

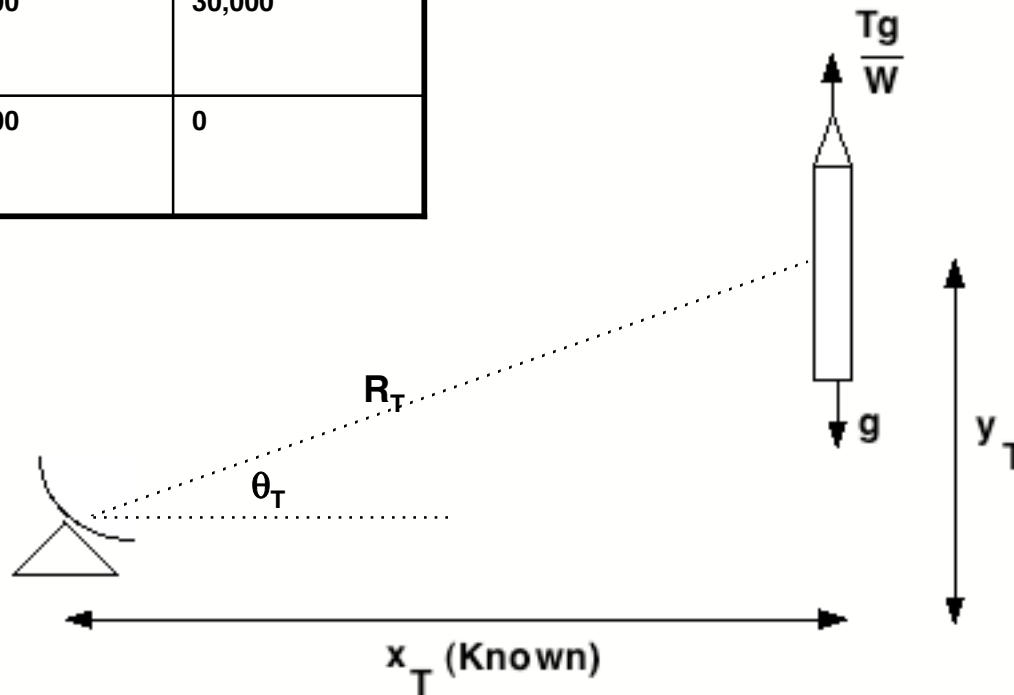
A Tracking Disaster

A Tracking Disaster Overview

- **Problem Setup**
- **Three-State Linear Polynomial Kalman Filter**
- **Six-State Linear Polynomial Kalman Filter**
- **Interesting Experiments**
- **Summary**

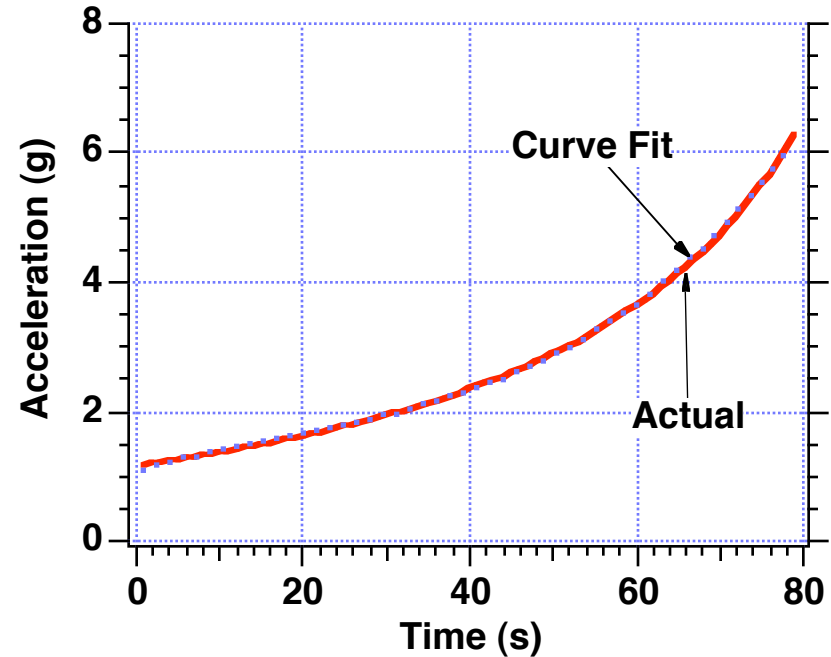
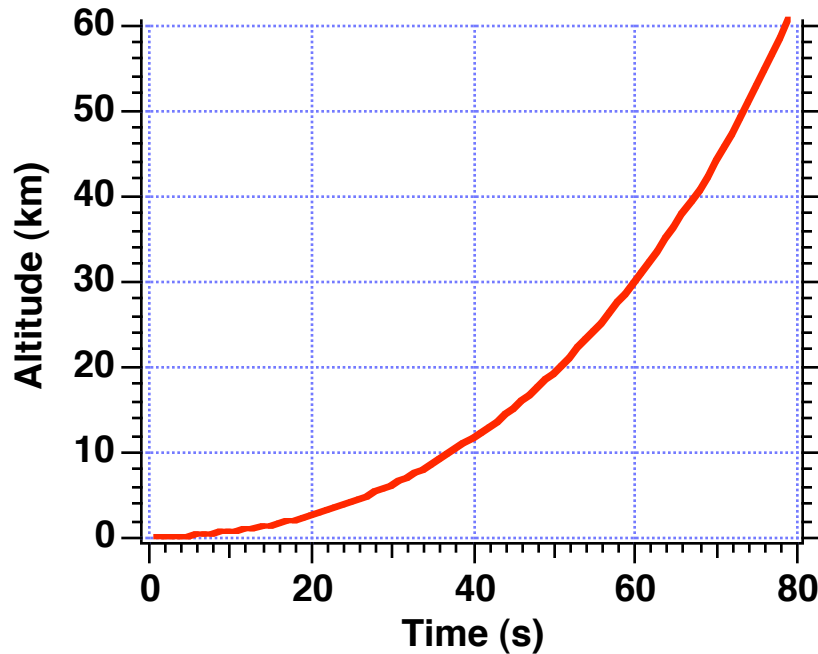
Problem Setup

Time (s)	Weight (lbs)	Thrust (lbs)
0	14,000	30,000
80	4,000	30,000
80	4,000	0



If x_T is known, measuring R_T and θ_T is equivalent to measuring y_T

Target Altitude and Acceleration Profiles



Acceleration curve fit $n_T = \ddot{y}_T = \frac{33.8 + 1.24t - 0.0228t^2 + 0.000422t^3}{32.2} \rightarrow \ddot{n}_T = \text{constant}$

Potential perfect filter has six states

$$\begin{bmatrix} y_T \\ \dot{y}_T \\ \ddot{y}_T = n_T \\ \dot{n}_T \\ \ddot{n}_T \\ \ddot{\ddot{n}}_T = q_T \end{bmatrix}$$

Three-State Linear Polynomial Kalman Filter

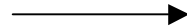
Recall Important Matrices for Different Order Polynomial Kalman Filters

Important matrices for 3-state Kalman filter →

Order	System Dynamics	Fundamental	Measurement	Noise
0	$F=1$	$\Phi_k = 1$	$H=1$	$R_k = \sigma_n^2$
1	$F = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}$	$\Phi_k = \begin{bmatrix} 1 & T_s \\ 0 & 1 \end{bmatrix}$	$H = [1 \ 0]$	$R_k = \sigma_n^2$
2	$F = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{bmatrix}$	$\Phi_k = \begin{bmatrix} 1 & T_s & .5T_s^2 \\ 0 & 1 & T_s \\ 0 & 0 & 1 \end{bmatrix}$	$H = [1 \ 0 \ 0]$	$R_k = \sigma_n^2$

