

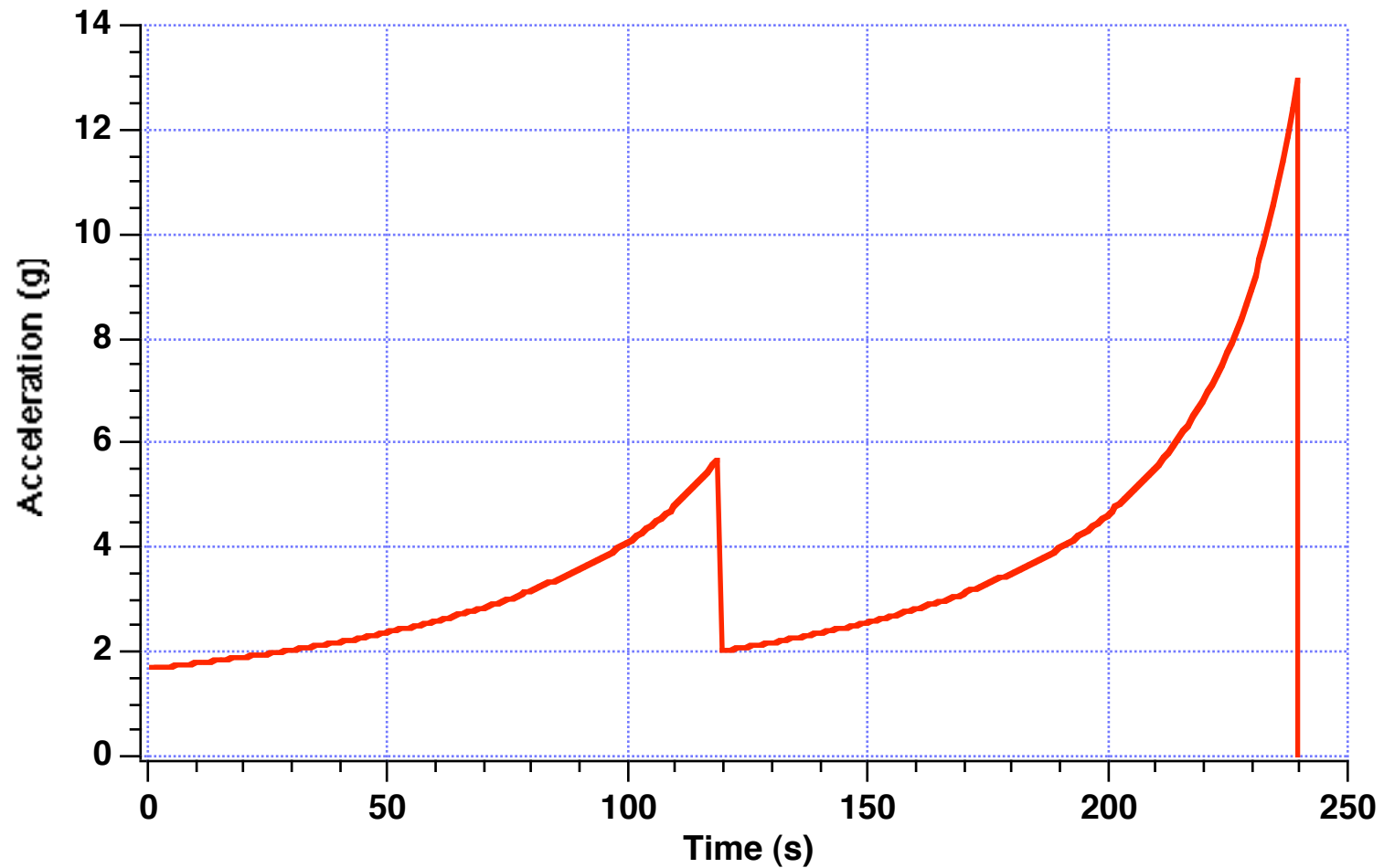
# **Boost Phase Filtering Options: Is Simpler Better?**

# Overview

- **Simple ICBM model**
  - **Simple guidance equations for flat earth**
- **Range and angle measurement model**
  - Creating pseudo measurements**
- **2-State template based Kalman filter**
  - Performance and robustness**
- **3-State polynomial Kalman filter**
  - **Performance and comparison with 2-state filter**
- **Summary**

# Simple ICBM Model

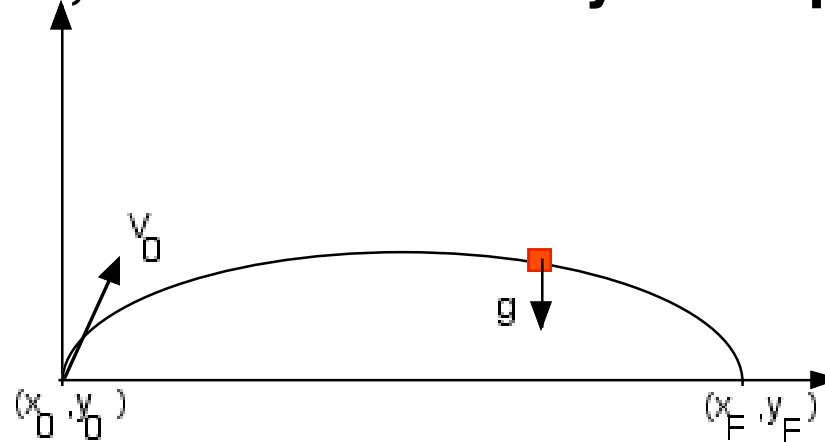
# Acceleration Profile of Generic Two-Stage APS Liquid ICBM



## **Two Kalman Filter Design Possibilities**

- **Two-state template based Kalman filter**
  - **Magnitude of ICBM acceleration profile is known (template)**
  - **Estimate ICBM position and velocity based on range angle measurements**
- **General purpose three-state Kalman filter**
  - **Assume no a priori information is available**
  - **Estimate ICBM position, velocity and acceleration based on range and angle measurements only**

## How ICBMs Guide to Their Intended Target (Flat Earth, Constant Gravity Example)



Given Initial ICBM location  $x_0, y_0$  and destination  $x_F, y_F$  and desired arrival time  $t_F$  we desire initial velocity vector required to hit target at desired arrival time

Future location of ICBM can be calculated from High School physics

$$x_F = x_0 + \dot{x}_0 t_F$$

$$y_F = y_0 + \dot{y}_0 t_F - 0.5 g t_F^2$$

Solve for initial velocity as if ICBM launched as cannon ball

$$\dot{x}_0 = \frac{x_F - x_0}{t_F}$$

$$\dot{y}_0 = \frac{y_F - y_0 + 0.5 g t_F^2}{t_F}$$

Flat earth solution to Lambert's problem

# Lambert Guidance - 1

At each instant of time compute desired (Lambert) velocity components

$$t_{go} = t_F - t$$

$$V_{Lambert_x} = \frac{x_F - x}{t_{go}}$$

$$V_{Lambert_y} = \frac{y_F - y + 0.5gt_{go}^2}{t_{go}}$$

Calculate velocity to be gained (Lambert velocity minus current velocity)

$$\Delta V_x = V_{Lambert_x} - V_x$$

$$\Delta V_y = V_{Lambert_y} - V_y$$

$$\Delta V = \sqrt{\Delta V_x^2 + \Delta V_y^2}$$

In Lambert guidance we align ICBM thrust vector with velocity to be gained vector

$$a_{T_x} = \frac{\Delta V_x}{\Delta V} a_T$$

$$a_{T_y} = \frac{\Delta V_y}{\Delta V} a_T$$

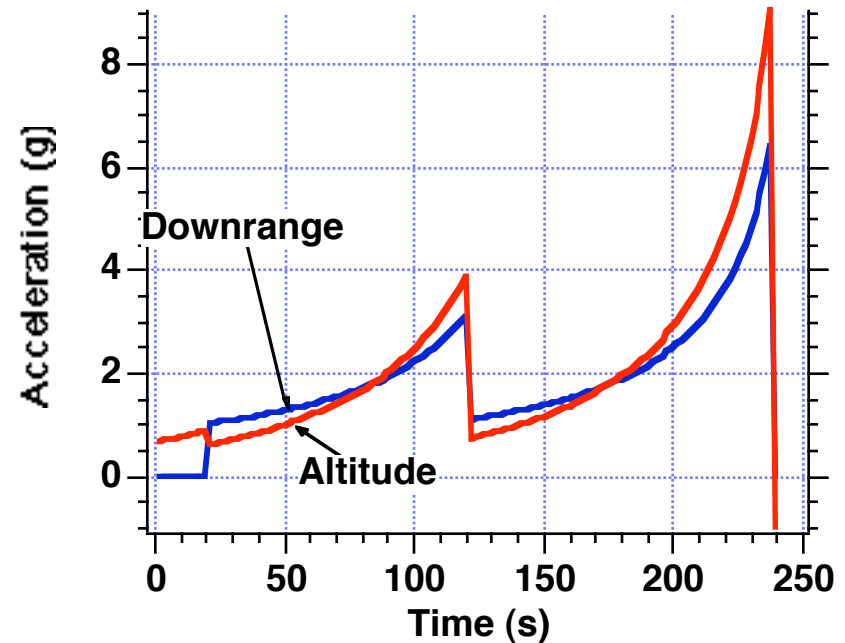
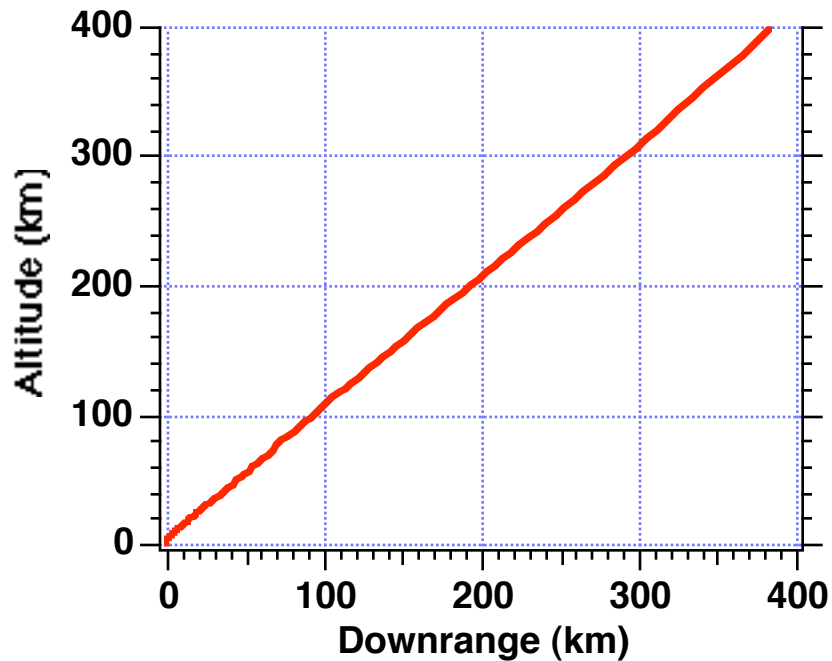
## Lambert Guidance - 2

