  - Adhesion “budget” gets used



“Looks like the simulation model is quite complex? Can we do this online?”



# Approaches to obtain braking distance distributions

- Error-propagation:
  - Conservative: assumes normal distribution for all parameters
  - Complex: requires explicit function formulation and partial differentiation
- (Standard) Monte-Carlo-Simulation:
  - Efficient (in terms of confidence): returns shortest (also asymmetric) confidence interval
  - Inefficient (in terms of computational effort):
    - For rare event  $\varepsilon \ll 1$ ,  $N \approx \frac{100}{\varepsilon}$  trials required
    - Typical according to CSM:  $\varepsilon \in [10^{-7} \dots 10^{-9}] \Rightarrow N \approx 10^{11}$

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- ERA proposes to precalculate braking curves for limited number of train formations
  - Freight trains to be handled using braked weight and correction factor

“OK, basic Monte-Carlo is too complex to be calculated for each freight train. I fear a correction factor may be too conservative for well maintained wagon fleets. Are there any means to overcome this?”



# ETCS: $\gamma$ vs. $\lambda$ braked trains

- Typical  $\gamma$ -braked trains:
  - Multiple units, other fixed formations
  - Braking curve specification via deceleration values
- Typical  $\lambda$ -braked trains:
  - Any in-service configurable trains, especially freight trains
  - Braking curve using correction factors ( $K_{dry,rst}$ ,  $K_{wet,rst}$ ) to calculate based on brake weight

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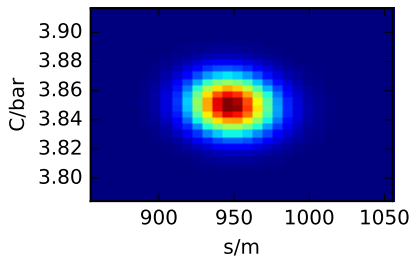
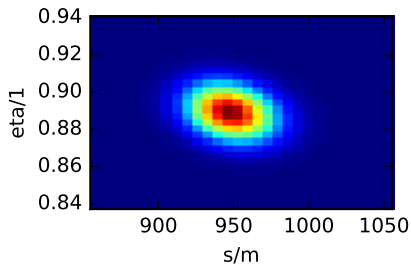
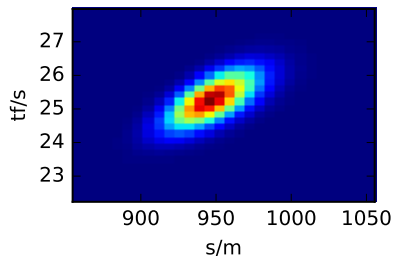
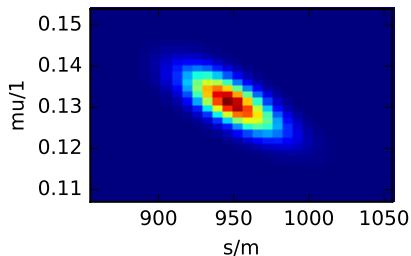
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  - The distribution of braking distances for freight trains of the same braked weight may be large:
    - Empty/loaded selection vs. automatic load detection
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    - Tread vs. disc brake

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  - The distribution of braking distances for freight trains of the same braked weight may be large:
    - Empty/loaded selection vs. automatic load detection
    - Maintenance state
    - Tread vs. disc brake
  - It may be of advantage to run certain  $\lambda$  trains as  $\gamma$  trains

## Towards a grey box approach

# Select dominating parameters



# Use importance sampling

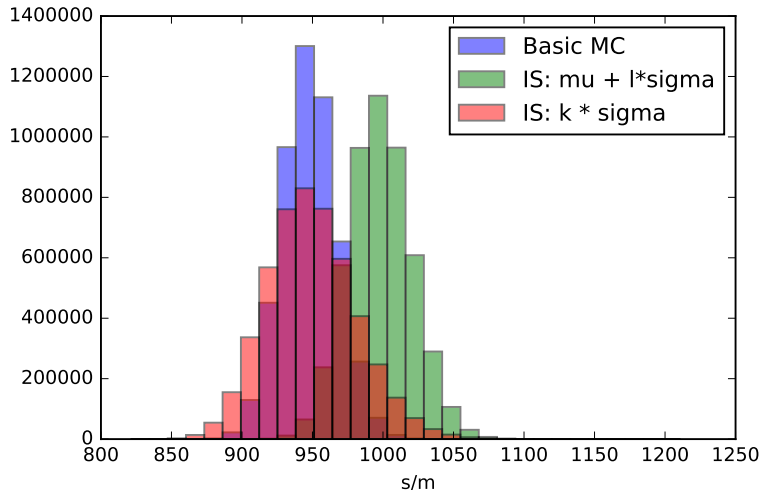
Importance sampling (IS) increases the probability of “desired” outcomes in Monte-Carlo-Simulations.