Recall That The Discrete Process Noise Matrix Varies With System Order

Process noise matrix for 3-state Kalman filter



Order	Continuous Q	Fundamental	Discrete Q
0	$Q = \Phi_s$	$\Phi_k = 1$	$Q_k = \Phi_s$
1	$Q = \Phi_s \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}$	$\Phi_k = \begin{bmatrix} 1 & T_s \\ 0 & 1 \end{bmatrix}$	$Q_k = \Phi_s \begin{bmatrix} \frac{T_s^3}{3} & \frac{T_s^2}{2} \\ \frac{T_s^2}{2} & T_s \end{bmatrix}$
2	$Q = \Phi_s \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix}$	$\Phi_k = \begin{bmatrix} 1 & T_s & .5T_s^2 \\ 0 & 1 & T_s \\ 0 & 0 & 1 \end{bmatrix}$	$Q_k = \Phi_s \begin{bmatrix} \frac{T_s^5}{20} & \frac{T_s^4}{8} & \frac{T_s^3}{6} \\ \frac{T_s^4}{8} & \frac{T_s^3}{3} & \frac{T_s^2}{2} \\ \frac{T_s^3}{6} & \frac{T_s^2}{2} & T_s \end{bmatrix}$

Filter gains are obtained from Riccati equations

$$M_k = \Phi_k P_{k-1} \Phi_k^T + Q_k$$

$$K_k = M_k H^T (H M_k H^T + R_k)^{-1}$$

$$P_k = (I - K_k H) M_k$$

Three-State Linear Polynomial Kalman Filter-1

```

GLOBAL DEFINE
      INCLUDE 'quickdraw.inc'

END
IMPLICIT REAL*8(A-H,O-Z)
REAL*8 P(3,3),Q(3,3),M(3,3),PHI(3,3),HMAT(1,3),HT(3,1),PHIT(3,3)
REAL*8 RMAT(1,1),IDN(3,3),PHIP(3,3),PHIPPHIT(3,3),HM(1,3)
REAL*8 HMHT(1,1),HMHTR(1,1),HMHTRINV(1,1),MHT(3,1),K(3,1)
REAL*8 KH(3,3),IKH(3,3)
INTEGER ORDER,STEP
OPEN(1,STATUS='UNKNOWN',FILE='DATFIL')
OPEN(2,STATUS='UNKNOWN',FILE='COVFIL')
G=32.2
XNT=193.2
YT=0.
YTD=0.
ORDER=3
PHIS=XNT*XNT/120.
TS=1.
YH=0.
YDH=0.
YDDH=0
SIGNOISE=100.
DO 14 I=1,ORDER
DO 14 J=1,ORDER
PHI(I,J)=0.
P(I,J)=0.
Q(I,J)=0.
IDN(I,J)=0.
CONTINUE
RMAT(1,1)=SIGNOISE**2
IDN(1,1)=1.
IDN(2,2)=1.
IDN(3,3)=1.
P(1,1)=100.**2
P(2,2)=100.**2
P(3,3)=64.4**2

```

Process noise

**Sampling time, initial state estimates
and measurement noise**

Measurement noise matrix

Initial covariance matrix

14

Three-State Linear Polynomial Kalman Filter-2

```

PHI(1,1)=1
PHI(1,2)=TS
PHI(1,3)=.5*TS*TS
PHI(2,2)=1
PHI(2,3)=TS
PHI(3,3)=1

```

**Non zero elements of
fundamental matrix**

```

HMAT(1,1)=1.
HMAT(1,2)=0.
HMAT(1,3)=0.

```

Measurement matrix

```

CALL MATTRN(PHI,ORDER,ORDER,PHIT)
CALL MATTRN(HMAT,1,ORDER,HT)

```

```

Q(1,1)=PHIS*TS**5/20
Q(1,2)=PHIS*TS**4/8
Q(1,3)=PHIS*TS**3/6
Q(2,1)=Q(1,2)
Q(2,2)=PHIS*TS**3/3
Q(2,3)=PHIS*TS*TS/2
Q(3,1)=Q(1,3)
Q(3,2)=Q(2,3)
Q(3,3)=PHIS*TS

```

Process noise matrix

```

S=0.
H=.01
T=0.

```

10 IF(T>79.)GOTO 999

```

S=S+H

```

```

YTOLD=YT

```

```

YTDOLD=YTD

```

```

STEP=1

```

```

GOTO 200

```

66

```

STEP=2

```

```

YT=YT+H*YTD

```

```

YTD=YTD+H*YTDD

```

```

T=T+H

```

```

GOTO 200

```

55

```

CONTINUE

```

```

YT=.5*(YTOLD+YT+H*YTD)

```

```

YTD=.5*(YTDOLD+YTD+H*YTDD)

```

**Second-order Runge-Kutta
Integration for booster**

Three-State Linear Polynomial Kalman Filter-3

```

IF(S<(TS-.00001))GOTO 10
S=0.
CALL MATMUL(PHI,ORDER,ORDER,P,ORDER,ORDER,PHIP)
CALL MATMUL(PHIP,ORDER,ORDER,PHIT,ORDER,ORDER,PHIPPHIT)
CALL MATADD(PHIPPHIT,ORDER,ORDER,Q,M)
CALL MATMUL(HMAT,1,ORDER,M,ORDER,ORDER,HM)
CALL MATMUL(HM,1,ORDER,HT,ORDER,1,HMHT)
CALL MATADD(HMHT,ORDER,ORDER,RMAT,HMHTR)
HMHTRINV(1,1)=1./HMHTR(1,1)
CALL MATMUL(M,ORDER,ORDER,HT,ORDER,1,MHT)
CALL MATMUL(MHT,ORDER,1,HMHTRINV,1,1,K)
CALL MATMUL(K,ORDER,1,HMAT,1,ORDER,KH)
CALL MATSUB(IDN,ORDER,ORDER,KH,IKH)
CALL MATMUL(IKH,ORDER,ORDER,M,ORDER,ORDER,P)
CALL GAUSS(YNOISE,SIGNOISE)
YS=YT+YNOISE
XK1=K(1,1)
XK2=K(2,1)
XK3=K(3,1)
RES=YS-YH-TS*YDH-.5*TS*TS*YDDH
YH=YH+YDH*TS+.5*TS*TS*YDDH+XK1*RES
YDH=YDH+YDDH*TS+XK2*RES
YDDH=YDDH+XK3*RES
ERRY=YT-YH
ERRYD=YTD-YDH
ERRYDD=YTDD-YDDH
    
```

