

# Miscellaneous Topics on Measurements

# Miscellaneous Topics on Measurements Overview

- **Using additional measurements**
- **Asynchronous data rate**
- **Batch processing**

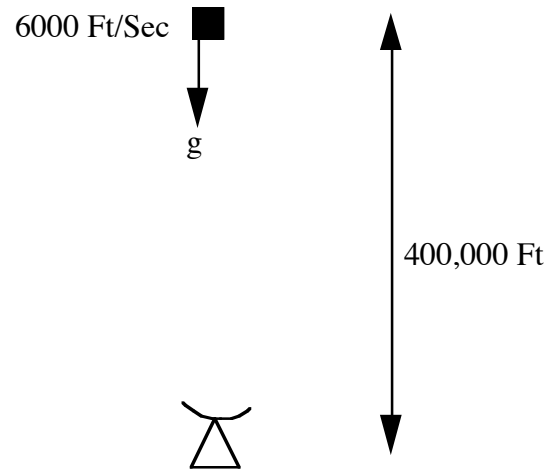
# Using Additional Measurements

# Using Additional Measurements Overview

- **Review of one measurement problem for tracing a falling object**
  - **Polynomial Kalman filter design with gravity compensation**
- **Incorporating a second measurement**
  - **New Kalman filter design**
  - **How to build filter when second measurement is at a different data rate**

# **Review of One Measurement Problem For Tracking a Falling Object**

# Radar Tracking Falling Object



**From basic physics**

$$x = 400000 - 6000t - \frac{gt^2}{2} \longleftarrow \text{Second-order process}$$

**Velocity of object can be found by differentiating**

$$\dot{x} = -6000 - gt$$

**Radar measures altitude with standard deviation of 1000 ft**

**Desire to track object and estimate altitude and velocity**

# Filter Design Making Use of A Priori Information - 1

## Model of the real world

$$\dot{\mathbf{x}} = \mathbf{F}\mathbf{x} + \mathbf{G}\mathbf{u} + \mathbf{w}$$

## Kalman filter

$$\hat{\mathbf{x}}_k = \Phi_k \hat{\mathbf{x}}_{k-1} + \mathbf{G}_k \mathbf{u}_{k-1} + \mathbf{K}_k (z_k - \mathbf{H} \Phi_k \hat{\mathbf{x}}_{k-1} - \mathbf{H} \mathbf{G}_k \mathbf{u}_{k-1})$$

## Where

$$\mathbf{G}_k = \int_0^{T_s} \Phi(\tau) \mathbf{G} d\tau$$

In our problem we know the model and measurement equation

$$\ddot{x} = -g \quad x^* = x + v$$

Expressing gravitational information in state space form

$$\begin{bmatrix} \dot{x} \\ \dot{\ddot{x}} \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} x \\ \dot{x} \end{bmatrix} + \begin{bmatrix} 0 \\ -1 \end{bmatrix} g \quad x^* = \begin{bmatrix} 1 & 0 \end{bmatrix} \begin{bmatrix} x \\ \dot{x} \end{bmatrix} + v$$

By inspection

$$\mathbf{F} = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \longrightarrow \Phi_k = \begin{bmatrix} 1 & T_s \\ 0 & 1 \end{bmatrix} \quad \mathbf{H} = \begin{bmatrix} 1 & 0 \end{bmatrix}$$

## Filter Design Making Use of A Priori Information - 2

Recall

$$\dot{\mathbf{x}} = \mathbf{F}\mathbf{x} + \mathbf{G}\mathbf{u} + \mathbf{w} \longrightarrow \begin{bmatrix} \dot{x} \\ \ddot{x} \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} x \\ \dot{x} \end{bmatrix} + \begin{bmatrix} 0 \\ -1 \end{bmatrix} g$$

We can also see that

$$\mathbf{G} = \begin{bmatrix} 0 \\ -1 \end{bmatrix} \quad \mathbf{u} = g$$

Therefore

$$\mathbf{G}_k = \int_0^{T_s} \Phi(\tau) \mathbf{G} \, d\tau = \int_0^{T_s} \begin{bmatrix} 1 & \tau \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 0 \\ -1 \end{bmatrix} d\tau = \begin{bmatrix} -\frac{T_s^2}{2} \\ -T_s \end{bmatrix}$$

Since formula of Kalman filter is given by

$$\hat{\mathbf{x}}_k = \Phi_k \hat{\mathbf{x}}_{k-1} + \mathbf{G}_k \mathbf{u}_{k-1} + \mathbf{K}_k (z_k - \mathbf{H} \Phi_k \hat{\mathbf{x}}_{k-1} - \mathbf{H} \mathbf{G}_k \mathbf{u}_{k-1})$$

Substitution yields

$$\begin{bmatrix} \hat{x}_k \\ \hat{\dot{x}}_k \end{bmatrix} = \begin{bmatrix} 1 & T_s \\ 0 & 1 \end{bmatrix} \begin{bmatrix} \hat{x}_{k-1} \\ \hat{\dot{x}}_{k-1} \end{bmatrix} + \begin{bmatrix} -.5T_s^2 \\ -T_s \end{bmatrix} g + \begin{bmatrix} \mathbf{K}_{1k} \\ \mathbf{K}_{2k} \end{bmatrix} \left[ x_k^* - \begin{bmatrix} 1 & 0 \end{bmatrix} \begin{bmatrix} 1 & T_s \\ 0 & 1 \end{bmatrix} \begin{bmatrix} \hat{x}_{k-1} \\ \hat{\dot{x}}_{k-1} \end{bmatrix} - \begin{bmatrix} 1 & 0 \end{bmatrix} \begin{bmatrix} -.5T_s^2 \\ -T_s \end{bmatrix} g \right]$$



## Filter Design Making Use of A Priori Information - 3

### Multiplying out the terms

$$\hat{x}_k = \hat{x}_{k-1} + \hat{\dot{x}}_{k-1}T_s - .5gT_s^2 + K_{1k}(x_k^* - \hat{x}_{k-1} - \hat{\dot{x}}_{k-1}T_s + .5gT_s^2)$$

$$\hat{\dot{x}}_k = \hat{\dot{x}}_{k-1} - gT_s + K_{2k}(x_k^* - \hat{x}_{k-1} - \hat{\dot{x}}_{k-1}T_s + .5gT_s^2)$$

### If we define the residual as

$$RES_k = x_k^* - \hat{x}_{k-1} - \hat{\dot{x}}_{k-1}T_s + .5gT_s^2$$

### The Kalman filter simplifies to

$$\hat{x}_k = \hat{x}_{k-1} + \hat{\dot{x}}_{k-1}T_s - .5gT_s^2 + K_{1k}RES_k$$

$$\hat{\dot{x}}_k = \hat{\dot{x}}_{k-1} - gT_s + K_{2k}RES_k$$

# Riccati Equations

## Riccati equations with process noise

$$\mathbf{M}_k = \Phi_k \mathbf{P}_{k-1} \Phi_k^T + \mathbf{Q}_k$$

$$\mathbf{H} = [1 \quad 0]$$

$$\Phi_k = \begin{bmatrix} 1 & T_s \\ 0 & 1 \end{bmatrix}$$

$$\mathbf{K}_k = \mathbf{M}_k \mathbf{H}^T (\mathbf{H} \mathbf{M}_k \mathbf{H}^T + \mathbf{R}_k)^{-1}$$

$$\mathbf{P}_k = (\mathbf{I} - \mathbf{K}_k \mathbf{H}) \mathbf{M}_k$$

## Ramp Signal

$$\begin{bmatrix} \dot{x} \\ \ddot{x} \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} x \\ \dot{x} \end{bmatrix} + \begin{bmatrix} 0 \\ u_s \end{bmatrix} \longrightarrow \mathbf{Q} = \mathbf{E} \left[ \begin{bmatrix} 0 \\ u_s \end{bmatrix} \begin{bmatrix} 0 & u_s \end{bmatrix} \right] = \begin{bmatrix} 0 & 0 \\ 0 & \Phi_s \end{bmatrix}$$

## Measurement Noise Matrix is Scalar

$$\mathbf{R}_k = \sigma_p^2$$

## Deriving discrete process noise matrix

$$\mathbf{Q}_k = \int_0^{T_s} \Phi(\tau) \mathbf{Q} \Phi^T(\tau) dt$$

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

# MATLAB Version of First-Order Polynomial Kalman Filter with Gravity Compensation-1

```
TS=1;  
PHIS=0.;  
A0=400000.;  
A1=-6000.;  
A2=-16.1;  
XH=0.;  
XDH=0.;  
SIGNOISE=1000.;  
ORDER=2;  
T=0.;  
S=0.;  
H=.001;  
PHI=[1 TS ;0 1];  
P=[99999999 0;0 99999999];  
IDNP=eye(ORDER);  
Q=zeros(ORDER);  
RMAT=SIGNOISE^2;  
Q(1,1)=TS*TS*TS*PHIS/3.;  
Q(1,2)=.5*TS*TS*PHIS;  
Q(2,1)=Q(1,2);  
Q(2,2)=PHIS*TS;  
HMAT=[1 0];  
HT=HMAT';  
PHIT=PHI';  
count=0;  
for T=0:TS:30
```

**Fundamental and initial covariance matrices**

**Process noise and measurement matrices**

**Riccati equations**

```
PHIP=PHI*P;  
PHIPPHIT=PHIP*PHIT;  
M=PHIPPHIT+Q;  
HM=HMAT*M;  
HMHT=HM*HT;  
HMHTR=HMHT+RMAT;  
HMHTRINV=inv(HMHTR);  
MHT=M*HT;  
K=MHT*HMHTRINV;  
KH=K*HMAT;  
IKH=IDNP-KH;  
P=IKH*M;
```

# MATLAB Version of First-Order Polynomial Kalman Filter with Gravity Compensation-2

```
XNOISE=SIGNOISE*randn;
```

```
X=A0+A1*T+A2*T*T;
```

```
XD=A1+2*A2*T;
```

```
XS=X+XNOISE;
```

```
RES=XS-XH-TS*XDH+16.1*TS*TS;
```

```
XH=XH+XDH*TS-16.1*TS*TS+K(1,1)*RES;
```

```
XDH=XDH-32.2*TS+K(2,1)*RES;
```

```
SP11=sqrt(P(1,1));
```

```
SP22=sqrt(P(2,2));
```

```
XHERR=X-XH;
```

```
XDHERR=XD-XDH;
```

```
SP11P=-SP11;
```

```
SP22P=-SP22;
```

```
count=count+1;
```

```
ArrayT(count)=T;
```

```
ArrayX(count)=X;
```

```
ArrayXH(count)=XH;
```

```
ArrayXD(count)=XD;
```

```
ArrayXDH(count)=XDH;
```

```
ArrayXHERR(count)=XHERR;
```

```
ArraySP11(count)=SP11;
```

```
ArraySP11P(count)=SP11P;
```

```
ArrayXDHERR(count)=XDHERR;
```

```
ArraySP22(count)=SP22;
```

