

Robust Vehicle Lane Keeping Control with Networked Proactive Adaptation

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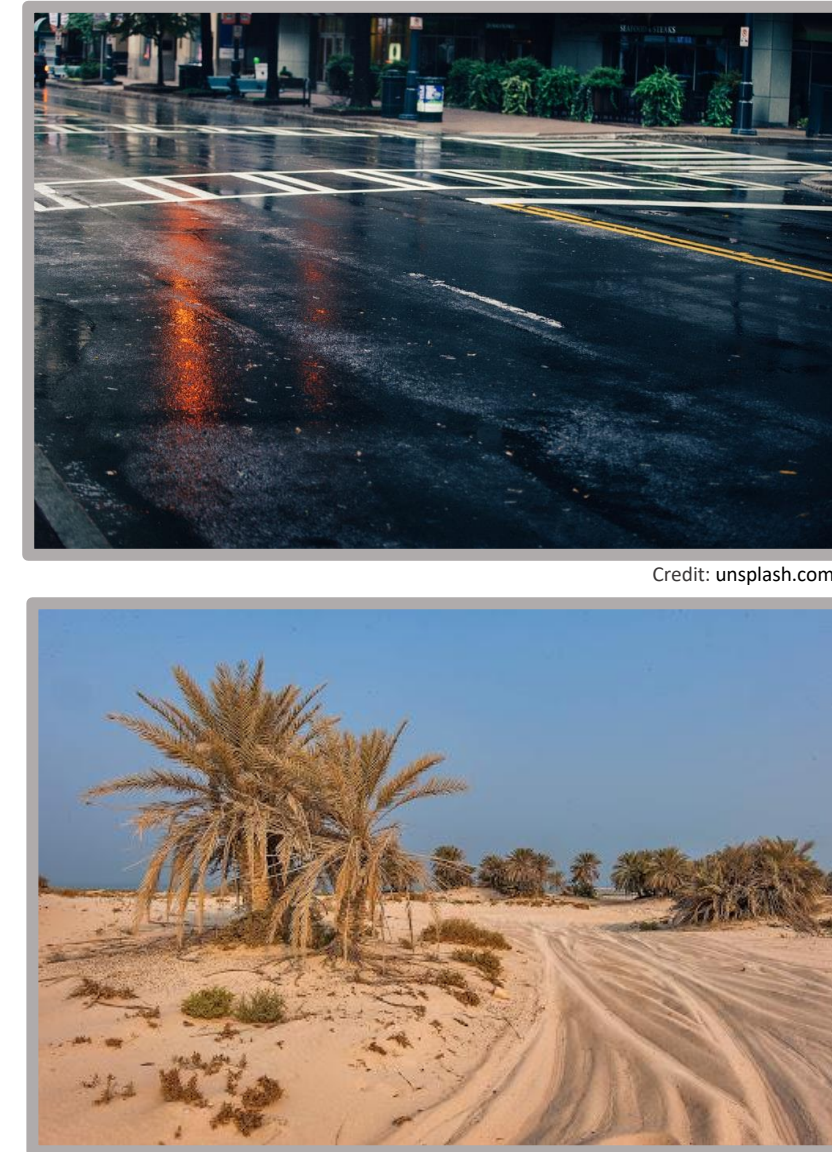
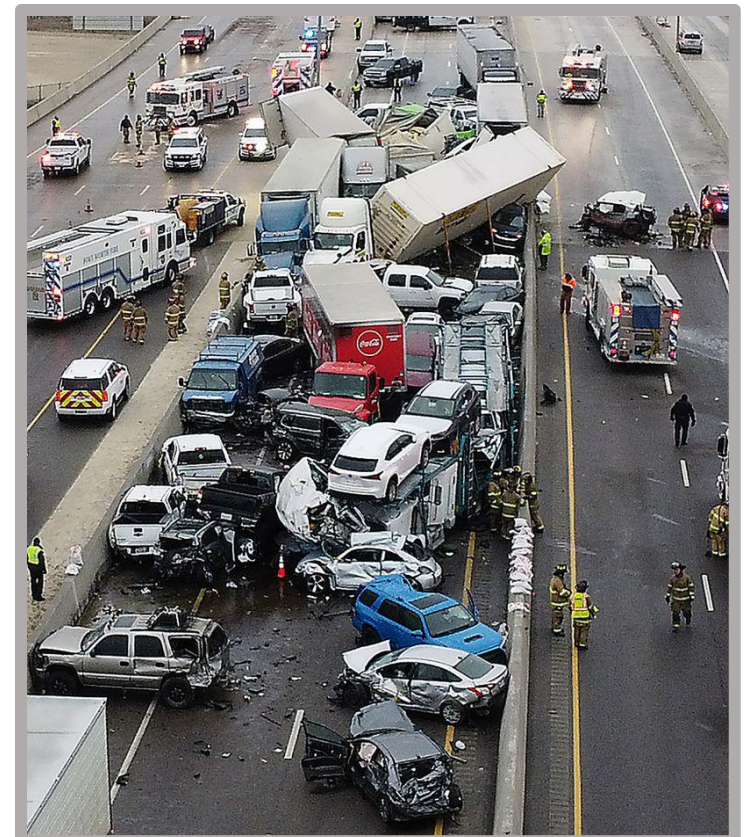