Riccati equations

Measurement noise

Three-state filter

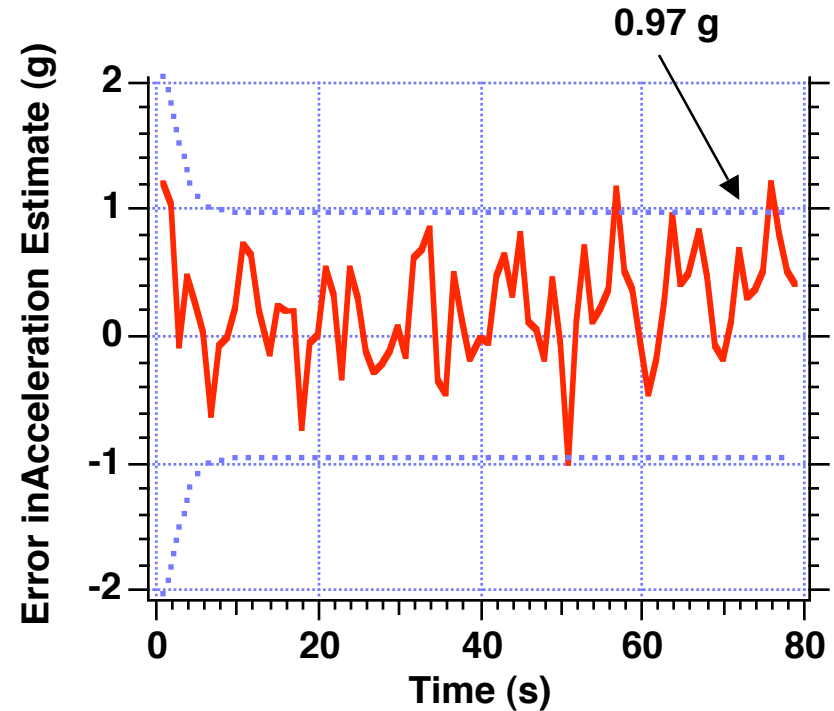
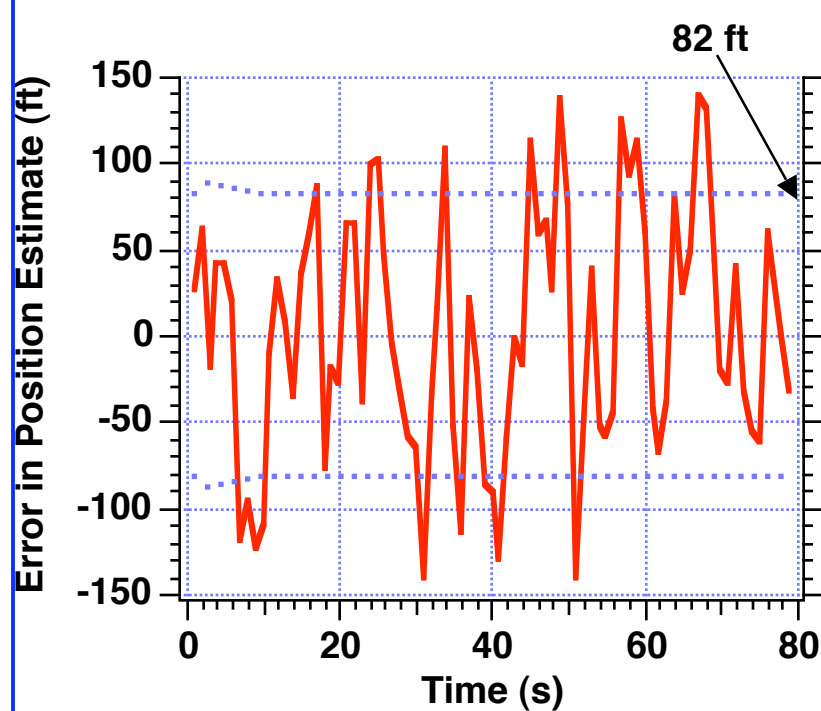
Three-State Linear Polynomial Kalman Filter-4

```
1      SP11=SQRT(P(1,1))
      SP22=SQRT(P(2,2))
      SP33=SQRT(P(3,3))
      WRITE(9,*)T,YTDD/32.2,YDDH/32.2
      WRITE(1,*)T,YTDD/32.2,YDDH/32.2,YTD,YDH,ZDH
      WRITE(2,*)T,ERRY,SP11,-SP11,ERRYD,SP22,-SP22,ERRYDD/32.2
      ,SP33/32.2,-SP33/32.2
200    GOTO 10
      CONTINUE
      IF(T<80.)THEN
          WGT=-125.*T+14000.
          TRST=30000.
      ELSE
          WGT=4000.
          TRST=0.
      ENDIF
      AT=G*TRST/WGT
      YTDD=AT-G
999    IF(STEP-1)66,66,55
      CONTINUE
      PAUSE
      CLOSE(1)
      CLOSE(2)
      END
```

Booster differential equation

Three-State Kalman Filter is Able To Track Boosting Target

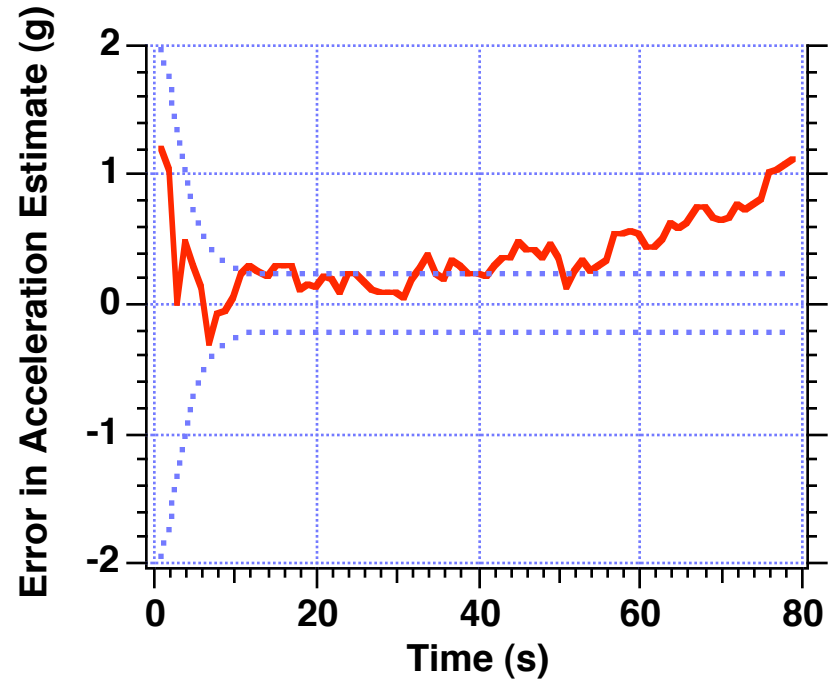
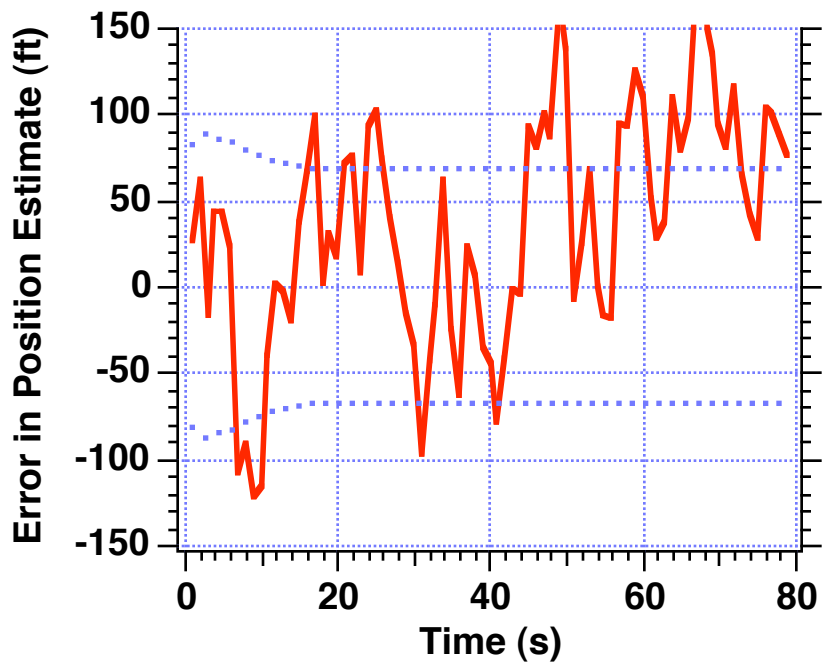
$$\Phi_s = \frac{193.2^2}{120}$$



*100 ft measurement noise and 1 s sampling time

Reducing Process Noise With Three-State Kalman Filter Causes Divergence

$$\Phi_s = \frac{32.2^2}{120}$$



*100 ft measurement noise and 1 s sampling time

Six-State Linear Polynomial Kalman Filter

Measurement and Real World Plant Models

Model of Real World (6 states required to fit booster acceleration profile)

$$\begin{bmatrix} \dot{y}_T \\ \ddot{y}_T \\ \dot{n}_T \\ \ddot{n}_T \\ \dot{\ddot{n}}_T \\ \dot{q}_T \end{bmatrix} = \begin{bmatrix} 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} y_T \\ \dot{y}_T \\ n_T \\ \dot{n}_T \\ \ddot{n}_T \\ \ddot{\ddot{n}}_T \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ u_s \end{bmatrix} \longrightarrow \mathbf{F} = \begin{bmatrix} 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix} \quad \mathbf{Q} = \begin{bmatrix} 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & \Phi_s \end{bmatrix}$$

Measurement Equation

$$y_T^* = [1 \ 0 \ 0 \ 0 \ 0 \ 0] \begin{bmatrix} y_T \\ \dot{y}_T \\ n_T \\ \dot{n}_T \\ \ddot{n}_T \\ \ddot{\ddot{n}}_T \end{bmatrix} + u_n \longrightarrow \mathbf{H} = [1 \ 0 \ 0 \ 0 \ 0 \ 0] \quad R = \Phi_s$$

Fundamental and Process Noise Matrices

Fundamental Matrix

$$\Phi_k = \begin{bmatrix} 1 & T_s & 0.5T_s^2 & T_s^3/6 & T_s^4/24 & T_s^5/120 \\ 0 & 1 & T_s & 0.5T_s^2 & T_s^3/6 & T_s^4/24 \\ 0 & 0 & 1 & T_s & 0.5T_s^2 & T_s^3/6 \\ 0 & 0 & 0 & 1 & T_s & 0.5T_s^2 \\ 0 & 0 & 0 & 0 & 1 & T_s \\ 0 & 0 & 0 & 0 & 0 & 1 \end{bmatrix}$$

