RT-KalmanNet:

A Robust Tailored Neural Network Aided Kalman Filtering Approach

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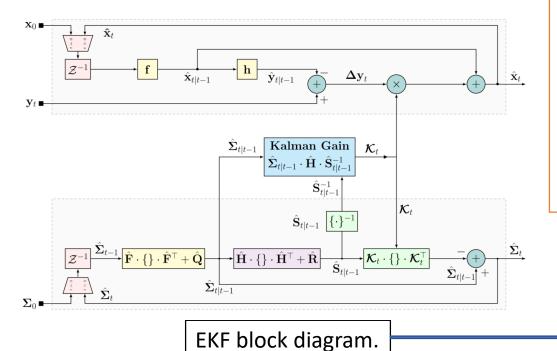
State of Art: EKF — a model-based estimator



State space model:

$$\mathbf{x}_{t} = \mathbf{f}(\mathbf{x}_{t-1}) + \mathbf{w}_{t}, \quad \mathbf{w}_{t} \sim \mathcal{N}(\mathbf{0}, \mathbf{Q}), \quad \mathbf{x}_{t} \in \mathbb{R}^{m}, \quad (1a)$$

$$\mathbf{y}_{t} = \mathbf{h}(\mathbf{x}_{t}) + \mathbf{v}_{t}, \qquad \mathbf{v}_{t} \sim \mathcal{N}(\mathbf{0}, \mathbf{R}), \ \mathbf{y}_{t} \in \mathbb{R}^{n}.$$
 (1b)

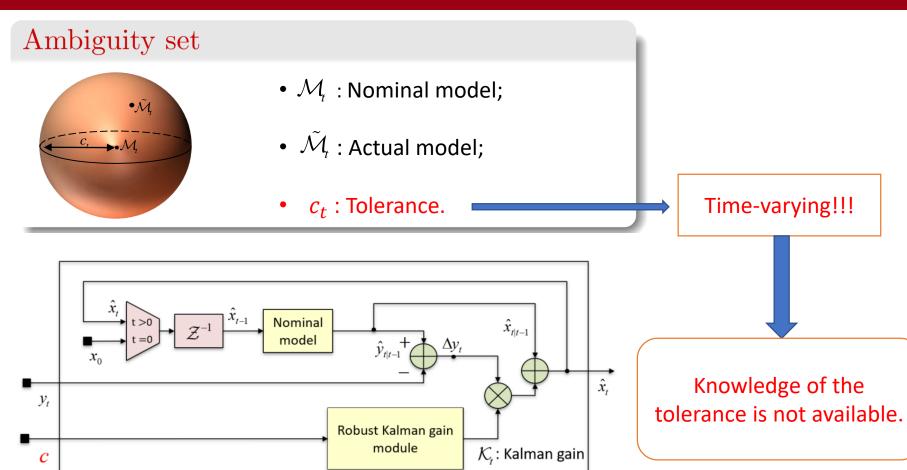


- 1. Knowledge of the noise is not available.
- 2. The functions f and h does not match the true one:
- the representation of continuous time dynamics in discrete time;
- Acquisition using misaligned sensors;
- other forms of mismatches...



State of Art: Robust EKF

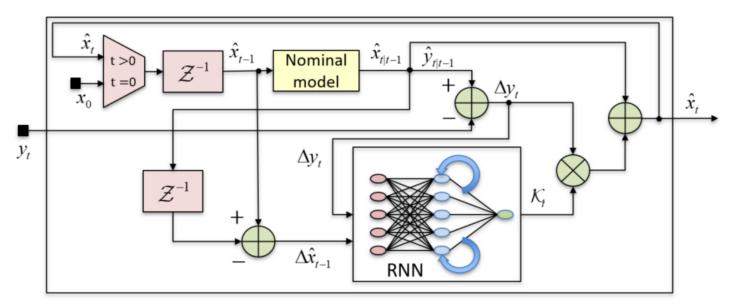




Robust EKF block diagram.

State of Art: KalmanNet



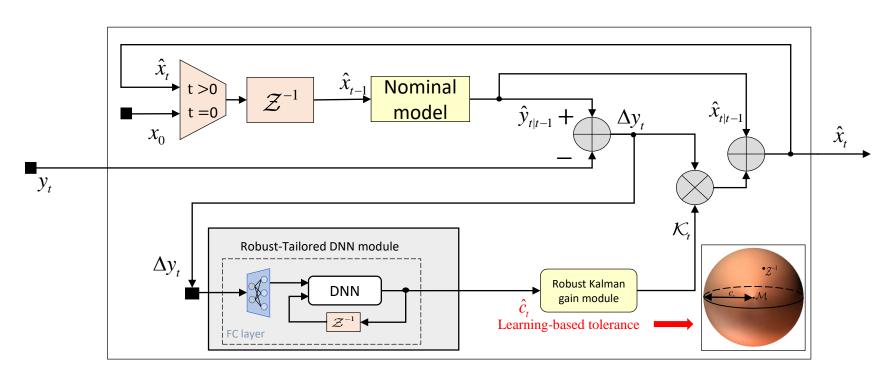


KalmanNet block diagram

Disadvantages

- Vulnerable to model uncertainty
- Substantial over-parameterization
- High complexity
- State label measurable





RT-KalmanNet diagram.



Implement the RT-KalmanNet and test it using the Synthetic Non-Linear Model:

- Compare the running time between REKF, KalmanNet and RT-KalmanNet;
- Compare the performance between REKF, KalmanNet and RT-KalmanNet.



Thank you for your attention!

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