MECH876 – Intelligent Vehicle Control Systems

Unscented Kalman Filter based Super Twisting Control for Half-car Suspension system

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- The presentation aims to provide the key concept, procedure and obtained results without being mathematically intensive.
- Details of the project can be found in the report.

Introduction – Project Motivation



Active suspension control:

- Better maneuver
- Safer
- Higher driving comfort
- Lower maintenance cost



Introduction – Project Objectives

(Primary) To stabilize both vertical and pitch motions of a half-car suspension system

• (Auxiliary) To guarantee road holding and driving comfort

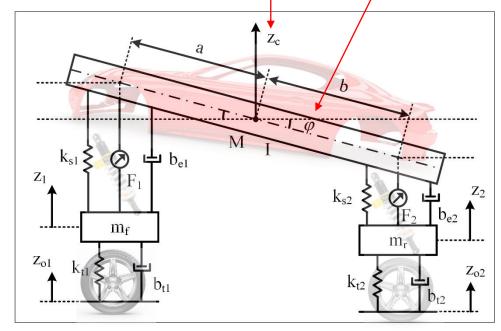


Figure 1: Modeling diagram of a half-car suspension system*

Why Half-car? - Fidelity

- Quarter-car: not sophisticated enough
- Full-car: too complicated

4 DOF system dynamics

- 4 Outputs: vertical z_c, pitch φ, front z₁, rear z₂
- 2 Inputs: force F₁, force F₂

Underactuated system

→ cannot control all 4 outputs!

(*) W. Sun, H. Pan, and H. Gao, "Filter-based adaptive vibration control for active vehicle suspensions with electrohydraulic actuators," IEEE Transactions on Vehicular Technology, vol. 65, no. 6, pp. 4619-4626, 2015.



Introduction- Obstacles and Solutions

(1) Robustness:

- Nonlinearity
- Parameter uncertainty
- External disturbance

Robust controller Super twisting sliding mode control

 k_{t2}

Figure 1: Modeling diagram of a half-car suspension system

(2) State availability:

Requirement of direct displacement & velocity feedback of vertical & pitch motion