Discrete Process Noise Matrix

$$\mathbf{Q}_k = \int_0^{T_s} \Phi(\tau) \mathbf{Q} \Phi^T(\tau) d\tau = \Phi_s \begin{bmatrix} \frac{T_s^{11}}{158400} & \frac{T_s^{10}}{28800} & \frac{T_s^9}{6480} & \frac{T_s^8}{1920} & \frac{T_s^7}{840} & \frac{T_s^6}{720} \\ \frac{T_s^{10}}{28800} & \frac{T_s^9}{5184} & \frac{T_s^8}{1152} & \frac{T_s^7}{336} & \frac{T_s^6}{144} & \frac{T_s^5}{120} \\ \frac{T_s^9}{6480} & \frac{T_s^8}{1152} & \frac{T_s^7}{252} & \frac{T_s^6}{72} & \frac{T_s^5}{30} & \frac{T_s^4}{24} \\ \frac{T_s^8}{1920} & \frac{T_s^7}{336} & \frac{T_s^6}{72} & \frac{T_s^5}{20} & \frac{T_s^4}{8} & \frac{T_s^3}{6} \\ \frac{T_s^7}{840} & \frac{T_s^6}{144} & \frac{T_s^5}{30} & \frac{T_s^4}{8} & \frac{T_s^3}{3} & \frac{T_s^2}{2} \\ \frac{T_s^6}{720} & \frac{T_s^5}{120} & \frac{T_s^4}{24} & \frac{T_s^3}{6} & \frac{T_s^2}{2} & T_s \end{bmatrix}$$

Six-State Linear Polynomial Kalman Filter-1

```
GLOBAL DEFINE
      INCLUDE 'quickdraw.inc'
END
IMPLICIT REAL *8(A-H,O-Z)
REAL *8 P(6,6),Q(6,6),M(6,6),PHI(6,6),HMAT(1,6),HT(6,1),PHIT(6,6)
REAL *8 RMAT(1,1),IDN(6,6),PHIP(6,6),PHIPPHIT(6,6),HM(1,6)
REAL *8 HMHT(1,1),HMHTR(1,1),HMHTRINV(1,1),MHT(6,1),K(6,1)
REAL *8 KH(6,6),IKH(6,6)
INTEGER ORDER,STEP
OPEN(1,STATUS='UNKNOWN',FILE='DATFIL')
OPEN(2,STATUS='UNKNOWN',FILE='COVFIL')
G=32.2
XNTDDD=.03
YT=0.
YTD=0.
ORDER=6
PHIS=XNTDDD*XNTDDD/120.
TS=1.
TS2=TS*TS
TS3=TS2*TS
TS4=TS3*TS
TS5=TS4*TS
TS6=TS5*TS
TS7=TS6*TS
TS8=TS7*TS
TS9=TS8*TS
TS10=TS9*TS
TS11=TS10*TS
YH=0.
YDH=0.
YDDH=0
YDDDH=0
YDDDDH=0.
YDDDDDH=0.
SIGNOISE=100.
```

Process noise and sampling time

Initial state estimates
and measurement noise

Six-State Linear Polynomial Kalman Filter-2

14

```
DO 14 I=1,ORDER
DO 14 J=1,ORDER
PHI(I,J)=0.
P(I,J)=0.
Q(I,J)=0.
IDN(I,J)=0.
CONTINUE
```

```
RMAT(1,1)=SIGNOISE**2 ]
```

Measurement noise matrix

```
IDN(1,1)=1.
IDN(2,2)=1.
IDN(3,3)=1.
IDN(4,4)=1.
IDN(5,5)=1.
IDN(6,6)=1.
```

```
P(1,1)=100.**2
P(2,2)=100.**2
P(3,3)=64.4**2
P(4,4)=2.5**2
P(5,5)=.25**2
P(6,6)=.025**2
```

Initial covariance matrix

```
PHI(1,1)=1.
PHI(1,2)=TS
PHI(1,3)=.5*TS2
PHI(1,4)=TS3/6.
PHI(1,5)=TS4/24.
PHI(1,6)=TS5/120.
PHI(2,2)=1.
PHI(2,3)=TS
PHI(2,4)=.5*TS2
PHI(2,5)=TS3/6.
PHI(2,6)=TS4/24.
PHI(3,3)=1.
PHI(3,4)=TS
PHI(3,5)=TS2/2.
```

**Non zero elements of
fundamental matrix**

Six-State Linear Polynomial Kalman Filter-3

PHI(3,6)=TS3/6.
PHI(4,4)=1.
PHI(4,5)=TS
PHI(4,6)=TS2/2.
PHI(5,5)=1.
PHI(5,6)=TS
PHI(6,6)=1.

HMAT(1,1)=1.
HMAT(1,2)=0.
HMAT(1,3)=0.
HMAT(1,4)=0.
HMAT(1,5)=0.
HMAT(1,6)=0.

CALL MATTRN(PHI,ORDER,ORDER,PHIT)
CALL MATTRN(HMAT,1,ORDER,HT)

Q(1,1)=PHIS*TS11/158400.
Q(1,2)=PHIS*TS10/28800.
Q(1,3)=PHIS*TS9/6480.
Q(1,4)=PHIS*TS8/1920.
Q(1,5)=PHIS*TS7/840.
Q(1,6)=PHIS*TS6/720.
Q(2,1)=Q(1,2)
Q(2,2)=PHIS*TS9/5184.
Q(2,3)=PHIS*TS8/1152.
Q(2,4)=PHIS*TS7/336.
Q(2,5)=PHIS*TS6/144.
Q(2,6)=PHIS*TS5/120.
Q(3,1)=Q(1,3)
Q(3,2)=Q(2,3)
Q(3,3)=PHIS*TS7/252.
Q(3,4)=PHIS*TS6/72.
Q(3,5)=PHIS*TS5/30.
Q(3,6)=PHIS*TS4/24.
Q(4,1)=Q(1,4)
Q(4,2)=Q(2,4)
Q(4,3)=Q(3,4)

Measurement matrix

Process noise matrix

Six-State Linear Polynomial Kalman Filter-4

```

Q(4,4)=PHIS*TS5/20.
Q(4,5)=PHIS*TS4/8.
Q(4,6)=PHIS*TS3/6.
Q(5,1)=Q(1,5)
Q(5,2)=Q(2,5)
Q(5,3)=Q(3,5)
Q(5,4)=Q(4,5)
Q(5,5)=PHIS*TS3/3.
Q(5,6)=PHIS*TS2/2.
Q(6,1)=Q(1,6)
Q(6,2)=Q(2,6)
Q(6,3)=Q(3,6)
Q(6,4)=Q(4,6)
Q(6,5)=Q(5,6)
Q(6,6)=PHIS*TS
S=0.
H=.01
T=0.
10 IF(T>79.)GOTO 999
   S=S+H
   YTOLD=YT
   YTDOLD=YTD
   STEP=1
   GOTO 200
66  STEP=2
   YT=YT+H*YTD
   YTD=YTD+H*YTDD
   T=T+H
   GOTO 200
55  CONTINUE
   YT=.5*(YTOLD+YT+H*YTD)
   YTD=.5*(YTDOLD+YTD+H*YTDD)

```

**Second-order Runge-Kutta
Integration for booster**

Six-State Linear Polynomial Kalman Filter-5

IF(S<(TS-.00001))GOTO 10

S=0.

CALL MATMUL(PHI,ORDER,ORDER,P,ORDER,ORDER,PHIP)
 CALL MATMUL(PHIP,ORDER,ORDER,PHIT,ORDER,ORDER,PHIPPHIT)
 CALL MATADD(PHIPPHIT,ORDER,ORDER,Q,M)
 CALL MATMUL(HMAT,1,ORDER,M,ORDER,ORDER,HM)
 CALL MATMUL(HM,1,ORDER,HT,ORDER,1,HMHT)
 CALL MATADD(HMHT,ORDER,ORDER,RMAT,HMHTR)
 HMHTRINV(1,1)=1/HMHTR(1,1)
 CALL MATMUL(M,ORDER,ORDER,HT,ORDER,1,MHT)
 CALL MATMUL(MHT,ORDER,1,HMHTRINV,1,1,K)
 CALL MATMUL(K,ORDER,1,HMAT,1,ORDER,KH)
 CALL MATSUB(IDN,ORDER,ORDER,KH,IKH)
 CALL MATMUL(IKH,ORDER,ORDER,M,ORDER,ORDER,P)

Riccati equations

CALL GAUSS(YNOISE,SIGNOISE)

YS=YT+YNOISE

Measurement noise

XK1=K(1,1)

XK2=K(2,1)

XK3=K(3,1)

XK4=K(4,1)

XK5=K(5,1)

XK6=K(6,1)

RES=YS-YH-TS*YDH-.5*TS*TS*YDDH-TS3*YDDDH/6.-TS4*YDDDDH/
 24.-TS5*YDDDDDH/120.

