

ASEN 6080 HW 4

- dan Zuber, 108577813

- a. make sure you have a working LKF with SNC.

It works! See PDF

- b. Implement the sequential filter smoothing algorithm, including the covariance matrix.

See PDF!

- c. Compare processing a set of data without process noise to processing with a batch after 1 iteration.

See PDF!

My smoothed LKF results without process noise look almost identical to a Batch filter after one iteration.

This makes sense, as the smoothed covariance without process noise is essentially just the a priori covariance mapped back in time, while the batch maps the estimated initial covariance forward in time. Since theoretically $P_{LKF, F} = P_{Batch, 0}$, the results will be identical.

However, the smoother exhibits some numerical issues; in particular, the reported covariance matrices become non-positive definite near the start of the orbit. This is likely due to saturation of the LKF, which results in a covariance estimate that's too snug to reflect reality. This issue is avoided when we introduce SNC, and hence some semblance of a noise floor to avoid saturation.

d. Take the results of the CKF with SNC using the optimal σ from HW 3 and smooth them. Generate plots of the inertial state errors over time with covariance envelope + provide the RMS state errors.

I used a σ of $1e-8 \text{ km/s}^2$.

See PDF for plots

Component-wise smoothed RMS:

$$\begin{bmatrix} 7.015 \times 10^{-3}, 4.555 \times 10^{-3}, 8.598 \times 10^{-3}, \\ 6.668 \times 10^{-6}, 3.434 \times 10^{-6}, 4.274 \times 10^{-6} \end{bmatrix}^T$$

3D smoothed RMS: 1.199×10^{-2}

e. compare the RMS values from part d to those from HW3 and discuss.

Component-wise SNC RMS:

$$\left[1.722 \times 10^{-2}, 1.550 \times 10^{-2}, 2.282 \times 10^{-2}, \right. \\ \left. 2.884 \times 10^{-5}, 1.492 \times 10^{-5}, 2.152 \times 10^{-5} \right]^T$$

3D SNC RMS: 3.252×10^{-2}

The smoothed state error RMS is much less than with just SNC, with most components, and the 3D RMS being more than halved.

This makes sense, as the point of a smoother is to generate a better state estimate at each measurement time, i.e. reduce the state error. Further, smoothing removes the initial spike in LKF state uncertainty, which contributes heavily to the LKF state error. Removing that spike then intuitively reduces the state error RMS over the whole dataset.

ASEN 6080 HW 3 Problem 1 Main Script

Table of Contents

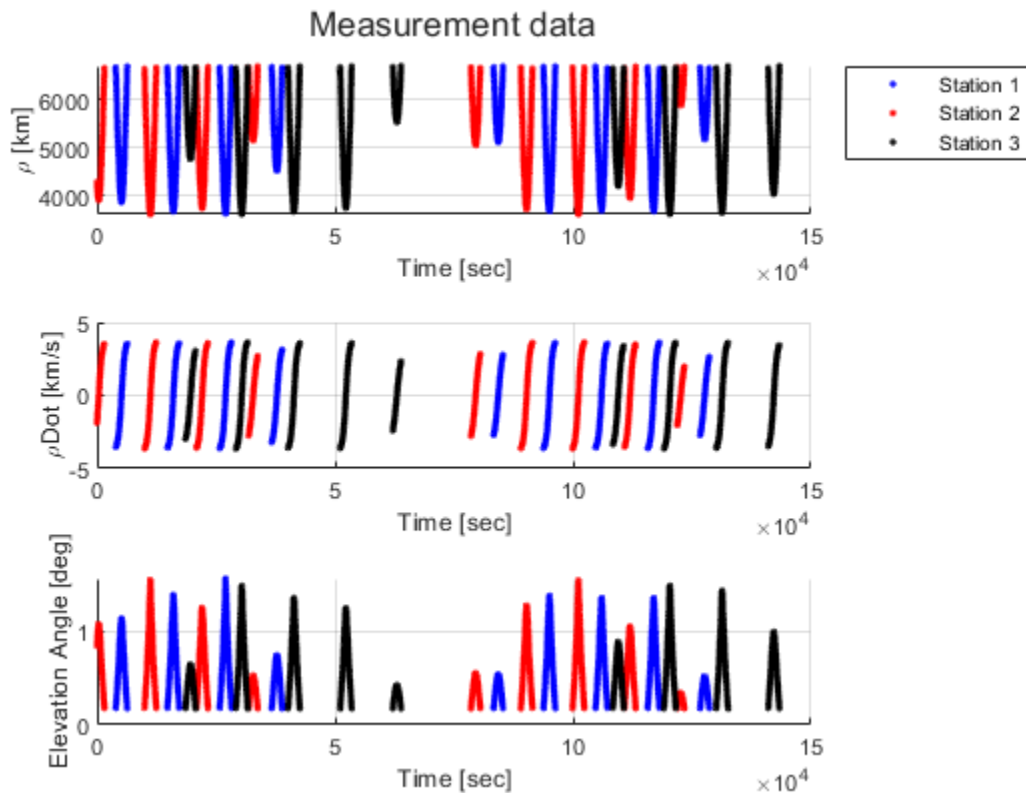
Housekeeping	1
Setup	1
Make Truth Data	1
Problem 1a: Filter setup	2
Problem 1a: Prove SNC works with optimal sigma	2
Problem 1b/d+e: Implement sequential filter smoothing algorithm and compare RMS values	2
Problem 1c: Smoothing without process noise	2

By: Ian Faber

Housekeeping

Setup

Make Truth Data



Problem 1a: Filter setup

Problem 1a: Prove SNC works with optimal sigma

Based on plots, $\sigma = 1e-8$ balances both postfit and 3D RMS

1a. Running LKF and Smoother with SNC for $\sigma = 1.000e-08$ km/s²

*Running LKF:
Prefit RMS: 242.0616, Postfit RMS: 1.0070. Hit max LKF iterations. Runs so far: 1
Final prefit RMS: 242.0616. Hit maximum number of 1 runs
Final postfit RMS: 1.0070. Hit maximum number of 1 runs*

Problem 1b/d+e: Implement sequential filter smoothing algorithm and compare RMS values

1b/d/e. Smoothing LKF SNC results

Running Smoother...

*LKF w/ SNC State RMS Errors:
Component-wise: [1.722e-02, 1.550e-02, 2.282e-02, 2.684e-05, 1.492e-05, 2.152e-05]
3D: 3.252e-02*

*Smoother w/ SNC State RMS Errors:
Component-wise: [7.015e-03, 4.555e-03, 8.598e-03, 6.668e-06, 5.434e-06, 4.274e-06]
3D: 1.199e-02*

Problem 1c: Smoothing without process noise

1c. Running LKF w/ Smoother and Batch, both without Process noise

*Running LKF:
Prefit RMS: 242.0616, Postfit RMS: 93.4748. Hit max LKF iterations. Runs so far: 1
Final prefit RMS: 242.0616. Hit maximum number of 1 runs
Final postfit RMS: 93.4748. Hit maximum number of 1 runs*