High cost of sensor installation

Complexity in physical design

I need direct Sta

State estimation

Unscented Kalman filter

need relative

displacemer

Higher-order sliding mode observer



Methodology – Design Procedure

Control-Oriented Modeling



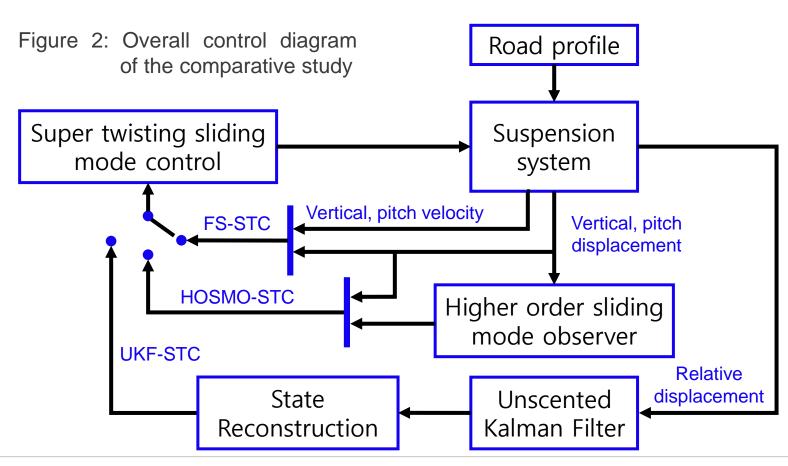
(Observer/Filter) + Controller



Comparative Simulation

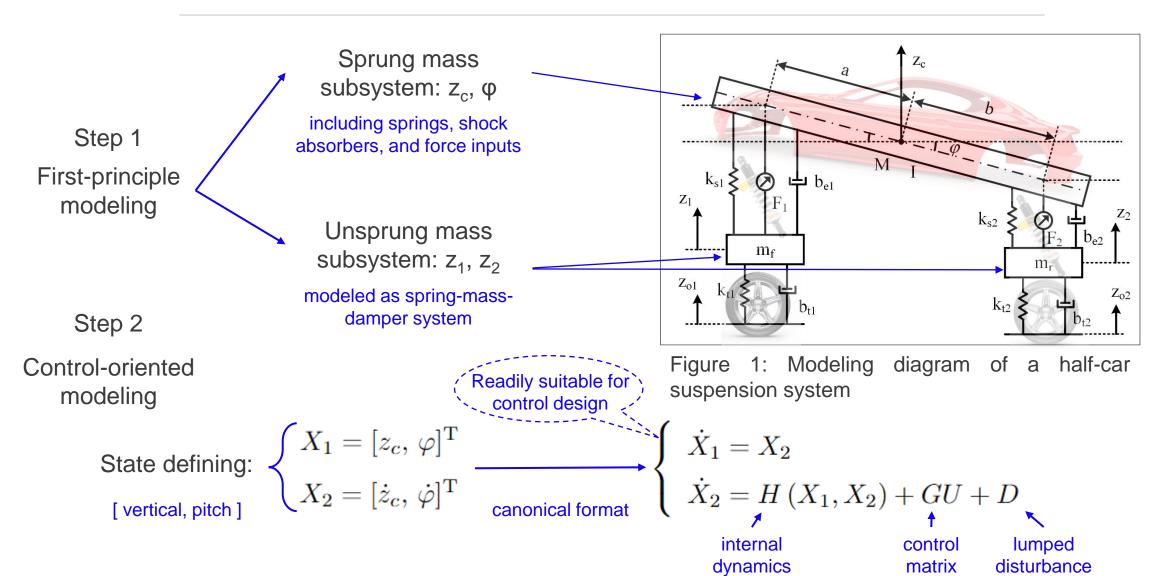
Three controller candidates:

- FS-STC (Full state feedback): requires both direct displacement & velocity measurements of vertical and pitch motions
- HOSMO-STC (Semi state feedback): requires only direct displacement measurement of vertical and pitch motions
- UKF-STC (No state feedback): no direct measurement required, needs only relative displacement measurements



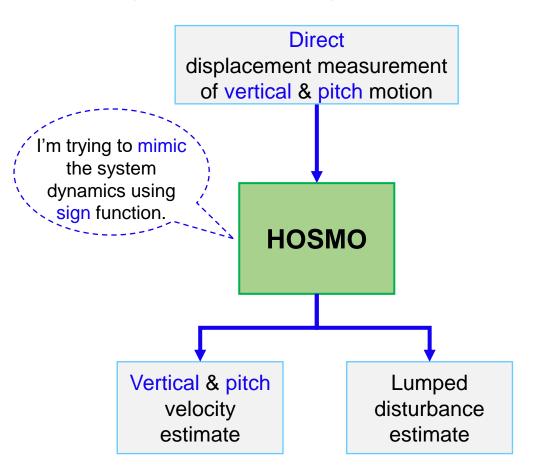


Design Procedure – Control-oriented Modeling

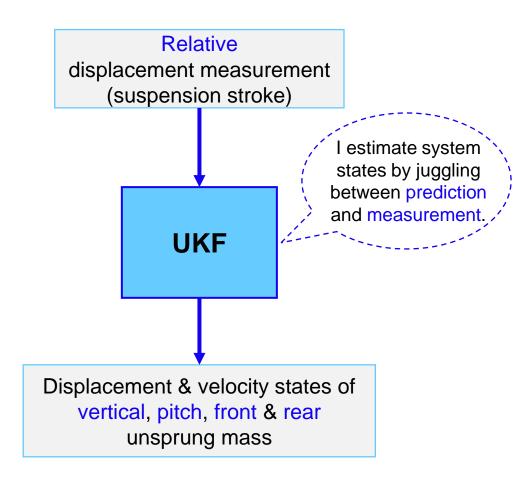


Design Procedure – Observer/Filter Design

Higher-order Sliding Mode Observer



Unscented Kalman Filter





Design Procedure – Controller Design

Step 1

Standard Sliding Mode Control design

Step 2

Sliding surface design

Control law derivation

To design how the system states will behave after the sliding mode occurs

To ensure the sliding mode happens despite disturbance and dynamics

$$S = C_1 X_1 + X_2$$

FS-STC

- Straight forward; requires both direct displacement & velocity state measurement
- Control gains must be sufficiently large to compensate for unknown disturbance