MOTIVATION

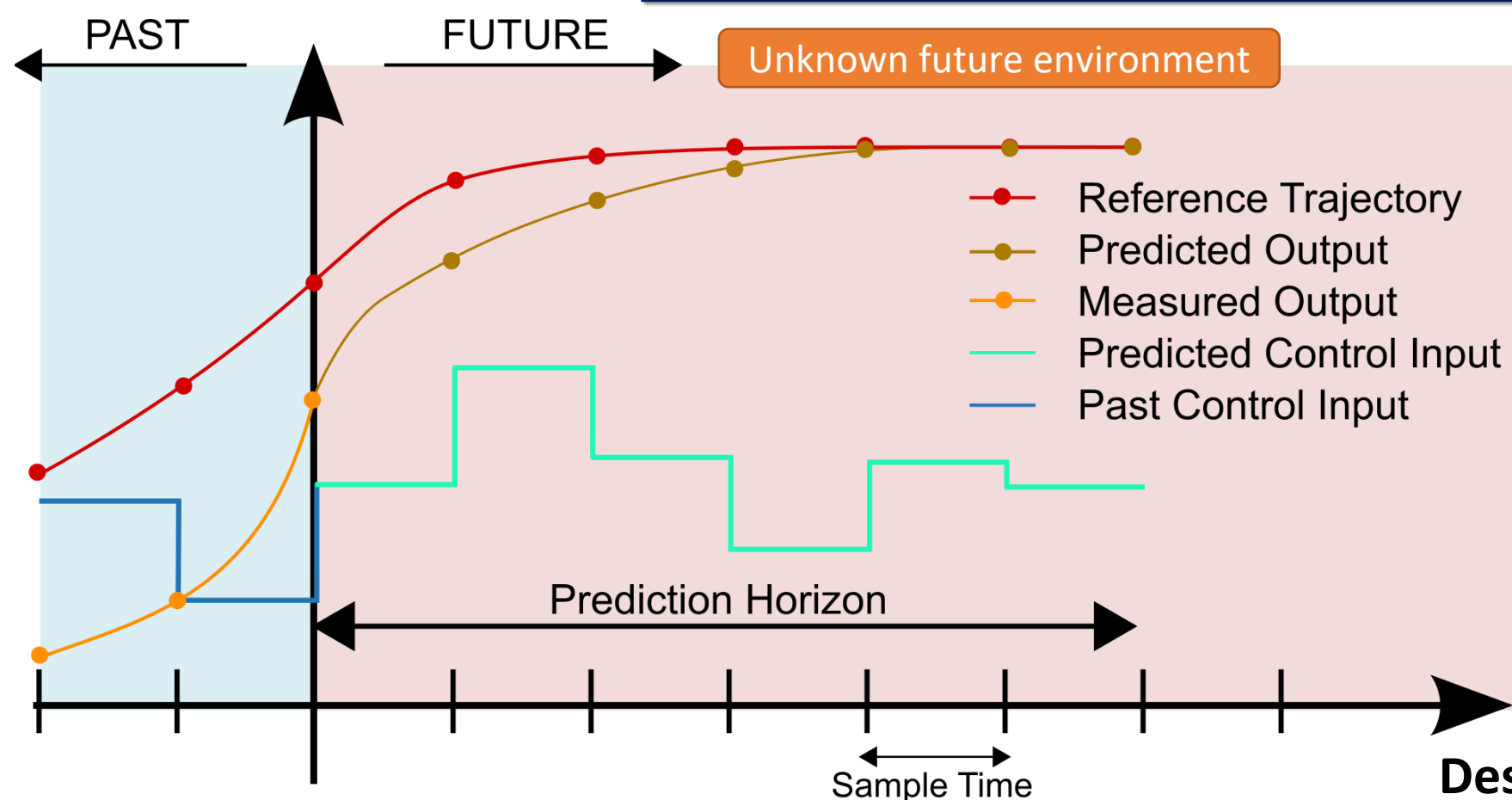
Unforeseen environment

Complex environment

Feb 11, 2021
133-car pile up on icy highway