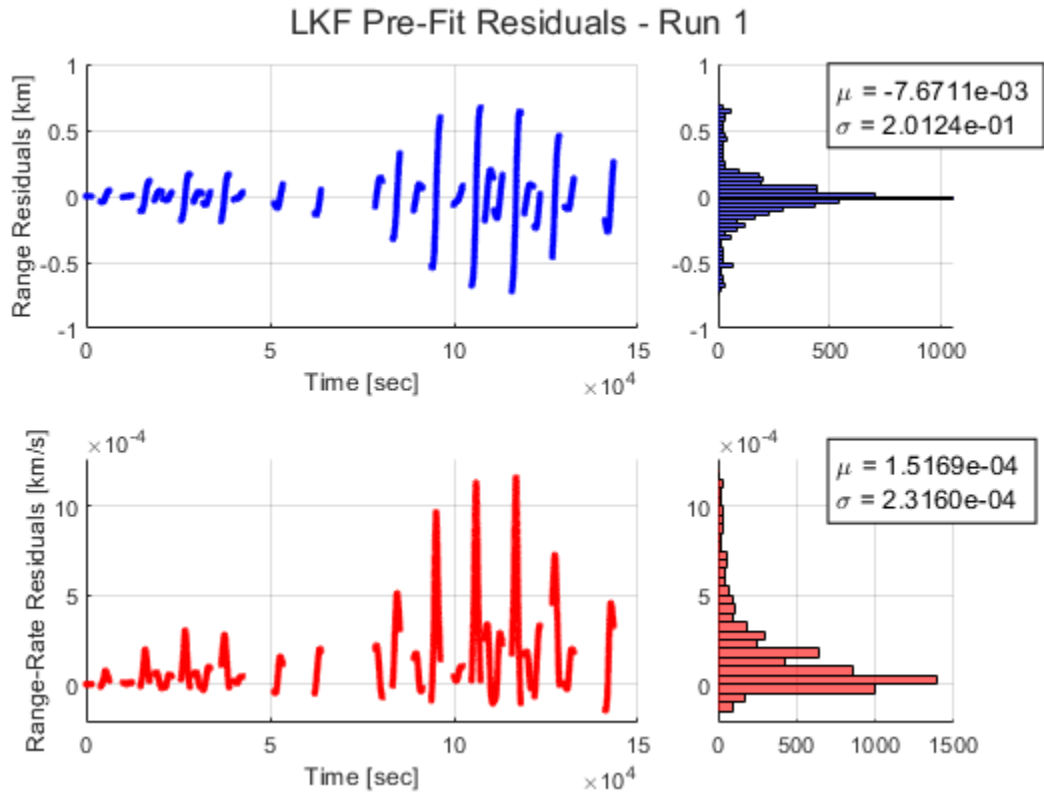
Running Smoother...

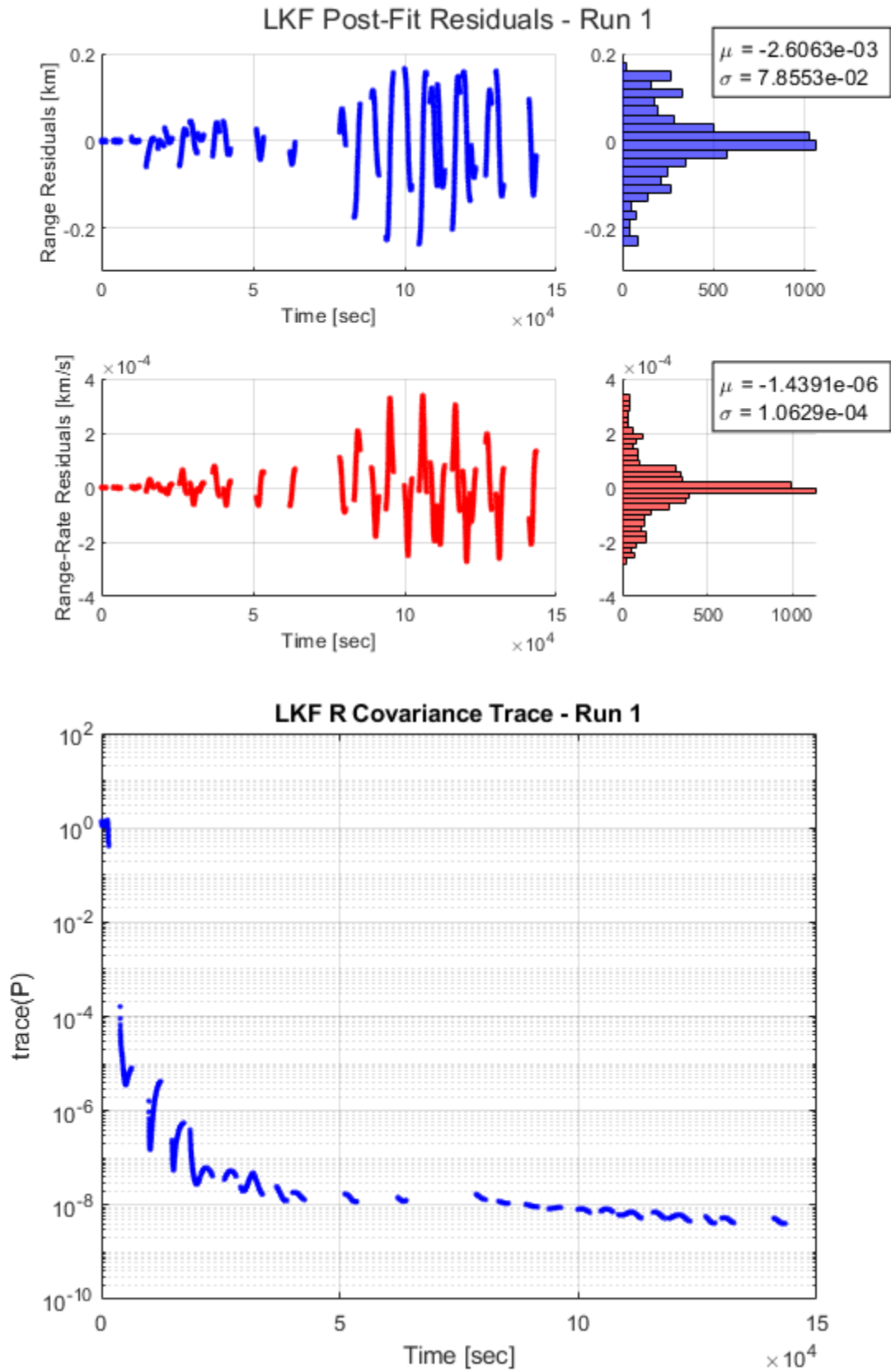
*Running Batch Filter:
Prefit RMS: 242.0619, Postfit RMS: 93.5517. Hit max Batch iterations. Runs*