$$\hat{S} = C_1 X_1 + \hat{X}_2$$

HOSMO-STC

- Requires only direct displacement state
- Uses estimates of velocity state and lumped disturbance
- Observer gains must be large enough for disturbance estimation

$$\hat{S} = C_1 \hat{X}_1 + \hat{X}_2$$

UKF-STC

- No direct measurement required
- Reconstructs & utilizes system states and disturbance
- Control gains are small

State-feedback dependency reduction direction



Simulation setups:

Environment: MATLAB – Simulink

Sampling time: 0.1 (ms)

Solver: ODE 4 Runge-Kutta

Disturbance inputs: road with 2 (cm)-bump at 1 (s)

Controller gains: same level for all controllers

Observer/Filter gains: trial-and-error

UKF-STC outperforms FS-STC and HOSMO-STC in convergence time, accuracy, and chattering attenuation.

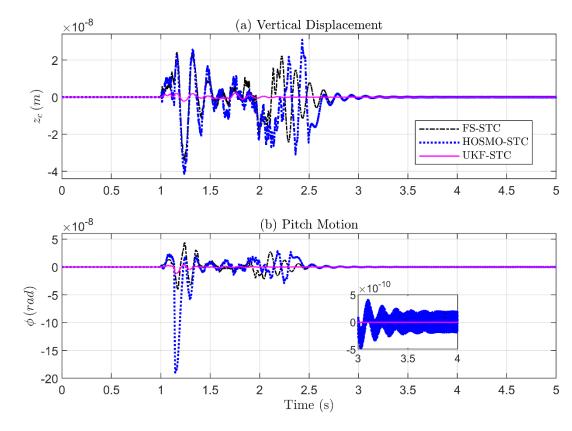


Figure 3: Stabilization performance of all control candidates under bumped road input at 1 (s).



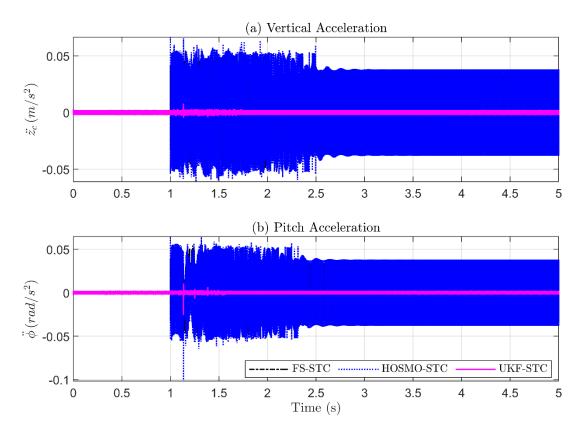


Figure 4: Driving comfort evaluation in term of acceleration minimization of all control candidates under bumped road input at.

- UKF-STC outperforms FS-STC and HOSMO-STC in acceleration minimization, chattering attenuation, trajectory consistency.
- FS-STC & HOSMO-STC exhibits chattering phenomenon even after the bump, indicating a sacrifice of smoothness for robustness.



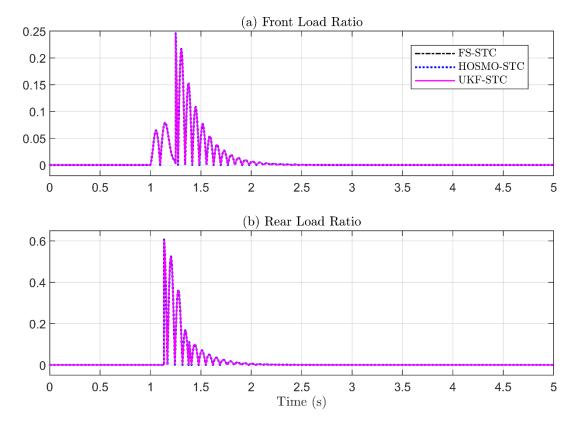


Figure 5: Road holding assessment with dynamic-static load ratio of all control candidates under bumped road input.

