Strengthening the Rail Mode of Transport by Condition Based Preventive Maintenance

Knowledge for Tomorrow

René Schenkendorf & Jörn Groos & Lars Johannes September 04, 2015



Research Programs of DLR

- Aeronautics
- Space
- Energy
- Security
- Transport



Figure : Next Generation Train (\mathbb{R}) , DLR





Institute of Transportation Systems

Residence:	Braunschweig, Berlin	
Since:	2001	
Director:	Prof. DrIng. Karsten Lemmer	
Employees:	About 150 employees from various scientific disciplines	
Fields of Research:	Automotive Railway Systems Traffic Management	
Range of Tasks:	Basic research Creating concepts and strategies Prototype development	
Quality:	DIN EN ISO 9001 VDA 6.2 ISO 17025 (RailSiTe®)	



Institute of Transportation Systems



DLR

1 Condition Based Preventive Maintenance

2 Localization

8 Prognostics



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Motivation

- conflicting demands: profitability, availability, safety, and punctuality
- potential solution: optimized scheduling of maintenance actions taking account of the actual infrastructure condition and its expected degradation
- critical railroad infrastructure: railway track (misaligned track sections + railsurface failures)







(b) Squat



(c) Corrugation



Preventive Maintenance Framework





Preventive Maintenance Framework



In-Line trains equipped with **low-cost sensor systems** are a key element for a **continuous** condition monitoring.





In-Line Trains == Moving Sensor Systems

- rail irregularities ⇒ vehicle response/vibration
- autonomous train-born measurement systems including...
- inertial measurement unit (IMU), acceleration sensors, microphone ... and other low-cost sensors



(d) RailDriVE



(e) Data Logger



(f) Acceleration Sensor



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LOCALIZATION

Track Selective: Location/Position of the train on the correct track

State-of-the-Art (Measurement Trains)

- localization based on the train's odometer
- uncertainty up to dozens of meters
- \Rightarrow **NO** automated and precise (below 10m) georeferencing

(railway network == large area with insufficient GNSS reception)



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Multi-Sensor Concept (In-Line Trains)

- GNSS (Global Navigation Satellite System) receiver
- odometer + speed sensor (Doppler radar)
- balise-antenna
- digital map of the railroad network
- \Rightarrow track selective accuracy



σ -Accuracies of Different Sensors

GNSS RTK0.02 - 0.20 m (depending on baseline)Balise0.20 mOdometer0.4 % (of covered distance)Speed0.8 % (of covered distance)





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 \Rightarrow Data Fusion via **Extended Kalman Filter** for Localization





GNSS (red dots) vs. Multi-Sensor concept (orange dots)





GNSS (red dots) vs. Multi-Sensor concept (orange dots)





Failure (Corrugation)



Localization



Failure (Squat)



Localization





Railway Network



(a) Misalignment



(b) Squat



(c) Corrugation



Railway Network



 $\downarrow \\ \text{In-Line Trains} + \text{Low-Cost Sensors} + \text{Localization} \\ \downarrow \\ \text{Key Performance Indicators (KPIs) of Monitored Track Segments} \\ \downarrow \\ \text{Expected Future Degradation?} \Rightarrow \text{Prognostics} \\ \end{cases}$















Local Degradation Models

- various approaches
- common: number of influencing parameters
- e.g. soil/rail quality, operating conditions, weather...
- $\Rightarrow \theta \equiv$ parameter vector
- in general, θ is uncertain (random variable)

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Proc. IMechE Vol. 226 Part O: J. Risk and Reliability

Monte Carlo simulation of railway track geometry deterioration and restoration

L M Quiroga and E Schnieder

Institute for Traffic Safety and Automation Technologies, TU Braunschweig, Braunschweig, Germany

The manuscript was received on 23 December 2010 and was accepted after revision for publication on 8 July 2011.

DOI: 10.1177/1748006X11418422





More than degradation



CrossMark

A stochastic model for railway track asset management

John Andrews ^{a,*}, Darren Prescott ^a, Florian De Rozières ^b ^a Nottingham Transportation Engineering Centre, University of Nottingham, UK ^b Grenoble Institute of Technology, France









Holistic Approach

- incorporation of several track segments
- difficult to parameterize
- complex and cpu-intensive analyzes
- Polynomial Chaos Expansion (PCE) might help to decrease computational burden

















Polynomial Chaos Expansion

- handy surrogate model, $\hat{g}(\theta)$
- needs to be parameterized

$$y = g(heta) pprox \hat{g}(heta) = \sum_{i=0}^{l_{pce}} a_i \Psi_i(heta)$$

 $\Psi_i(\theta)$ - proper orthogonal functions (Hermite Polynomials)



Polynomial Chaos Expansion

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- needs to be parameterized

$$y = g(\theta) \approx \hat{g}(\theta) = \sum_{i=0}^{l_{pce}} a_i \Psi_i(\theta)$$
$$a_i = \frac{\int_{\Omega} g(\theta) \Psi_i(\theta) p df_{\theta} d\theta}{\int_{\Omega} \Psi_i(\theta)^2 p df_{\theta} d\theta} = \frac{\int_{\Omega} g'(\theta) p df_{\theta} d\theta}{\int_{\Omega} \Psi_i(\theta)^2 p df_{\theta} d\theta}$$



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$$\int_{\Omega} g'(\theta) p df_{\theta} d\theta = E[g'(\theta)] \approx \sum_{i=1}^{L} w_i g'(\Theta_i)$$



An ideal approximation method should provide:

- good approximation power
- workable computational load

 $E[g'(\theta)] \approx \sum_{i=1}^{L} w_i g'(\Theta_i)$ by Numerical Integration Methods

Point Estimate Method (PEM)

- Generator Function, $GF[\cdot]$, makes the difference
- describes how sample points are directly determined in \mathbb{R}^n by:
 - permutation
 - change of sign-combinations



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 - change of sign-combinations

 $\int_{\Omega} g'(\theta) p df_{\theta} d\theta \approx w_0 g'(GF_0) + w_1 \sum g'(GF_1) + w_2 \sum g'(GF_2)$

Any statement about the ...

- approximation power ?
- computational load ?





- correct approximation for monomials of order 5
- PEM implies $\mathbf{2n}^2 + \mathbf{1}$ sample points $(\theta \in \mathbb{R}^n)$

 PEM provides a workable compromise on accuracy and computational load





Illustration: In-silico example

• many degradation models include exponential terms

$$y = g(\theta, t) = \theta_1 e^{-\theta_2(e^{-\theta_3 t})}$$

$$egin{array}{rcl} heta_1 &\sim & \mathcal{N}(5,1) \ heta_2 &\sim & \mathcal{N}(2,1) \ heta_3 &\sim & \mathcal{N}(3,1) \end{array}$$

$$y = g(\theta, t) = \theta_1 e^{-\theta_2 (e^{-\theta_3 t})}$$



Model response, y





Model response, y





Time in weeks



Summary

- In-line trains as moving sensors
- continuous track monitoring via low-cost sensor systems
- precise localization is mandatory
- \Rightarrow Multi-sensor concept for localization
- efficient algorithm to take account of uncertain parameters
- \Rightarrow combination of PCE and PEM



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