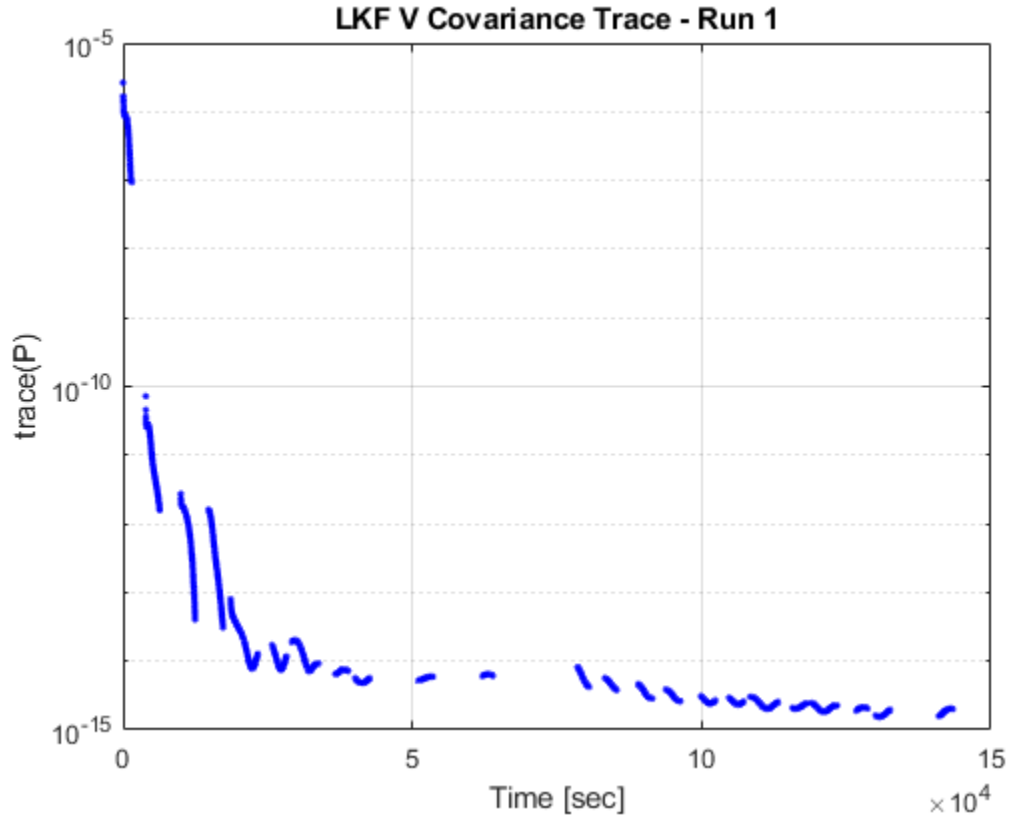
so far: 1

Final prefit RMS: 242.0619. Hit maximum number of 1 runs

Final postfit RMS: 93.5517. Hit maximum number of 1 runs



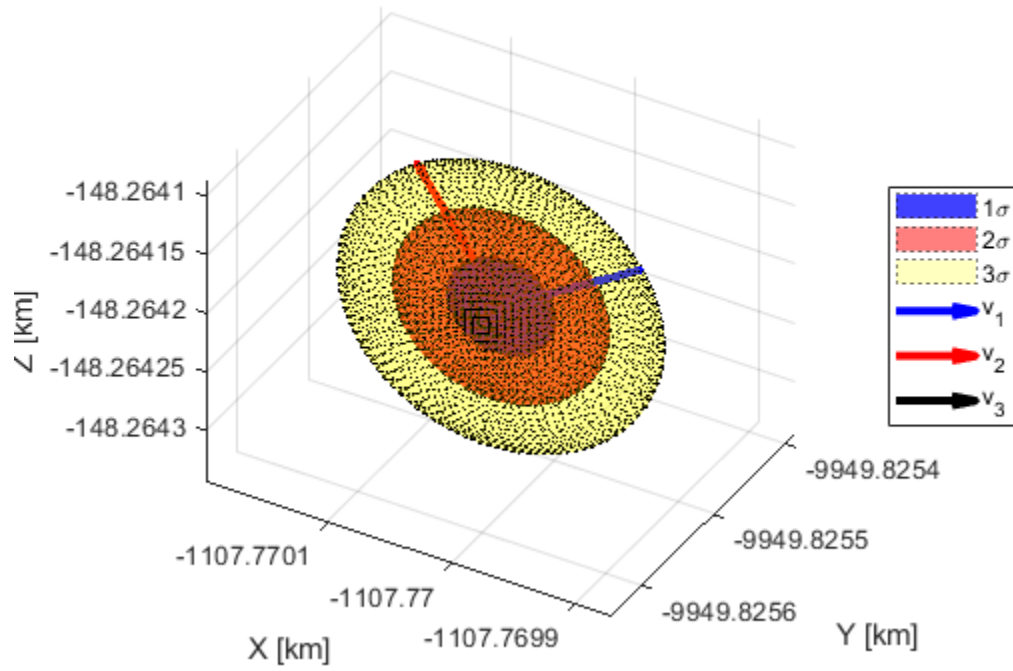




Final LKF Position Covariance Ellipsoid, $t = 143370.000$ sec

$$\mu = [-1.108e+03, -9.950e+03, -1.483e+02]^T \text{ km}$$

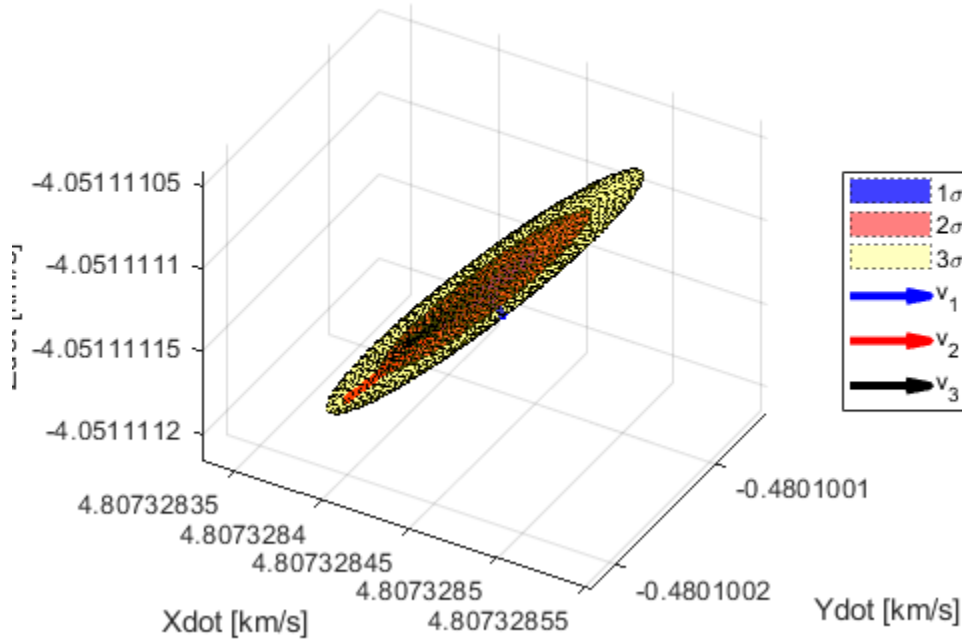
$$\sigma_X = 4.648e-05 \text{ km}, \sigma_Y = 9.785e-06 \text{ km}, \sigma_Z = 4.305e-05 \text{ km}$$



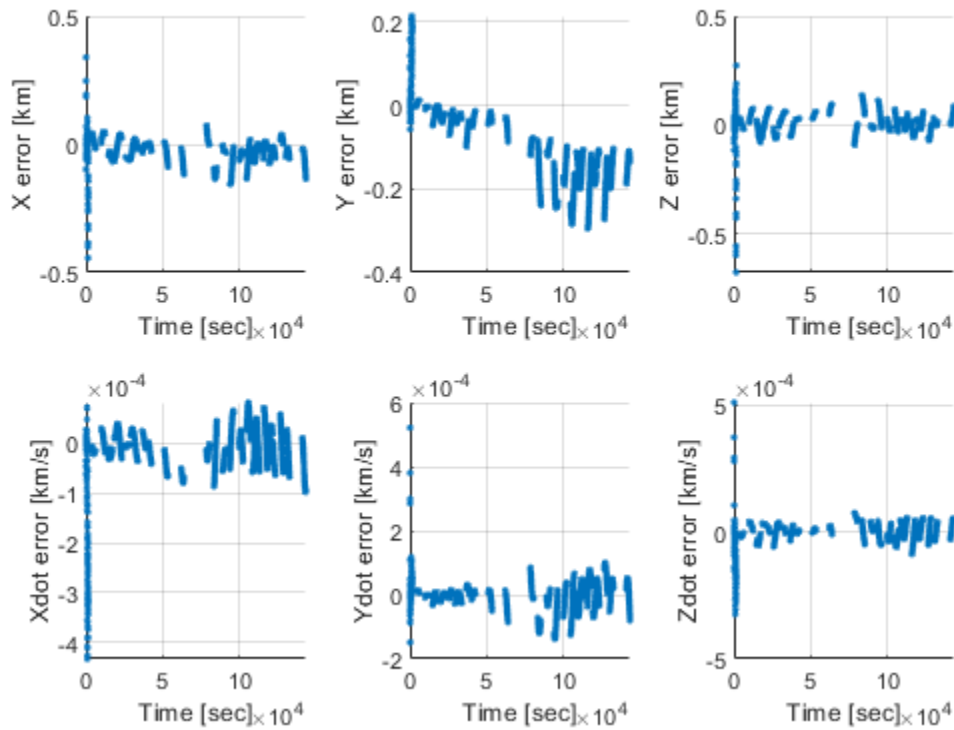
Final LKF Velocity Covariance Ellipsoid, $t = 143370.000$ sec