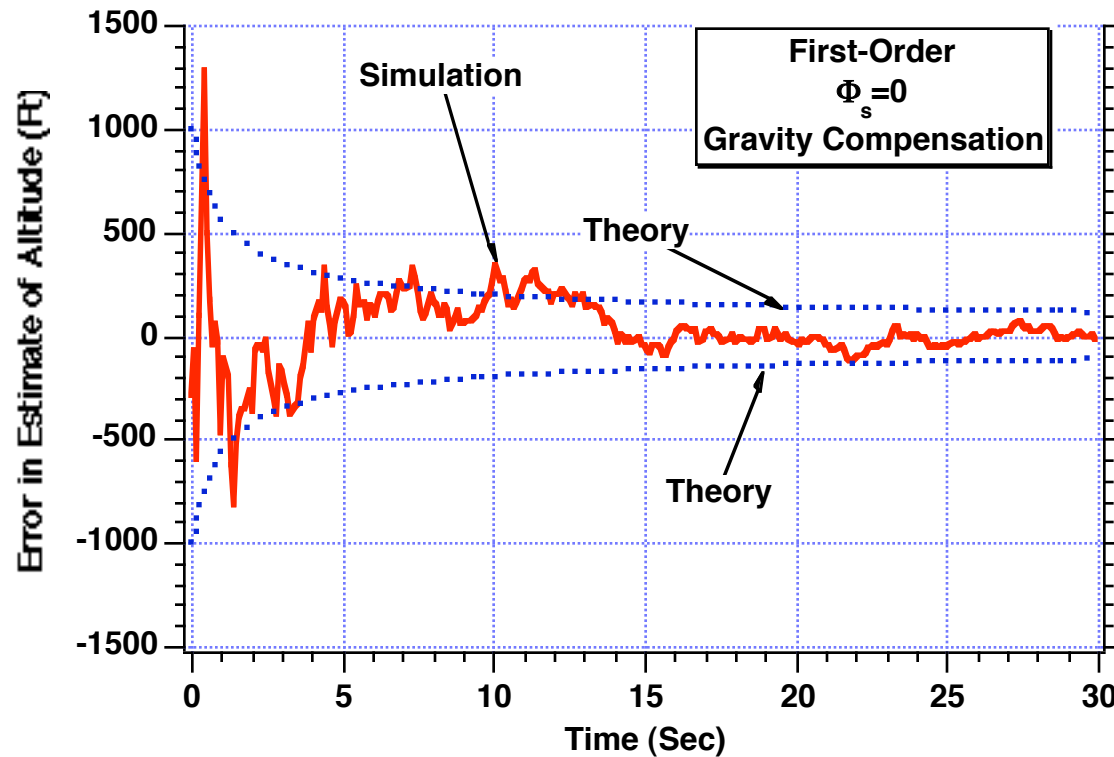
```
ArraySP22P(count)=SP22P;
```

```
end
```

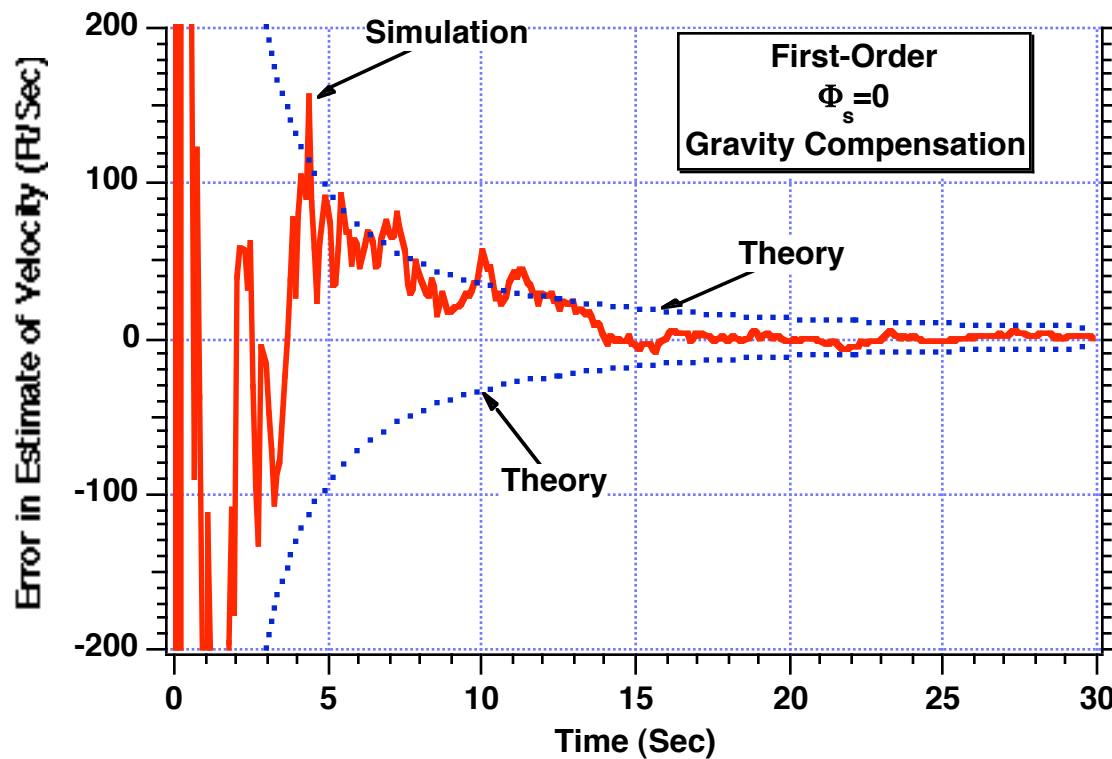
Filter has gravity compensation



## Typical Altitude Error Results For Case in Which There Are Only Position Measurements

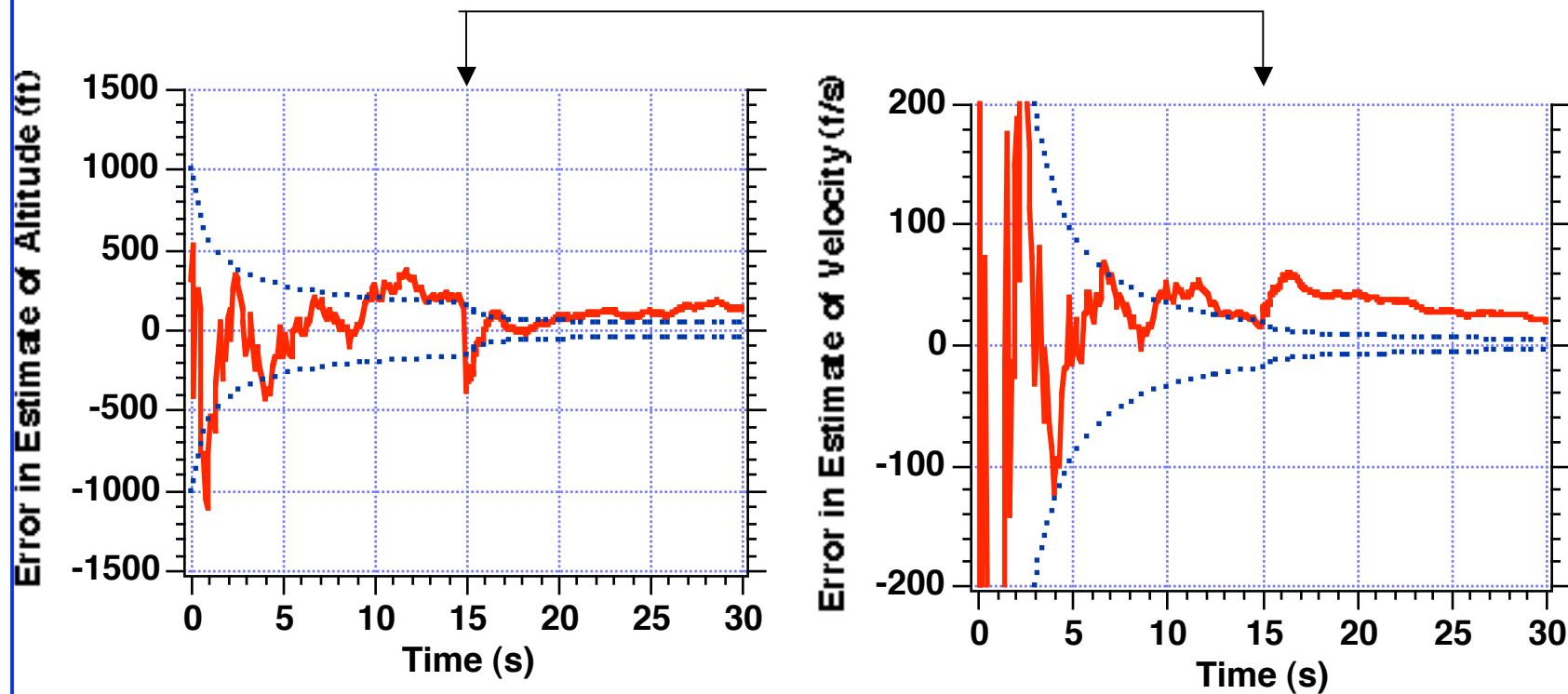


# Typical Velocity Error Results For Case in Which There Are Only Position Measurements



# Going To Higher Data Rate During Flight Caused Transient Problem

Data rate changes



$T_s = 0.1$  s for first 15 s and then  $T_s = 0.01$  for last 15 s

# Multiple Data Rate Simulation-1

```
GLOBAL DEFINE
      INCLUDE 'quickdraw.inc'
END
IMPLICIT REAL*8(A-H,O-Z)
REAL*8 P(2,2),Q(2,2),M(2,2),PHI(2,2),HMAT(1,2),HT(2,1),PHIT(2,2)
REAL*8 RMAT(1,1),IDN(2,2),PHIP(2,2),PHIPPHIT(2,2),HM(1,2)
REAL*8 HMHT(1,1),HMHTR(1,1),HMHTRINV(1,1),MHT(2,1),K(2,1)
REAL*8 KH(2,2),IKH(2,2)
INTEGER ORDER
TS=.1
PHIS=0.
A0=400000.
A1=-6000.
A2=-16.1
XH=0.
XDH=0.
SIGNOISE=1000.
ORDER=2
OPEN(1,STATUS='UNKNOWN',FILE='DATFIL')
OPEN(2,STATUS='UNKNOWN',FILE='COVFIL')
T=0.
S=0.
H=.001
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
```



## Multiple Data Rate Simulation-2

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

```
IDN(1,1)=1.
```

```
IDN(2,2)=1.
```

```
P(1,1)=999999999999.
```

```
P(2,2)=999999999999.
```

```
HMAT(1,1)=1.
```

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

```
T=0.
```

```
IF(T>30.)GOTO 999
```

```
IF(T<15.)THEN
```

```
    TS=.1
```

```
    TSOLD=TS
```

```
ELSE
```

```
    TS=.01
```

```
ENDIF
```

```
PHI(1,1)=1.
```

```
PHI(1,2)=TSOLD
```

```
PHI(2,2)=1.
```

```
Q(1,1)=TSOLD*TSOLD*TSOLD*PHIS/3.
```

```
Q(1,2)=.5*TSOLD*TSOLD*PHIS
```

```
Q(2,1)=Q(1,2)
```

```
Q(2,2)=PHIS*TSOLD
```

```
CALL MATTRN(PHI,ORDER,ORDER,PHIT)
```

```
CALL MATTRN(HMAT,1,ORDER,HT)
```

```
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)
```