- Typical IS approaches:
  - Stratification: select only relevant strata of the sampling range
  - Scaling: Scale random variable
  - Translation: Move random variable to more relevant part of sampling space
  - Change of random variable: Replace random variable by one more likely to produce outcomes in the relevant range
  - Adaptive approaches
- Effect: higher number of samples in region of interest
- Correction factor: Likelihood ratio  $L(y) = \frac{\tilde{f}(y)}{f(y)}$



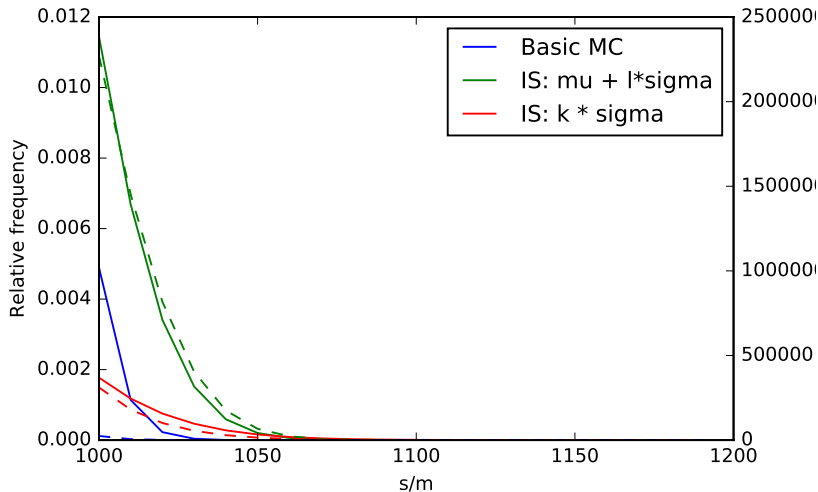
# Application of IS to braking curves

Change identified random variables, in the case at hand  $\mu_B$



# Application of IS to braking curves

Analyse for rare events, here braking distances in excess of 1100 m.  $N = 5 \cdot 10^7$



# Application of IS to braking curves

Analyse for rare events, here braking distances in excess of 1100 m.  $N = 5 \cdot 10^7$

$s$	$n_{\mathcal{U}}$	$p_{\mathcal{U}}$	$n_{IS,1}$	$p_{IS,1}$	$n_{IS,2}$	$p_{IS,2}$
1000	24400	$4.89 \cdot 10^{-3}$	$2.27 \cdot 10^6$	$1.14 \cdot 10^{-2}$	$3.11 \cdot 10^5$	$1.77 \cdot 10^{-3}$
1050	2	$4 \cdot 10^{-7}$	$6.66 \cdot 10^4$	$2.02 \cdot 10^{-4}$	$1.48 \cdot 10^4$	$1.59 \cdot 10^{-4}$
1100	0	0	115	$2.04 \cdot 10^{-7}$	419	$7.50 \cdot 10^{-6}$
1150	0	0	0	0	15	$3.88 \cdot 10^{-7}$
1160	0	0	0	0	7	$2.05 \cdot 10^{-7}$
1170	0	0	0	0	5	$1.46 \cdot 10^{-7}$
1180	0	0	0	0	4	$1.16 \cdot 10^{-7}$
1190	0	0	0	0	1	$2.90 \cdot 10^{-8}$

# Section 5

## A black box approach

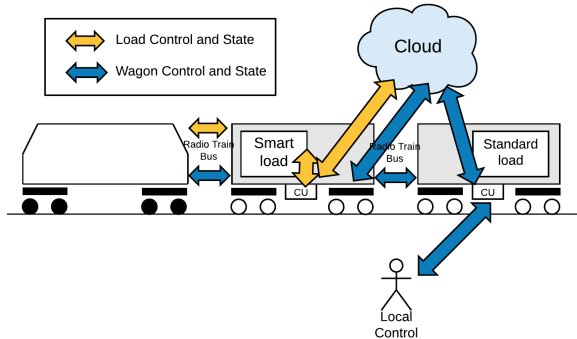
“Well, this reduces the required Monte Carlo iterations by far, however handling braking curves for each wagon during brake assessment doesn't appear feasible.”



