

# Kalman Filtering

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## 1 Introduction

The task is to compute the coordinates and velocities of a moving vehicle. The calculations were done with a constant velocity kinematic model. Measurement values in the Excel document, *Measurement.xls*, give us the position and speed of the vehicle measured in 2 second intervals. Other values given are:

- PSD (power spectral density) of the random acceleration:  $0.01 \text{ m}^2\text{s}^3$
- Standard deviation of measured coordinates (both components): 3 m
- Standard deviation of measured abs. velocity: 0.5 m/s
- Standard deviation of initial velocity (both components): 3 m/s
- Standard deviation of initial coordinates (both components): 10 m

## 2 Methodology

Matlab was used to solve for the following parameters and all of the following equations were taken from the technical report: Kalman Filtering.<sup>[1]</sup> The steps and equations from the Kalman filtering algorithm are documented in the Matlab code. We start off with the matrix differential equation,

$$\dot{\mathbf{x}}(t) = \mathbf{F}(t)\mathbf{x}(t) + \mathbf{G}(t)\mathbf{u}(t) \quad (1)$$

where  $\mathbf{x} = [e \ n \ v_e \ v_n]^T$ , the east and north positions and their velocity components.  $\mathbf{F}$ ,  $\mathbf{G}$ , and  $\mathbf{u}$  are given by,

$$\mathbf{F} = \begin{bmatrix} 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} \quad \mathbf{G} = \begin{bmatrix} 0 & 0 \\ 0 & 0 \\ 1 & 0 \\ 0 & 1 \end{bmatrix} \quad \mathbf{u} = \begin{bmatrix} \omega_{ae} \\ \omega_{an} \end{bmatrix}$$

We do not know  $\mathbf{u}$  because it represents white noise. The discrete solution to Equation (1) is

$$\mathbf{x}_k = \mathbf{T}_{k-1,k}\mathbf{x}_{k-1} + \mathbf{w}_{k-1,k}$$

where values of  $k$  represent discretization of time. The transition matrix,  $\mathbf{T}_{k-1,k}$ , can be approximated by,

$$\mathbf{T}_{k-1,k} = \mathbf{I} + \mathbf{F}_k \Delta t = \begin{bmatrix} 1 & 0 & \Delta t & 0 \\ 0 & 1 & 0 & \Delta t \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

The covariance of  $\mathbf{w}_k$ ,  $\mathbf{Q}_k$ , is evaluated by use of the power spectral density (PSD) of the random acceleration, which are initially given. We can now represent the covariance of  $\mathbf{w}_k$  as,

$$\mathbf{Q}_k = \mathbf{Q}_G \Delta t + (\mathbf{F} \mathbf{Q}_G + \mathbf{Q}_G \mathbf{F}^T) \frac{\Delta t^2}{2} + \mathbf{F} \mathbf{Q}_G \mathbf{F}^T \frac{\Delta t^3}{3}$$

where

$$\mathbf{Q}_G = \mathbf{G} \mathbf{Q} \mathbf{G}^T \quad \mathbf{Q} = \begin{bmatrix} q_{ae} & 0 \\ 0 & q_{an} \end{bmatrix}$$

## 2.1 Kalman Filter

The main steps of the discrete Kalman Filter algorithm are:

1. Initialization:

$$\mathbf{x}_0, \quad \mathbf{Q}_{x0} = \text{var}[x_0]$$

2. Time propagation

$$\mathbf{x}_k^- = \mathbf{T}_{k-1,k} \mathbf{x}_{k-1}, \quad \mathbf{Q}_{x,k}^- = \mathbf{T}_{k-1,k} \mathbf{Q}_{x,k-1} \mathbf{T}_{k-1,k}^T + \mathbf{Q}_k$$

3. Gain calculation:

$$\mathbf{K}_k = \mathbf{Q}_{x,k}^- \mathbf{H}_k^T [\mathbf{R}_k + \mathbf{H}_k \mathbf{Q}_{x,k}^- \mathbf{H}_k^T]^{-1}$$

4. Measurement update

$$\mathbf{x}_k = \mathbf{x}_k^- + \mathbf{K}_k [\tilde{L}_k - \mathbf{h}_k(\mathbf{x}_k^-)]$$