Where  $a_T$  is the ICBM acceleration magnitude (Thrust/Weight)

We iterate at each guidance update until  $\Delta V$  goes to zero and then we thrust terminate



# Trajectory and Acceleration Profile Components of ICBM Boost Phase Portion of 5000 km Trajectory



# Measurements and Pseudo Measurements

# Pseudo Measurement Equations

## Actual range and angle from radar to ICBM

$$r_T = \sqrt{(x_T - x_R)^2 + (y_T - y_R)^2}$$

$$\theta_T = \tan^{-1}\left(\frac{y_T - y_R}{x_T - x_R}\right)$$

In order to avoid building EKF we will consider pseudo position measurements

$$x_T^* = r_T^* \cos \theta_T^* + x_R$$

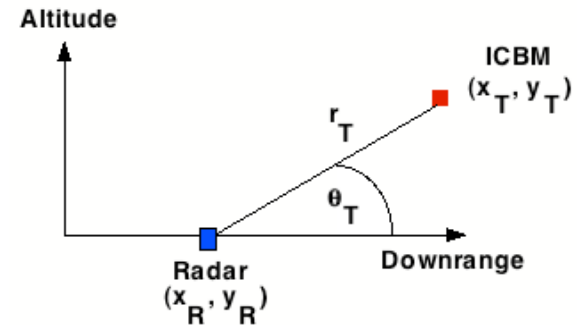
$$y_T^* = r_T^* \sin \theta_T^* + y_R$$

These pseudo measurements can serve as inputs to 2 decoupled linear Kalman filters

One can show (next 2 slides) pseudo measurement variances are given by

$$\sigma_x^2 = \cos^2 \theta_T \sigma_R^2 + r_T^2 \sin^2 \theta_T \sigma_\theta^2$$

$$\sigma_y^2 = \sin^2 \theta_T \sigma_R^2 + r_T^2 \cos^2 \theta_T \sigma_\theta^2$$



# Deriving Variance For Pseudo Measurement Noise-1

In Cartesian frame model of real world is linear but actual measurements are nonlinear with respect to states. However pseudo measurements are linearly related to states

Recall from previous slide

$$x_T = r \cos \theta + x_R$$

$$y_T = r \sin \theta + y_R$$

Find total differential from calculus

$$\Delta x_T = \frac{\partial x_T}{\partial r} \Delta r + \frac{\partial x_T}{\partial \theta} \Delta \theta = \cos \theta \Delta r - r \sin \theta \Delta \theta$$

$$\Delta y_T = \frac{\partial y_T}{\partial r} \Delta r + \frac{\partial y_T}{\partial \theta} \Delta \theta = \sin \theta \Delta r + r \cos \theta \Delta \theta$$

Square both equations

$$\Delta x_T^2 = \cos^2 \theta \Delta r^2 - 2r \sin \theta \cos \theta \Delta r \Delta \theta + r^2 \sin^2 \theta \Delta \theta^2$$

$$\Delta y_T^2 = \sin^2 \theta \Delta r^2 + 2r \sin \theta \cos \theta \Delta r \Delta \theta + r^2 \cos^2 \theta \Delta \theta^2$$

## Deriving Variance For Pseudo Measurement Noise -2

Taking expectations of both sides assuming range and angle measurements are not correlated

$$E(\Delta x_T^2) = \cos^2\theta E(\Delta r^2) + r^2 \sin^2\theta E(\Delta\theta^2)$$

$$E(\Delta y_T^2) = \sin^2\theta E(\Delta r^2) + r^2 \cos^2\theta E(\Delta\theta^2)$$

Since

$$\sigma_{x_T}^2 = E(\Delta x_T^2)$$

$$\sigma_{y_T}^2 = E(\Delta y_T^2)$$

$$\sigma_r^2 = E(\Delta r^2)$$

$$\sigma_\theta^2 = E(\Delta\theta^2)$$

We can say

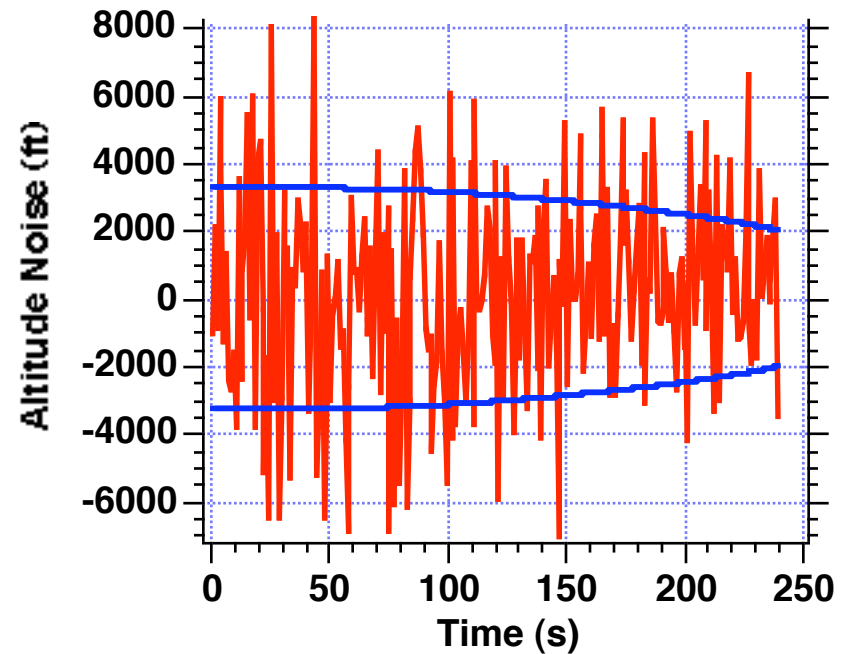
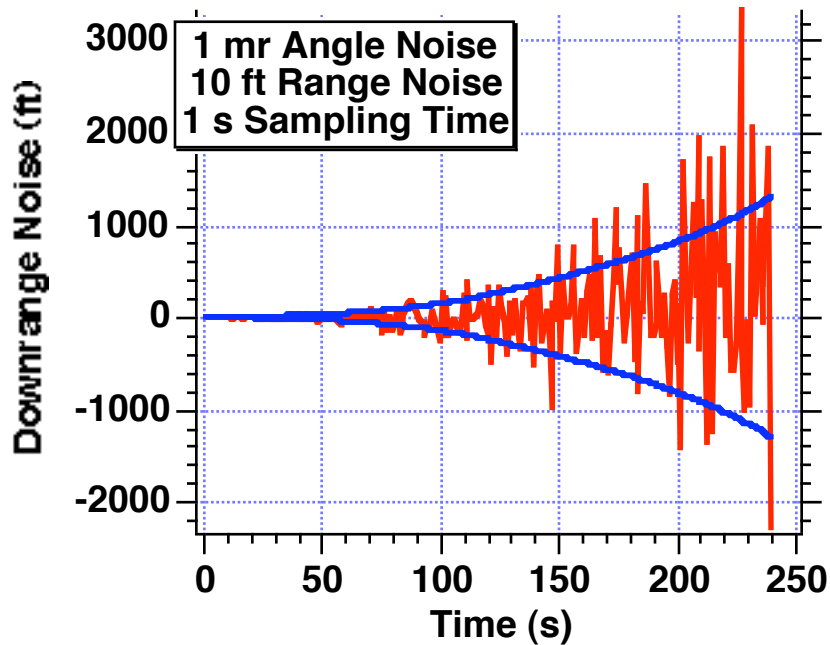
$$\sigma_{x_T}^2 = \cos^2\theta \sigma_r^2 + r^2 \sin^2\theta \sigma_\theta^2$$

$$\sigma_{y_T}^2 = \sin^2\theta \sigma_r^2 + r^2 \cos^2\theta \sigma_\theta^2$$

**\*We are pretending noise is on x and y rather than r and  $\theta$**

**These pseudo measurement variances can be input to Ricatti equations of 2 decoupled linear Kalman filters**

# Effective Position Noise



Theoretical calculations for pseudo measurement noise variance appear to be correct

## **2-State Template Base Kalman Filter**

# Two 2-State Template Based Decoupled Linear Polynomial Kalman Filters

## Downrange Filter

$$\text{Res}_{x_k} = x_{T_k}^* - \hat{x}_{T_{k-1}} - \hat{\dot{x}}_{T_{k-1}} T_s - 0.5a_{Tx_{k-1}} T_s^2$$

$$\hat{x}_{T_k} = \hat{x}_{T_{k-1}} + \hat{\dot{x}}_{T_{k-1}} T_s + 0.5a_{Tx_{k-1}} T_s^2 + K_{1x_k} \text{Res}_{x_k}$$

$$\hat{\dot{x}}_{T_k} = \hat{\dot{x}}_{T_{k-1}} + a_{Tx_{k-1}} T_s + K_{2x_k} \text{Res}_{x_k}$$

Note that template implies we know magnitude and direction of  $a_T$  perfectly

## Altitude Filter

$$\text{Res}_{y_k} = y_{T_k}^* - \hat{y}_{T_{k-1}} - \hat{\dot{y}}_{T_{k-1}} T_s - 0.5a_{Ty_{k-1}} T_s^2$$

$$\hat{y}_{T_k} = \hat{y}_{T_{k-1}} + \hat{\dot{y}}_{T_{k-1}} T_s + 0.5a_{Ty_{k-1}} T_s^2 + K_{1y_k} \text{Res}_{y_k}$$

$$\hat{\dot{y}}_{T_k} = \hat{\dot{y}}_{T_{k-1}} + a_{Ty_{k-1}} T_s + K_{2y_k} \text{Res}_{y_k}$$