YH=YH+YDH*TS+.5*TS*TS*YDDH+TS3*YDDDH/6.+TS4*YDDDDH/24.
 +TS5*YDDDDDH/120.+XK1*RES

YDH=YDH+YDDH*TS+.5*TS2*YDDDH+TS3*YDDDDH/6.+
 TS4*YDDDDDH/24.+XK2*RES

YDDH=YDDH+YDDDH*TS+.5*TS2*YDDDDH+TS3*YDDDDDH/6.+XK3*RES

YDDDH=YDDDH+YDDDDH*TS+.5*TS2*YDDDDDH+XK4*RES

YDDDDH=YDDDDH+TS*YDDDDDH+XK5*RES

YDDDDDH=YDDDDDH+XK6*RES

ERRY=YT-YH

ERRYD=YTD-YDH

ERRYDD=YTDD-YDDH

SP11=SQRT(P(1,1))

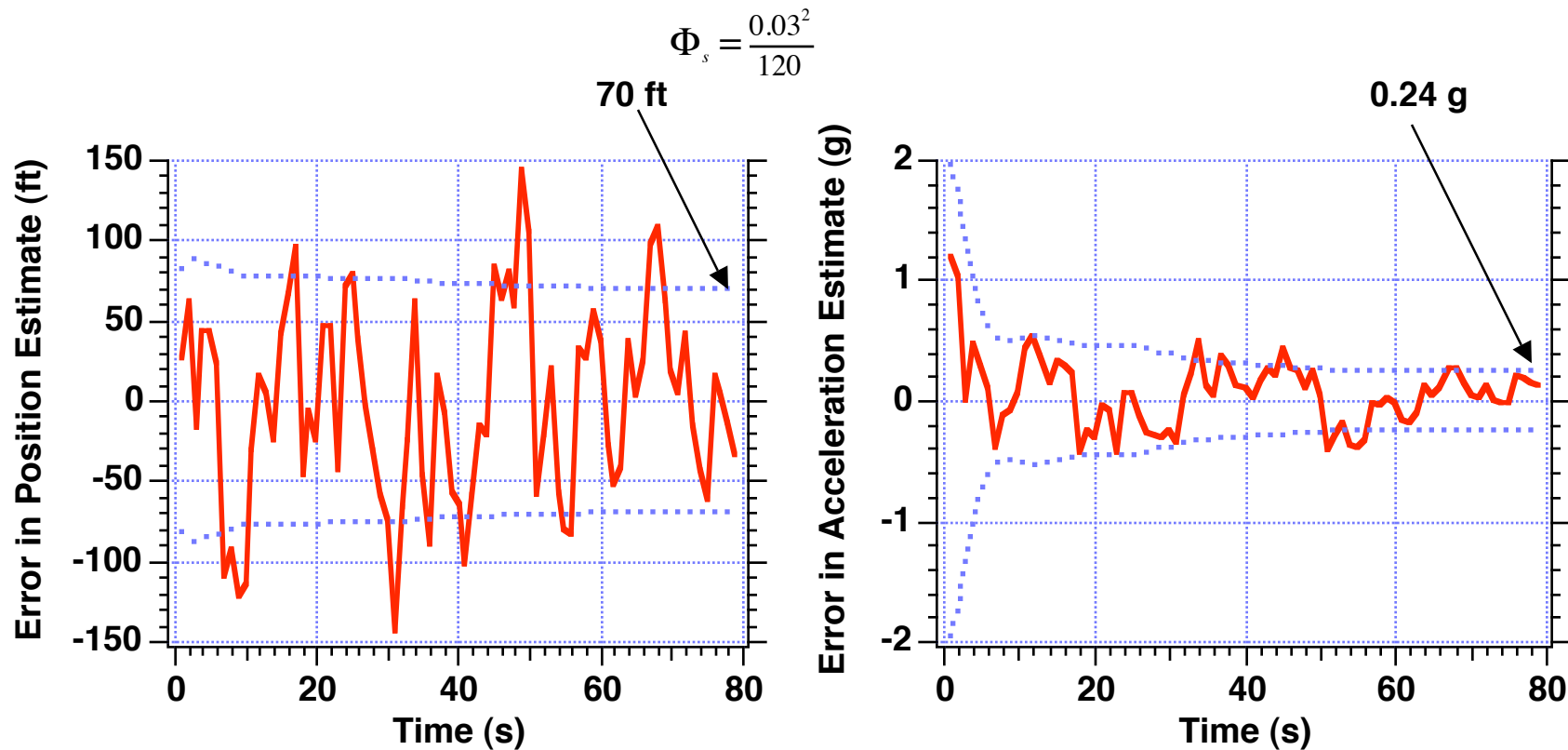
Six-state filter

Six-State Linear Polynomial Kalman Filter-6

```
SP22=SQRT(P(2,2))
SP33=SQRT(P(3,3))
WRITE(9,*)T,YTDD/32.2,YDDH/32.2
WRITE(1,*)T,YTDD/32.2,YDDH/32.2,YTD,YDH,ZDH
WRITE(2,*)T,ERRY,SP11,-SP11,ERRYD,SP22,-SP22,ERRYDD/32.2
1                                     ,SP33/32.2,-SP33/32.2
GOTO 10
200 CONTINUE
IF(T<80.)THEN
    WGT=-125.*T+14000.
    TRST=30000.
ELSE
    WGT=4000.
    TRST=0.
ENDIF
AT=G*TRST/WGT
YTDD=AT-G
999 IF(STEP-1)66,66,55
CONTINUE
PAUSE
CLOSE(1)
CLOSE(2)
END
```

Booster differential equation

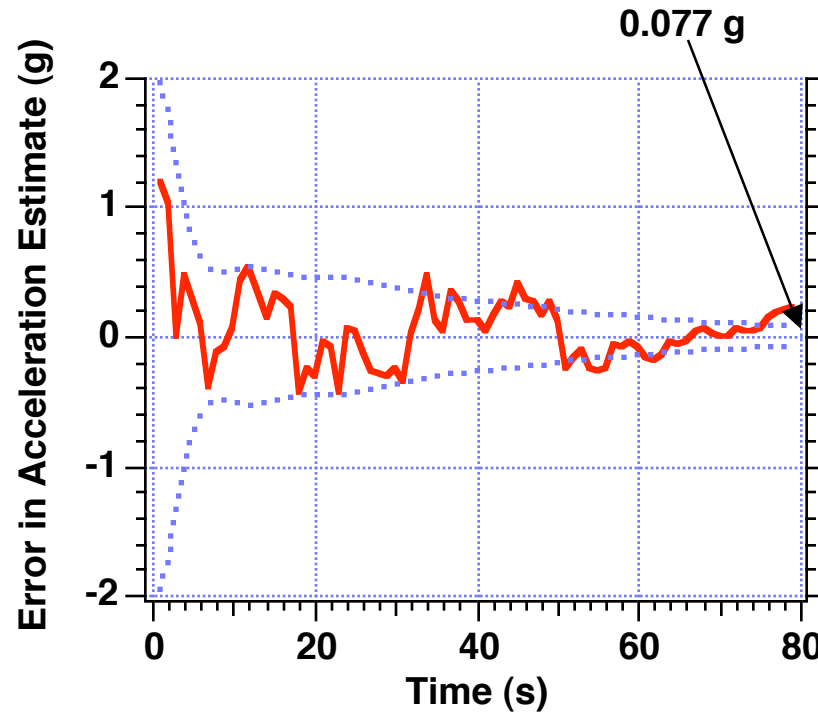
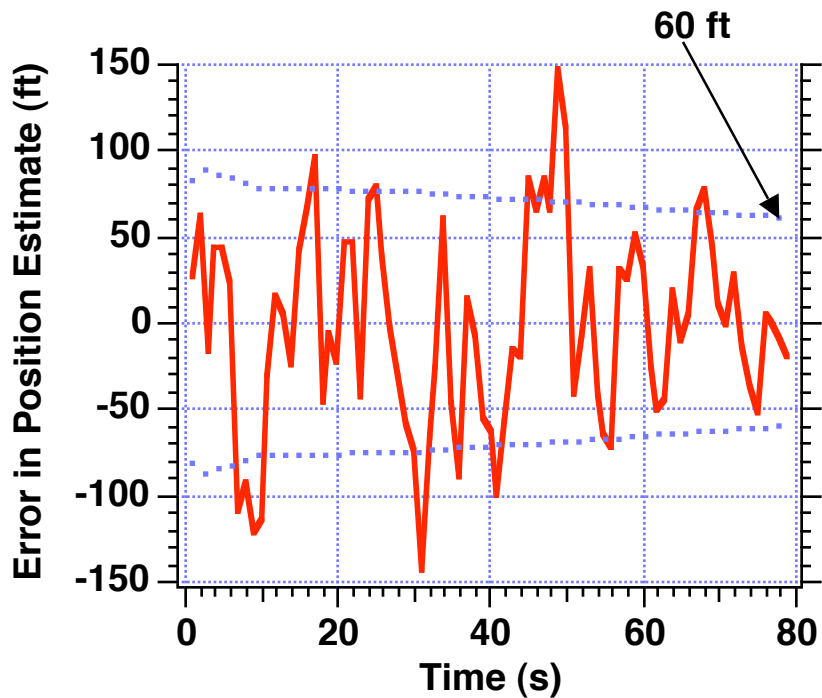
Six-State Kalman Filter is Able To Track Boosting Target Better