Road – holding ratio =
$$\left| \frac{\text{Dynamic tire load}}{\text{Static tire load}} \right| < 1$$

Keeps the tires on the ground!

All controllers satisfy the road-holding requirement.



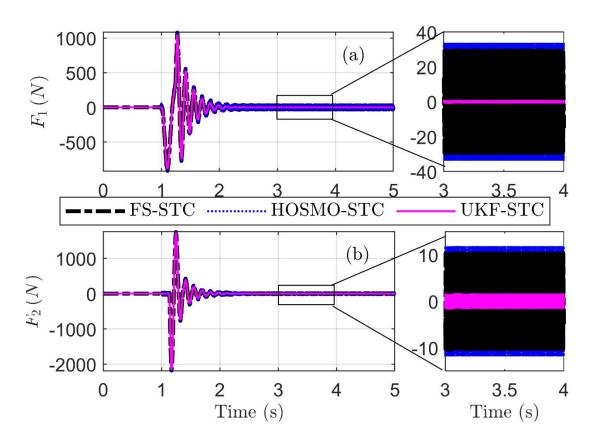


Figure 6: Control inputs in the form of ideal force generator of all control candidates under bumped road input.

- Similar shape of control trajectory for all controllers.
- UKF-STC exhibits much less chattering!

 → advantage if the actuator dynamics is included in control design.



HOSMO State Estimation Results

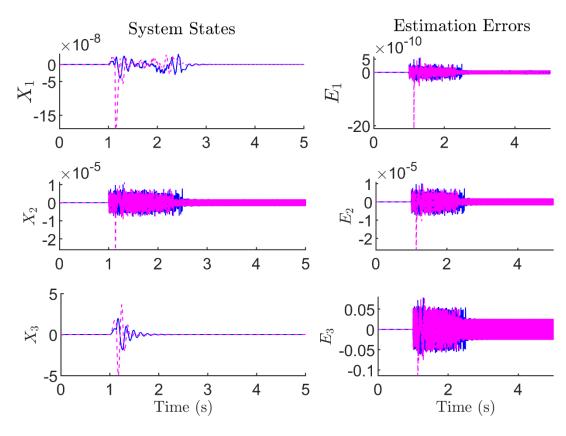


Figure 7: State & lumped disturbance estimation performance of HOSMO under bumped road input.

RMS Error

 $[7.9749, 10.593] \times 10^{-11}$

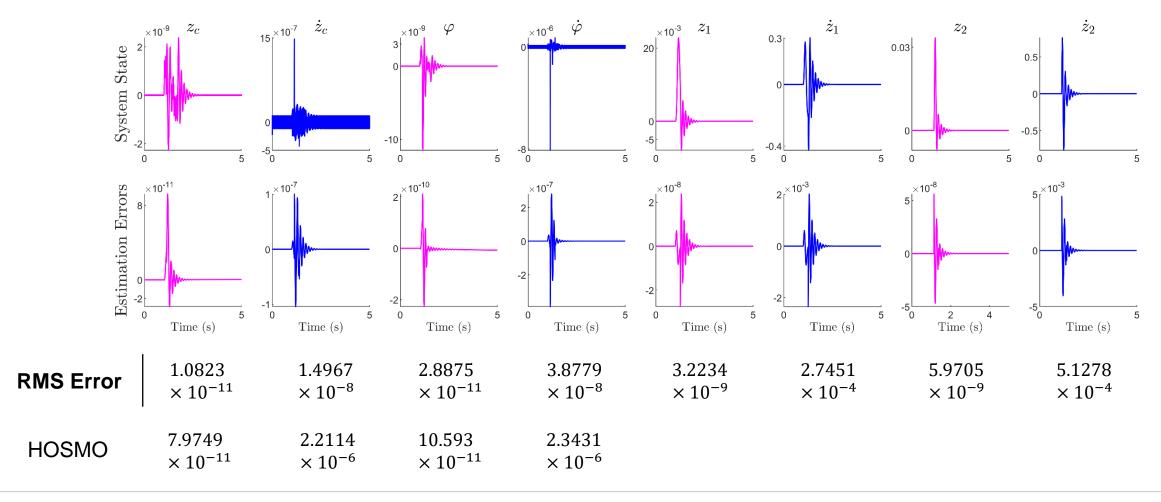
 $[2.2114, 2.3431] \times 10^{-6}$

 $[1.87, 1.90] \times 10^{-2}$



UKF State Estimation Results

Figure 8: State estimation precision of UKF under bumped road input.