## Process Noise

$$\mathbf{Q}_k = \Phi_s \begin{bmatrix} \frac{T_s^3}{3} & \frac{T_s^2}{2} \\ \frac{T_s^2}{2} & T_s \end{bmatrix}$$

We will be adjusting  $\Phi_s$  to tune the filters



# 2-State Template Based Decoupled Linear Polynomial Kalman Filters -1

```
GLOBAL DEFINE
    INCLUDE 'quickdraw.inc'
```

```
END
IMPLICIT REAL*8 (A-H)
IMPLICIT REAL*8 (O-Z)
REAL*8 M11,M12,M22,K1,K2
REAL*8 M11P,M12P,M22P,K1P,K2P
LOGICAL QBOOST,QGRAV
INTEGER STEP
OPEN(1,STATUS='UNKNOWN',FILE='DATFIL')
OPEN(2,STATUS='UNKNOWN',FILE='NOISEFIL')
OPEN(3,STATUS='UNKNOWN',FILE='COVFIL')
```

```
QBOOST=.TRUE.
QGRAV=.FALSE.
```

```
TF=1300.
RT1F=5000.*3280.
RT2F=0.
GAMTDEG=89.9
```

Desired Destination and Time of Arrival

```
VT=1.
RT1=0.*3280.
RT2=0.
XR=1000.*3280.
YR=0.
VT1=VT*COS(GAMTDEG/57.3)
VT2=VT*SIN(GAMTDEG/57.3)
```

```
G=32.2
H=.01
TS=1.
S=0.
```

```
SCOUNT=0.
SIGR=10.
SIGTH=.001
```

Range and angle measurement errors

```
PHIN=100.
RT=SQRT((RT1-XR)**2+(RT2-YR)**2)
THET=ATAN2(RT2-YR,RT1-XR)
```

Filter process noise  
Actual range and angle

```
SIGRT1=SQRT((COS(THET)*SIGR)**2+(RT*SIN(THET)*SIGTH)**2)
SIGRT2=SQRT((SIN(THET)*SIGR)**2+(RT*COS(THET)*SIGTH)**2)
```

Standard deviation of pseudo measurements

```
K1=0.
K2=0.
```

```
RT1H=0.
VT1H=0.
RT2H=0.
VT2H=0.
```

Filter initialization

```
T=0.
```

# 2-State Template Based Decoupled Linear Polynomial Kalman Filters -2

```
P11=99999999999.
P12=0.
P22=99999999999.
P11P=99999999999.
P12P=0.
P22P=99999999999.
```

Initial covariance matrix

```
AX=0.
AY=0.
XN=0.
ERR=0.
```

10 IF(T>240.)GOTO 999

```
RT1OLD=RT1
RT2OLD=RT2
VT1OLD=VT1
VT2OLD=VT2
```

```
STEP=1
GOTO 200
```

66

```
STEP=2
RT1=RT1+H*VT1
RT2=RT2+H*VT2
VT1=VT1+H*AT1
VT2=VT2+H*AT2
```

```
T=T+H
```

```
GOTO 200
```

55

```
RT1=(RT1OLD+RT1)/2+.5*H*VT1
RT2=(RT2OLD+RT2)/2+.5*H*VT2
VT1=(VT1OLD+VT1)/2+.5*H*AT1
VT2=(VT2OLD+VT2)/2+.5*H*AT2
IF(QBOOST)THEN
```

```
TGOLAM=TF-T
VXLAM=(RT1F-RT1)/TGOLAM
VYLAM=(RT2F-RT2+16.1*TGOLAM*TGOLAM)/TGOLAM
```

```
DEL VX=VXLAM-VT1
DEL VY=VYLAM-VT2
DEL V=SQRT(DEL VX**2+DEL VY**2)
```

```
IF(TRST>0..AND.DELV>10.)THEN
    AX=AT*DEL VX/DEL V
    AY=AT*DEL VY/DEL V
```

```
ELSEIF(DEL V<10.)THEN
    TRST=0.
    QBOOST=.FALSE.
```

```
AX=0.
AY=0.
VT1OLD=VXLAM
VT2OLD=VYLAM
```

2nd order Runge-Kutta integration of ICBM for boost phase

Lambert guidance

## 2-State Template Based Decoupled Linear Polynomial Kalman Filters -3

```

ELSE
    QBOOST=.FALSE.
    AX=0.
    AY=0.
    WRITE(9,*)DELV
    PAUSE
ENDIF
ENDIF
IF(T<20.)THEN
    AX=0.
    AY=AT
ENDIF
SCOUNT=SCOUNT+H
IF(SCOUNT.LT.(TS-.00001))GOTO 10
SCOUNT=0.
XN=XN+1.
XK1=2.*(2.*XN-1.)/(XN*(XN+1.))
XK2=6./(XN*(XN+1.)*TS)
TS2=TS*TS
TS3=TS2*TS
TS4=TS3*TS
TS5=TS4*TS
RT=SQRT((RT1-XR)**2+(RT2-YR)**2)
THET=ATAN2(RT2-YR,RT1-XR)
SIGRT1=SQRT((COS(THET)*SIGR)**2+(RT*SIN(THET)*SIGTH)**2)
SIGRT2=SQRT((SIN(THET)*SIGR)**2+(RT*COS(THET)*SIGTH)**2)
SIGN2=SIGRT1*SIGRT1
M11=P11+2.*TS*P12+TS2*P22+TS3*PHIN/3.
M12=P12+TS*P22+.5*PHIN*TS2
M22=P22+PHIN*TS
K1=M11/(M11+SIGN2)
K2=M12/(M11+SIGN2)
P11=(1.-K1)*M11
P12=(1.-K1)*M12
P22=-K2*M12+M22

```

**Go straight up for first 20 s**

**Least squares filter gains**

**True angle and range**

**Pseudo measurement standard deviations**

**Ricatti equations for downrange filter**

## 2-State Template Based Decoupled Linear Polynomial Kalman Filters -4

```

SIGN2P=SIGRT2*SIGRT2
M11P=P11P+2.*TS*P12P+TS2*P22P+TS3*PHIN/3.
M12P=P12P+TS*P22P+.5*PHIN*TS2
M22P=P22P+PHIN*TS
K1P=M11P/(M11P+SIGN2P)
K2P=M12P/(M11P+SIGN2P)
P11P=(1.-K1P)*M11P
P12P=(1.-K1P)*M12P
P22P=-K2P*M12P+M22P
CALL GAUSS(THETNOISE,SIGTH)
CALL GAUSS(RTNOISE,SIGR)
THETMEAS=THET+THETNOISE
RTMEAS=RT+RTNOISE
RT1S=RTMEAS*COS(THETMEAS)+XR
RT2S=RTMEAS*SIN(THETMEAS)+YR
IF(XN<10.)THEN
    XK1PZ=XK1
    XK2PZ=XK2
ELSE
    XK1PZ=K1
    XK2PZ=K2
ENDIF
IF(.NOT.QGRAV)THEN
    AT1H=AT1*(1+ERR)
    AT2H=AT2*(1+ERR)
ELSE
    IF(XN<10.)THEN
        AT1H=AT1
        AT2H=AT2
    ELSE
        VTH=SQRT(VT1H**2+VT2H**2)
        ATH=AT*(1.+ERR)
        AT1H=ATH*VT1H/VTH
        AT2H=ATH*VT2H/VTH-G
    ENDIF
ENDIF
ENDIF

```

**Ricatti equations for altitude filter**

**Actual angle and range measurements**

**Pseudo measurements**

**Use least squares gains for first 10 measurements**

**Know direction of acceleration**

**Gravity turn assumption for acceleration vector**

## 2-State Template Based Decoupled Linear Polynomial Kalman Filters -5

```

RESX=RT1S-RT1H-TS*VT1H-.5*TS2*AT1H
RT1H=XK1PZ*RESX+RT1H+TS*VT1H+.5*TS2*AT1H
VT1H=XK2PZ*RESX+VT1H+TS*AT1H
IF(XN<10.)THEN
    XK1PPZ=XK1
    XK2PPZ=XK2
ELSE
    XK1PPZ=K1P
    XK2PPZ=K2P
ENDIF
RESY=RT2S-RT2H-TS*VT2H-.5*TS2*AT2H
RT2H=XK1PPZ*RESY+RT2H+TS*VT2H+.5*TS2*AT2H
VT2H=XK2PPZ*RESY+VT2H+TS*AT2H
RT1KM=RT1/3280.
RT2KM=RT2/3280.
RT1NOISE=RT1-RT1S
RT2NOISE=RT2-RT2S
ERRRT1=RT1-RT1H
ERRVT1=VT1-VT1H
SP11=SQRT(P11)
SP22=SQRT(P22)
ERRRT2=RT2-RT2H
ERRVT2=VT2-VT2H
SP11P=SQRT(P11P)
SP22P=SQRT(P22P)
WRITE(9,*)T,RT1KM,RT2KM,VT1,VT1H,VT2,VT2H,AT1/G,AT1H/G,
1           AT2/G,AT2H/G
WRITE(1,*)T,RT1KM,RT2KM,VT1,VT1H,VT2,VT2H,AT1/G,AT1H/G,
1           AT2/G,AT2H/G
WRITE(2,*)T,RT1NOISE,SIGRT1,-SIGRT1,RT2NOISE,SIGRT2,-SIGRT2
WRITE(3,*)T,ERRRT1,SP11,-SP11,ERRVT1,SP22,-SP22,ERRRT2,SP11P,
1           -SP11P,ERRVT2,SP22P,-SP22P
GOTO 10
200 CONTINUE