5. Covariance update

$$\mathbf{Q}_{x,k} = [\mathbf{I} - \mathbf{K}_k \mathbf{H}_k] \mathbf{Q}_{x,k}^-$$

## 2.2 Smoothing

After the Kalman Filter algorithm, we then smooth the results using the smoothed estimations for previous epochs,

$$\hat{\mathbf{x}}_k = \mathbf{x}_k + \mathbf{D}_k [\hat{\mathbf{x}}_{k+1} - \mathbf{x}_{k+1}^-]$$

$$\mathbf{D}_k = \mathbf{Q}_{x,k} \mathbf{T}_{k-1,k}^T (\mathbf{Q}_{x,k+1}^-)^{-1}$$

with a covariance matrix of,

$$\hat{\mathbf{Q}}_{x,k} = \mathbf{Q}_{x,k} + \mathbf{D}_k [\hat{\mathbf{Q}}_{x,k+1} - \mathbf{Q}_{x,k+1}^-] \mathbf{D}_k^T$$

### 3 Results

A comparison of the filtered and smoothed coordinates with the original and true measurements are shown in Figure 1. Table 1 and 2 are the coordinates after filtering and smoothing.

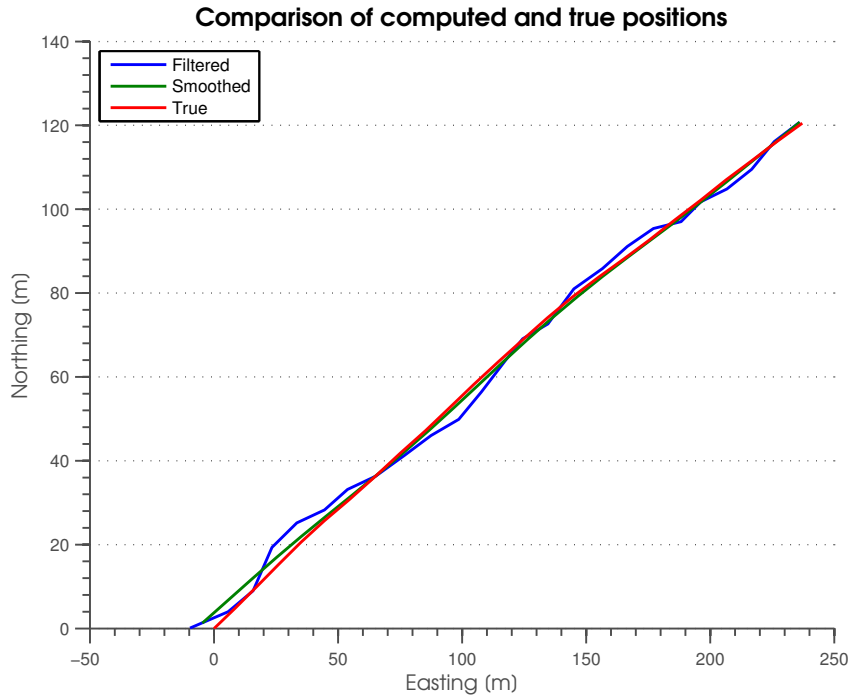


Figure 1: Comparison of filtered, smoothed, original, and true coordinates.

### 4 Analysis and Discussion

- What happens if we change the values of

1. PSD
2. standard deviations of measurements
3. standard deviations of initial state variables

- Can we draw any conclusion/implications from the results?

We can conclude that the smoothed coordinates are a better approximation to the true values than the filtered coordinates.