10

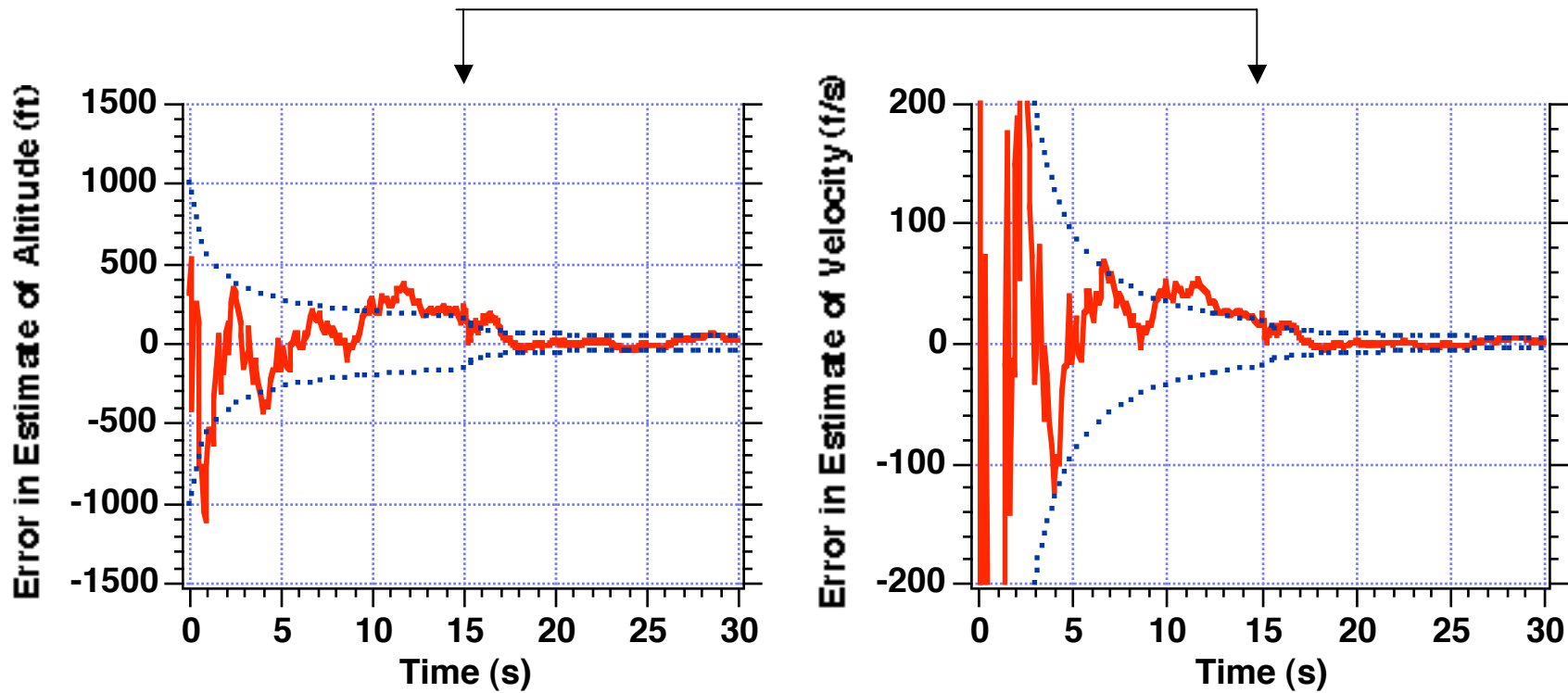
## Multiple Data Rate Simulation-3

```
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(XNOISE,SIGNOISE)
X=A0+A1*T+A2*T*T
XD=A1+2*A2*T
XS=X+XNOISE
RES=XS-XH-TSOLD*XDH+16.1*TSOLD*TSOLD
XH=XH+XDH*TSOLD-16.1*TSOLD*TSOLD+K(1,1)*RES
XDH=XDH-32.2*TSOLD+K(2,1)*RES
SP11=SQRT(P(1,1))
SP22=SQRT(P(2,2))
XHERR=X-XH
XDHERR=XD-XDH
WRITE(9,*)T,XD,XDH,K(1,1),K(2,1)
WRITE(1,*)T,X,XH,XD,XDH
WRITE(2,*)T,XHERR,SP11,-SP11,XDHERR,SP22,-SP22
T=T+TS
TSOLD=TS
GOTO 10
CONTINUE
PAUSE
CLOSE(1)
END
```

999

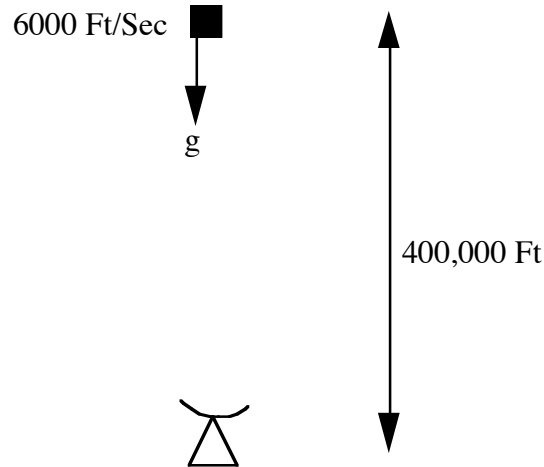
# Adjusting Fundamental Matrix Allows Us To Change Data Rates Without Transient Effects

Data rate changes



# Incorporating a Second Measurement

# Radar Tracking Falling Object



**From basic physics**

$$x = 400000 - 6000t - \frac{gt^2}{2} \quad \leftarrow \text{Second-order process}$$

**Velocity of object can be found by differentiating**

$$\dot{x} = -6000 - gt$$

**Radar measures altitude with standard deviation of 1000 ft**

**Radar measures velocity with standard deviation of 10 ft/s**

**Desire to track object and estimate altitude and velocity**

# Filter Design With Two Measurements - 1

## Model of the real world

$$\dot{\mathbf{x}} = \mathbf{F}\mathbf{x} + \mathbf{G}\mathbf{u} + \mathbf{w}$$

## Kalman filter

$$\hat{\mathbf{x}}_k = \Phi_k \hat{\mathbf{x}}_{k-1} + \mathbf{G}_k \mathbf{u}_{k-1} + \mathbf{K}_k (\mathbf{z}_k - \mathbf{H} \Phi_k \hat{\mathbf{x}}_{k-1} - \mathbf{H} \mathbf{G}_k \mathbf{u}_{k-1})$$

## Where

$$\mathbf{G}_k = \int_0^{T_s} \Phi(\tau) \mathbf{G} d\tau$$

In our problem we know the model and measurement equation

$$\begin{aligned} \ddot{x} &= -g & x^* &= x + v_1 \\ \dot{x}^* &= \dot{x} + v_2 \end{aligned}$$

Expressing gravitational information in state space form

$$\begin{bmatrix} \dot{x} \\ \dot{x}^* \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} x \\ \dot{x} \end{bmatrix} + \begin{bmatrix} 0 \\ -1 \end{bmatrix} g \quad \begin{bmatrix} x^* \\ \dot{x}^* \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ \dot{x} \end{bmatrix} + \begin{bmatrix} v_1 \\ v_2 \end{bmatrix}$$

By inspection

$$\mathbf{F} = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \longrightarrow \Phi_k = \begin{bmatrix} 1 & T_s \\ 0 & 1 \end{bmatrix}$$

$$\mathbf{H} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

## Filter Design With Two Measurements - 2

### Recall

$$\dot{\mathbf{x}} = \mathbf{F}\mathbf{x} + \mathbf{G}\mathbf{u} + \mathbf{w} \longrightarrow \begin{bmatrix} \dot{x} \\ \ddot{x} \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} x \\ \dot{x} \end{bmatrix} + \begin{bmatrix} 0 \\ -1 \end{bmatrix} g$$

We can also see that

$$\mathbf{G} = \begin{bmatrix} 0 \\ -1 \end{bmatrix} \quad \mathbf{u} = g$$

Therefore

$$\mathbf{G}_k = \int_0^{T_s} \Phi(\tau) \mathbf{G} \, d\tau = \int_0^{T_s} \begin{bmatrix} 1 & \tau \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 0 \\ -1 \end{bmatrix} \, d\tau = \begin{bmatrix} \frac{-T_s^2}{2} \\ -T_s \end{bmatrix}$$

Since formula of Kalman filter is given by

$$\hat{\mathbf{x}}_k = \Phi_k \hat{\mathbf{x}}_{k-1} + \mathbf{G}_k \mathbf{u}_{k-1} + \mathbf{K}_k (\mathbf{z}_k - \mathbf{H} \Phi_k \hat{\mathbf{x}}_{k-1} - \mathbf{H} \mathbf{G}_k \mathbf{u}_{k-1})$$

Substitution yields

$$\begin{bmatrix} \hat{x}_k \\ \hat{\dot{x}}_k \end{bmatrix} = \begin{bmatrix} 1 & T_s \\ 0 & 1 \end{bmatrix} \begin{bmatrix} \hat{x}_{k-1} \\ \hat{\dot{x}}_{k-1} \end{bmatrix} + \begin{bmatrix} -.5T_s^2 \\ -T_s \end{bmatrix} g + \begin{bmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{bmatrix} \left\{ \begin{bmatrix} x^* \\ \dot{x}^* \end{bmatrix} - \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & T_s \\ 0 & 1 \end{bmatrix} \begin{bmatrix} \hat{x}_{k-1} \\ \hat{\dot{x}}_{k-1} \end{bmatrix} - \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} -.5T_s^2 \\ -T_s \end{bmatrix} g \right\}$$

## Filter Design With Two Measurements - 3

### Multiplying out the terms

$$\hat{x}_k = \hat{x}_{k-1} + T_s \hat{\dot{x}}_{k-1} - .5gT_s^2 + K_{11}(x_k^* - \hat{x}_{k-1} - T_s \hat{\dot{x}}_{k-1} + .5gT_s^2) + K_{12}(\dot{x}_k^* - \hat{\dot{x}}_{k-1} + gT_s)$$

$$\hat{\dot{x}}_k = \hat{\dot{x}}_{k-1} - gT_s + K_{21}(x_k^* - \hat{x}_{k-1} - T_s \hat{\dot{x}}_{k-1} + .5gT_s^2) + K_{22}(\dot{x}_k^* - \hat{\dot{x}}_{k-1} + gT_s)$$

### If we define the residual as

$$RES_{1k} = x_k^* - \hat{x}_{k-1} - T_s \hat{\dot{x}}_{k-1} + .5gT_s^2$$

$$RES_{2k} = \dot{x}_k^* - \hat{\dot{x}}_{k-1} + gT_s$$

### The Kalman filter simplifies to

$$\hat{x}_k = \hat{x}_{k-1} + T_s \hat{\dot{x}}_{k-1} - .5gT_s^2 + K_{11}RES_{1k} + K_{12}RES_{2k}$$

$$\hat{\dot{x}}_k = \hat{\dot{x}}_{k-1} - gT_s + K_{21}RES_{1k} + K_{22}RES_{2k}$$



# Riccati Equations For Two Measurements

## Riccati equations with process noise

$$\mathbf{M}_k = \Phi_k \mathbf{P}_{k-1} \Phi_k^T + \mathbf{Q}_k \quad \mathbf{H} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \quad \Phi_k = \begin{bmatrix} 1 & T_s \\ 0 & 1 \end{bmatrix}$$

$$\mathbf{K}_k = \mathbf{M}_k \mathbf{H}^T (\mathbf{H} \mathbf{M}_k \mathbf{H}^T + \mathbf{R}_k)^{-1}$$

$$\mathbf{P}_k = (\mathbf{I} - \mathbf{K}_k \mathbf{H}) \mathbf{M}_k$$

## Ramp Signal

$$\begin{bmatrix} \dot{x} \\ \ddot{x} \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} x \\ \dot{x} \end{bmatrix} + \begin{bmatrix} 0 \\ u_s \end{bmatrix} \longrightarrow \mathbf{Q} = \mathbf{E} \left[ \begin{bmatrix} 0 \\ u_s \end{bmatrix} \begin{bmatrix} 0 & u_s \end{bmatrix} \right] = \begin{bmatrix} 0 & 0 \\ 0 & \Phi_s \end{bmatrix}$$

## Measurement Noise Matrix Has Changed

$$\mathbf{R}_k = \begin{bmatrix} \sigma_p^2 & 0 \\ 0 & \sigma_v^2 \end{bmatrix}$$

## Deriving discrete process noise matrix

$$\mathbf{Q}_k = \int_0^{T_s} \Phi(\tau) \mathbf{Q} \Phi^T(\tau) dt$$

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

# FORTRAN Version of Two Measurement Filter-1

```
GLOBAL DEFINE
      INCLUDE 'quickdraw.inc'

END
IMPLICIT REAL*8(A-H,O-Z)
REAL*8 P(2,2),Q(2,2),M(2,2),PHI(2,2),HMA T(2,2),HT(2,2),PHIT(2,2)
REAL*8 RMA T(2,2),IDN(2,2),PHIP(2,2),PHIPPHIT(2,2),HM(2,2)
REAL*8 HMHT(2,2),HMHTR(2,2),HMHTRINV(2,2),MHT(2,2),K(2,2)
REAL*8 KH(2,2),IKH(2,2)
INTEGER ORDER
C IF NSAMP=0 WE HAVE BOTH MEASUREMENTS AT THE SAME RATE
C IF NSAMP=INFINITY WE ONLY HAVE POSITION MEASUREMENT
NSAMP=5
TS=.1
PHIS=0.
A0=400000.
A1=-6000.
A2=-16.1
XH=0.
XDH=0.
SIGNOISE=1000.
SIGVEL=10.
ORDER=2
OPEN(1,STATUS='UNKNOWN',FILE='DATFIL')
OPEN(2,STATUS='UNKNOWN',FILE='COVFIL')
T=0.
S=0.
H=.001
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.
RMA T(I,J)=0.
HMA T(I,J)=0.
CONTINUE
RMAT(1,1)=SIGNOISE**2
RMA T(2,2)=SIGVEL**2
```