```

Downrange Kalman filter

Use least squares gains for first 10 measurements

Altitude Kalman filter

Actual and theoretical errors in estimates

## 2-State Template Based Decoupled Linear Polynomial Kalman Filters -6

```

IF(T<120.)THEN
    WGT=-2622*T+440660.
    TRST=725850.
ELSEIF(T<240.)THEN
    WGT=-642.*T+168120.
    TRST=182250.
ELSE
    WGT=5500.
    TRST=0.
ENDIF
VT=SQRT(VT1**2+VT2**2)
GAM=ATAN2(VT2,VT1)
AT=G*TRST/WGT
AT1=AX
AT2=-G+AY
IF(STEP-1)66,66,55
CONTINUE
RT1KM=RT1/3280.
RT2KM=RT2/3280.
WRITE(9,*)T,RT1KM,RT2KM,VT1,VT1H,VT2,VT2H,AT1/G,AT1H/G,
1          AT2/G,AT2H/G
WRITE(1,*)T,RT1KM,RT2KM,VT1,VT1H,VT2,VT2H,AT1/G,AT1H/G,
1          AT2/G,AT2H/G
PAUSE
CLOSE(1)
CLOSE(2)
CLOSE(3)
END

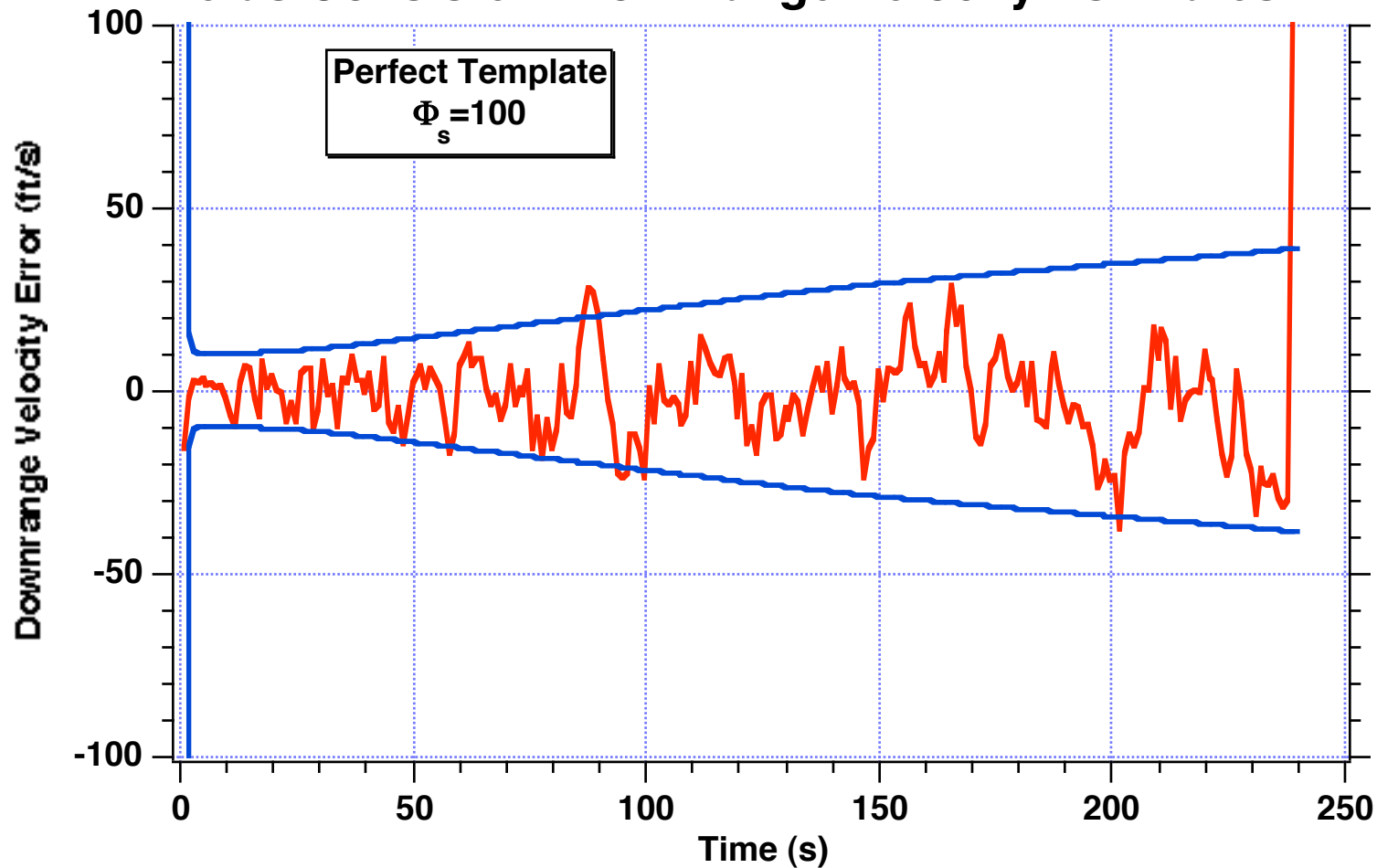
```