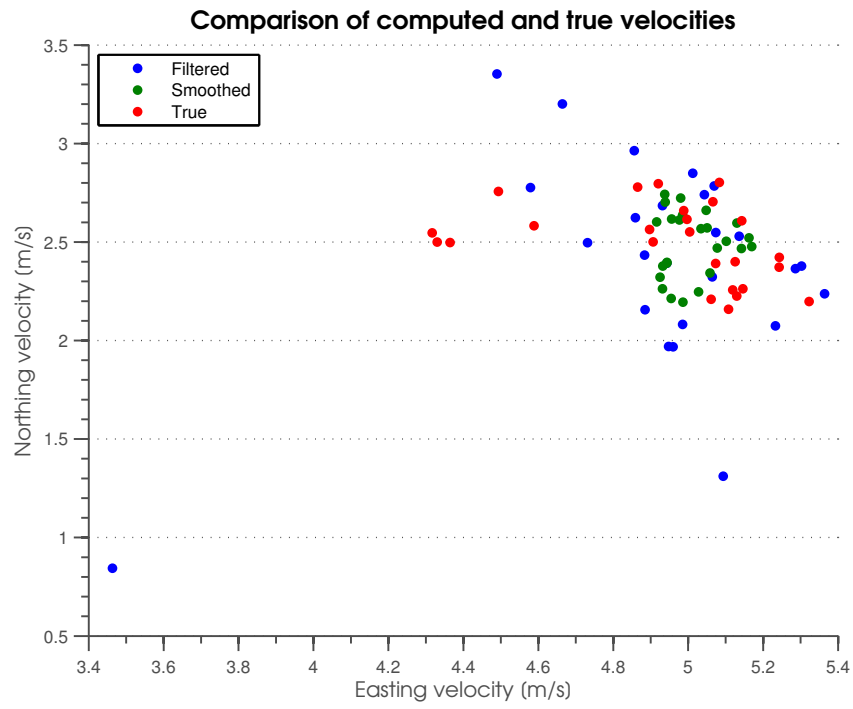


Figure 2: Comparison of filtered, smoothed, and true velocities.

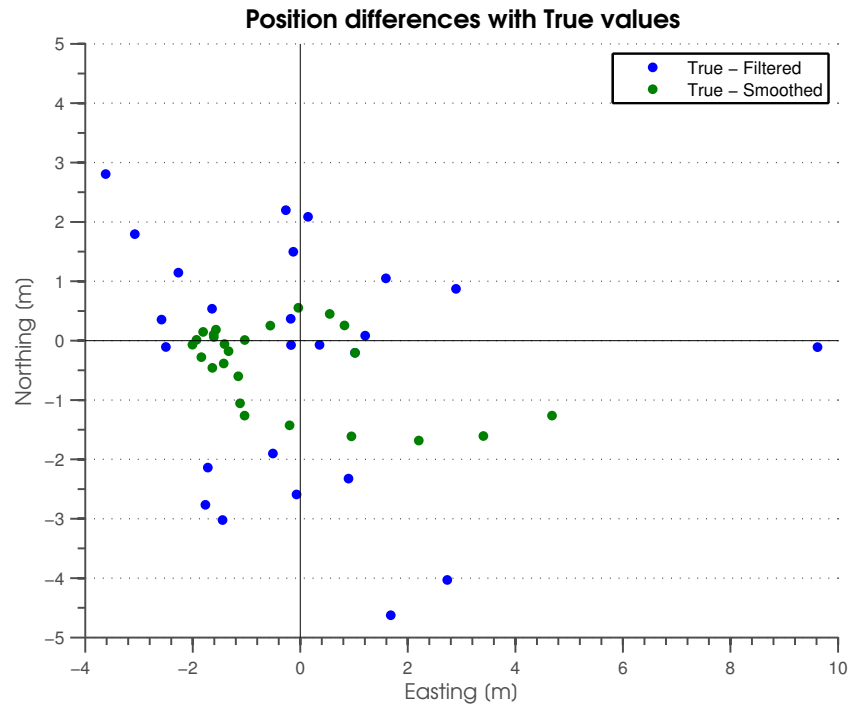


Figure 3: Plot of differences between true values and the filtered and smoothed coordinates.