Position and velocity  
measurements at different  
times

Measurement noise

Measurement noise matrix

# FORTRAN Version of Two Measurement Filter-2

```

IDN(1,1)=1.
IDN(2,2)=1.
P(1,1)=99999999999.
P(2,2)=99999999999.
PHI(1,1)=1.
PHI(1,2)=TS
PHI(2,2)=1.
Q(1,1)=TS*TS*TS*PHIS/3.
Q(1,2)=.5*TS*TS*PHIS
Q(2,1)=Q(1,2)
Q(2,2)=PHIS*TS
HMAT(1,1)=1.
N=0
DO 10 T=0.,.30.,TS
IF(N>=NSAMP)THEN
    HMA T(2,2)=1.
ENDIF
CALL MA TIRN(PHI,ORDER,ORDER,PHIT)
CALL MA TIRN(HMA T,ORDER,ORDER,HT)
CALL MA TMUL(PHI,ORDER,ORDER,P,ORDER,ORDER,PHIP)
CALL MA TMUL(PHIP,ORDER,ORDER,PHIT,ORDER,ORDER,PHIPPHIT)
CALL MA TADD(PHIPPHIT,ORDER,ORDER,Q,M)
CALL MA TMUL(HMA T,ORDER,ORDER,M,ORDER,ORDER,HM)
CALL MA TMUL(HM,ORDER,ORDER,HT,ORDER,ORDER,HMHT)
CALL MA TADD(HMHT,ORDER,ORDER,RMA T,HMHTR)
DET=HMHTR(1,1)*HMHTR(2,2)-HMHTR(1,2)*HMHTR(2,1)
HMHTRINV(1,1)=HMHTR(2,2)/DET
HMHTRINV(1,2)=-HMHTR(1,2)/DET
HMHTRINV(2,1)=-HMHTR(2,1)/DET
HMHTRINV(2,2)=HMHTR(1,1)/DET
CALL MA TMUL(M,ORDER,ORDER,HT,ORDER,ORDER,MHT)
CALL MA TMUL(MHT,ORDER,ORDER,HMHTRINV,ORDER,ORDER,K)
CALL MA TMUL(K,ORDER,ORDER,HMA T,ORDER,ORDER,KH)
CALL MA TSUB(IDN,ORDER,ORDER,KH,IKH)
CALL MA TMUL(IKH,ORDER,ORDER,M,ORDER,ORDER,P)

```

Change measurement matrix when velocity  
Measurement is available

# FORTRAN Version of Two Measurement Filter-3

```
CALL GAUSS(XNOISE,SIGNOISE)
CALL GAUSS(XDNOISE,SIGVEL)
X=A0+A1*T+A2*T*T
XD=A1+2*A2*T
XS=X+XNOISE
XDS=XD+XDNOISE
RES1=XS-XH-TS*XDH+16.1*TS*TS
RES2=0.
IF(N>=NSAMP)THEN
    N=0
    RES2=XDS-XDH+32.2*TS
ENDIF
XH=XH+XDH*TS-16.1*TS*TS+K(1,1)*RES1+K(1,2)*RES2
XDH=XDH-32.2*TS+K(2,1)*RES1+K(2,2)*RES2
SP11=SQRT(P(1,1))
SP22=SQRT(P(2,2))
XHERR=X-XH
XDHERR=XD-XDH

WRITE(9,*)T,XD,XDH,K(1,1),K(2,1)
WRITE(1,*)T,X,XH,XD,XDH
WRITE(2,*)T,XHERR,SP11,-SP11,XDHERR,SP22,-SP22,HMAT(2,2),RES2
N=N+1
HMAT(2,2)=0.
CONTINUE
PAUSE
CLOSE(1)
END
```

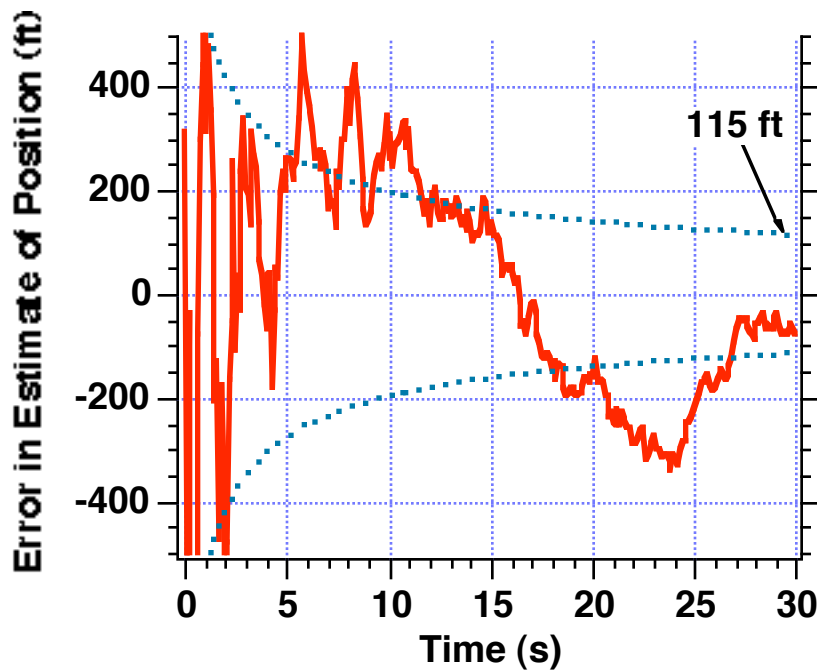
New residual when velocity measurement is available

Reset measurement matrix

10

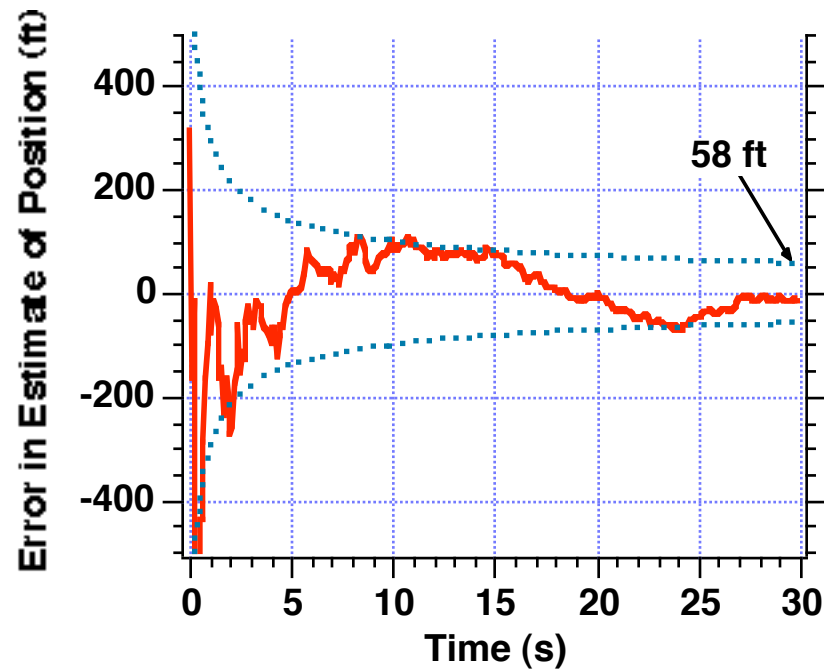
# Addition of Velocity Measurement Reduces Position Estimation Error By Factor of Two

## Position Measurements Only



There are 10 position measurements per second

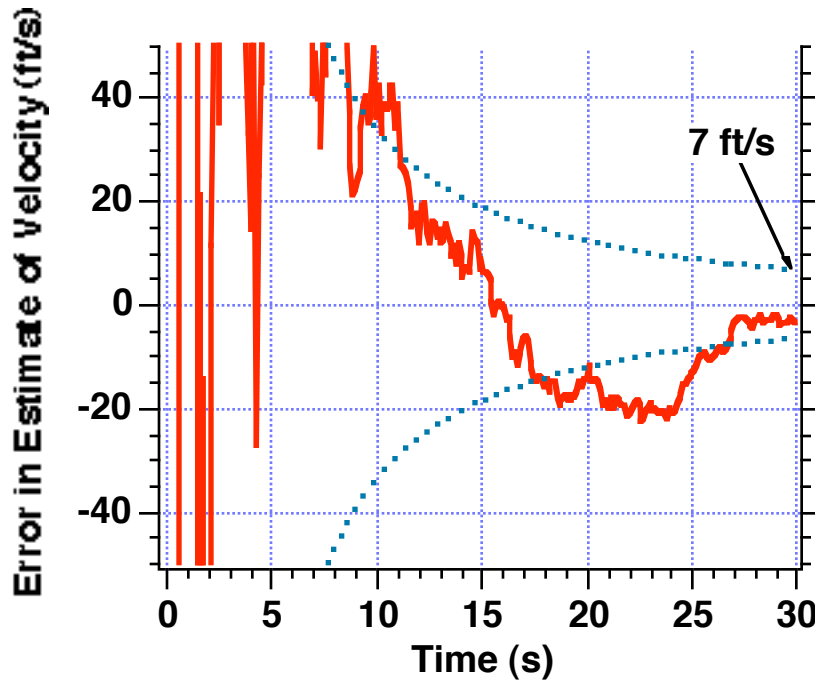
## Position and Velocity Measurements



There are 10 position and velocity measurements per second

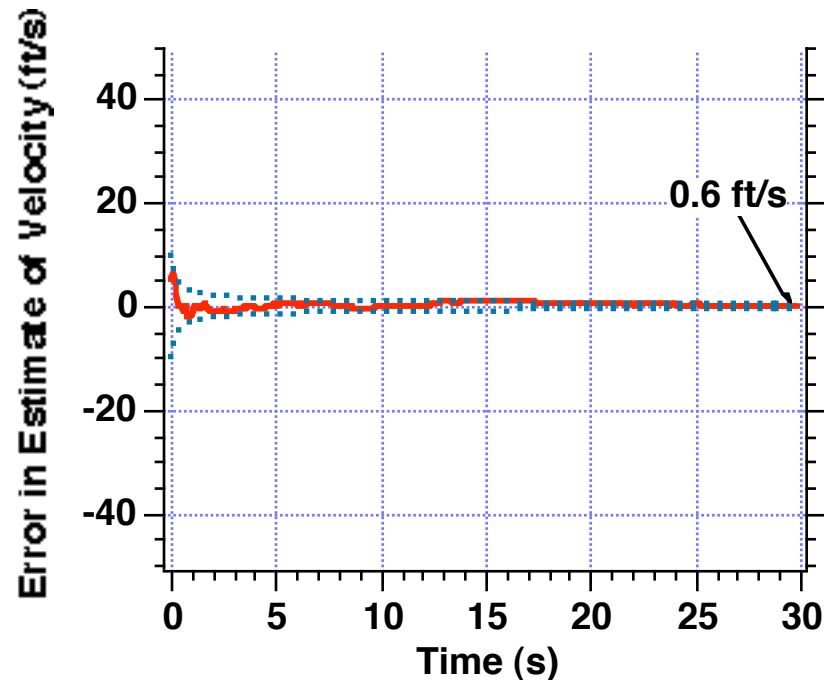
# Addition of Velocity Measurement Reduces Velocity Estimation Error By Order of Magnitude