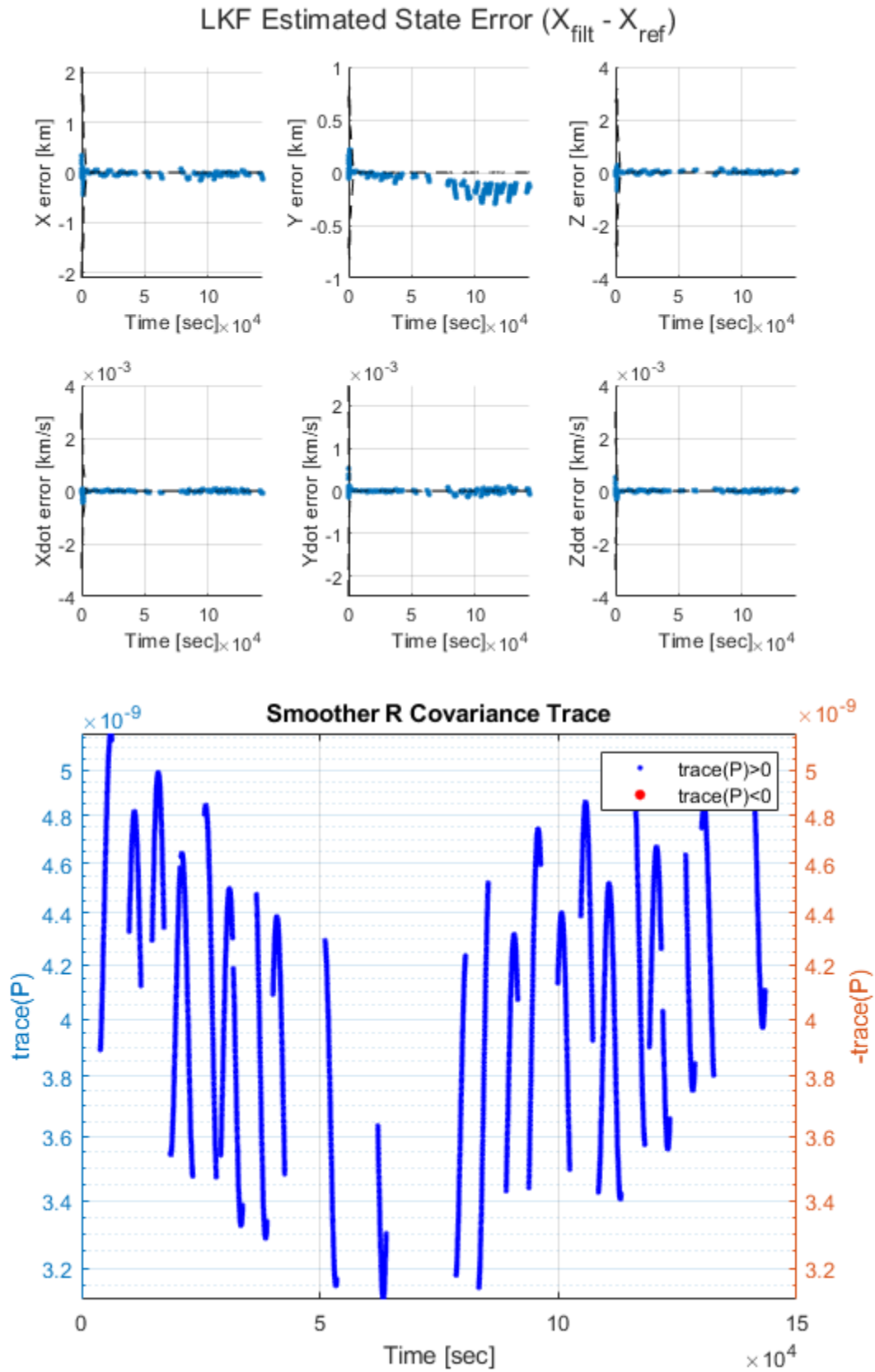
$$\mu = [4.807e+00, -4.801e-01, -4.051e+00]^T \text{ km/s}$$

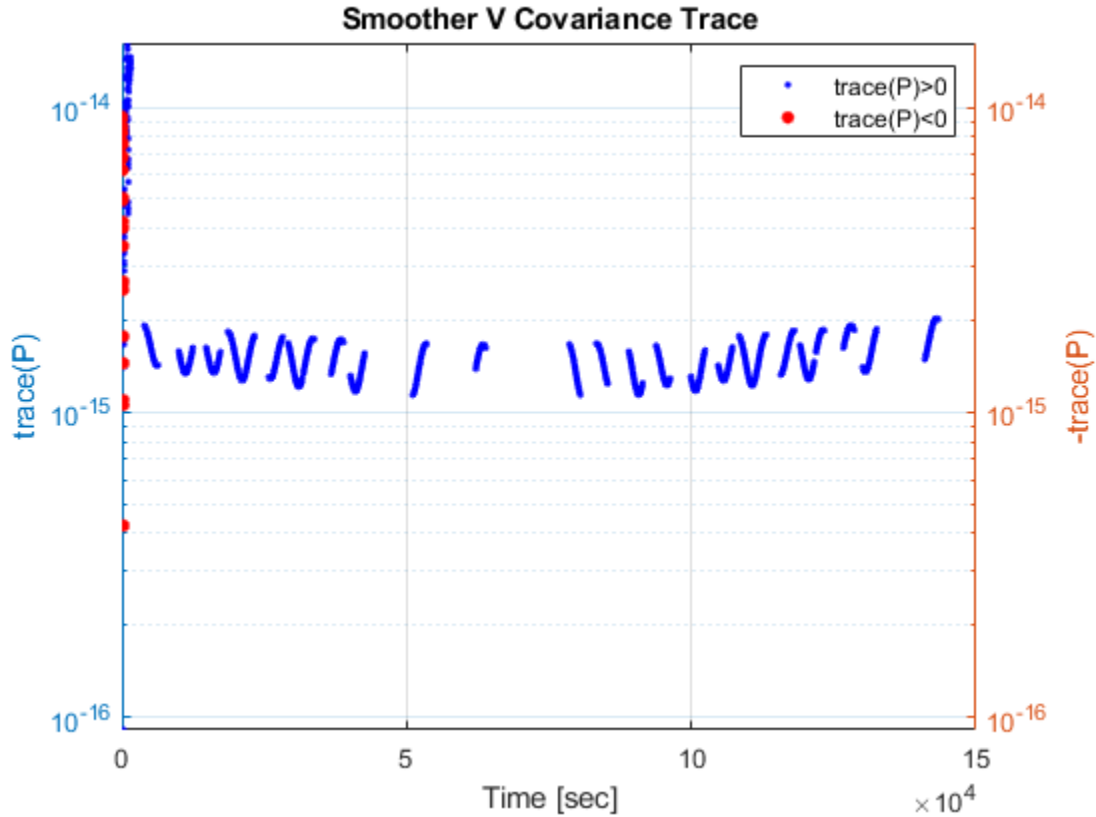
$$\sigma_{\dot{X}} = 2.519e-08 \text{ km/s}, \sigma_{\dot{Y}} = 2.291e-08 \text{ km/s}, \sigma_{\dot{Z}} = 2.928e-08 \text{ km/s}$$



LKF Estimated State Error ($X_{\text{filt}} - X_{\text{ref}}$)