ICBM

Acceleration

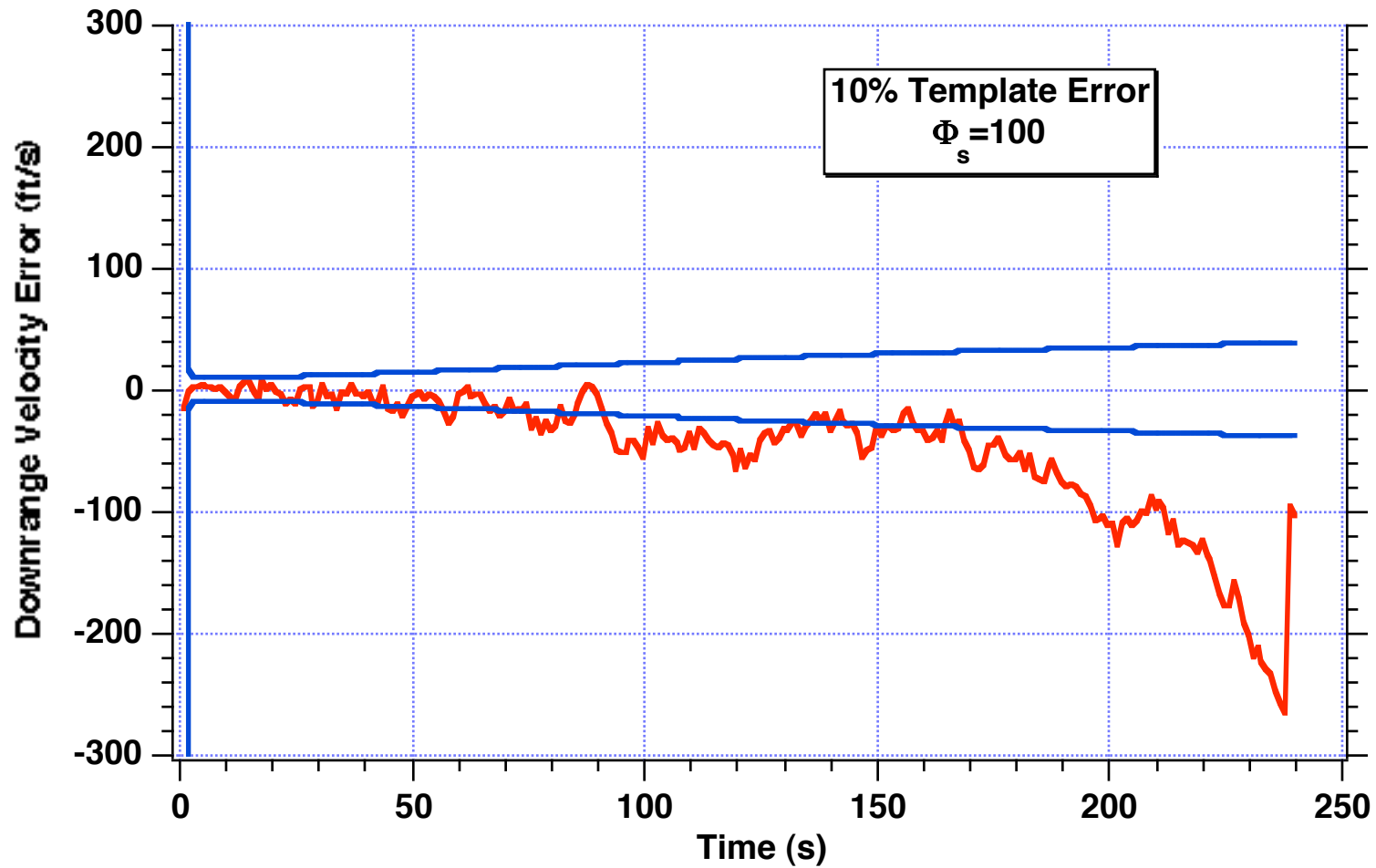
Differential equations

## With Perfect Acceleration Template Two-State Kalman Filter Yields Consistent Downrange Velocity Estimates



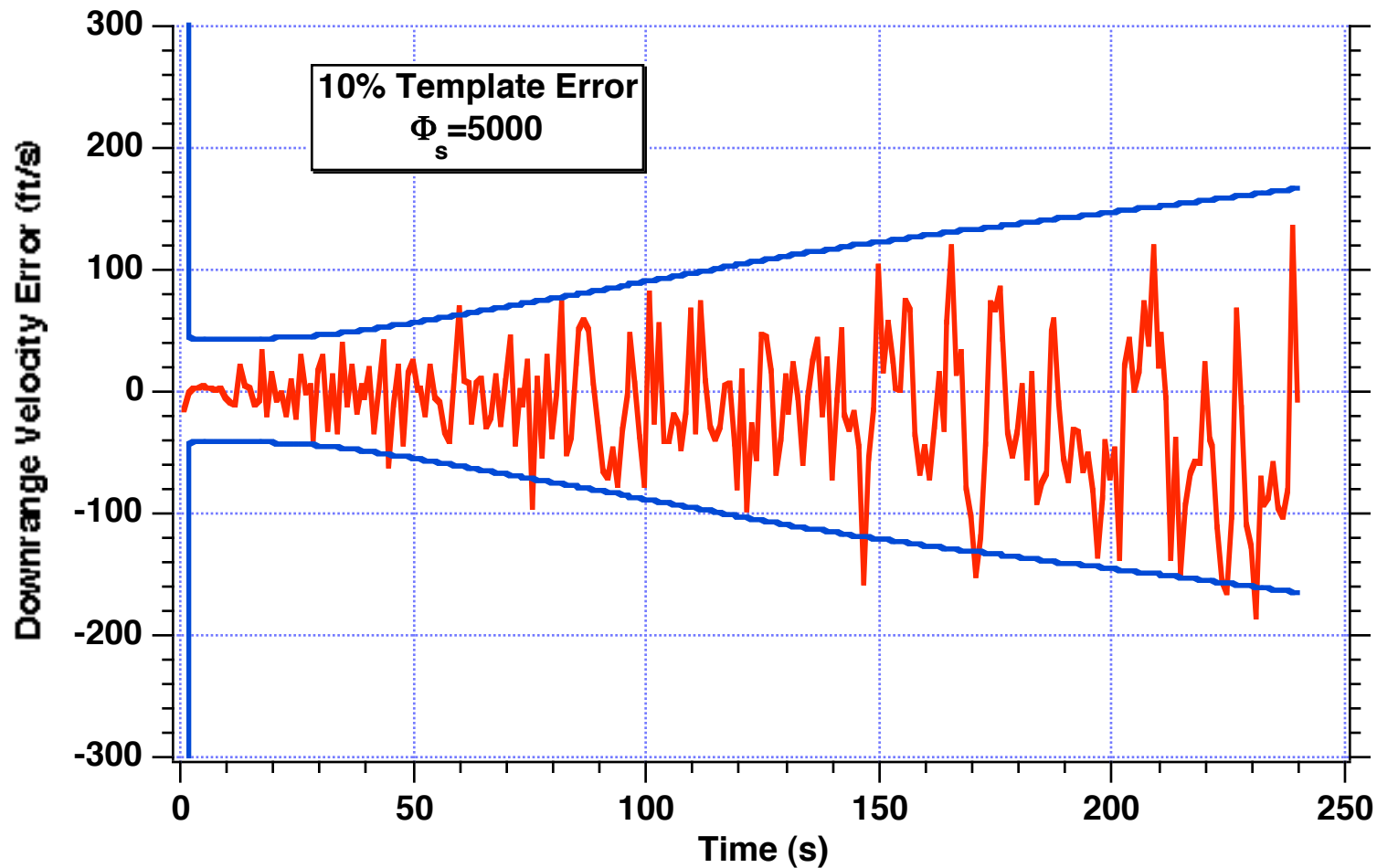
Don't forget we need to know target acceleration magnitude and direction for filter to work

# 10% Acceleration Errors Cause Two-State Template Based Kalman Filter to Diverge





## More Process Noise is Required by Two-State Kalman Filter to Eliminate Divergence When There is Acceleration Error



10% template error means that there is a 10% error in knowledge of ICBM acceleration magnitude

# Gravity Turn Assumption For Template Based Kalman Filter

Assume we know magnitude of acceleration perfectly **but not its direction**

Gravity turn assumption\*

$$a_{T_x} = a_T \frac{\dot{x}_T}{V_T}$$

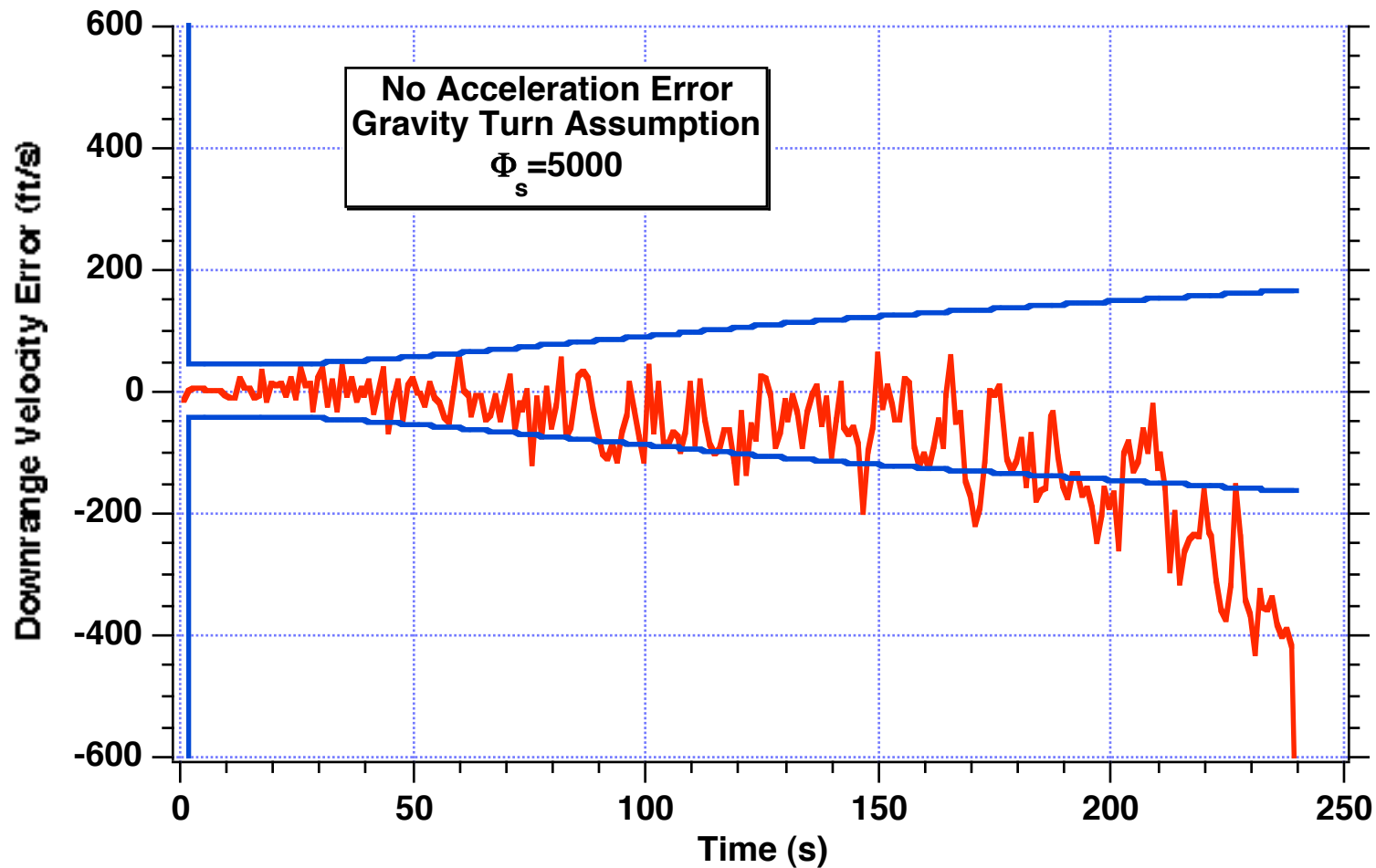
$$a_{T_y} = a_T \frac{\dot{y}_T}{V_T}$$

Where  $a_T$  is acceleration magnitude and  $V_T$  is total velocity or

$$V_T = \sqrt{\dot{x}_T^2 + \dot{y}_T^2}$$

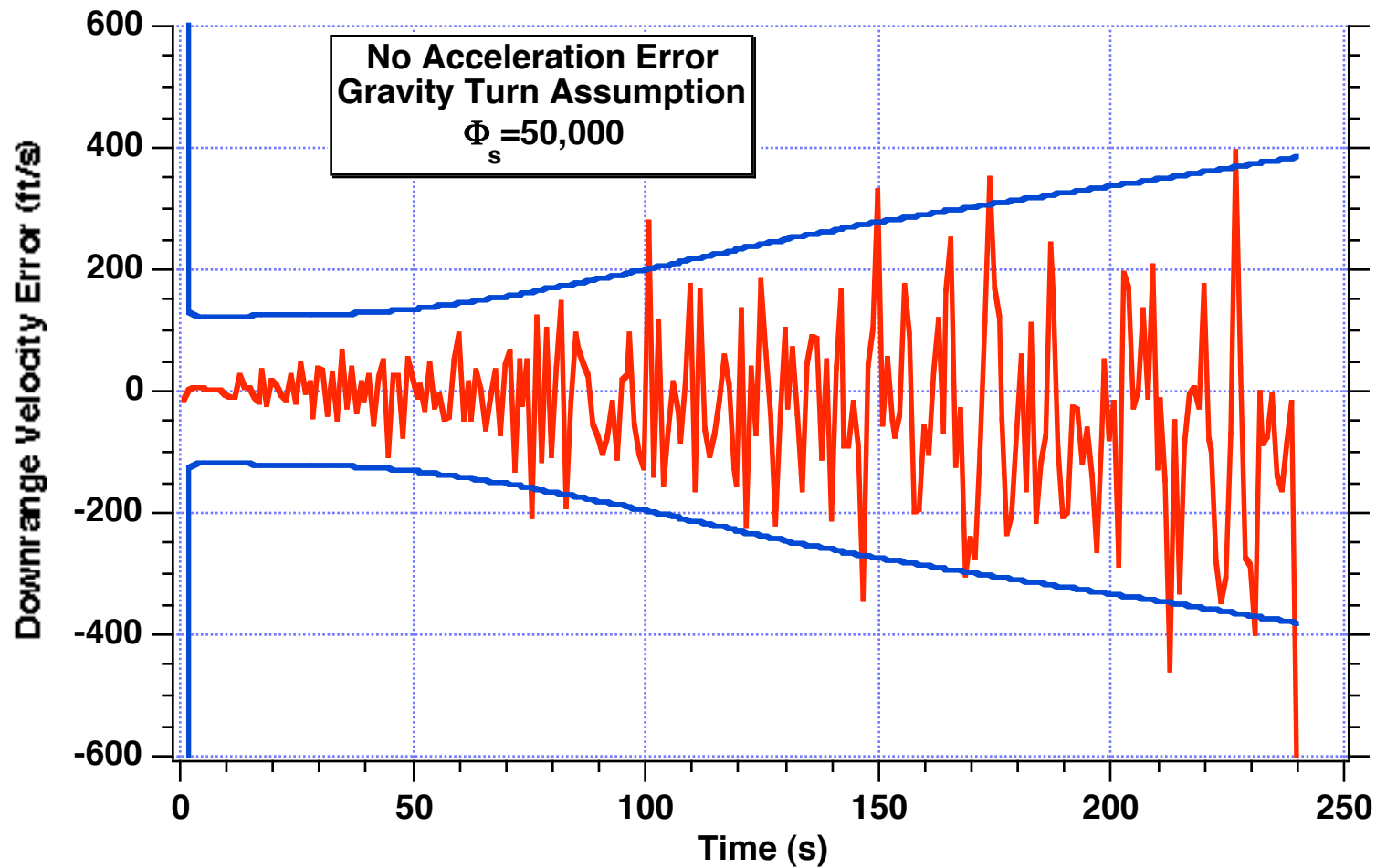
\*Recall ICBM is performing Lambert guidance rather than gravity turn in this example