LACK OF INFORMATION



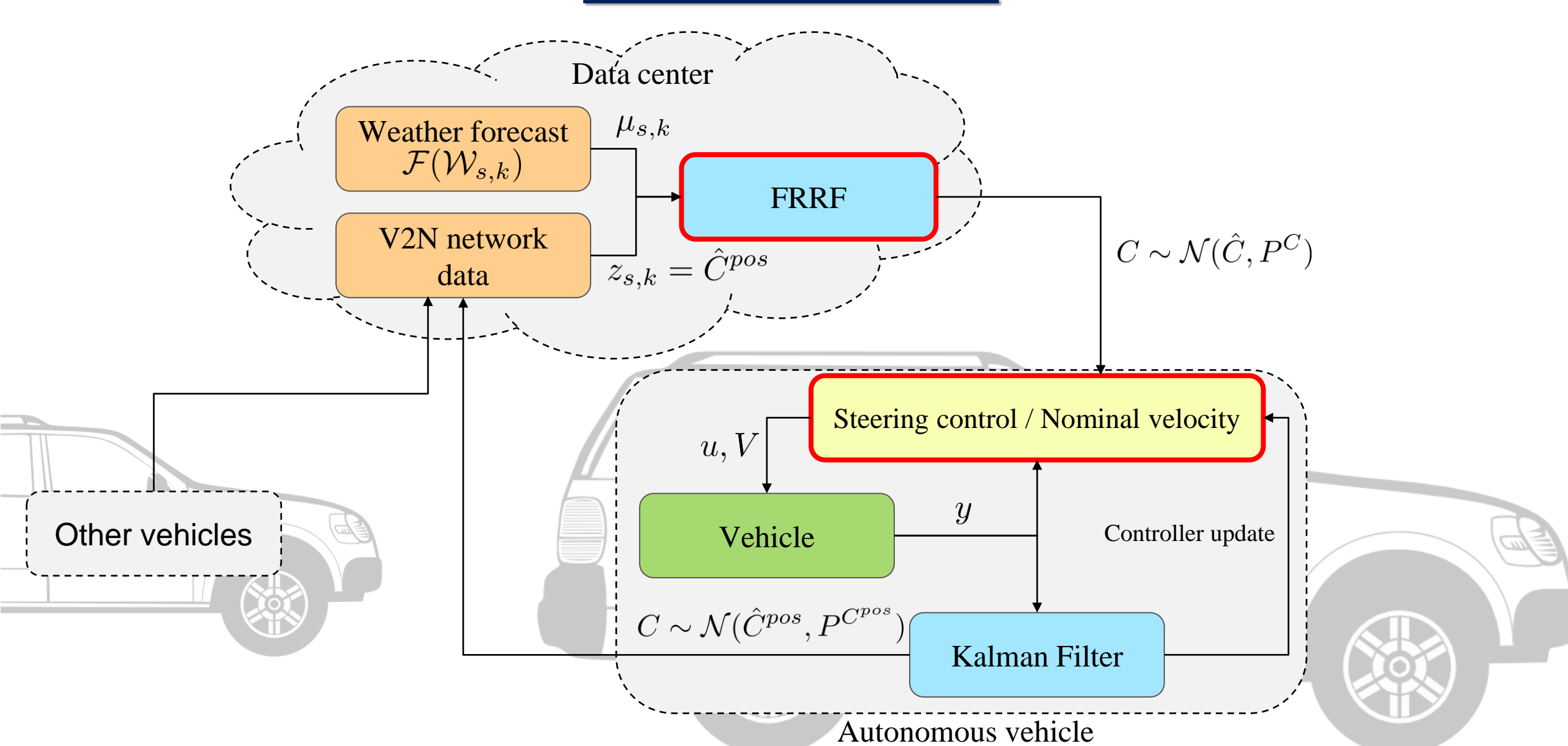
Can we access 'future environmental information'?