*100 ft measurement noise and 1 s sampling time

Reducing Process Noise To Zero With Six-State Kalman Filter Does Not Cause Divergence

$$\Phi_s = \frac{0^2}{120}$$



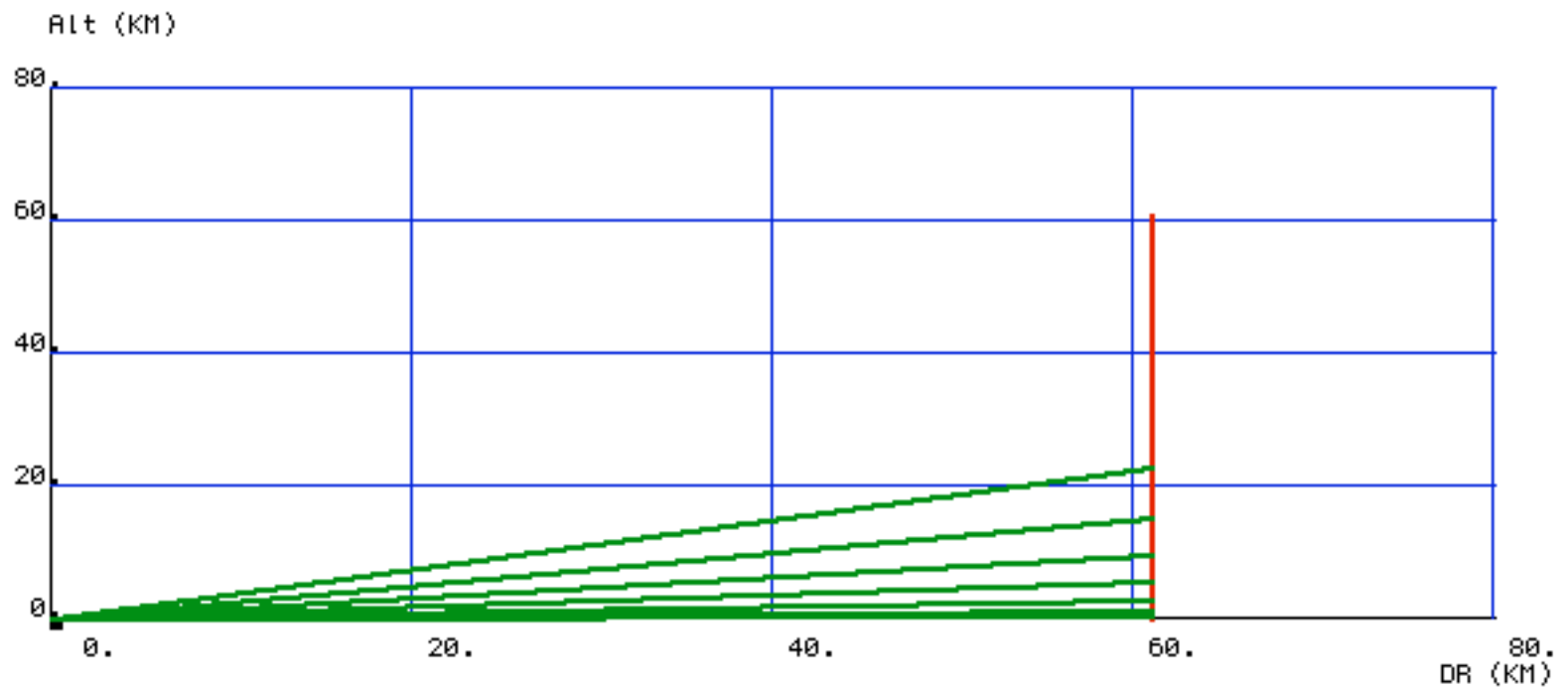
*100 ft measurement noise and 1 s sampling time

Tracking Accuracy Summary

	Position Error	Acceleration Error
Nominal 3SKF	82 ft	0.97 g
Nominal 6SKF	70 ft	0.24 g
6SKF Zero Process Noise	60 ft	0.077 g

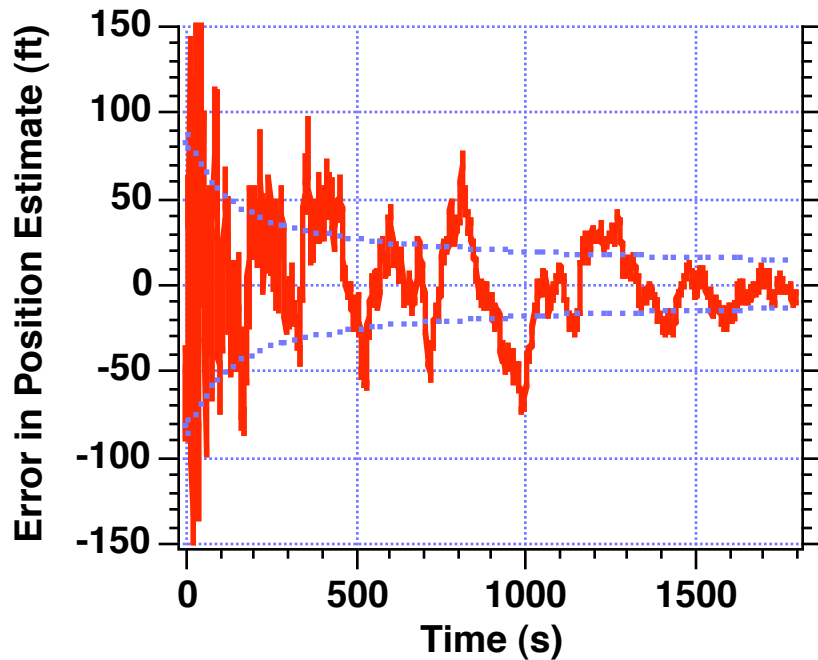
Interesting Experiments

If Booster Takes Off Half Hour Late Radar Loses Track With Six-State Filter and Zero Process Noise