## Gravity Turn Assumption Causes Two-State Kalman Filter Divergence - Even When Acceleration Magnitude is Known Perfectly



Gravity turn assumption causes divergence problems in template based Kalman filter because target is actually performing Lambert guidance

# More Process Noise is Required to Prevent Gravity Turn Two-State Kalman Filter Divergence - Even When ICBM Acceleration Magnitude is Known Perfectly



# 3-State Polynomial Kalman Filter

# 3-State Downrange and Altitude Decoupled Linear Polynomial Kalman Filters - **No Template Required**

## Downrange Filter

$$\text{Res}_{x_k} = x_{T_k}^* - \hat{x}_{T_{k-1}} - \hat{\dot{x}}_{T_{k-1}} T_s - 0.5 \hat{\ddot{x}}_{T_{k-1}} T_s^2$$

$$\hat{x}_{T_k} = \hat{x}_{T_{k-1}} + \hat{\dot{x}}_{T_{k-1}} T_s + 0.5 \hat{\ddot{x}}_{T_{k-1}} T_s^2 + K_{1x_k} \text{Res}_{x_k}$$

$$\hat{\dot{x}}_{T_k} = \hat{\dot{x}}_{T_{k-1}} + \hat{\ddot{x}}_{T_{k-1}} T_s + K_{2x_k} \text{Res}_{x_k}$$

$$\hat{\ddot{x}}_{T_k} = \hat{\ddot{x}}_{T_{k-1}} + K_{3x_k} \text{Res}_{x_k}$$

**Note that acceleration is estimated and **no a priori information is required****

## Altitude Filter

$$\text{Res}_{y_k} = y_{T_k}^* - \hat{y}_{T_{k-1}} - \hat{\dot{y}}_{T_{k-1}} T_s - 0.5 \hat{\ddot{y}}_{T_{k-1}} T_s^2$$

$$\hat{y}_{T_k} = \hat{y}_{T_{k-1}} + \hat{\dot{y}}_{T_{k-1}} T_s + 0.5 \hat{\ddot{y}}_{T_{k-1}} T_s^2 + K_{1y_k} \text{Res}_{y_k}$$

$$\hat{\dot{y}}_{T_k} = \hat{\dot{y}}_{T_{k-1}} + \hat{\ddot{y}}_{T_{k-1}} T_s + K_{2y_k} \text{Res}_{y_k}$$

$$\hat{\ddot{y}}_{T_k} = \hat{\ddot{y}}_{T_{k-1}} + K_{3y_k} \text{Res}_{y_k}$$

## Process Noise

$$Q = \Phi_s \begin{bmatrix} \frac{T_s^5}{20} & \frac{T_s^4}{8} & \frac{T_s^3}{6} \\ \frac{T_s^4}{8} & \frac{T_s^3}{3} & \frac{T_s^2}{2} \\ \frac{T_s^3}{6} & \frac{T_s^2}{2} & T_s \end{bmatrix}$$

**We will be adjusting  $\Phi_s$  to tune the filter**

# 3-State Decoupled Linear Polynomial Kalman Filters-1

```

GLOBAL DEFINE
    INCLUDE 'quickdraw.inc'
END
IMPLICIT REAL*8 (A-H)
IMPLICIT REAL*8 (O-Z)
REAL*8 M11,M12,M13,M22,M23,M33,K1,K2,K3
REAL*8 M11P,M12P,M13P,M22P,M23P,M33P,K1P,K2P,K3P
LOGICAL QBOOST
INTEGER STEP
OPEN(1,STATUS='UNKNOWN',FILE='DATFIL')
OPEN(2,STATUS='UNKNOWN',FILE='NOISEFIL')
OPEN(3,STATUS='UNKNOWN',FILE='COVFIL')
QBOOST=.TRUE.
TF=1300.
RT1F=5000*3280.
RT2F=0.
GAMTDEG=89.9
VT=1.
RT1=0.
RT2=0.
XR=1000.*3280.
YR=0.
VT1=VT*COS(GAMTDEG/57.3)
VT2=VT*SIN(GAMTDEG/57.3)
G=32.2
H=.01
TS=1.
S=0.
SCOUNT=0.
SIGR=10.
SIGTH=.001
PHIN=60.*G*G/240.
RT=SQRT((RT1-XR)**2+(RT2-YR)**2)
THET=ATAN2(RT2-YR,RT1-XR)
SIGRT1=SQRT((COS(THET)*SIGR)**2+(RT*SIN(THET)*SIGTH)**2)
SIGRT2=SQRT((SIN(THET)*SIGR)**2+(RT*COS(THET)*SIGTH)**2)
K1=0.
K2=0.
K3=0.
RT1H=0.
VT1H=0.
RT2H=0.
VT2H=0.
AT1H=0.
AT2H=0.
T=0.

```

**Desired Destination and Time of Arrival**

**Range and angle measurement errors**

**Filter process noise**

**Actual range and angle**

**Standard deviation of pseudo measurements**

**Filter initialization**

# 3-State Decoupled Linear Polynomial Kalman Filters-2

```
P11=9999999999.
P12=0.
P13=0.
P22=9999999999.
P23=0.
P33=9999999999.
P11P=9999999999.
P12P=0.
P13P=0.
P22P=9999999999.
P23P=0.
P33P=9999999999.
AX=0.
AY=0.
XN=0.
```

**Initial covariance matrix**

10 IF(T>240.)GOTO 999

```
RT1OLD=RT1
RT2OLD=RT2
VT1OLD=VT1
VT2OLD=VT2
```

STEP=1  
GOTO 200

66 STEP=2  
RT1=RT1+H\*VT1  
RT2=RT2+H\*VT2  
VT1=VT1+H\*AT1  
VT2=VT2+H\*AT2  
T=T+H

GOTO 200

55 RT1=(RT1OLD+RT1)/2+.5\*H\*VT1  
RT2=(RT2OLD+RT2)/2+.5\*H\*VT2  
VT1=(VT1OLD+VT1)/2+.5\*H\*AT1  
VT2=(VT2OLD+VT2)/2+.5\*H\*AT2  
IF(QBOOST)THEN

```
TGOLAM=TF-T
VXLAM=(RT1F-RT1)/TGOLAM
VYLAM=(RT2F-RT2+16.1*TGOLAM*TGOLAM)/TGOLAM
DELVX=VXLAM-VT1
DELVY=VYLAM-VT2
DELV=SQRT(DELVX**2+DELVY**2)
IF(TRST>0..AND.DELV>10.)THEN
```

```
AX=AT*DELVX/DELV
AY=AT*DELVY/DELV
```

```
ELSEIF(DELV<10.)THEN
TRST=0.
QBOOST=.FALSE.
AX=0.
AY=0.
VT1OLD=VXLAM
VT2OLD=VYLAM
```

**2nd order Runge-Kutta integration  
of ICBM for boost phase**

**Lambert guidance**



# 3-State Decoupled Linear Polynomial Kalman Filters-3

```

ELSE
    QBOOST=.FALSE.
    AX=0.
    AY=0.
    WRITE(9,*)DELV
    PAUSE
ENDIF
ENDIF
IF(T<20.)THEN
    AX=0.
    AY=AT
ENDIF
SCOUNT=SCOUNT+H
IF(SCOUNT.LT.(TS-.00001))GOTO 10
SCOUNT=0.
XN=XN+1.
XK1=3*(3*XN*XN-3*XN+2)/(XN*(XN+1)*(XN+2))
XK2=18*(2*XN-1)/(XN*(XN+1)*(XN+2)*TS)
XK3=60/(XN*(XN+1)*(XN+2)*TS*TS)
TS2=TS*TS
TS3=TS2*TS
TS4=TS3*TS
TS5=TS4*TS
RT=SQRT((RT1-XR)**2+(RT2-YR)**2)
THET=ATAN2(RT2-YR,RT1-XR)
SIGRT1=SQRT((COS(THET)*SIGR)**2+(RT*SIN(THET)*SIGTH)**2)
SIGRT2=SQRT((SIN(THET)*SIGR)**2+(RT*COS(THET)*SIGTH)**2)
SIGN2=SIGRT1*SIGRT1
M11=P11+TS*P12+.5*TS2*P13+TS*(P12+TS*P22+.5*TS2*P23)
M11=M11+.5*TS2*(P13+TS*P23+.5*TS2*P33)+TS5*PHIN/20.
M12=P12+TS*P22+.5*TS2*P23+TS*(P13+TS*P23+.5*TS2*P33)+TS4*PHIN/8.
M13=P13+TS*P23+.5*TS2*P33+PHIN*TS3/6.
M22=P22+TS*P23+TS*(P23+TS*P33)+PHIN*TS3/3.
M23=P23+TS*P33+.5*TS2*PHIN
M33=P33+PHIN*TS
BOT=M11+SIGN2
K1=M11/BOT
K2=M12/BOT
K3=M13/BOT
FACT=1.-K1
P11=FACT*M11
P12=FACT*M12
P13=FACT*M13
P22=-K2*M12+M22
P23=-K2*M13+M23
P33=-K3*M13+M33

```

Go straight up for  
first 20 s

Least squares filter gains

True angle and range

Pseudo measurement  
standard deviations

Ricatti equations for  
downrange filter

# 3-State Decoupled Linear Polynomial Kalman Filters-4

1

```

SIGN2P=SIGRT2*SIGRT2
M11P=P11P+TS*P12P+.5*TS2*P13P+TS*(P12P+TS*P22P+.5*TS2*P23P)
M11P=M11P+.5*TS2*(P13P+TS*P23P+.5*TS2*P33P)+TS5*PHIN/20.
M12P=P12P+TS*P22P+.5*TS2*P23P+TS*(P13P+TS*P23P+.5*TS2*P33P)
      +TS4*PHIN/8.
M13P=P13P+TS*P23P+.5*TS2*P33P+PHIN*TS3/6.
M22P=P22P+TS*P23P+TS*(P23P+TS*P33P)+PHIN*TS3/3.
M23P=P23P+TS*P33P+.5*TS2*PHIN
M33P=P33P+PHIN*TS
BOTP=M11P+SIGN2P
K1P=M11P/BOTP
K2P=M12P/BOTP
K3P=M13P/BOTP
FACTP=1.-K1P
P11P=FACTP*M11P
P12P=FACTP*M12P
P13P=FACTP*M13P
P22P=-K2P*M12P+M22P
P23P=-K2P*M13P+M23P
P33P=-K3P*M13P+M33P