Preceding vehicles can share their measurement! (Network-enabled)

Design \mathcal{L}_1 adaptive control

- Prediction quality is determined by limited prior knowledge

OUR SOLUTION



FIXED RANK RESILIENT FILTERING (FRRF)

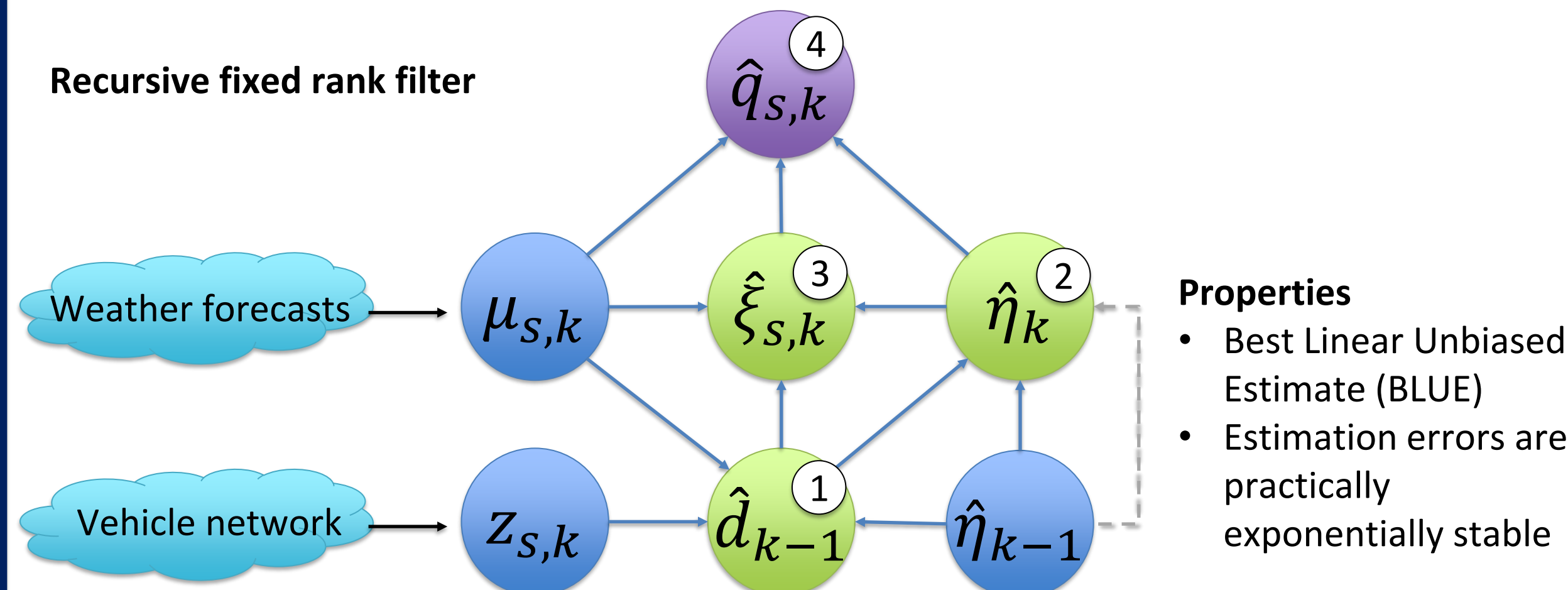
$$q_{s,k} = \mu_{s,k} + S_{s,k}\eta_k + \xi_{s,k} \quad \text{Variable of interest}$$

$$z_k = [z_{s_{1k},k}, z_{s_{2k},k}, \dots, z_{s_{n_k},k}]^T \quad \text{Measurements for some areas}$$