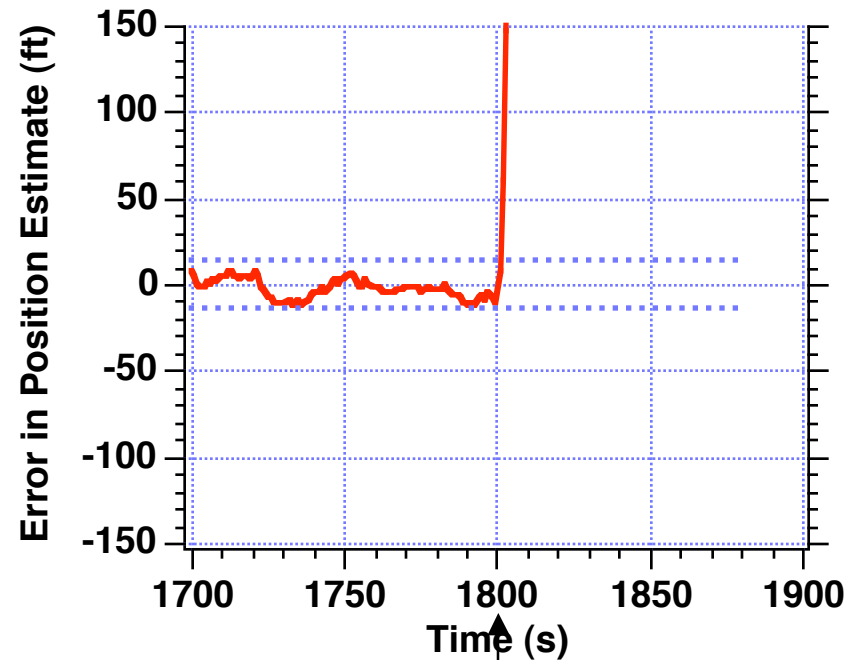


With Zero Process Noise 6-State Filter Diverges After Booster Launch

While Booster on Launch Pad For Half an Hour

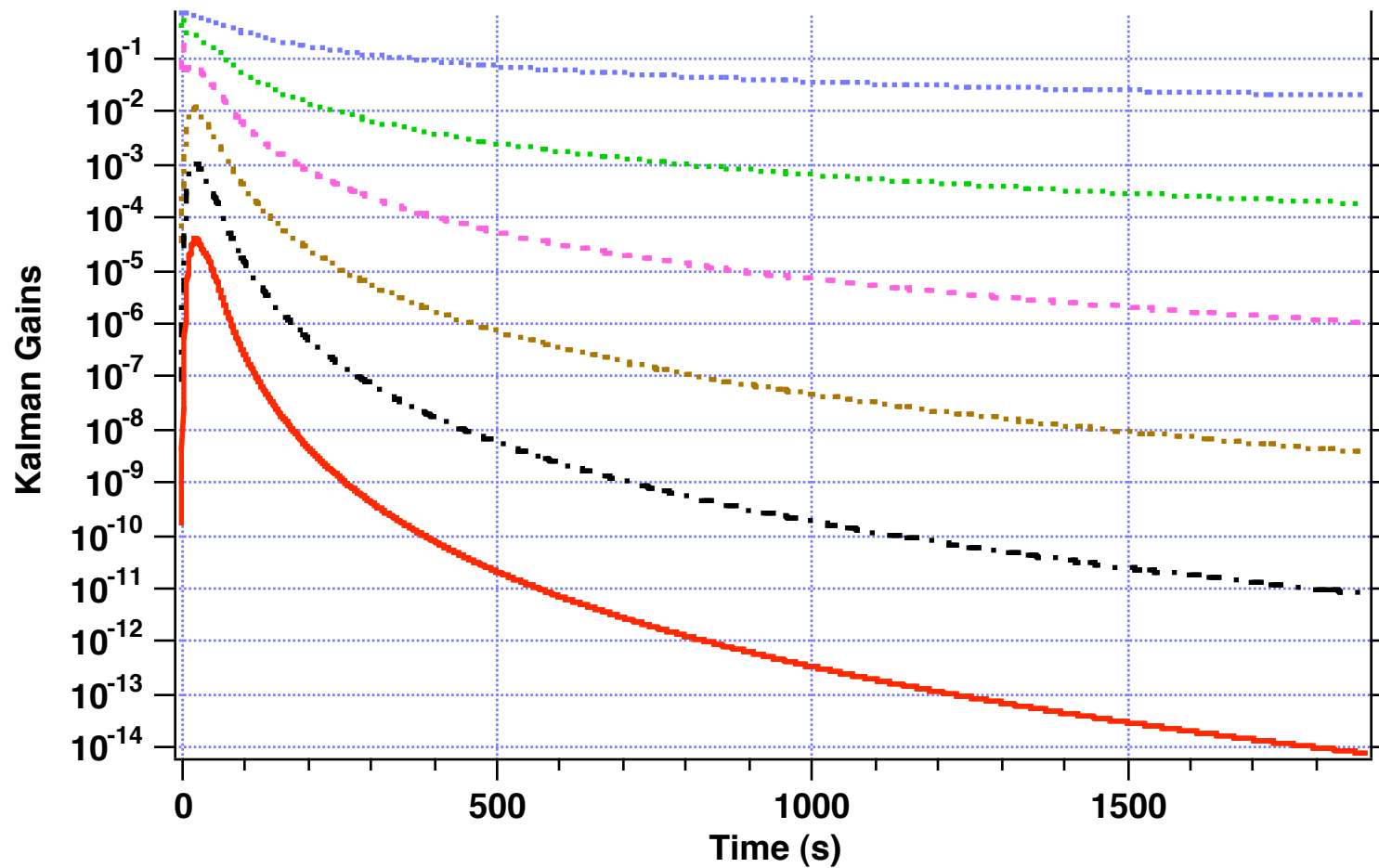


From 100 s Before Launch Until a Little After Launch



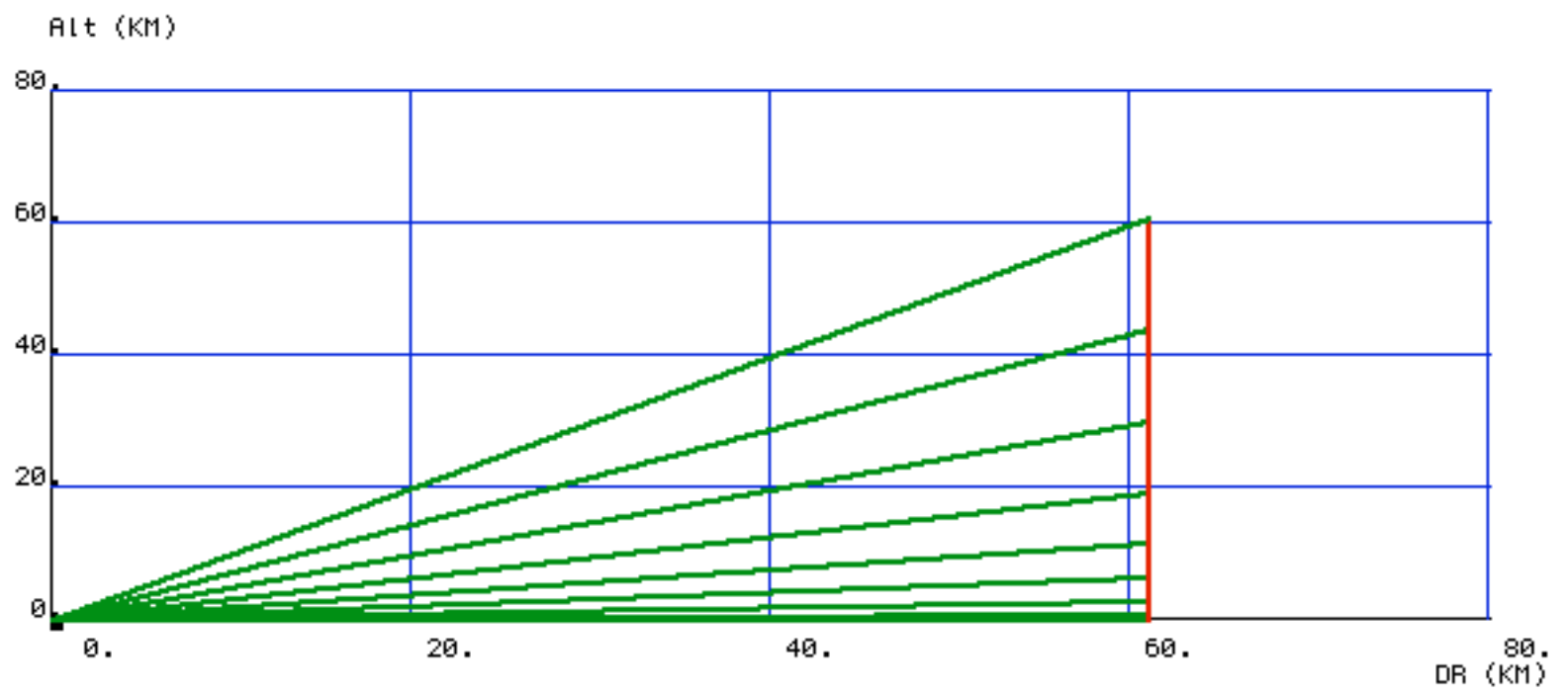
*100 ft measurement noise and 1 s sampling time