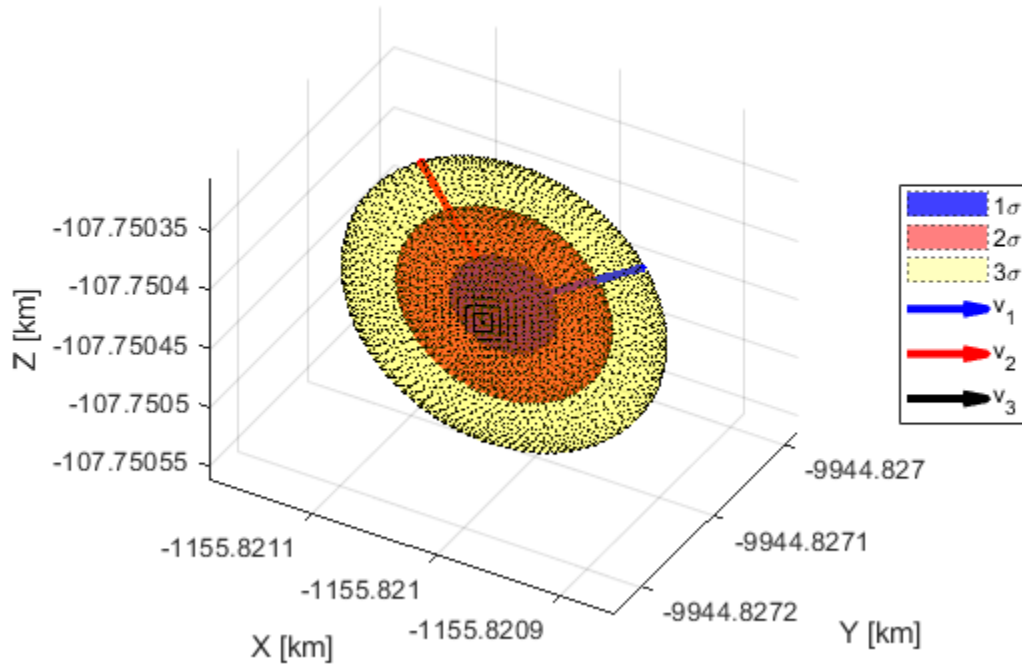




Final Smoother Position Covariance Ellipsoid, $t = 143360.000$ sec

$$\mu = [-1.156e+03, -9.945e+03, -1.078e+02]^T \text{ km}$$

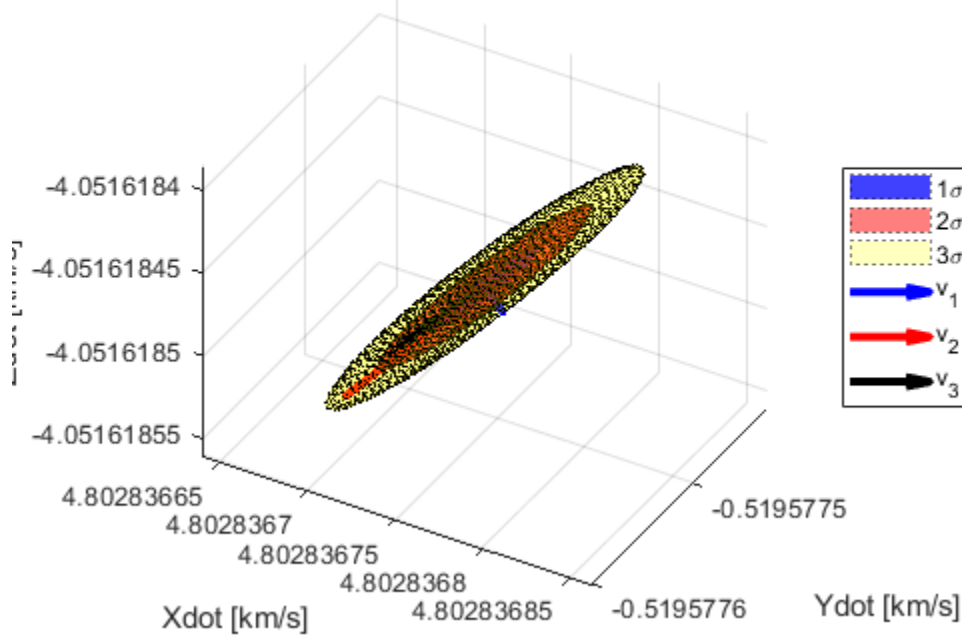
$$\sigma_X = 4.641e-05 \text{ km}, \sigma_Y = 9.827e-06 \text{ km}, \sigma_Z = 4.306e-05 \text{ km}$$



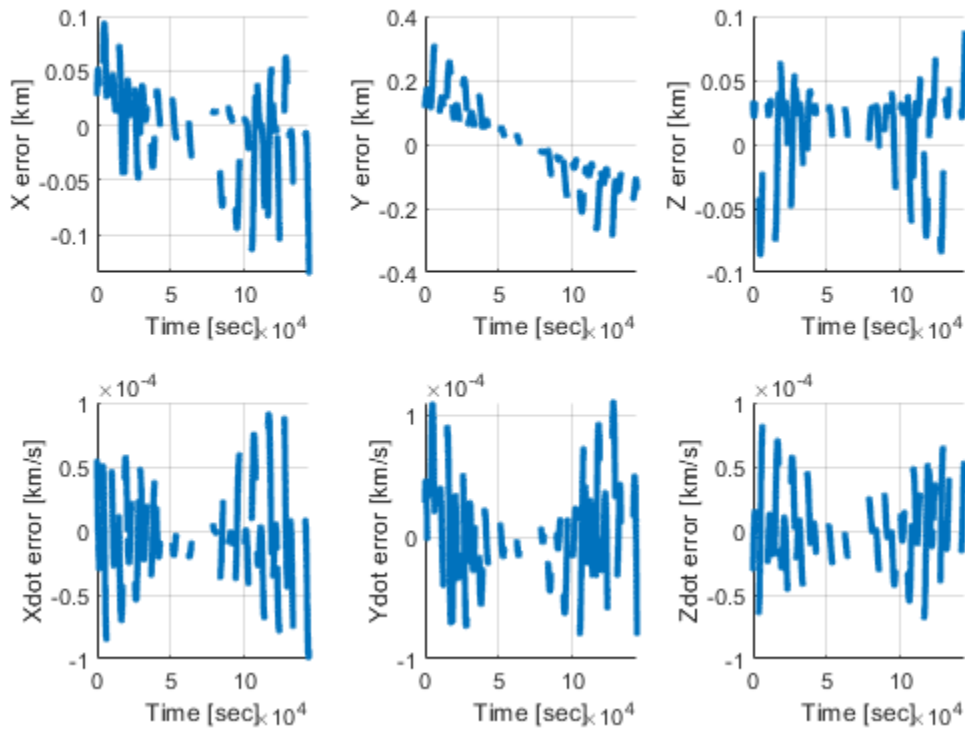
Final Smoother Velocity Covariance Ellipsoid, $t = 143360.000$ sec