$$z_{s,k} = q_{s,k} + \epsilon_{s,k},$$

$$\eta_{k+1} = H_k\eta_k + G_k d_k + \zeta_k \quad \text{Fixed rank model}$$

Recursive fixed rank filter



PROACTIVE CONTROLLER DESIGN

Vehicle lateral model

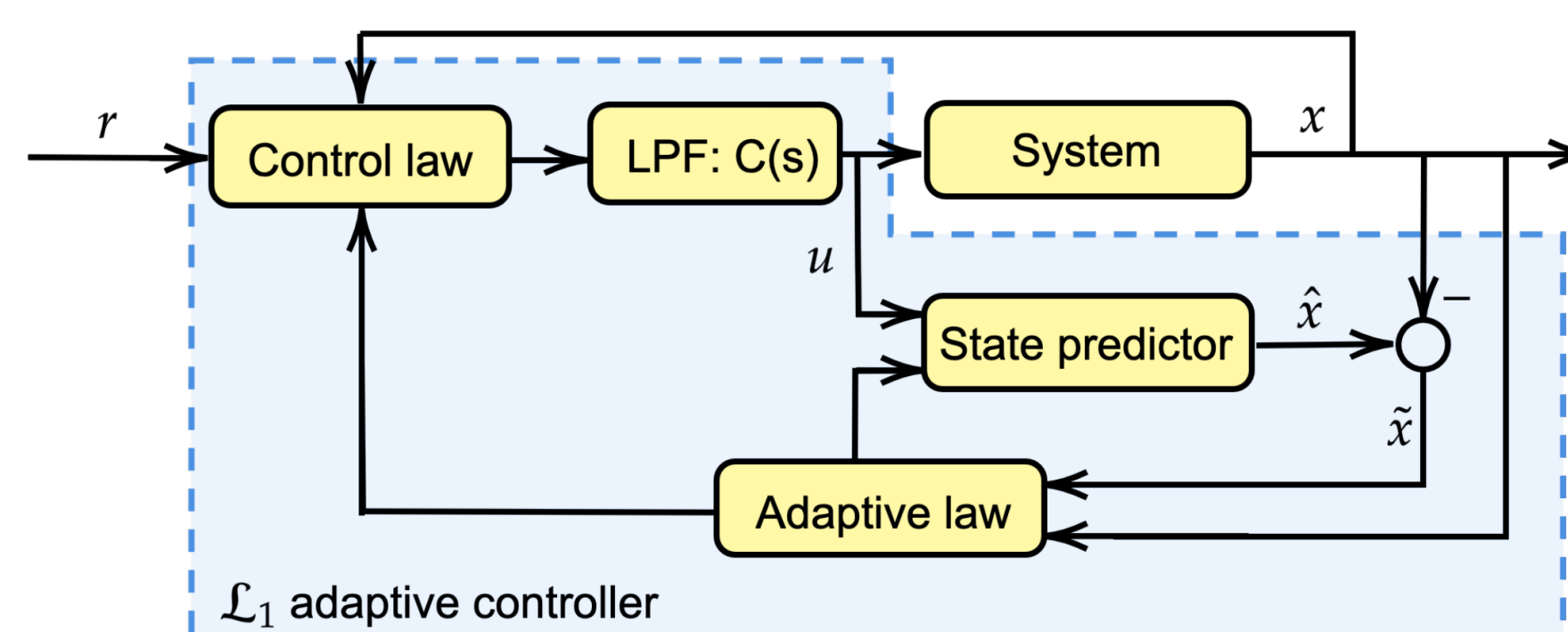
$$\dot{x} = A(V, C_f, C_r)x + b(C_f)u + g(V, C_f, C_r)\dot{\psi}^{des}$$

System matrices depend on the longitudinal velocity and cornering stiffness

Obtain the uncertainty model by propagating estimation uncertainty ($u = -Kx + u_{ad}$)

$$\dot{x}(t) = A_m x(t) + b_m(wu_{ad}(t) + \theta^T x(t) + \sigma(t))$$

$$y(t) = c^T x(t) \quad x(0) = x_0,$$



- The architecture decouples adaptation from robustness. We can choose adaptation gain as large as CPU permits.
- The low pass filter design trade-offs between the performance and robustness.