## Position Measurements Only



There are 10 position measurements per second

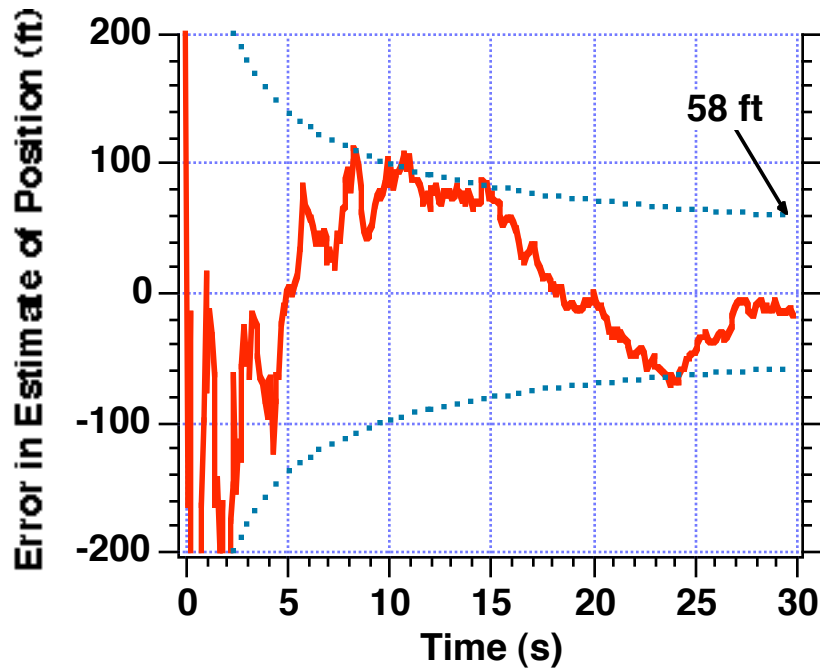
## Position and Velocity Measurements



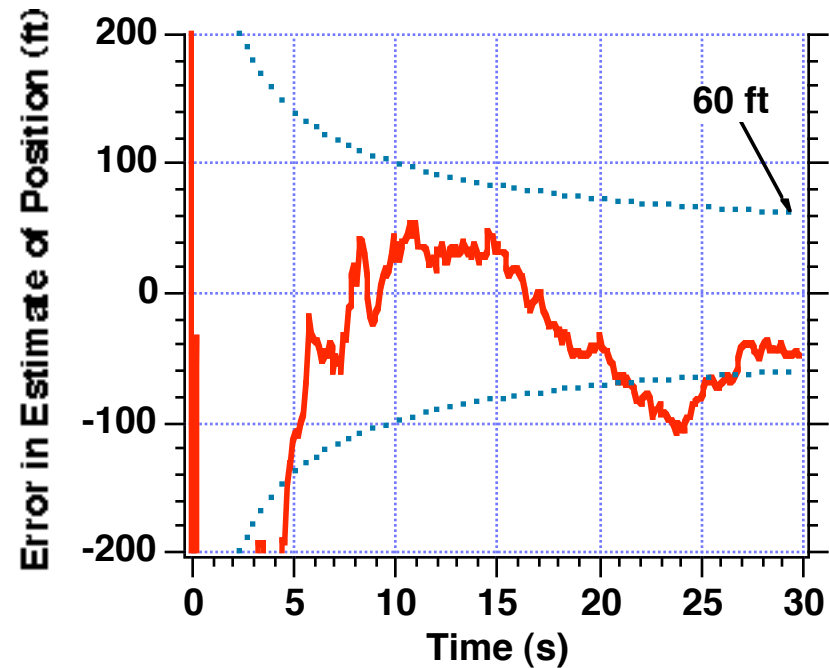
There are 10 position and velocity measurements per second

# Having Fewer Velocity Measurements Slightly Increases Position Estimation Error

10 Velocity Measurements Per Sec



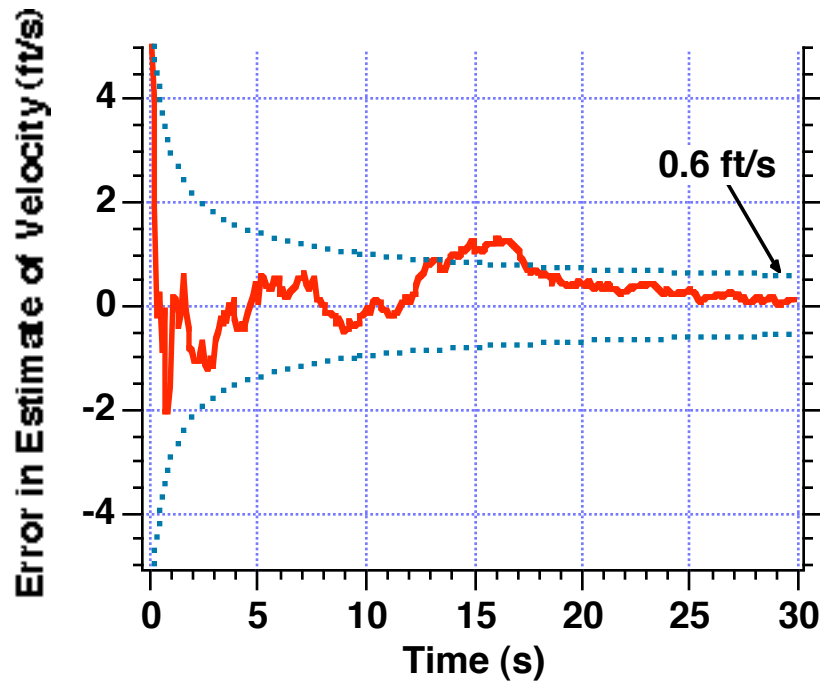
2 Velocity Measurements Per Sec



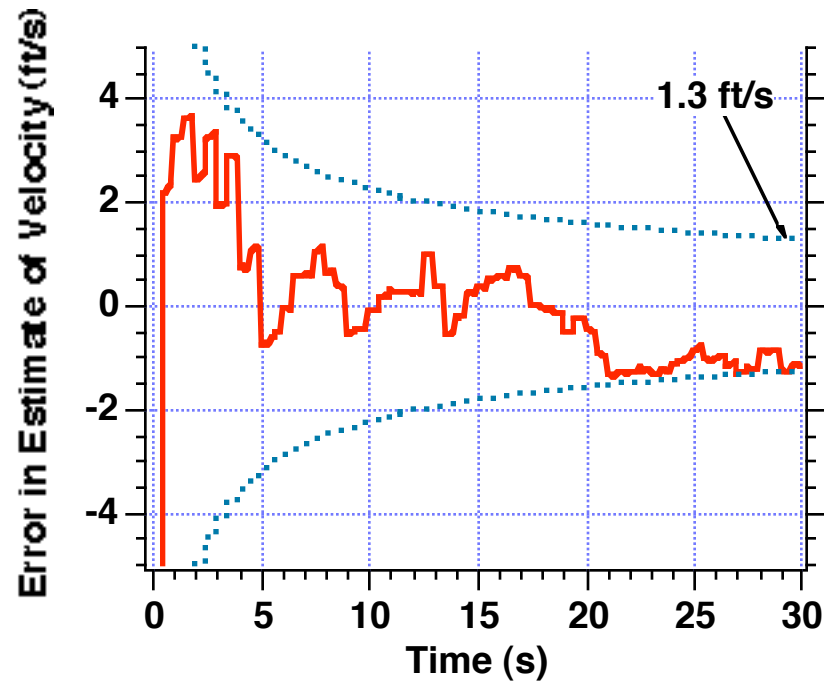
There are 10 position measurements per second

# Having Fewer Velocity Measurements Increases Velocity Estimation Error

10 Velocity Measurements Per Sec



2 Velocity Measurements Per Sec



There are 10 position measurements per second



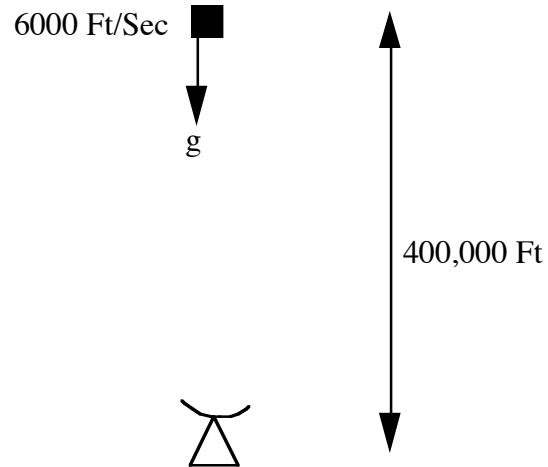
## **Using Additional Measurements Summary**

- **Addition of second measurement significantly reduces estimation errors**
- **Modeling of second measurement at different data rate is easy**

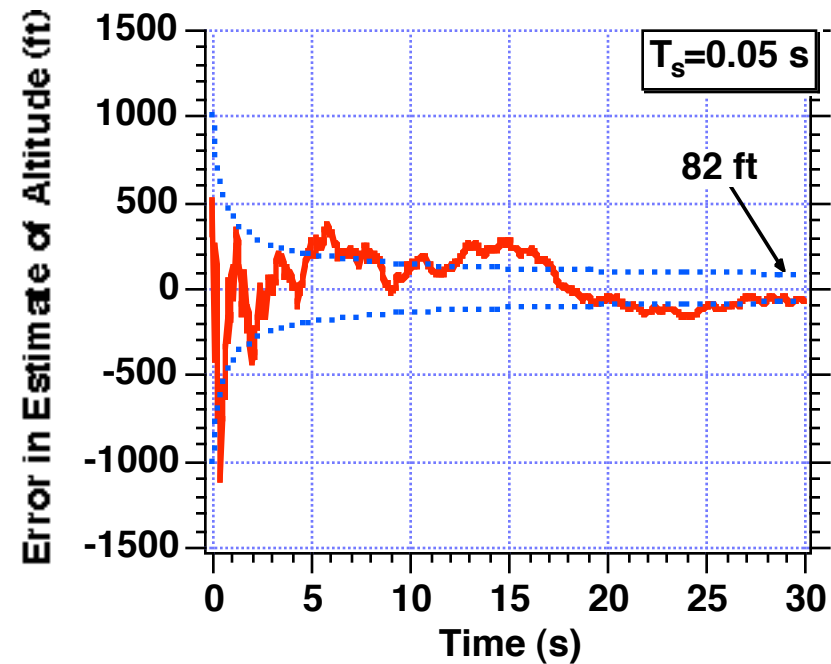
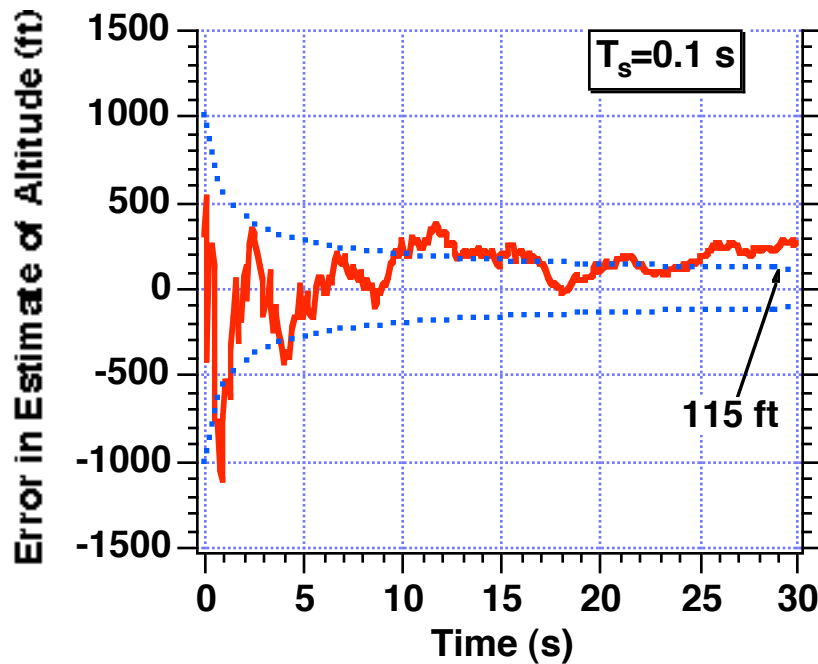
# Asynchronous Data Rate