Filtered Estimations and standard deviations				
t	$x_e \pm \sigma_e$	$x_n \pm \sigma_n$	$v_e \pm \sigma_{v_e}$	$v_n \pm \sigma_{v_n}$
0	$-9.62 \pm 10.00$	$0.11 \pm 10.00$	$3.46 \pm 3.00$	$0.84 \pm 3.00$
2	$5.71 \pm 1.71$	$4.00 \pm 1.71$	$5.09 \pm 0.90$	$1.31 \pm 2.52$
4	$15.74 \pm 1.33$	$8.98 \pm 1.64$	$4.95 \pm 0.51$	$1.97 \pm 1.05$
6	$23.36 \pm 1.23$	$19.32 \pm 1.53$	$4.49 \pm 0.37$	$3.35 \pm 0.58$
8	$33.40 \pm 1.19$	$25.17 \pm 1.42$	$4.66 \pm 0.30$	$3.20 \pm 0.40$
10	$44.56 \pm 1.17$	$28.26 \pm 1.31$	$4.93 \pm 0.27$	$2.68 \pm 0.31$
12	$53.82 \pm 1.15$	$33.15 \pm 1.24$	$4.86 \pm 0.25$	$2.62 \pm 0.27$
14	$65.20 \pm 1.13$	$36.34 \pm 1.19$	$5.06 \pm 0.24$	$2.32 \pm 0.25$
16	$76.79 \pm 1.11$	$41.27 \pm 1.16$	$5.30 \pm 0.23$	$2.38 \pm 0.24$
18	$87.52 \pm 1.09$	$46.04 \pm 1.14$	$5.29 \pm 0.23$	$2.37 \pm 0.24$
20	$98.70 \pm 1.09$	$49.89 \pm 1.13$	$5.36 \pm 0.23$	$2.24 \pm 0.24$
22	$107.87 \pm 1.08$	$56.46 \pm 1.13$	$5.14 \pm 0.23$	$2.53 \pm 0.24$
24	$117.40 \pm 1.08$	$63.85 \pm 1.13$	$5.01 \pm 0.23$	$2.85 \pm 0.24$
26	$124.41 \pm 1.08$	$69.08 \pm 1.13$	$4.58 \pm 0.23$	$2.78 \pm 0.24$
28	$134.62 \pm 1.08$	$72.63 \pm 1.13$	$4.73 \pm 0.23$	$2.50 \pm 0.24$
30	$145.07 \pm 1.08$	$81.00 \pm 1.13$	$4.86 \pm 0.23$	$2.96 \pm 0.24$
32	$156.58 \pm 1.08$	$85.84 \pm 1.13$	$5.07 \pm 0.23$	$2.78 \pm 0.24$
34	$166.60 \pm 1.08$	$91.13 \pm 1.12$	$5.04 \pm 0.23$	$2.74 \pm 0.24$
36	$177.15 \pm 1.08$	$95.37 \pm 1.12$	$5.07 \pm 0.23$	$2.55 \pm 0.24$
38	$188.30 \pm 1.08$	$97.05 \pm 1.12$	$5.23 \pm 0.23$	$2.07 \pm 0.24$
40	$196.25 \pm 1.08$	$101.75 \pm 1.13$	$4.89 \pm 0.23$	$2.16 \pm 0.24$
42	$206.72 \pm 1.08$	$104.80 \pm 1.13$	$4.96 \pm 0.23$	$1.97 \pm 0.24$
44	$216.74 \pm 1.08$	$109.52 \pm 1.13$	$4.99 \pm 0.23$	$2.08 \pm 0.24$
46	$225.83 \pm 1.08$	$116.11 \pm 1.13$	$4.88 \pm 0.23$	$2.43 \pm 0.24$
48	$236.10 \pm 1.08$	$120.72 \pm 1.13$	$4.94 \pm 0.23$	$2.39 \pm 0.24$

Table 1: The easting and northing coordinates and velocities are given with their standard deviations.

- Are results reliable and accurate?

I believe the results are reliable and accurate because the standard deviations dropped when the coordinates were smoothed and the state vector was a better approximation of the true values.

## References

- [1] M. Horemuž, “Kalman filtering,” Royal Institute of Technology, Tech. Rep., 2014.