Stability condition:

$$A_m, A_g = \begin{bmatrix} A_m + b_m\theta^T & b_m w \\ -k\theta^T & -kw \end{bmatrix} \text{ Hurwitz}$$

Filter gain

Common Lyapunov approach: Stability condition for all velocity ranges. This gives a freedom to choose velocity later.

$$A_m(V)P + PA_m^T(V) < 0$$

LONGITUDINAL VELOCITY DESIGN

$$\max_{k, V \in [V_{\min}, V_{\max}]} V$$

$$\text{s.t. } \|G(s)\|_{\mathcal{L}_1} \leq \lambda_{gp} \text{ for } \forall w \in \Omega$$

$$k \leq \bar{k}$$

Tracking performance constraint

Robustness constraint

- $k \uparrow$: Better tracking performance, time-delay margin decreases.
- Lateral controller is designed for all velocities.
- The velocity determines performance.

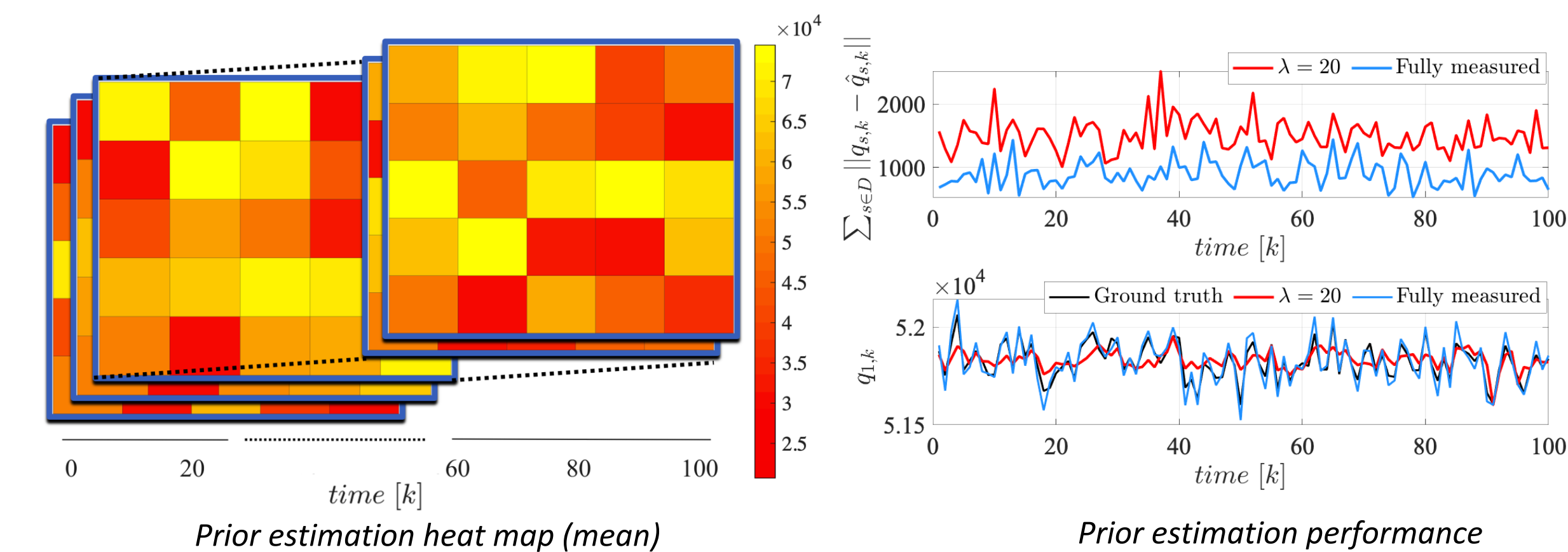
CONTROLLER UPDATE

$$k = \arg \min_k |k - k_*|$$

$$\text{s.t. } A_g(V) \text{ being Hurwitz, } k \leq \bar{k}$$

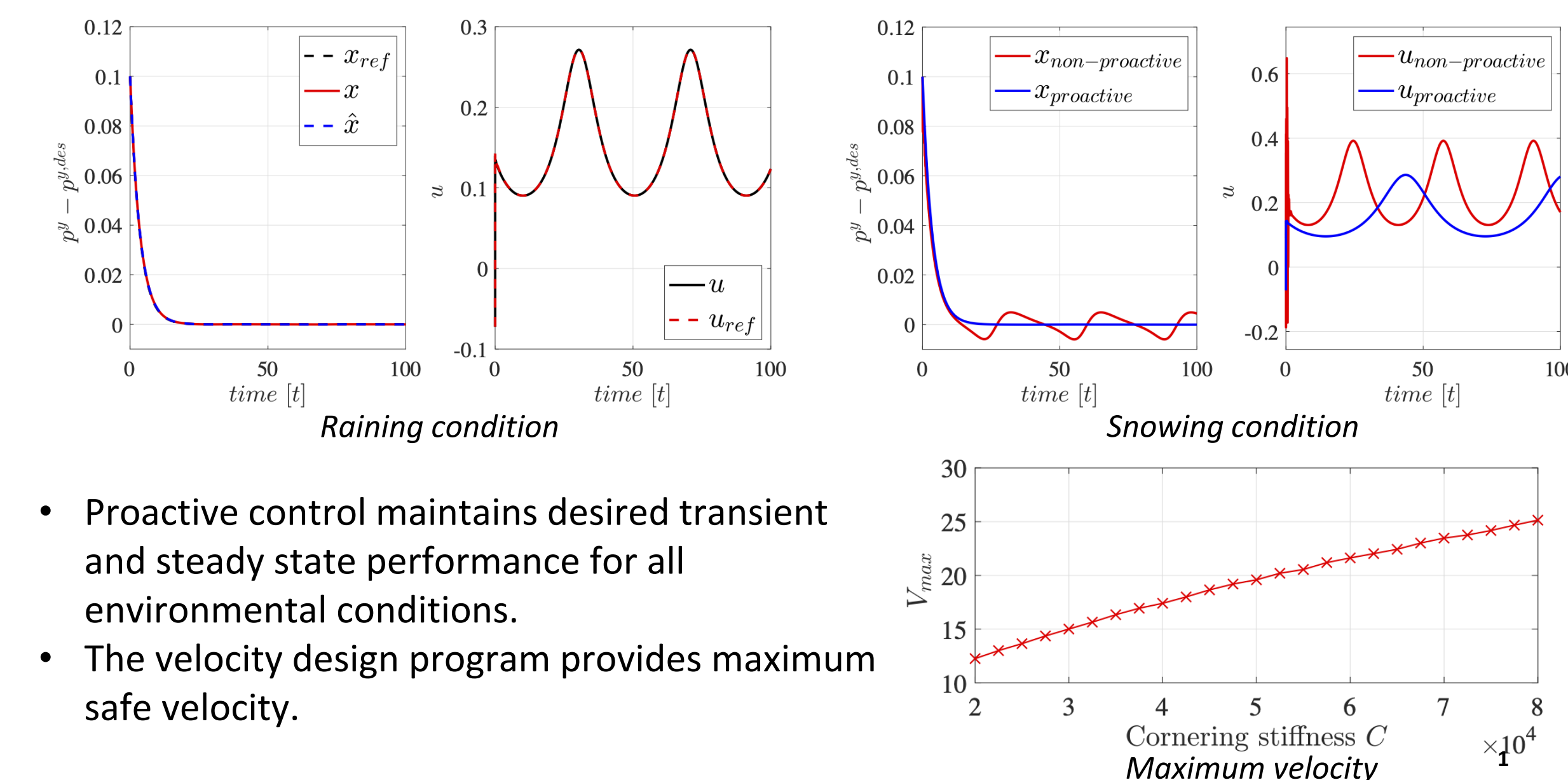
- Reference system A_m and A_g must be Hurwitz stable for all uncertainty ranges.
- A_m does not depend on the uncertainty bound, i.e., no tuning required.
- Given posterior distribution, we update control gain k to guarantee A_g is Hurwitz.

SIMULATION: FRRF



- FRRF generates prior estimation heatmap characterized by mean and variance.
- Estimation performance has been compared with full measurement and sparse measurements in the presence of modeling uncertainty.

SIMULATION: PROACTIVE ADAPTIVE CONTROL (CURVY ROAD)



- Proactive control maintains desired transient and steady state performance for all environmental conditions.
- The velocity design program provides maximum safe velocity.

REFERENCES

[Full version] H. Kim, W. Wan, N. Hovakimyan, L. Sha, and P. Voulgaris, "Robust Vehicle Lane Keeping Control with Networked Proactive Adaptation", arXiv e-prints, pp.arXiv-2009, 2020.
N. Hovakimyan, and C. Cao, " \mathcal{L}_1 adaptive control theory: Guaranteed robustness with fast adaptation", Society for Industrial and Applied Mathematics, 2010.