Comparative Simulation Results – Noisy Measurement

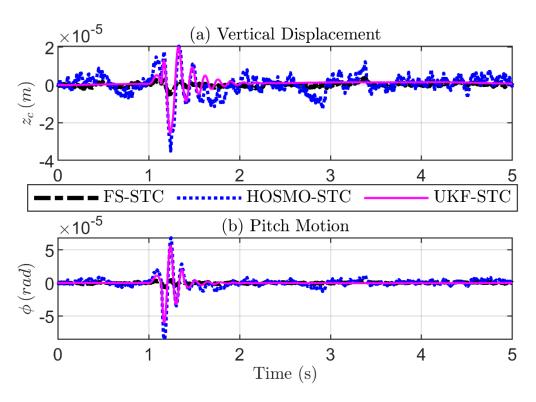


Figure 9: Stabilization performance of all control candidates under bumped road input with noisy measurement.

- Control precision degrades due to noise
- UKF-STC is still comparable to FS-STC and HOSMO-STC in terms of convergence time, accuracy.
- UKF-STC trajectory is more stable.



Controller Comparison

	Advantage	Disadvantage
FS-STC	Straight-forward control design	 Requires direct displacement & velocity me asurements of both vertical & pitch motion Large control gains associated with sign function → prone to chattering
HOSMO-STC	 Velocity state of vertical & pitch motion is estimated → less dependent on direct state-feedback information 	
	Lumped disturbance is estimated and used in control design	 Large observer gains associated with sign function → prone to chattering*
UKF-STC	 Uses only easily accessible relative displacement (suspension stroke) All system states can be estimated → internal dynamics & disturbance compensation → small control gains → less chattering 	 Control design is indirect by state estimates → performance strongly depends on UKF accuracy → careful parameter tuning



Conclusion & Future Work

Summary:

Objective: Vertical & pitch stabilization of a half-car suspension system & ensuring road

holding and driving comfort

Methodology: Super twisting sliding mode control (main controller), implementation of HOSMO

(observer) or UKF (filter) to be less dependent on state feedback

Result: UKF-STC demonstrates superior performance over FS-STC and HOSMO-STC

Limitations suggesting future work:

- Lack of intensive assessment on influence of noise and parameter uncertainty on control performance
- Missing tuning guideline for process & measurement noise covariance of UKF
- Exclusion of actuator dynamics in control design





경청해주셔서 감사합니다.



Appendix – First-principle Modeling

(modeled as spring-mass damper system)

Unsprung mass subsystem:
$$\begin{cases} m_f \ddot{z}_1 - k_{s1} \Delta y_1 - b_{e1} \Delta \dot{y}_1 + k_{t1} (z_1 - z_{o1}) + b_{t1} (\dot{z}_1 - \dot{z}_{o1}) = -F_1 \\ m_r \ddot{z}_2 - k_{s2} \Delta y_2 - b_{e2} \Delta \dot{y}_2 + k_{t2} (z_2 - z_{o2}) + b_{t2} (\dot{z}_2 - \dot{z}_{o2}) = -F_2 \end{cases}$$

Relative displacement
$$\Delta y_1 = z_c + a \sin \varphi - z_1, \quad \Delta \dot{y}_1 = \dot{z}_c + a \dot{\varphi} \cos \varphi - \dot{z}_1 \\ \Delta y_2 = z_c - b \sin \varphi - z_2, \quad \Delta \dot{y}_2 = \dot{z}_c - b \dot{\varphi} \cos \varphi - \dot{z}_2$$
 Relative velocity

Sprung mass subsystem:

(including springs, shock absorbers, and force inputs)