$$\mu = [4.803e+00, -5.196e-01, -4.052e+00]^T \text{ km/s}$$

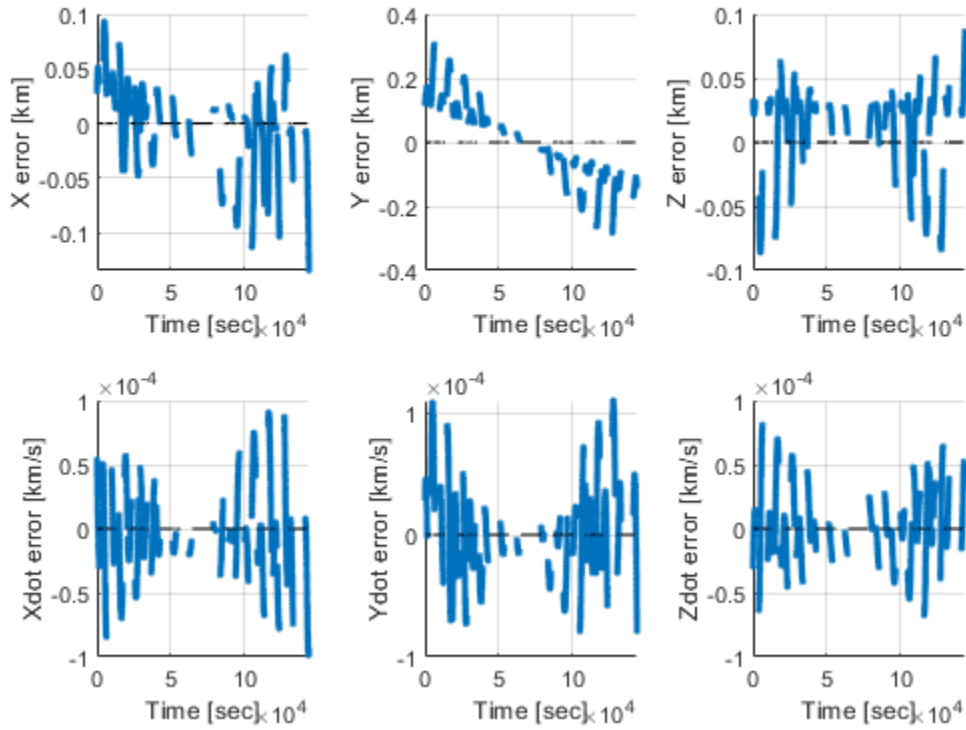
$$\sigma_{Xdot} = 2.524e-08 \text{ km/s}, \sigma_{Ydot} = 2.288e-08 \text{ km/s}, \sigma_{Zdot} = 2.928e-08 \text{ km/s}$$



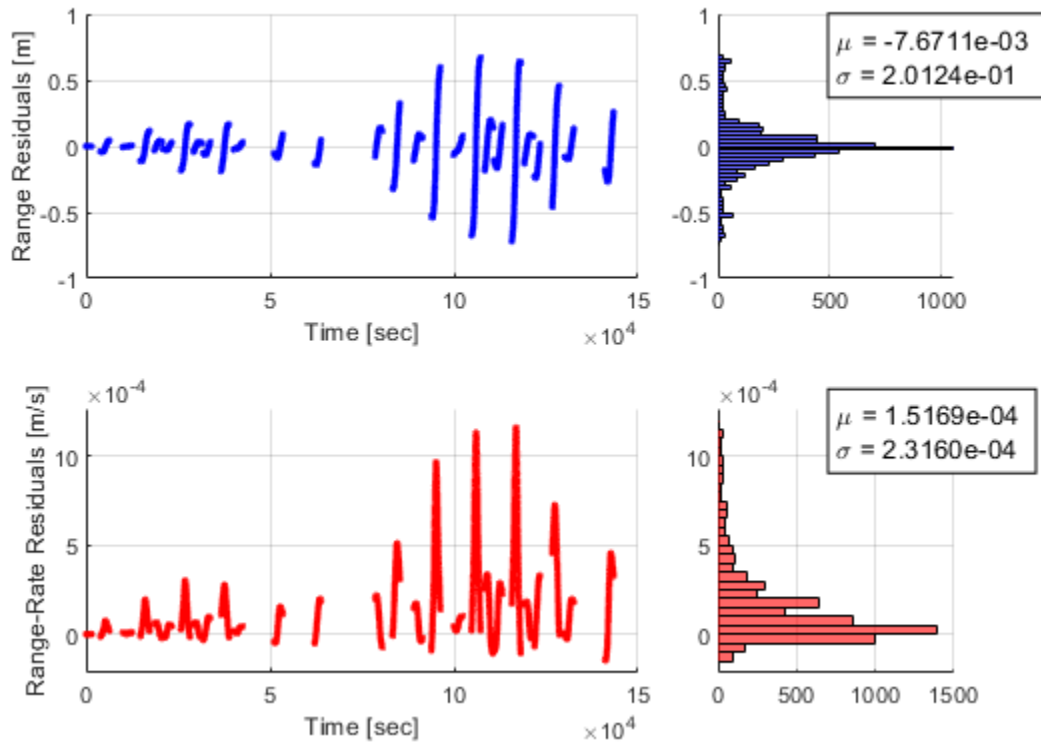
Smoother Estimated State Error ($X_{\text{smooth}} - X_{\text{ref}}$)

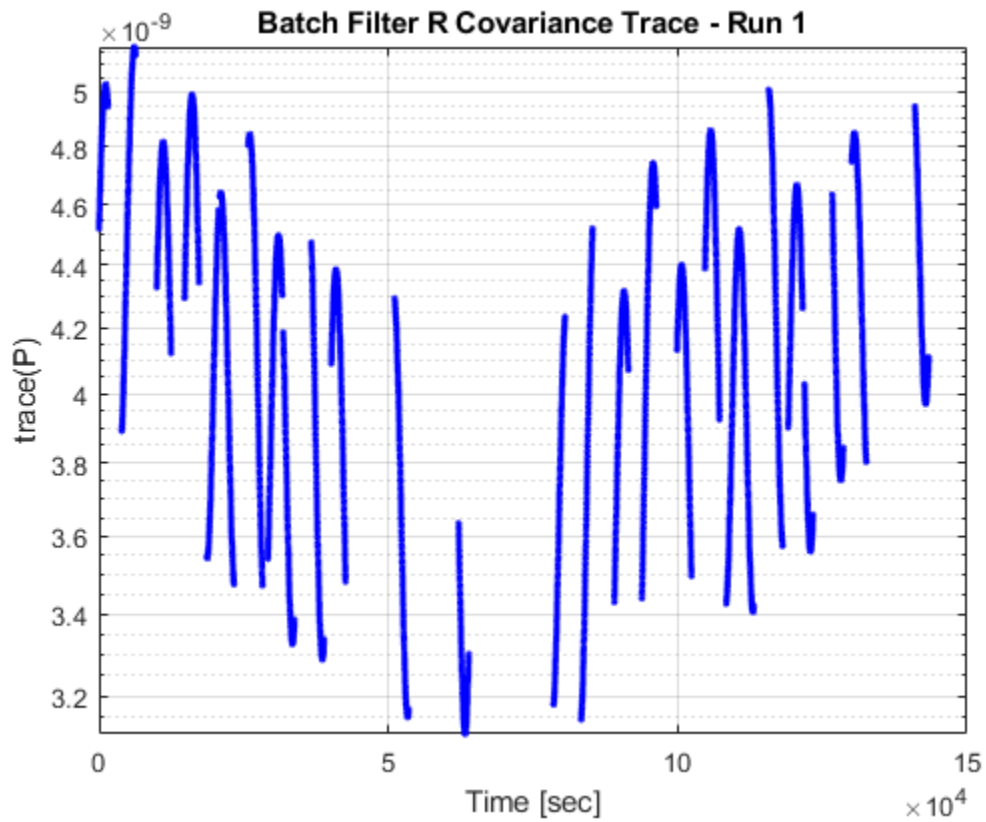
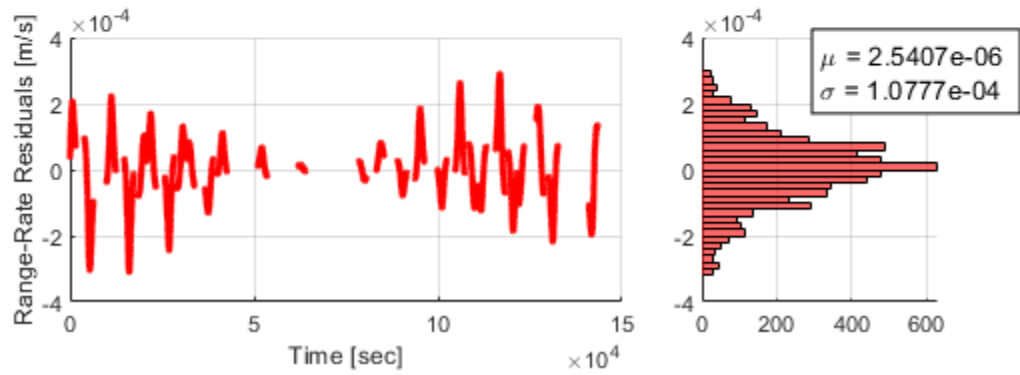
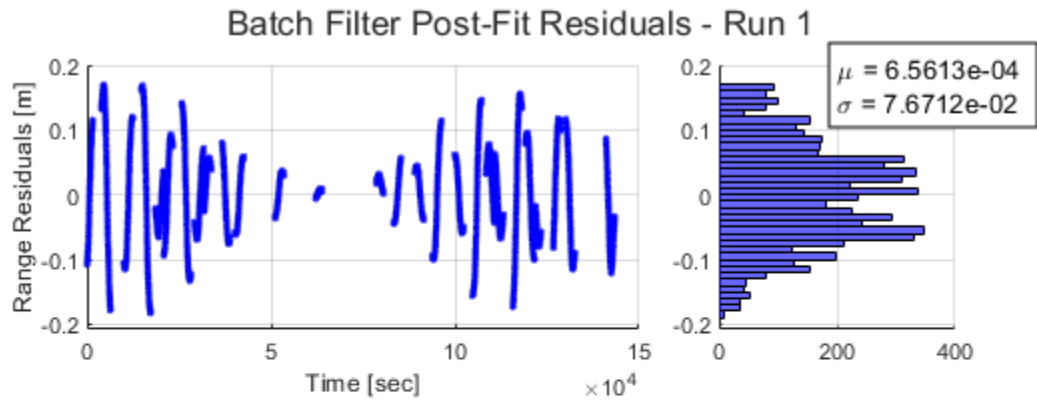