# Connect the wagon subsystem

The Wagon 4.0 offers sensing and connectivity as well as cloud representation.

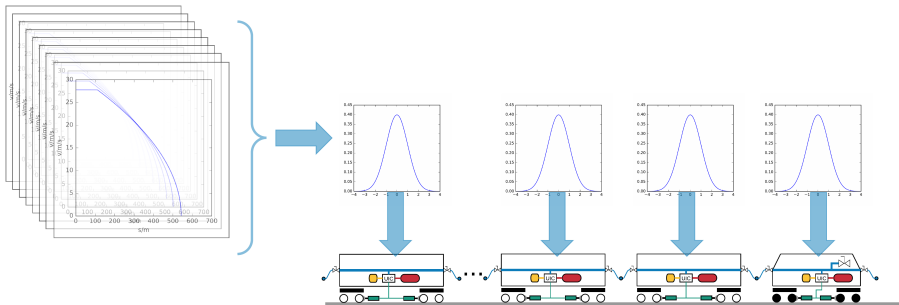
- Sensing: Accelerometers to record deceleration, brake cylinder pressure sensor to measure braking force
- Connectivity: send braking data to cloud



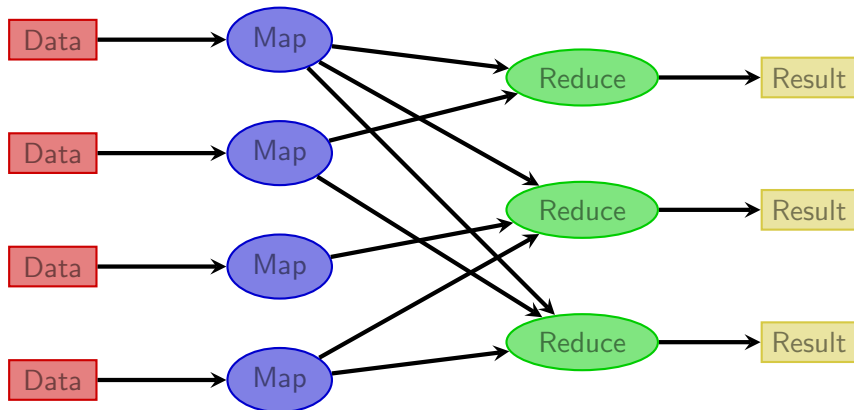
# Big data analytics

Big data analytics can be applied to separate train and wagon braking performance

- Record brake deceleration for wagons (in trains) in cloud
- Use big data analytics to derive individual wagon braking performance distribution