# Problem

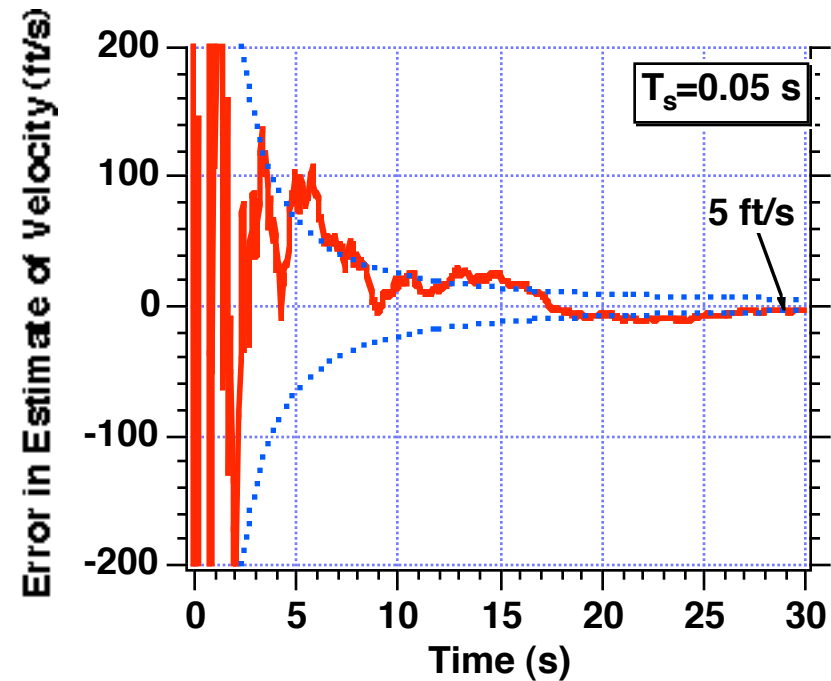
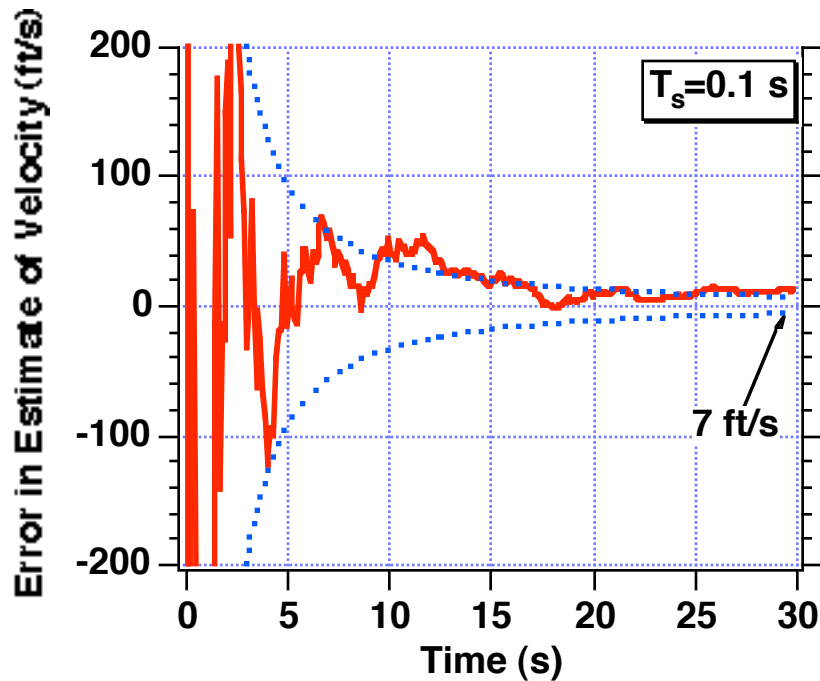
**In some applications data rate varies and is not known in advance  
We will revisit falling object problem of previous sections**



## Error in Estimate of Altitude Improves If Fixed Sampling Time is Smaller



## Error in Estimate of Velocity Improves If Fixed Sampling Time is Smaller



# Asynchronous Kalman Filter-1

```
GLOBAL DEFINE
      INCLUDE 'quickdraw.inc'

END
IMPLICIT REAL*8(A-H,O-Z)
REAL*8 P(2,2),Q(2,2),M(2,2),PHI(2,2),HMAT(1,2),HT(2,1),PHIT(2,2)
REAL*8 RMAT(1,1),IDN(2,2),PHIP(2,2),PHIPPHIT(2,2),HM(1,2)
REAL*8 HMHT(1,1),HMHTR(1,1),HMHTRINV(1,1),MHT(2,1),K(2,1)
REAL*8 KH(2,2),IKH(2,2)
INTEGER ORDER
LOGICAL QFIRST
QFIRST=.TRUE.
TS=.1
PHIS=0.
A0=400000.
A1=-6000.
A2=-16.1
XH=0.
XDH=0.
SIGNOISE=1000.
ORDER=2
OPEN(1,STATUS='UNKNOWN',FILE='DATFIL')
OPEN(2,STATUS='UNKNOWN',FILE='COVFIL')
T=0.
S=0.
H=.001
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
```

**Needed for logic**

# Asynchronous Kalman Filter-2

```
10 RMAT(1,1)=SIGNOISE**2
    IDN(1,1)=1.
    IDN(2,2)=1.
    P(1,1)=99999999999.
    P(2,2)=99999999999.
    HMAT(1,1)=1.
    HMAT(1,2)=0.
    T=0.
    IF(T>30.)GOTO 999
    CALL GAUSS(XNOISE,SIGNOISE)
    X=A0+A1*T+A2*T*T
    XD=A1+2*A2*T
    XS=X+XNOISE
    CALL UNIF(XNOISE)
    TSP=XNOISE*.1
    IF(QFIRST)THEN
        XH=XS
        XDH=0.
        TOLD=T
        QFIRST=.FALSE.
    ELSE
        TS=T-TOLD
        PHI(1,1)=1.
        PHI(1,2)=TS
        PHI(2,2)=1.
        Q(1,1)=TS*TS*TS*PHIS/3.
        Q(1,2)=.5*TS*TS*PHIS
        Q(2,1)=Q(1,2)
        Q(2,2)=PHIS*TS
```

Model of real world comes first

Sampling time uniformly distributed  
Between 0 and .1 s

Initialize Kalman filter

Calculate new sampling time

# Asynchronous Kalman Filter-3

```

CALL MATTRN(PHI,ORDER,ORDER,PHIT)
CALL MATTRN(HMAT,1,ORDER,HT)
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)
RES=XS-XH-TS*XDH+16.1*TS*TS
XH=XH+XDH*TS-16.1*TS*TS+K(1,1)*RES
XDH=XDH-32.2*TS+K(2,1)*RES
TOLD=T

```

**Reset**

```

SP11=SQRT(P(1,1))
SP22=SQRT(P(2,2))
XHERR=X-XH
XDHERR=XD-XDH
WRITE(9,*)T,XD,XDH,TS,TSP
WRITE(1,*)T,X,XH,XD,XDH,TS,TSP
WRITE(2,*)T,XHERR,SP11,-SP11,XDHERR,SP22,-SP22

```

```

ENDIF
T=T+TSP
GOTO 10
CONTINUE
PAUSE
CLOSE(1)
END

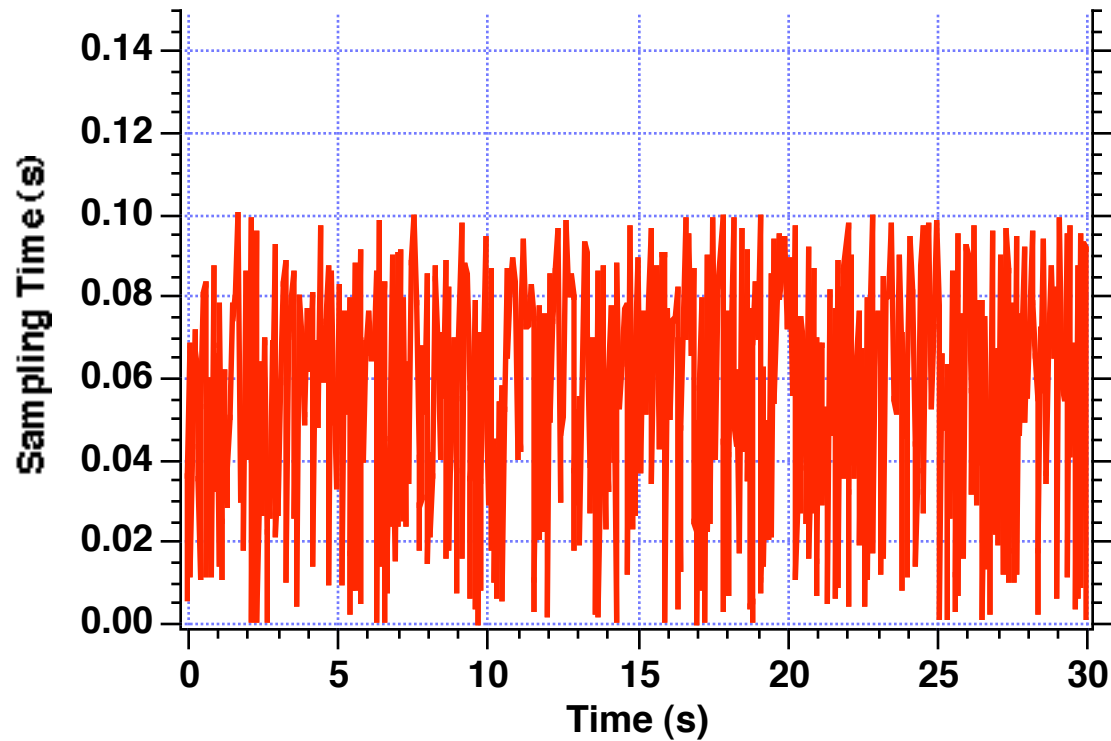
```

**Update time**

999

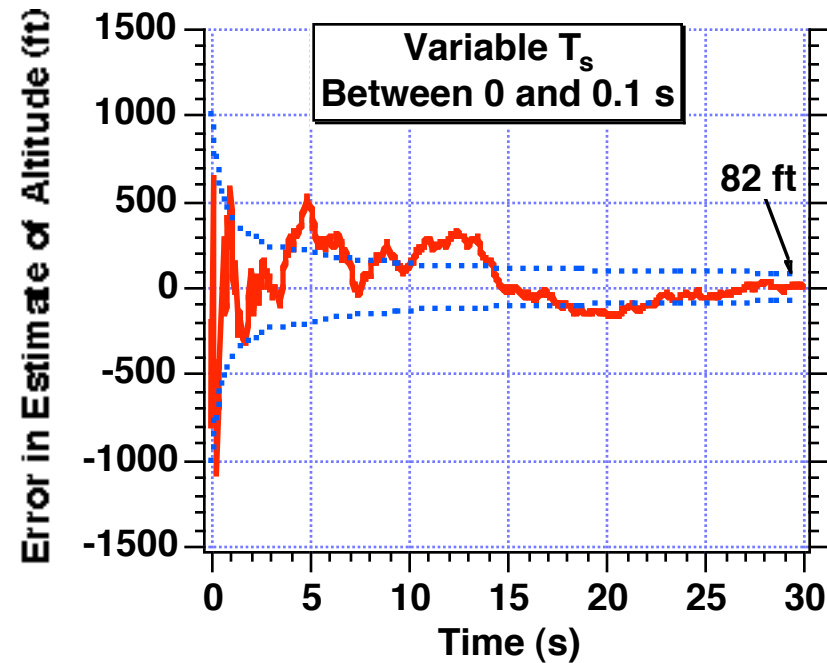
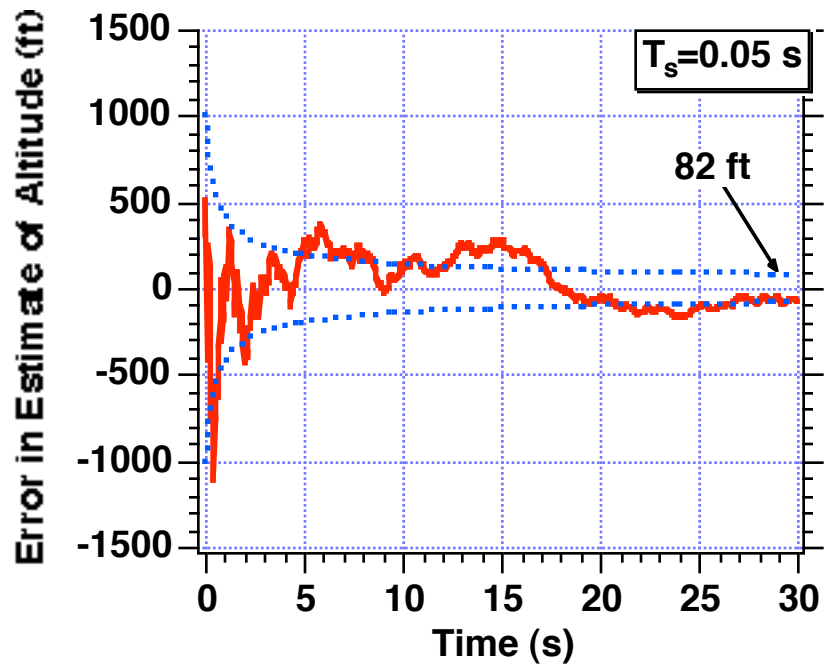