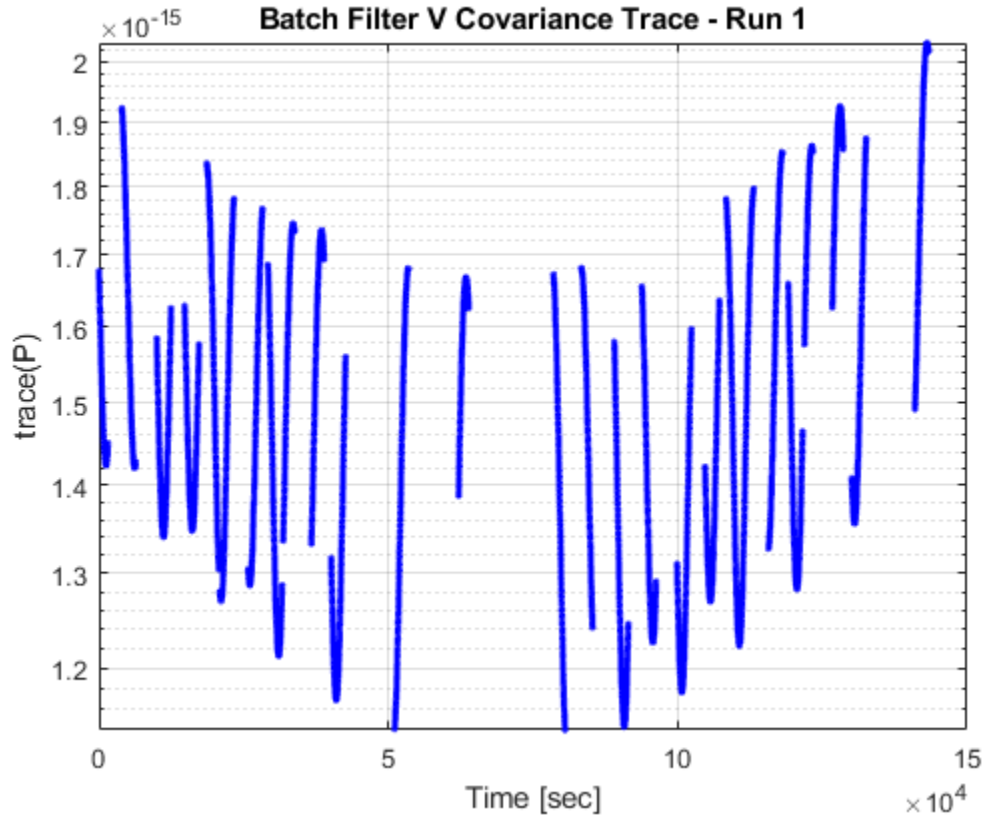
Smoother Estimated State Error ($X_{\text{smooth}} - X_{\text{ref}}$)