All Kalman Gains Go To Zero When There is No Process Noise



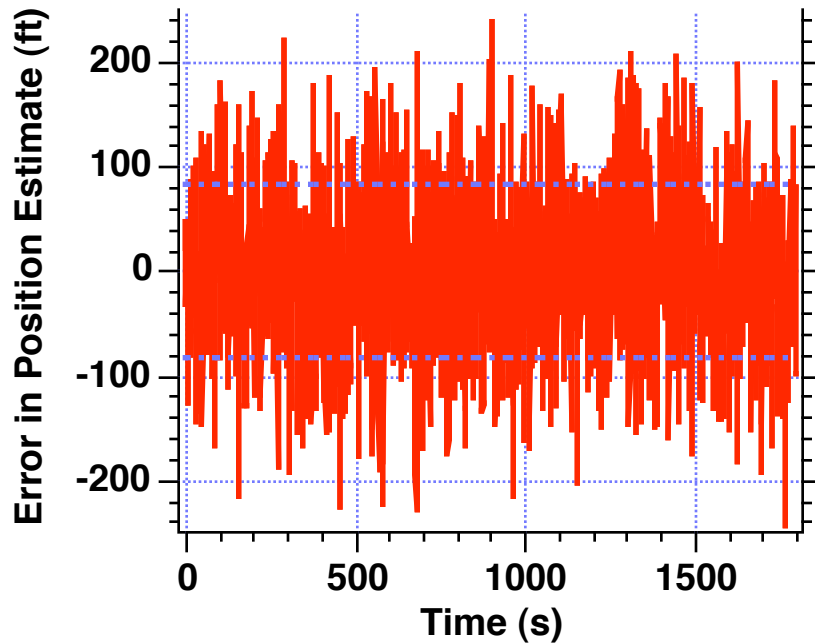
When Kalman gains are small, filter stops paying attention to measurements

If Booster Takes Off Half Hour Late Radar Does Not Lose Track With Three-State Filter With Process Noise

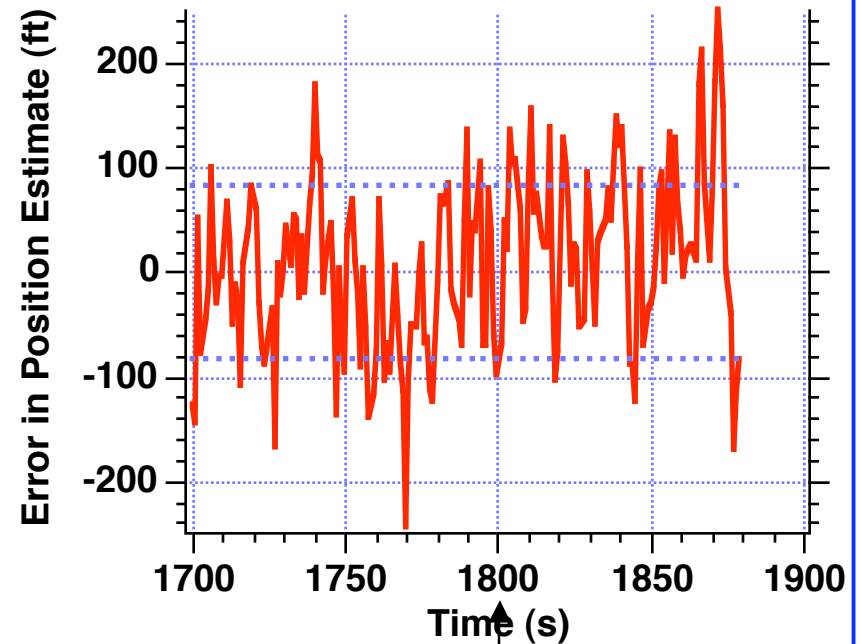


Three-State Filter With Process Noise is Able To Track Target After Half Hour Launch Delay

While Booster on Launch Pad For Half an Hour

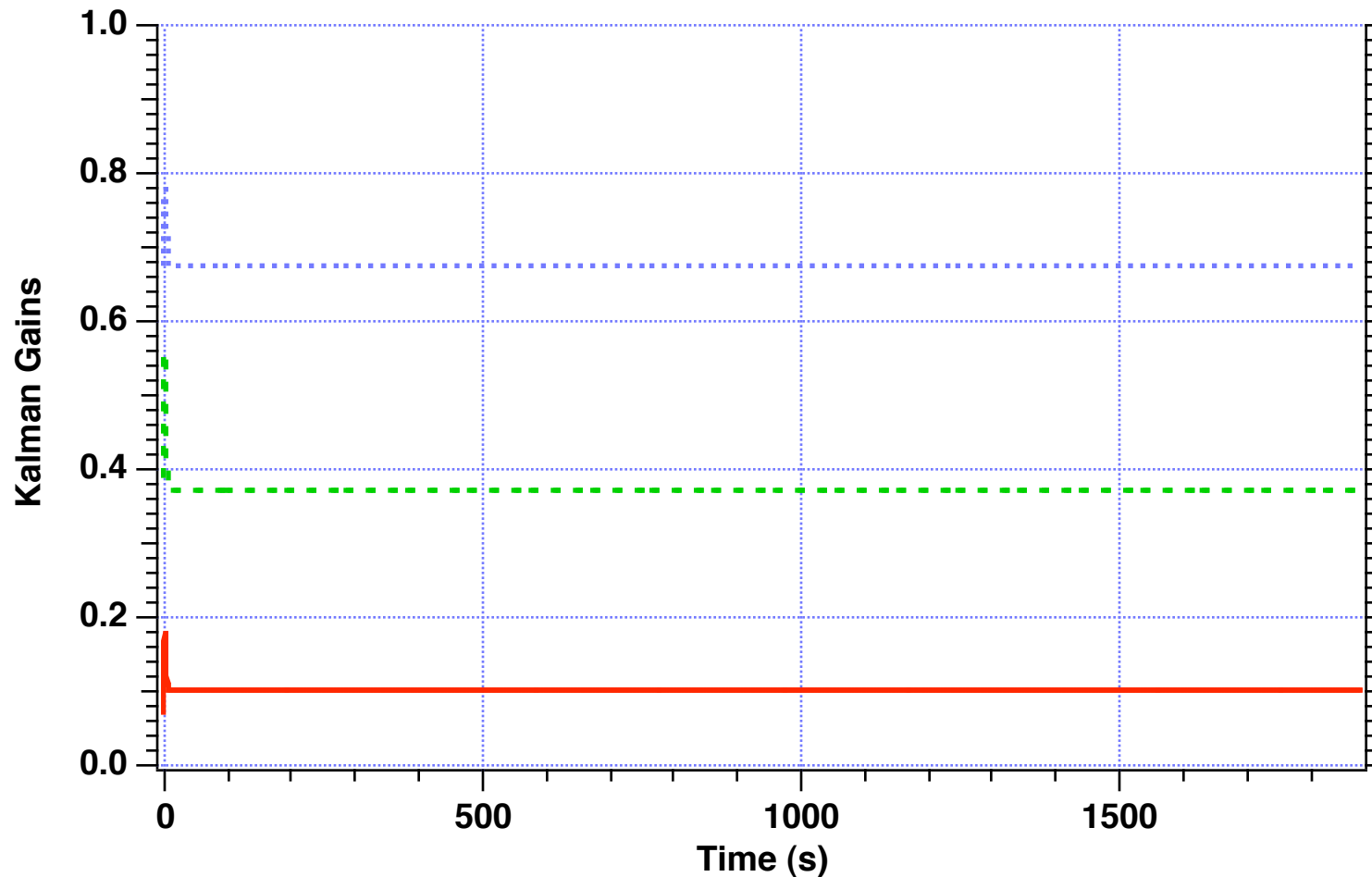


From 100 s Before Launch Until a Little After Launch



*100 ft measurement noise and 1 s sampling time

Kalman Gains Do Not Go To Zero For Three-State Filter With Process Noise



When Kalman gains do not get too small filter always pays attention to measurements

A Tracking Disaster Summary

**Making process noise too small or zero be very dangerous, even
when computer results are very encouraging**