## Sampling Time is Uniformly Distributed Between 0 and 0.1 s and 0.1 s

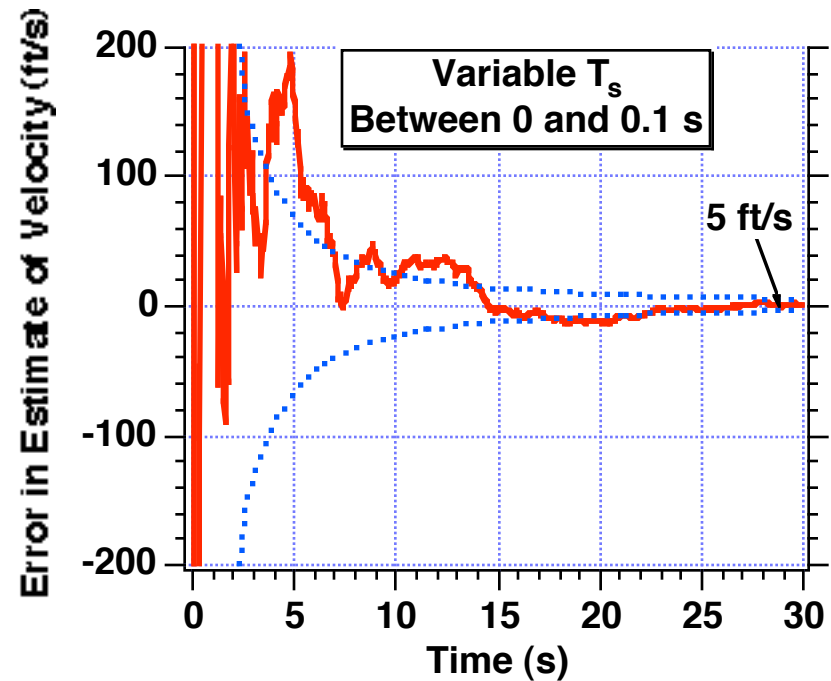
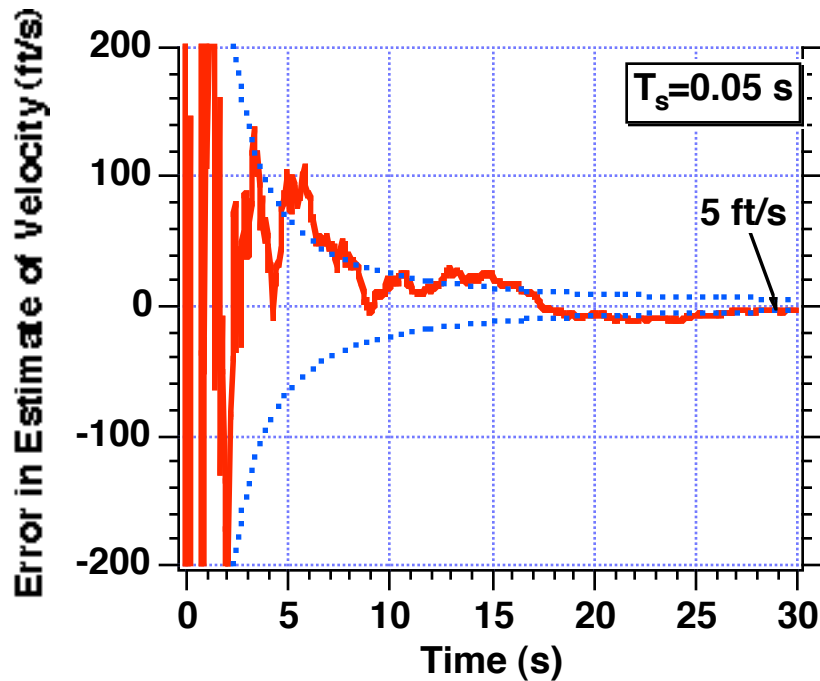


**Average sampling time is 0.05 s**

# Variable Sampling Time Filter Gives Good Position Estimates



# Variable Sampling Time Filter Gives Good Velocity Estimates

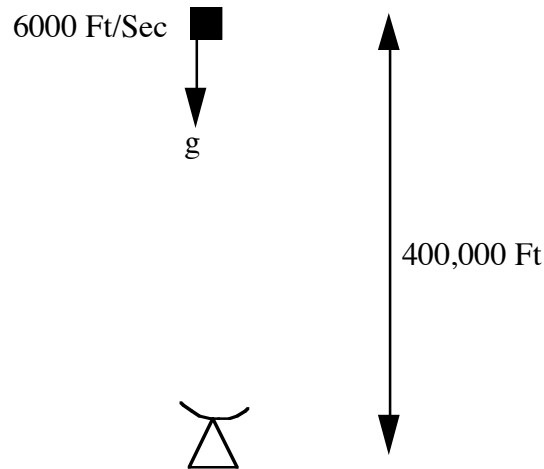


# Batch Processing

# Batch Processing Overview

- **Problem review**
  - Throwing away data
- **Preprocessing**
  - Wrong way
  - Right way

# Radar Tracking Falling Object



**From basic physics**

$$x = 400000 - 6000t - .5gt^2$$

**Second-order process**

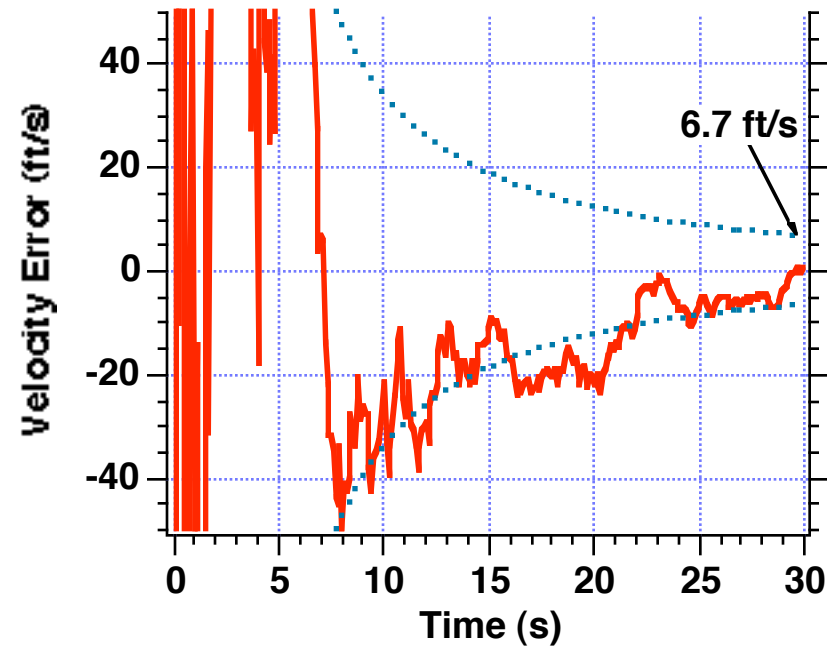
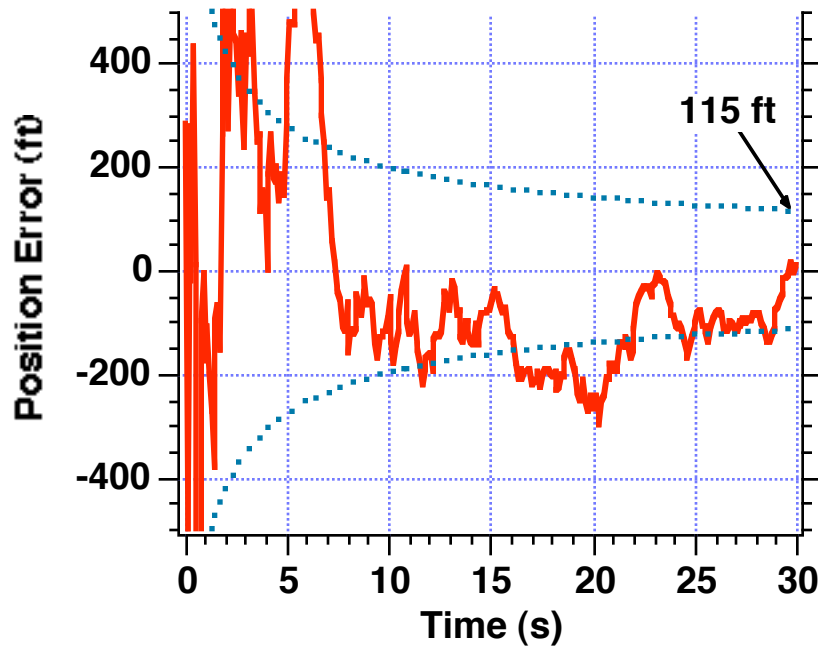
**Velocity of object can be found by differentiating**

$$\dot{x} = -6000 - gt$$

**Radar measures altitude with standard deviation of 1000 ft**

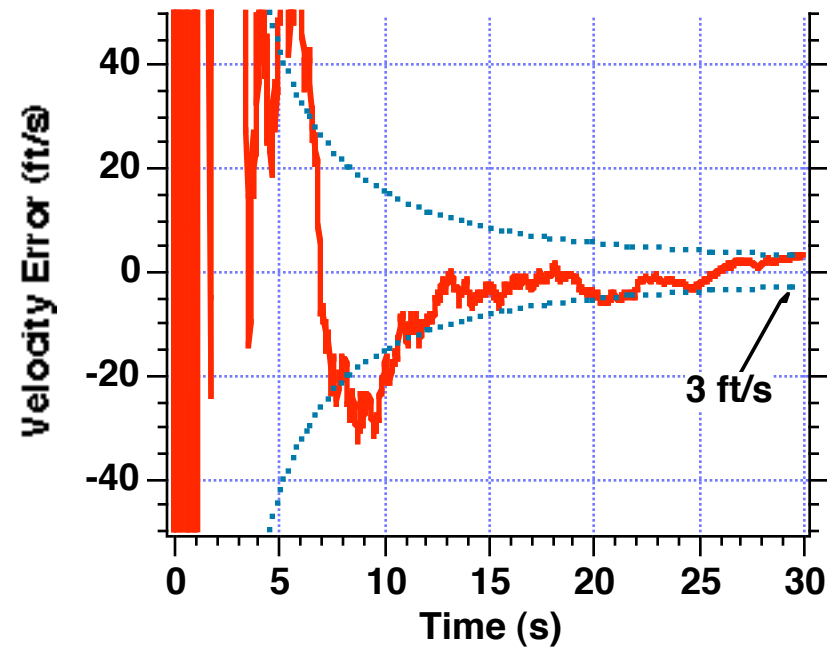
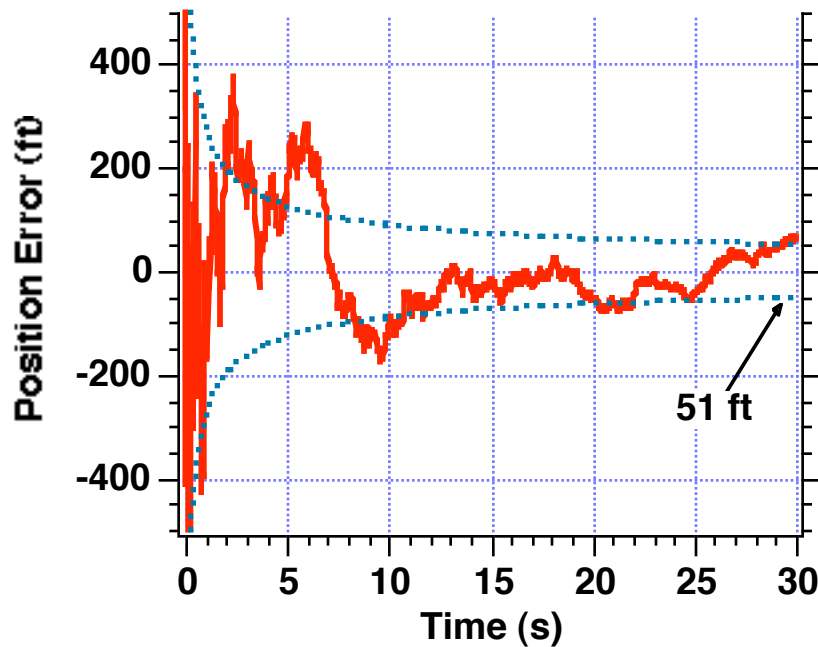
**Desire to track object and estimate altitude and velocity**