Batch Filter Pre-Fit Residuals - Run 1



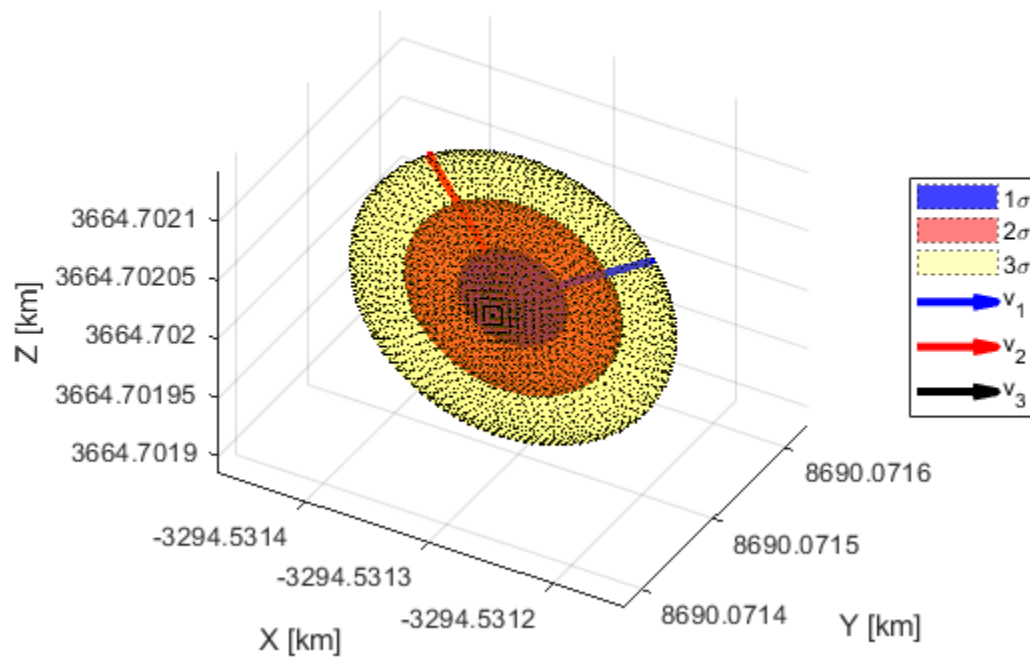




Final Batch Filter Position Covariance Ellipsoid, $t = 149280.000$ sec

$$\mu = [-3.295e+03, 8.690e+03, 3.665e+03]^T \text{ km}$$

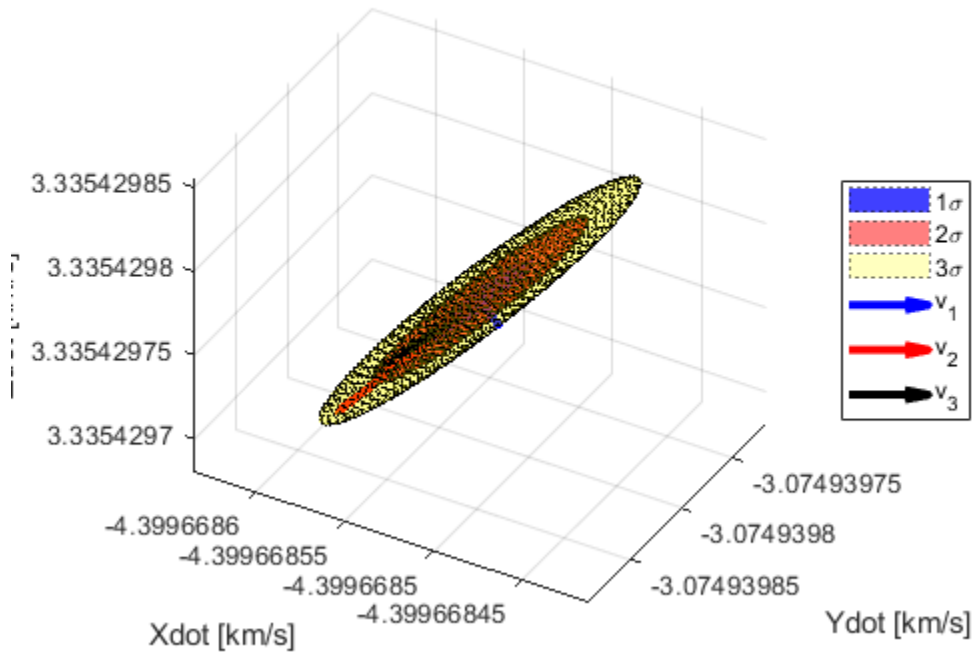
$$\sigma_X = 4.648e-05 \text{ km}, \sigma_Y = 9.785e-06 \text{ km}, \sigma_Z = 4.305e-05 \text{ km}$$



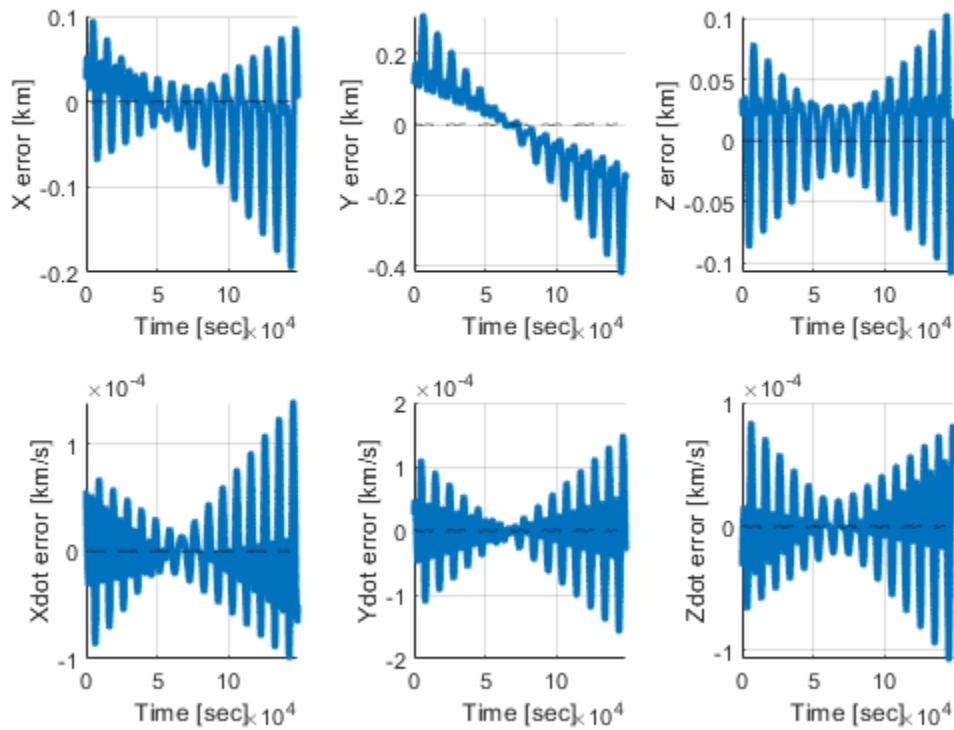
Final Batch Filter Velocity Covariance Ellipsoid, $t = 149280.000$ sec

$$\mu = [-4.400e+00, -3.075e+00, 3.335e+00]^T \text{ km/s}$$

$$\sigma_{Xdot} = 2.519e-08 \text{ km/s}, \sigma_{Ydot} = 2.291e-08 \text{ km/s}, \sigma_{Zdot} = 2.928e-08 \text{ km/s}$$



Batch Filter Estimated State Error ($X_{filt} - X_{ref}$)



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Table of Contents

.....	1
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```
function smoothOut = Smoother(filterOut)
% Function that implements a sequential smoothing algorithm for Stat OD
% problems
% - Inputs:
%   - filterOut: Output structure from an LKF run, as defined in
%                 LKF_SNC.m
% - Outputs:
%   - smoothOut: Smoother output structure with the following fields:
%     - tSmoothed: Smoothed estimate time
%     - xSmoothed: Smoothed deviation estimates:
%                   [xSmoothed_1, xSmoothed_2, ..., xSmoothed_k]
%     - PSmoothed: Smoothed state covariance estimates:
%                   [{PSmoothed_1}, {PSmoothed_2}, ..., {PSmoothed_k}]
%     - XSmoothed: Smoothed full state estimate, calculated as
%                   XSmoothed = Xstar + xSmoothed:
%                   [XSmoothed_1, XSmoothed_2, ..., XSmoothed_k]
%
% By: Ian Faber, 03/03/2025
%
% Process filterOut structure
t = filterOut.t;
xEst = filterOut.xEst;
PBarEst = filterOut.PBarEst;
PEst = filterOut.PEst;
XStar = filterOut.XStar;
Phi = filterOut.Phi;

% Initialize outputs
tSmoothed = [];
xSmoothed = [];
PSmoothed = [];
XSmoothed = [];

% Find value of l
l = size(xEst, 2);

% Run smoother
x_kp1l = xEst(:,l); % Deviation at t_kp1 (t_l) given measurements at t_l
P_kp1l = PEst{1}; % Covariance at t_kp1 (t_l) given measurements at t_l
for k = (l-1):-1:1
    % Pull out filter estimates
    x_kk = xEst(:,k); % Deviation at t_k given measurements at t_k
    P_kp1k = PBarEst{k+1}; % Covariance at t_kp1 given measurements at t_k
    P_kk = PEst{k}; % Covariance at t_k given measurements at t_k
    tSmoothed = [tSmoothed; t(k)];

    % Pull out Phi(t_kp1, t_k)
```

```

Phi_kp1 = Phi{k+1};

    % Smooth estimate
    % S_k = P_kk*Phi_kp1'*(Phi_kp1*P_kk*Phi_kp1' + Q_kp1)^-1;
    S_k = P_kk*Phi_kp1'*(P_kp1k)^-1;
    x_kl = x_kk + S_k*(x_kp1l - Phi_kp1*x_kk);

    % Smooth covariance
    P_kl = P_kk + S_k*(P_kp1l - P_kp1k)*S_k';

    % Save outputs
    xSmoothed = [xSmoothed, x_kl];
    PSmoothed = [PSmoothed; {P_kl}];
    XSmoothed = [XSmoothed, XStar(:,k) + x_kl];

    % Update for next run
    x_kp1l = x_kl;
    P_kp1l = P_kl;

end

    % Assign smoother outputs - need to flip them to be time ascending!
smoothOut.tSmoothed = flipud(tSmoothed);
smoothOut.xSmoothed = fliplr(xSmoothed);
smoothOut.PSmoothed = flipud(PSmoothed);
smoothOut.XSmoothed = fliplr(XSmoothed);

end

```

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