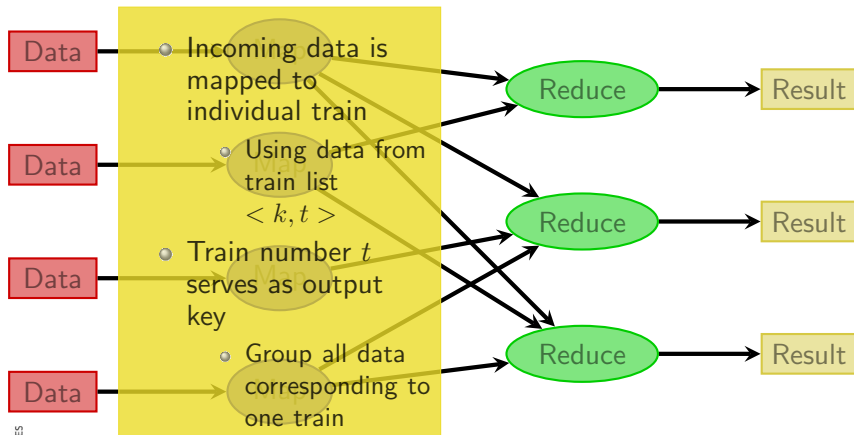


# MapReduce approach to braking curve estimation

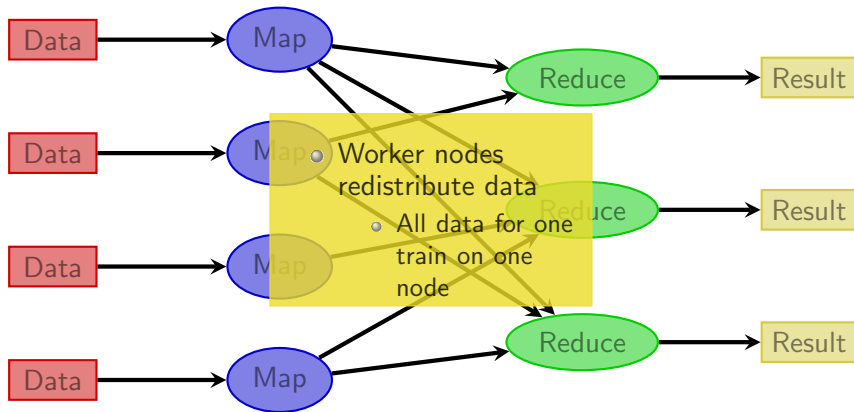




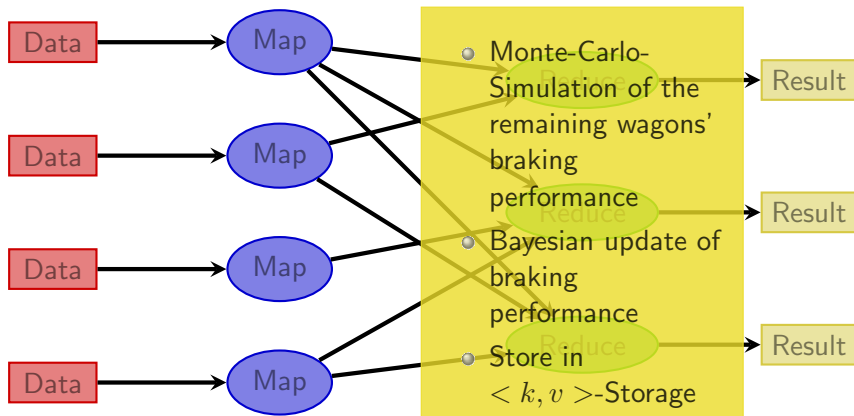
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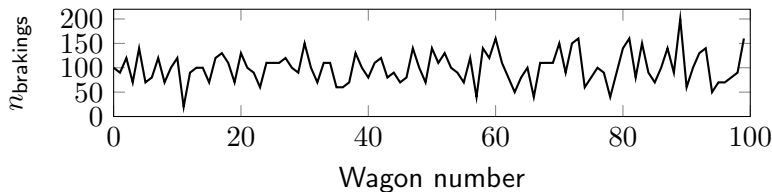
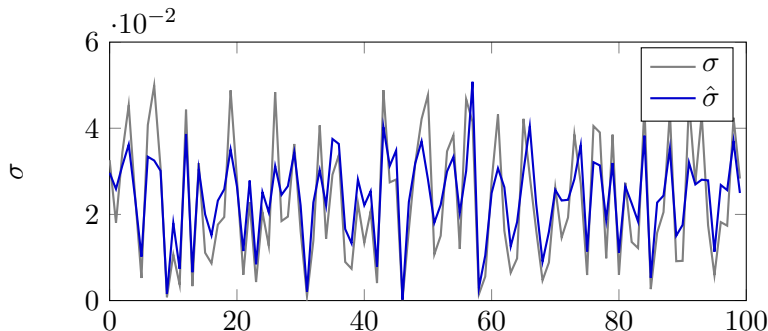
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# Simulated results of Big Data analysis



“Great, the approach to use Importance Sampling, IoT-technologies and Big Data analytics to gain the braking curves of each individually composed train improves our performance compared to running  $\lambda$ -trains.”



# Thank you!



Full paper available:



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[www.raphaelpfaff.net](http://www.raphaelpfaff.net)