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## Kalman Filter

```
clear all;clc;close all
addpath(['/Users/kevin/SkyDrive/KTH Work/',...
        'Period 4 2014/GNSS/Labs/L4 ',...
        '- Kalman filtering/']);
```

## Import numbers from excel

Measured values from sheet 1

```
data.s1 = xlsread('Measurement.xlsx');
meas.time = data.s1(1:25,1);
meas.east = data.s1(1:25,2);
meas.north = data.s1(1:25,3);
meas.speed = data.s1(1:25,4);
% True values from sheet 2
data.s2 = xlsread('Measurement.xlsx','True values');
true.time = data.s2(1:25,1);
true.east = data.s2(1:25,2);
true.north = data.s2(1:25,3);
true.vel_east = data.s2(1:25,4);
true.vel_north = data.s2(1:25,5);
```

## A priori statistics

```
PSD = 0.01; % - PSD (power spectral density) of the
% random acceleration
meas.sd_coord = 3; % m - Standard error of measured coordinates
meas.sd_abs_vel = 0.5; % m/s - Standard error of measured
% abs. velocity
sd_ini_vel = 3; % m/s - Standard error of initial velocity
sd_ini_coord = 10; % m - Standard error of initial coordinates
ve = 3.53; % m/s
vn = 0.86; % m/s
dt = 2; % time difference -> 2 sec between measurements
```

## State variables and Equations

```
xk = [meas.east(1) meas.north(1) ve vn];
```

---

```

xk = padarray(xk,24,0,'post');
xk = xk';
% Equation 4
F = zeros(4);
F(1,3) = 1;
F(2,4) = 1;
G = zeros(4,2);
G(3,1) = 1;
G(4,2) = 1;
% Equation 5
Tk = eye(length(F)) + dt * F;
% Equation 9
Q = [ PSD 0 ; 0 PSD];
% Equation 11
QG = G*Q*G';
% Equation 12
Qk = QG * dt + (F*QG + QG*F')*dt^2/2 + F*QG*F'*dt^3/3;
% Equation 15
% covariance matrix of initial state
Qx(:, :, 1) = diag([sd_ini_coord^2 ...
    sd_ini_coord^2 sd_ini_vel^2 sd_ini_vel^2]);
Rk = diag([meas.sd_coord meas.sd_coord meas.sd_abs_vel]);
Qxm_predicted = zeros(4,4,25);

```

## FOR LOOP

```

for i=1:25
    % Equation 16 Time propagation
    xkm_predicted(:,i) = Tk * xk(:,i);
    % Qx = cov(xk(:,i));
    Qxm_predicted(:, :, i) = Tk * Qx(:, :, i) * Tk' + Qk;
    vm_predicted = sqrt(xkm_predicted(3,i)^2 + ...
        xkm_predicted(4,i)^2); % should be equal to speed_meas(1)
    Hk(:, :, i) = [1, 0, 0, 0; ...
        0, 1, 0, 0; ...
        0, 0, xkm_predicted(3,i)/vm_predicted, ...
        xkm_predicted(4,i) /vm_predicted];
    % Equation 17 Gain
    Kk(:, :, i) = Qxm_predicted(:, :, i) * Hk(:, :, i)' * inv([Rk + ...
        Hk(:, :, i) * Qxm_predicted(:, :, i) * Hk(:, :, i)']);
    Lk(:, i) = [meas.east(i) meas.north(i) ...
        meas.speed(i)]';
    hkm_predicted = [xkm_predicted(1,i) xkm_predicted(2,i) ...
        sqrt(xkm_predicted(3,i)^2 + xkm_predicted(4,i)^2)]';

    xk(:, i+1) = xkm_predicted(:, i) + Kk(:, :, i) * [ Lk(:, i) ...
        - hkm_predicted ]; % Equation 18
    % Measurement update
    % Equation 19
    Qx(:, :, i+1) = [eye(length(Kk(:, :, i)) * ...
        Hk(:, :, i)) - Kk(:, :, i) * Hk(:, :, i)] * Qxm_predicted(:, :, i);
    % Equation 22
    % Lk = [meas.east(i) meas.north(i) ...

