

Health Monitoring of Hybrid Systems Using Hybrid Particle Petri Nets

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- 1.Context
- 2. Motivations
- 3. Methodology for behavioral diagnosis
- 4. Methodology for degradation diagnosis
- **5. Conclusion & future work**



- Complex dynamic systems exposed to failures
 - Maintenance and repair to reduce cost of products
 - **Reconfiguration** and **replanning** to maximize rewards of 'local' missions





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Hybrid systems

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- · Continuous and discrete observations
- Multi-modes systems (continuous and discrete behaviors)





Diagnosis

• Determine the current health state of the system

Prognosis

- Determine the future health states of the system
- Often related to the temporal prediction of the date at which the system is not operational anymore
- RUL/EOL prediction





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• Model-based diagnosis and prognosis



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- Interleave diagnosis and prognosis (InterDP)
 - To perform a more efficient health monitoring [Chanthery & Ribot, 2013]



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Modified Particle Petri Nets (MPPN)







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Continuous dynamics of the system (dynamics equation)

Behavioral modes of the system (discrete state)



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The occurrence of an unobservable event (supported by continuous observations)

The occurrence and the non occurrence of an observable event

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MPPN combine conditioned-firing and pseudo-firing (duplication of tokens)

- Use particle filters to model uncertainty about:
 - Model equations
 - Continuous observations







- Use particle filters to model uncertainty about:
 - **Model equations** • **Continuous observations** • 1. transitions firing 2. particle values updating M_0^S M_0^N Prediction OK ON Observations π^2 е (O_1^S, O_1^N) Correction State (p^S, p^N) \hat{M}_{k}^{S} \hat{M}_{k}^{N} Prediction KO₁ Observations π^3 (O_{k+1}^S, O_{k+1}^N) Correction State KO₂ OFF \hat{M}_{k+1}^N \hat{M}_{k+1}^S (p^S, p^N)

MPPN on line process

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 - Model equations



Behavioral diagnosis using MPPN [Gaudel et al. 2014]

- Based on the generation of a diagnoser
 - The diagnoser is also a MPPN
 - The diagnoser is an upgrade of the MPPN model



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The diagnosis results is the marking of the diagnoser at time k

Indicates the belief on the fault occurrences level-headed with particle weights

$$\Delta_{\!\scriptscriptstyle k}\!=\!\hat{M}_{\scriptscriptstyle k}$$

MPPN behavioral level





MPPN behavioral level







- Degradation is an hybrid information
- Health modes are composed of one symbolic place and one numerical place





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- Associate health mode degradation law to corresponding hybrid place
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Conclusion

- **Definition of a new HPPN formalism to monitor hybrid system** by enriching the MPPN formalism
- Application of the HPPN for behavioral and degradation diagnosis of hybrid systems
- Illustration of the proposed method on an academic example



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Future work

- Implement and **test** the proposed framework
- Introduce model-based prognosis method into the proposed framework
- Test the health management architecture on a **mobile or flying robot**



Main related work

- Chanthery, E., & Ribot, P. (2013). An integrated framework for diagnosis and prognosis of hybrid systems. In 3rd Workshop on Hybrid Autonomous System (HAS). Roma, Italy.
- Zouaghi, L., Alexopoulos, A., Wagner, A., & Badreddin, E. (2011a). Modified particle petri nets for hybrid dynamical systems monitoring under environmental uncertainties. In *IEEE/SICE International Symposium on System Integration (SII)* (pp. 497–502).
- Gaudel, Q., Chanthery, E., Ribot, P., & Le Corronc, E. (2014). Hybrid systems diagnosis using modified particle petri nets. In *Proceedings of the 25th International Workshop on Principles of Diagnosis (DX'14)*. Graz, Austria.
- Gaudel, Q., Chanthery, E., & Ribot, P. (2014). Health Monitoring of Hybrid Systems Using Hybrid Particle Petri Nets. In *Proceedings of the Annual Conference of the Prognostics And Health Management Society 2014 (PHM'14)*. Fort Worth, Texas, USA.



Conclusion

Health Monitoring of Hybrid System Using Hybrid Particle Petri Nets

Thank you! Any question?

Key words:

Hybrid systems

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- Model-based diagnosis under uncertainty
- Behavioral diagnosis
- Degradation diagnosis
- Health management architecture
- Hybrid particle Petri nets