```

**Ricatti equations for  
altitude filter**

```

CALL GAUSS(THETNOISE,SIGTH)
CALL GAUSS(RTNOISE,SIGR)
THETMEAS=THET+THETNOISE
RTMEAS=RT+RTNOISE
RT1S=RTMEAS*COS(THETMEAS)+XR
RT2S=RTMEAS*SIN(THETMEAS)+YR
IF(XN<10.)THEN

```

**Actual angle and range measurements**

**Pseudo measurements**

```

      XK1PZ=XK1
      XK2PZ=XK2
      XK3PZ=XK3
ELSE
      XK1PZ=K1
      XK2PZ=K2
      XK3PZ=K3

```

**Use least squares gains for  
first 10 measurements**

```

ENDIF
RESX=RT1S-RT1H-TS*VT1H-.5*TS2*AT1H
RT1H=XK1PZ*RESX+RT1H+TS*VT1H+.5*TS2*AT1H
VT1H=XK2PZ*RESX+VT1H+TS*AT1H
AT1H=XK3PZ*RESX+AT1H
IF(XN<10.)THEN

```

**Downrange Kalman filter**

```

      XK1PPZ=XK1
      XK2PPZ=XK2
      XK3PPZ=XK3
ELSE
      XK1PPZ=K1P
      XK2PPZ=K2P
      XK3PPZ=K3P

```

**Use least squares gains for  
first 10 measurements**

```

ENDIF

```

# 3-State Decoupled Linear Polynomial Kalman Filters-5

```

IF(XN<10.)THEN
    XK1PPZ=XK1
    XK2PPZ=XK2
    XK3PPZ=XK3
ELSE
    XK1PPZ=K1P
    XK2PPZ=K2P
    XK3PPZ=K3P
ENDIF
RESY=RT2S-RT2H-TS*VT2H-.5*TS2*AT2H
RT2H=XK1PPZ*RESY+RT2H+TS*VT2H+.5*TS2*AT2H
VT2H=XK2PPZ*RESY+VT2H+TS*AT2H
AT2H=XK3PPZ*RESY+AT2H
SP33=SQRT(P33)/32.2
ERRAT1=(AT1-AT1H)/32.2
SP33P=SQRT(P33P)/32.2
ERRAT2=(AT2-AT2H)/32.2
SP33=SQRT(P33)/32.2
ERRAT1=(AT1-AT1H)/32.2
SP33P=SQRT(P33P)/32.2
ERRAT2=(AT2-AT2H)/32.2
RT1KM=RT1/3280.
RT2KM=RT2/3280.
RT1NOISE=RT1-RT1S
RT2NOISE=RT2-RT2S
ERRRT1=RT1-RT1H
ERRVT1=VT1-VT1H
SP11=SQRT(P11)
SP22=SQRT(P22)
ERRRT2=RT2-RT2H
ERRVT2=VT2-VT2H
SP11P=SQRT(P11P)
SP22P=SQRT(P22P)
WRITE(9,*)T,RT1KM,RT2KM,AT1/G,AT1H/G,AT2/G,AT2H/G
WRITE(1,*)T,RT1KM,RT2KM,AT1/G,AT1H/G,AT2/G,AT2H/G
WRITE(2,*)T,RT1NOISE,SIGRT1,-SIGRT1,RT2NOISE,SIGRT2,-SIGRT2
WRITE(3,*)T,ERRRT1,SP11,-SP11,ERRVT1,SP22,-SP22,ERRAT1,SP33,
1      -SP33,ERRRT2,SP11P,-SP11P,ERRVT2,SP22P,-SP22P,ERRAT2,
2      SP33P,-SP33P
GOTO 10
200 CONTINUE

```

Use least squares gains for  
first 10 measurements

Altitude Kalman filter

Actual and theoretical  
errors in estimates

# 3-State Decoupled Linear Polynomial Kalman Filters-6

```
IF(T<120.)THEN
    WGT=-2622*T+440660.
    TRST=725850.
ELSEIF(T<240.)THEN
    WGT=-642.*T+168120.
    TRST=182250.
ELSE
    WGT=5500.
    TRST=0.
ENDIF
VT=SQRT(VT1**2+VT2**2)
GAM=ATAN2(VT2,VT1)
AT=G*TRST/WGT
AT1=AX
AT2=-G+AY
IF(STEP-1)66,66,55
CONTINUE
RT1KM=RT1/3280.
RT2KM=RT2/3280.
WRITE(9,*)T,RT1KM,RT2KM,AT1/G,AT1H/G,AT2/G,AT2H/G
WRITE(1,*)T,RT1KM,RT2KM,AT1/G,AT1H/G,AT2/G,AT2H/G
PAUSE
CLOSE(1)
CLOSE(2)
CLOSE(3)
END
```