```

---

```

%           sqrt((xk(3,i))^2 + (xk(4,i))^2)]';

%           Hk = inv(xk(:,i))*Lk;
final.xplot(:,i+1) = xk(:,i);
end

```

## Smoothing

```

xkhat(:,25) = xk(:,end);
nStep = 25;
count = nStep;
Qxkhat(:, :, 25) = Qx(:, :, end);
for i = 1:(nStep-1)
    Dk = Qx(:, :, count+1)*Tk'*inv(Qxm_predicted(:, :, count));
    xkhat(:, count-1) = xk(:, count) + Dk*[xkhat(:, count) ...
        - xkm_predicted(:, count)];
    Qxkhat(:, :, count-1) = Qx(:, :, count+1) ...
        + Dk*[Qxkhat(:, :, count) - Qxm_predicted(:, :, count)]*Dk';
    count = count - 1;
end

```

## Plot

```

final.x1 = xk(1,:); % final x values
final.y1 = xk(2,:); % final y values
meas.x2 = meas.east; % original x
meas.y2 = meas.north; % original y
true.x3 = true.east; % true x
true.y3 = true.north; % true y
figure('Units', 'pixels', ...
    'Position', [100 100 500 375]);
hold on;
yl_plot = plot(final.x1, final.y1, ... % Before Smoothing
    xkhat(1,:), xkhat(2,:), ... % Smoothing
    meas.x2, meas.y2, ... % Original
    true.x3, true.y3) % True Plot
hTitle = title('Kalman filtering');
hXLabel = xlabel('Easting [m]');
hYLabel = ylabel('Northing [m]');
hLegend = legend(...
    'Before Smoothing', ...
    'Smoothing', ...
    'Original', ...
    'True Plot', ...
    'location', 'best');
set(gca, ...
    'FontName', 'Helvetica');
set([hTitle, hXLabel, hYLabel], ...
    'FontName', 'AvantGarde');
set([hLegend, gca], ...
    'FontSize', 10);
set([hXLabel, hYLabel], ...
    'FontSize', 11);

```



---

```
set( hTitle
    'FontSize'    , 13          , ...
    'FontWeight' , 'bold'      );
set(gca, ...
    'Box'        , 'off'       , ...
    'TickDir'    , 'out'       , ...
    'TickLength' , [.02 .02]   , ...
    'XMinorTick' , 'on'        , ...
    'YMinorTick' , 'on'        , ...
    'YGrid'      , 'on'        , ...
    'XColor'     , [.3 .3 .3]   , ...
    'YColor'     , [.3 .3 .3]   , ...
    'LineWidth'  , 1           );
hold off;
```

