

Efficient Failure Detection for Mobile Robots Using Mixed-Abstraction Particle Filters

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Overview

- A mobile robot should
 - detect failures quickly,
 - be able to infer the cause of a failure, and
 - be efficient in the failure-free case.
- We extend the Monte Carlo Localization scheme to simultaneously track multiple system models. Samples are distributed adaptively to the most appropriate models.



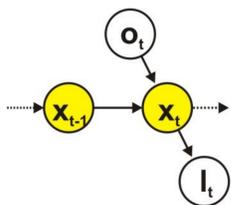
Mixed Abstraction Particle Filtering

- Sequential State Estimation

$$p(x_t) = \underbrace{\eta_t}_{\text{Normalization}} \cdot \underbrace{p(z_t | x_t)}_{\text{Observation model}} \cdot \underbrace{\int p(x_t | u_{t-1}, x_{t-1}) \cdot p(x_{t-1}) dx_{t-1}}_{\text{Motion model Recursive term}}$$

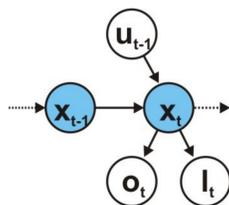
- We use **multiple system models** simultaneously, e.g.

- Standard Model



- Odometry-based motion model
- Cannot handle failures
- Is efficient in the failure-free case

- Dynamic Model



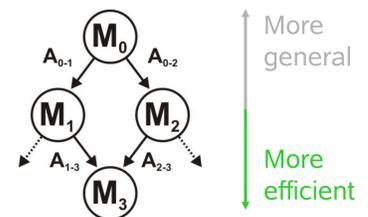
- Includes the motor commands
- Can handle certain failures, e.g.
 - Collisions
 - Deflating tires
- Is substantially more complex to compute

- The different models are related by considering the **assumptions** made about the state space.

- For the state space $\{x, y, \theta, p_l, p_r\}$ which includes the tire pressures $\{p_l, p_r\}$, the standard model would for example assume that these are equal and constant.

- The models are organized in a **model abstraction hierarchy** that uses the explicit model assumptions

- Edges: assumptions
- Nodes: models



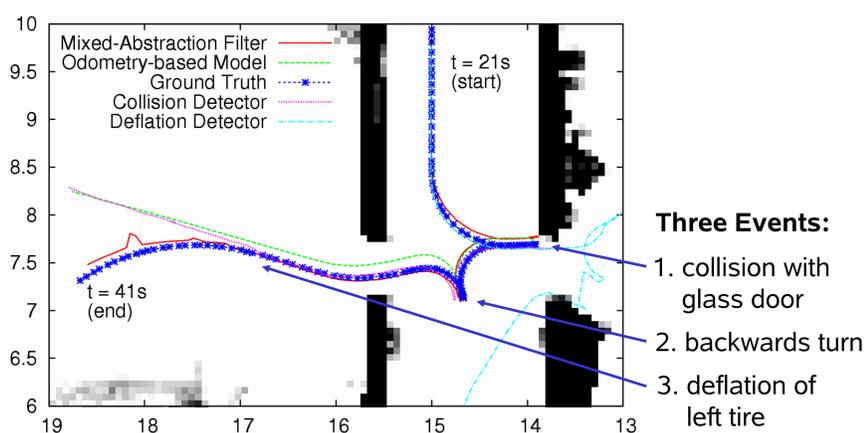
- The filter iteratively validates the model assumptions

$$v_t(A) := \frac{\sum_{\bar{A}} p(z_t | x_t^{[m]}) \frac{1}{|\bar{A}|}}{\sum p(z_t | x_t^{[m]}) \frac{1}{|\bar{A}|}}$$

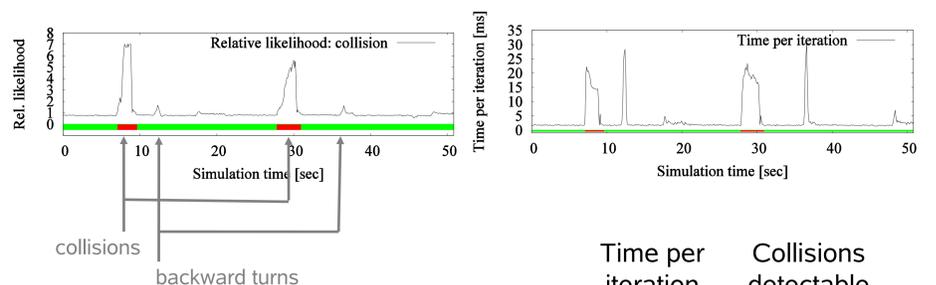
- and samples to the most efficient model which assumptions receive enough relative evidence

Experiments

- Combining the standard model with a dynamic model for detecting collisions and one for deflating tires.



- Analyzing the gain in efficiency



| | Time per iteration | Collisions detectable |
|---|--------------------|-----------------------|
| Standard model (20 particles) | 0.67 ms | ✗ |
| Standard model (200 particles) | 5.83 ms | ✗ |
| Dynamic model (300 particles) | 10.10 ms | ✓ |
| Mixed Abstraction (Standard: 20, Dyn.: 300) | 3.42 ms | ✓ |

Conclusions

- The proposed system is able to combine multiple system models.
- Computational resources are distributed adaptively by verifying explicit model assumptions.
- Usefulness has been demonstrated in several failure detection settings.