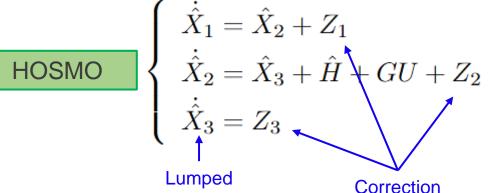
$$\begin{cases} M\ddot{z}_{c} = -\phi_{1}(t) + F_{1} + F_{2} \\ I\ddot{\varphi} = -\phi_{2}(t) + aF_{1} - bF_{2} \end{cases}$$

$$\phi_{1}(t) = k_{s1}\Delta y_{1} + k_{s2}\Delta y_{2} + b_{e1}\Delta \dot{y}_{1} + b_{e2}\Delta \dot{y}_{2}$$

$$\phi_{2}(t) = a(k_{s1}\Delta y_{1} + b_{e1}\Delta \dot{y}_{1}) - b(k_{s2}\Delta y_{2} + b_{e2}\Delta \dot{y}_{2})$$



Appendix – HOSMO Formulation



disturbance estimate

 $Z_1 = K_1 |E_1|^{2/3} \operatorname{sign}(E_1)$

$$Z_2 = K_2 |E_1|^{1/3} \text{sign}(E_1)$$

$$Z_3 = K_3 \mathrm{sign}\left(E_1\right)$$

terms

$$E_1 = X_1 - \hat{X}_1$$

First state estimation error

Observer gains K_i (i=1,2,3) must be appropriately selected for the convergence of the observer.

$$K_1 = \rho_1 L^{1/3}$$

 $K_2 = \rho_2 L^{1/2}$
 $K_3 = \rho_3 L$

$$Z_2 = K_2 |E_1|^{1/3} \mathrm{sign}\left(E_1\right)$$
 $\rho_i(i=1,2,3)$: tuning parameters

L: Lipschitz constant

must be sufficiently large for high precision!



Appendix – Standard UKF Formulation*

Initialization:

$$\hat{X}_0 = E[X_0], \quad P_0 = E[(X_0 - \hat{X}_0)(X_0 - \hat{X}_0)^T]$$

Sigma points calculation:

$$X_{k-1} = [\hat{X}_{k-1} \quad \hat{X}_{k-1} + \eta \sqrt{P_{k-1}} \quad \hat{X}_{k-1} - \eta \sqrt{P_{k-1}}]$$

Time update:

$$X_{k|k-1} = F(X_{k-1}, U_{k-1})$$

$$\hat{X}_{k}^{-} = \sum_{i=0}^{2N} \omega_{i}^{(m)} X_{i,k|k-1}$$

$$P_{k}^{-} = \sum_{i=0}^{2N} \omega_{i}^{(c)} [X_{i,k|k-1} - \hat{X}_{k}^{-}] [X_{i,k|k-1} - \hat{X}_{k}^{-}]^{T} + Q$$

$$Y_{k|k-1} = J(X_{k|k-1})$$

$$\hat{Y}_{k}^{-} = \sum_{i=0}^{2N} \omega_{i}^{(m)} Y_{i,k|k-1}$$

Scaling factors:
$$\begin{aligned} \eta &= \sqrt{N+\lambda} \\ \lambda &= N(\alpha^2-1) \end{aligned}$$

$$\omega_0^{(m)}=\lambda/(N+\lambda),\quad \omega_0^{(c)}=\lambda/(N+\lambda)+(1-\alpha^2+\beta)$$
 weights:
$$\omega_i^{(m)}=\omega_i^{(c)}=1/\{2(N+\lambda)\},\quad i=1,2,...,2N$$