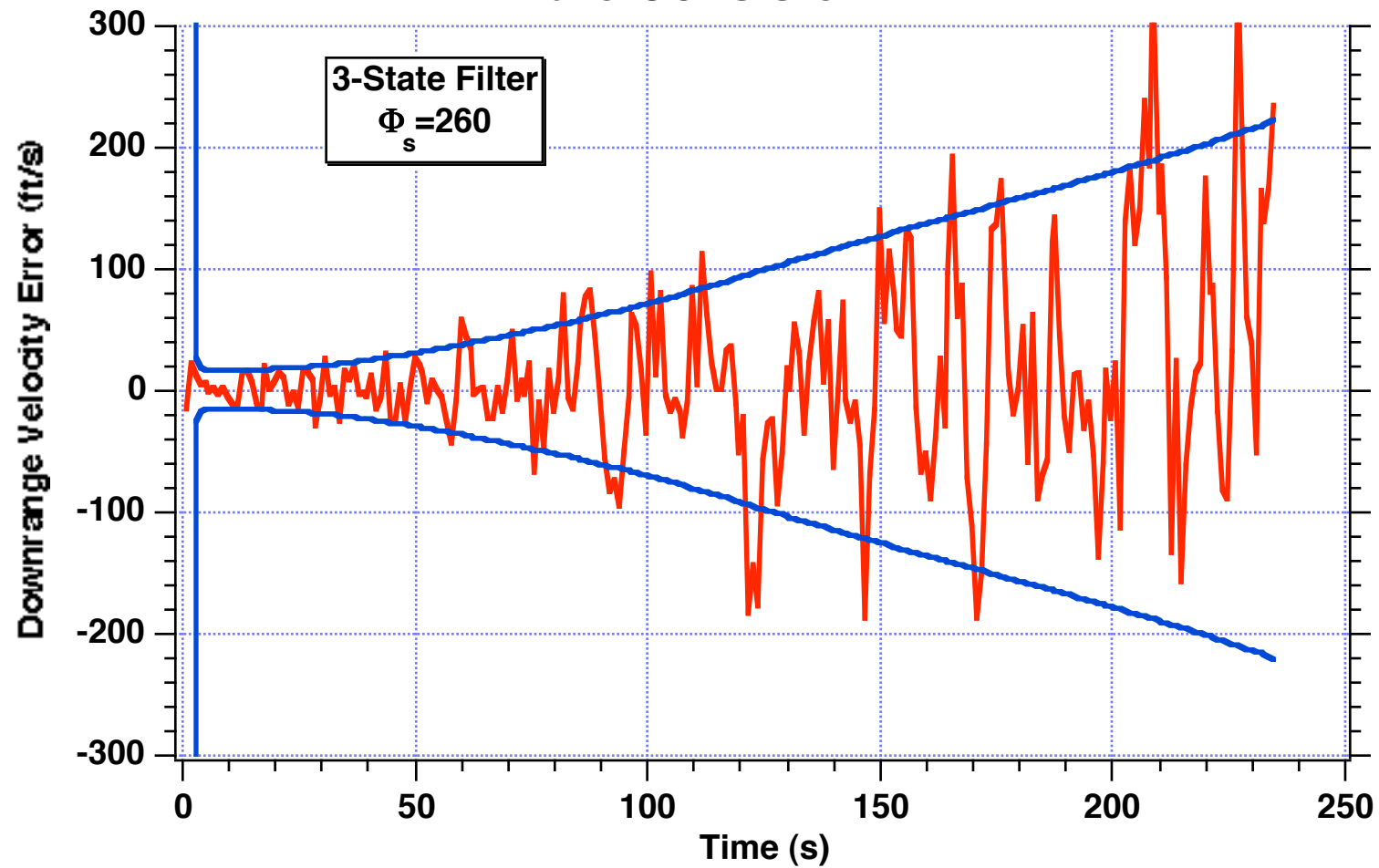
ICBM

Differential equations

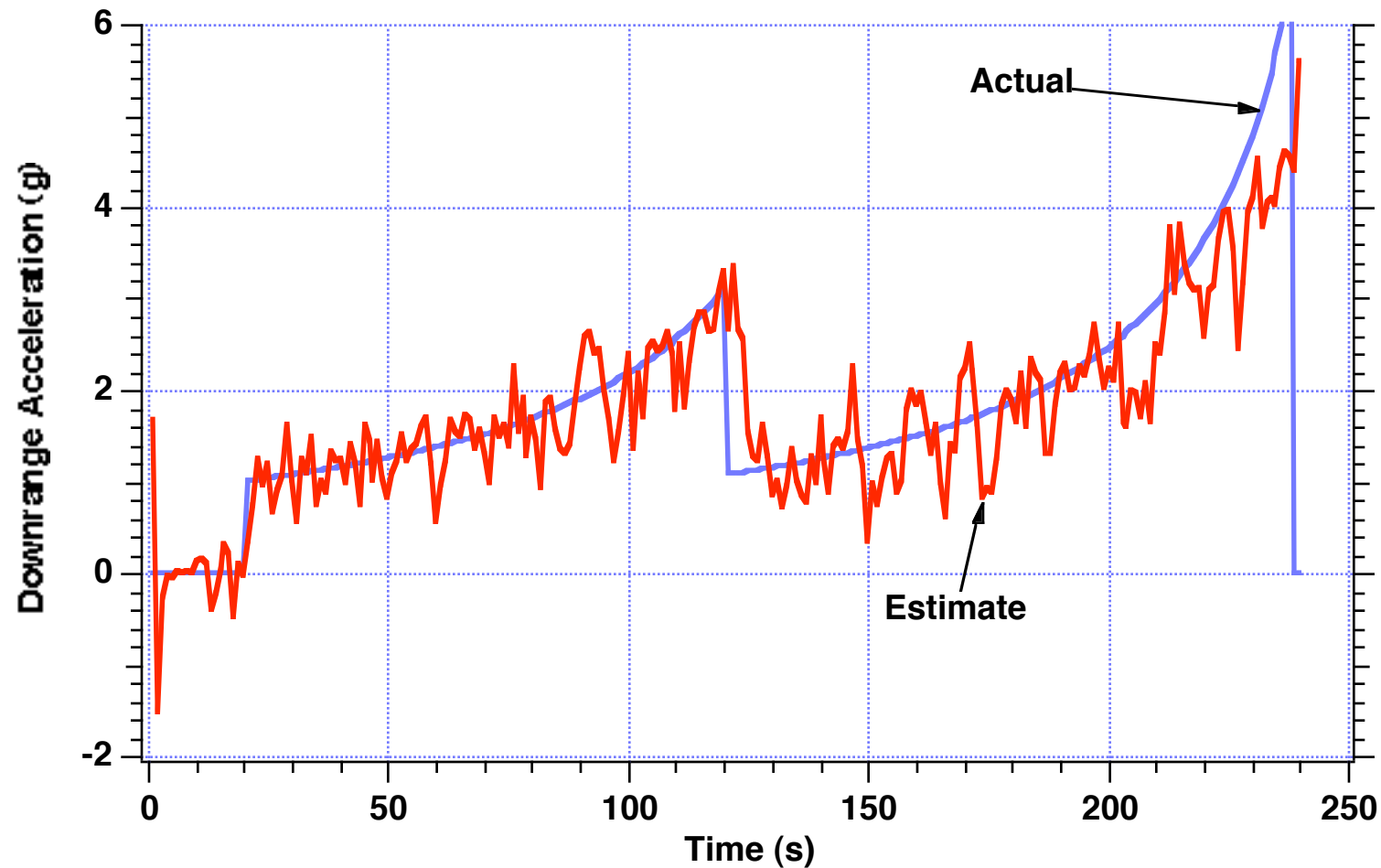
Acceleration

999

## Three-State Kalman Filter is Tuned so That Errors in Estimates are Consistent

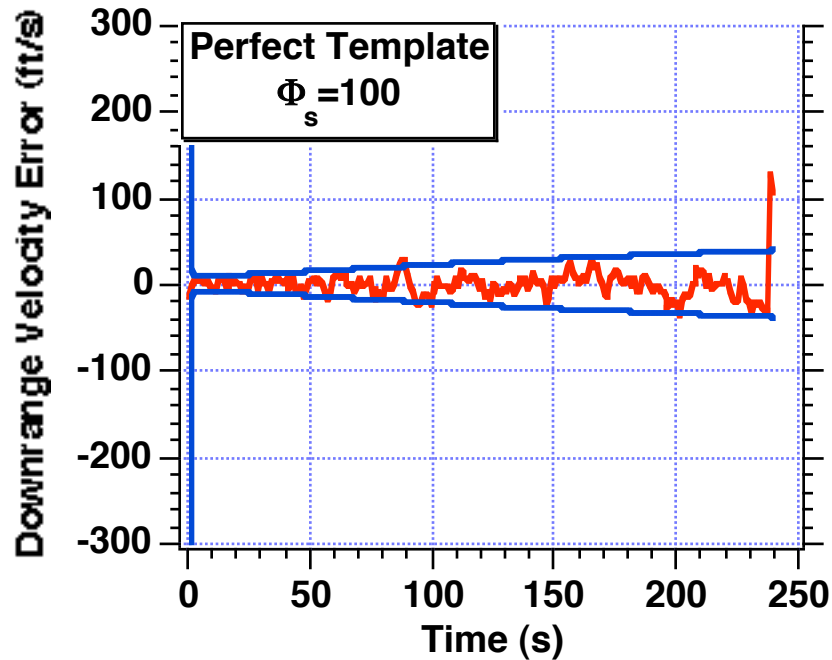


# Nominal Three-State Linear Polynomial Kalman Filter Tracks ICBM Acceleration Quite Well

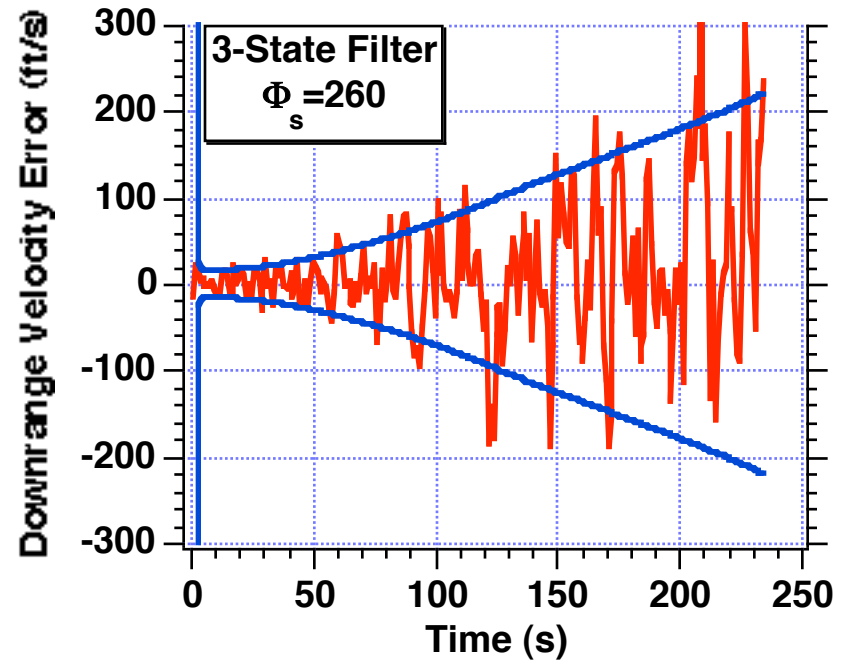


# Filter Comparison - 1

## 2-State Filter



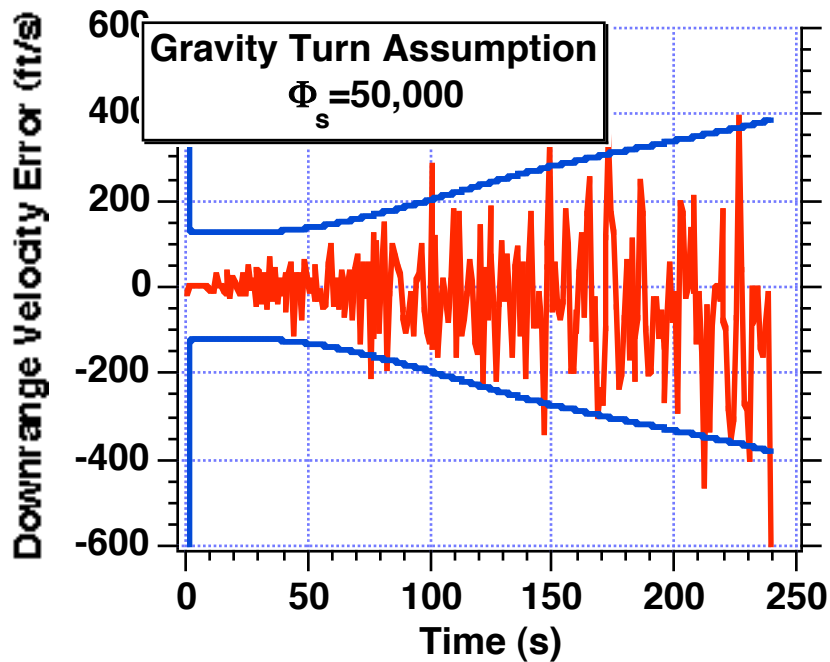
## 3-State Filter



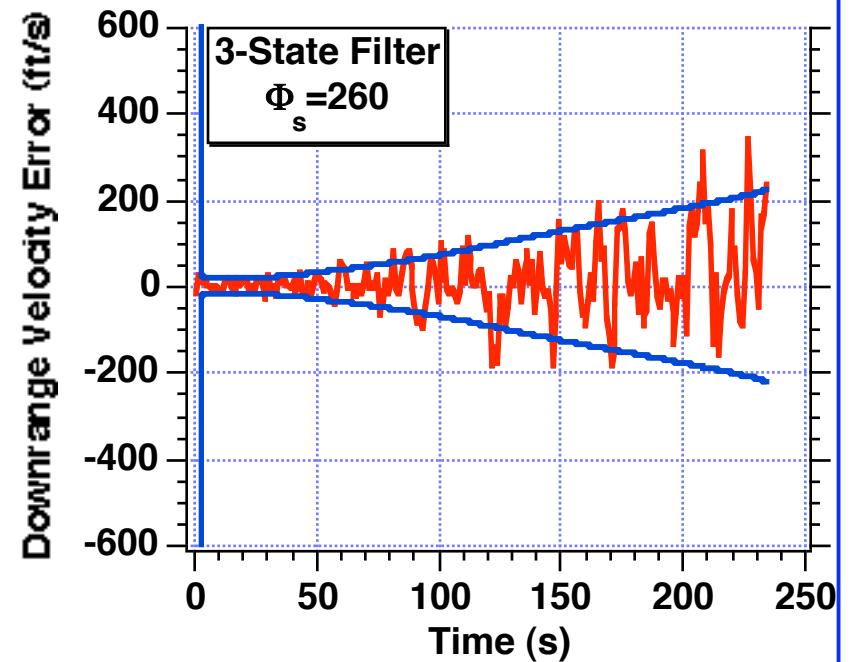
2-State filter yields better results than 3-State filter if it has a perfect knowledge of the target acceleration magnitude and direction

## Filter Comparison - 2

### 2-State Filter



### 3-State Filter



3-State filter yields better results than 2-State filter when realistic errors are considered