*Published with MATLAB® R2013a*

Smoothed Estimations and standard deviations				
t	$x_e \pm \sigma_e$	$x_n \pm \sigma_n$	$v_e \pm \sigma_{v_e}$	$v_n \pm \sigma_{v_n}$
0	$5.20 \pm 1.22$	$1.26 \pm 1.53$	$4.92 \pm 0.81$	$2.60 \pm 2.47$
2	$15.13 \pm 0.42$	$6.48 \pm 0.37$	$4.96 \pm 0.37$	$2.62 \pm 0.97$
4	$25.14 \pm 0.62$	$11.72 \pm 0.41$	$4.98 \pm 0.19$	$2.61 \pm 0.46$
6	$35.28 \pm 0.62$	$16.90 \pm 0.37$	$5.03 \pm 0.07$	$2.57 \pm 0.24$
8	$45.53 \pm 0.60$	$21.97 \pm 0.46$	$5.10 \pm 0.07$	$2.50 \pm 0.10$
10	$55.84 \pm 0.59$	$26.94 \pm 0.53$	$5.14 \pm 0.10$	$2.47 \pm 0.08$
12	$66.18 \pm 0.60$	$31.88 \pm 0.58$	$5.17 \pm 0.12$	$2.48 \pm 0.12$
14	$76.48 \pm 0.61$	$36.87 \pm 0.61$	$5.16 \pm 0.12$	$2.52 \pm 0.13$
16	$86.66 \pm 0.62$	$41.99 \pm 0.63$	$5.13 \pm 0.13$	$2.60 \pm 0.13$
18	$96.69 \pm 0.63$	$47.24 \pm 0.65$	$5.05 \pm 0.13$	$2.66 \pm 0.13$
20	$106.60 \pm 0.64$	$52.63 \pm 0.65$	$4.98 \pm 0.13$	$2.72 \pm 0.13$
22	$116.47 \pm 0.64$	$58.11 \pm 0.65$	$4.94 \pm 0.13$	$2.74 \pm 0.13$
24	$126.38 \pm 0.64$	$63.56 \pm 0.65$	$4.94 \pm 0.13$	$2.70 \pm 0.13$
26	$136.43 \pm 0.65$	$68.90 \pm 0.65$	$4.98 \pm 0.13$	$2.64 \pm 0.13$
28	$146.56 \pm 0.64$	$74.11 \pm 0.65$	$5.05 \pm 0.13$	$2.57 \pm 0.13$
30	$156.71 \pm 0.65$	$79.17 \pm 0.65$	$5.08 \pm 0.13$	$2.47 \pm 0.13$
32	$166.79 \pm 0.64$	$83.98 \pm 0.66$	$5.06 \pm 0.13$	$2.34 \pm 0.13$
34	$176.81 \pm 0.65$	$88.56 \pm 0.66$	$5.03 \pm 0.13$	$2.25 \pm 0.13$
36	$186.75 \pm 0.65$	$92.99 \pm 0.67$	$4.99 \pm 0.13$	$2.19 \pm 0.13$
38	$196.63 \pm 0.65$	$97.39 \pm 0.67$	$4.95 \pm 0.13$	$2.21 \pm 0.14$
40	$206.49 \pm 0.65$	$101.87 \pm 0.67$	$4.93 \pm 0.14$	$2.26 \pm 0.14$
42	$216.34 \pm 0.67$	$106.45 \pm 0.68$	$4.93 \pm 0.15$	$2.32 \pm 0.15$
44	$226.22 \pm 0.73$	$111.15 \pm 0.74$	$4.93 \pm 0.16$	$2.38 \pm 0.17$
46	$236.10 \pm 0.85$	$115.93 \pm 0.88$	$4.94 \pm 0.19$	$2.40 \pm 0.20$

Table 2: The easting and northing coordinates and velocities are given with their standard deviations.

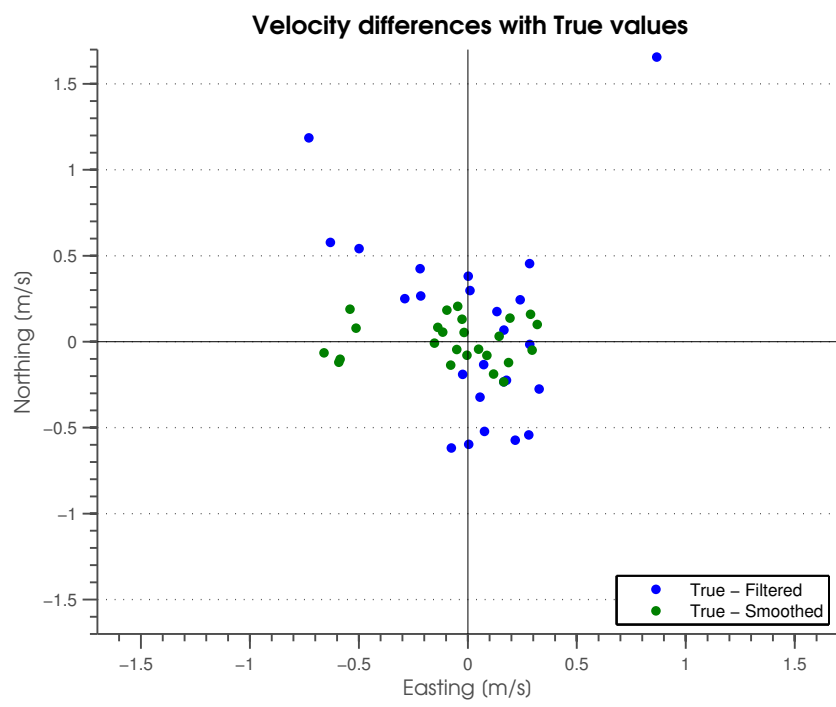


Figure 4: Plot of differences between true values and the filtered and smoothed velocities.