Measurement update:

$$P_{\tilde{Y}_{k}\tilde{Y}_{k}} = \sum_{i=0}^{2N} \omega_{i}^{(c)} [Y_{i,k|k-1} - \hat{Y}_{k}^{-}] [Y_{i,k|k-1} - \hat{Y}_{k}^{-}]^{T} + R$$

$$P_{X_{k}Y_{k}} = \sum_{i=0}^{2N} \omega_{i}^{(c)} [X_{i,k|k-1} - \hat{X}_{k}^{-}] [Y_{i,k|k-1} - \hat{Y}_{k}^{-}]^{T}$$

$$K_{k} = P_{X_{k}Y_{k}} P_{\tilde{Y}_{k}\tilde{Y}_{k}}^{-1}$$

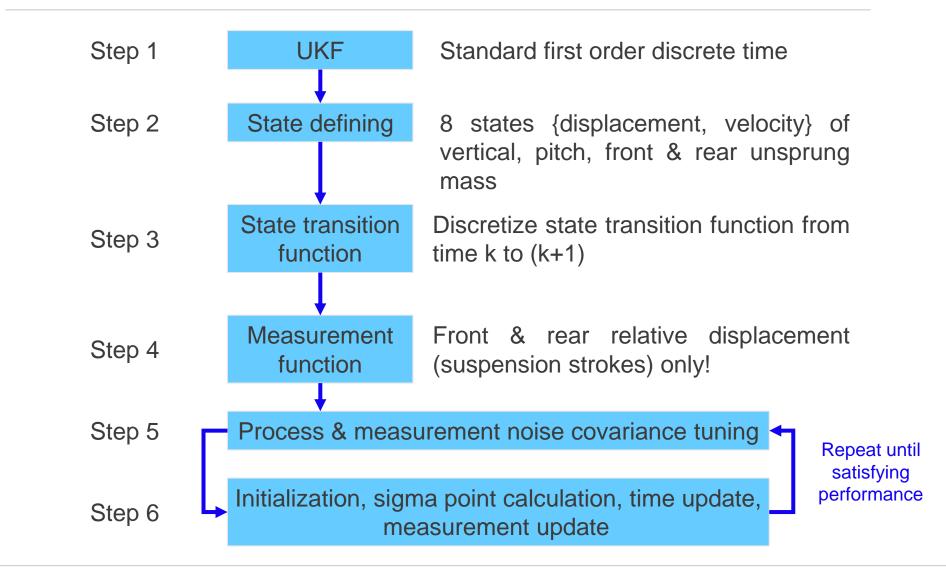
$$\hat{X}_{k} = \hat{X}_{k}^{-} + K_{k} (Y_{k} - \hat{Y}_{k}^{-})$$

$$P_{k} = P_{k}^{-} - K_{k} P_{\tilde{Y}_{k}\tilde{Y}_{k}} K_{k}^{T}$$

(*) R. Van Der Merwe and E. A. Wan, "The square-root unscented Kalman filter for state and parameter-estimation," in 2001 IEEE international conference on acoustics, speech, and signal processing. Proceedings (Cat. No. 01CH37221), vol. 6. IEEE, 2001, pp. 3461–3464.



Appendix – UKF Design Procedure in Simulink





Appendix – Controller Formulation

Sliding surface

Control law

FS-STC

$$S = C_1 X_1 + X_2$$

$$U = G^{-1} \left(-C_1 X_2 - H - \lambda_1 |S|^{1/2} sign\left(S\right) - \int_0^t \lambda_2 sign\left(S\right) d\tau \right)$$
 state feedback

HOSMO-STC

$$\hat{S} = C_1 X_1 + \hat{X}_2$$

Cause of chattering
$$Z_2 = K_2 |E_1|^{1/3} \mathrm{sign}\,(E_1)$$

$$U = G^{-1} \left(-C_1 \hat{X}_2 - \hat{X}_3 - \hat{H} + Z_2 - \lambda_1 \left| \hat{S} \right|^{1/2} sign(\hat{S}) + \int_0^t \lambda_2 sign(\hat{S}) d\tau \right)$$
 State Disturbance small estimate estimate

UKF-STC

$$\hat{S} = C_1 \hat{X}_1 + \hat{X}_2$$

$$\hat{S} = C_1 \hat{\bar{X}}_1 + \hat{\bar{X}}_2 \qquad U = G^{-1} (-C_1 \hat{\bar{X}}_2 - \hat{\bar{H}} - \hat{\bar{D}} - \lambda_1 |S|^{\frac{1}{2}} sign\left(S\right) - \int_0^t \lambda_2 sign\left(S\right) d\tau)$$

State estimate **Disturbance** estimate