## Simple Filtering Every 0.1 s Works



Data available every 0.02 s, **but** sent to KF every 0.1 s

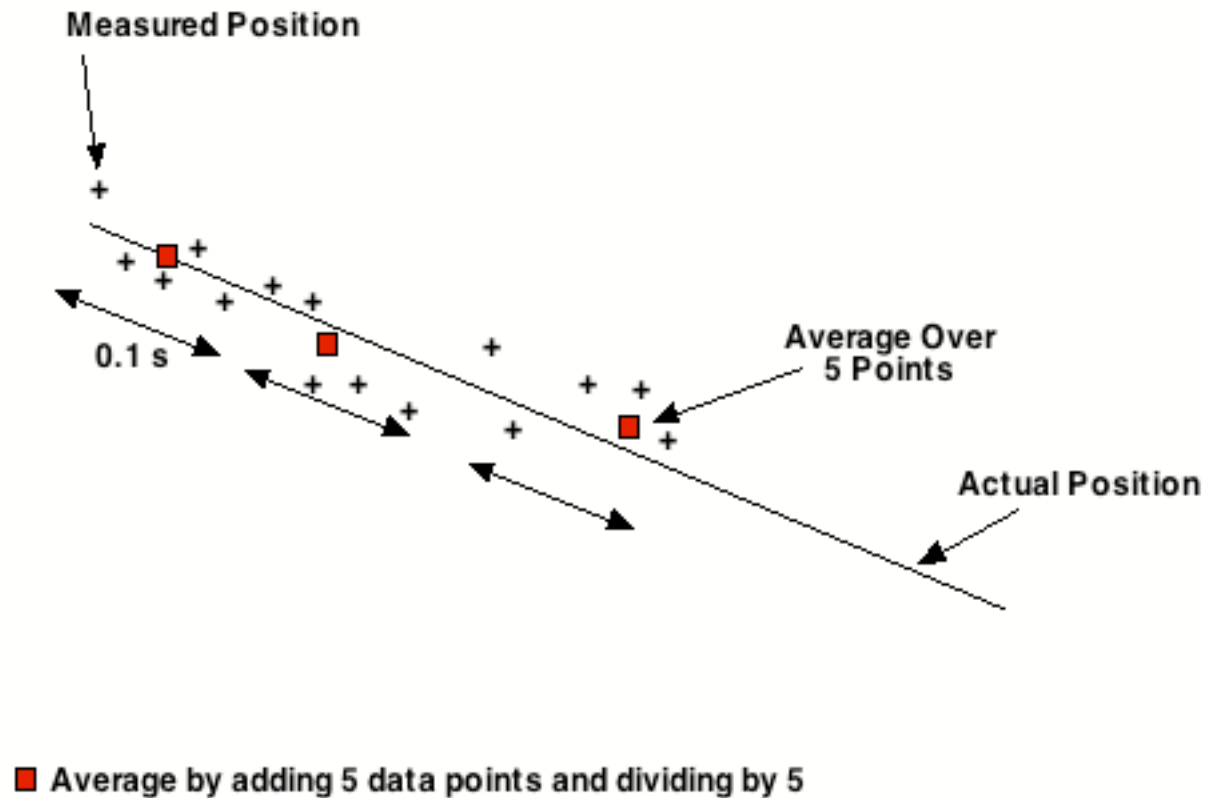
## Simple Filtering Every 0.02 s Works Even Better



Data available every 0.02 s, **and** sent to KF every 0.02 s



# Preprocessing Measurements By Averaging



# FORTRAN Code For Batch Processing Scheme-1

```
GLOBAL DEFINE
      INCLUDE 'quickdraw.inc'

END
IMPLICIT REAL*8 (A-H)
IMPLICIT REAL*8 (O-Z)
REAL*8 P(2,2),Q(2,2),M(2,2),PHI(2,2),HMAT(1,2),HT(2,1),PHIT(2,2)
REAL*8 RMAT(1,1),IDN(2,2),PHIP(2,2),PHIPPHIT(2,2),HM(1,2)
REAL*8 HMHT(1,1),HMHTR(1,1),HMHTRINV(1,1),MHT(2,1),K(2,1)
REAL*8 KH(2,2),IKH(2,2)
INTEGER ORDER
OPEN(1,STATUS='UNKNOWN',FILE='DATFIL')
OPEN(2,STATUS='UNKNOWN',FILE='COVFIL')
PHIS=0.
TS1=.02
TS2=.1
TF=30.
A0=400000.
A1=-6000.
A2=-16.1
XH=0.
XDH=0.
SIGNOISE=1000.
ORDER=2
T=0.
S1=0.
S2=0.
H=.001
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
IDN(1,1)=1.
IDN(2,2)=1.
P(1,1)=99999999999.
P(2,2)=99999999999.
```

14

## FORTRAN Code For Batch Processing Scheme-2

```
PHI(1,1)=1.  
PHI(1,2)=TS2  
PHI(2,2)=1.  
Q(1,1)=TS2*TS2*TS2*PHIS/3.  
Q(1,2)=.5*TS2*TS2*PHIS  
Q(2,1)=Q(1,2)  
Q(2,2)=PHIS*TS2  
HMAT(1,1)=1.  
HMAT(1,2)=0.  
E1=0.  
E2=0.  
E3=0.  
E4=0.  
DO 10 T=0.,TF,TS1  
    X=A0+A1*T+A2*T*T  
    XD=A1+2.*A2*T  
    CALL GAUSS(XNOISE,SIGNOISE)  
    XS=X+XNOISE  
    E=XS  
    EAV=(E4+E3+E2+E1+E)/5.  
    E4=E3  
    E3=E2  
    E2=E1  
    E1=E  
    S2=S2+TS1
```

**Average 5 measurements**

# FORTRAN Code For Batch Processing Scheme-3

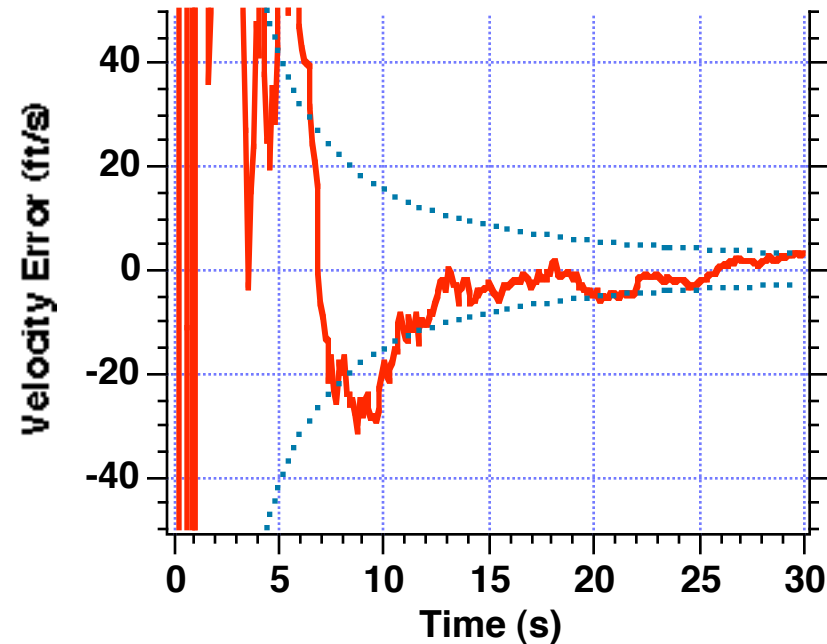
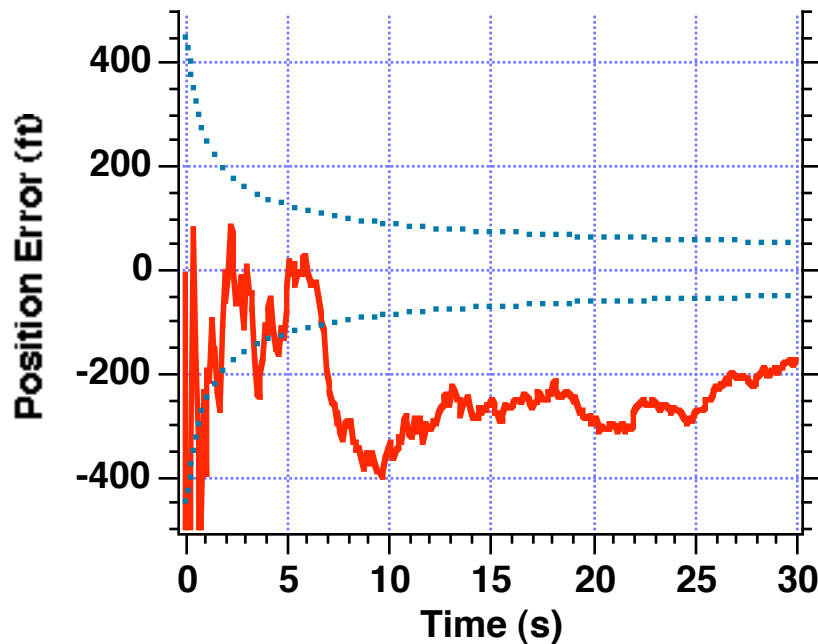
```

1          IF(S2>=(TS2-.00001))THEN
2              S2=0.
3              RMAT(1,1)=(SIGNOISE/2.24)**2
4              CALL MATTRN(PHI,ORDER,ORDER,PHIT)
5              CALL MATTRN(HMAT,1,ORDER,HT)
6              CALL MATMUL(PHI,ORDER,ORDER,P,ORDER,ORDER,PHIP)
7              CALL MATMUL(PHIP,ORDER,ORDER,PHIT,ORDER,ORDER,
8                  PHIPPHIT)
9              CALL MATADD(PHIPPHIT,ORDER,ORDER,Q,M)
10             CALL MATMUL(HMAT,1,ORDER,M,ORDER,ORDER,HM)
11             CALL MATMUL(HM,1,ORDER,HT,ORDER,1,HMHT)
12             CALL MATADD(HMHT,ORDER,ORDER,RMAT,HMHTR)
13             HMHTRINV(1,1)=1./HMHTR(1,1)
14             CALL MATMUL(M,ORDER,ORDER,HT,ORDER,1,MHT)
15             CALL MATMUL(MHT,ORDER,1,HMHTRINV,1,1,K)
16             CALL MATMUL(K,ORDER,1,HMAT,1,ORDER,KH)
17             CALL MATSUB(IDN,ORDER,ORDER,KH,IKH)
18             CALL MATMUL(IKH,ORDER,ORDER,M,ORDER,ORDER,P)
19             WRONG WAY
20             XS=EAV
21             CORRECT WAY
22             XS=EAV+.05*XDH-16.1*.05*.05
23             RES=XS-XH-TS2*XDH+16.1*TS2*TS2
24             XH=XH+XDH*TS2-16.1*TS2*TS2+K(1,1)*RES
25             XDH=XDH-32.2*TS2+K(2,1)*RES
26             SP11=SQRT(P(1,1))
27             SP22=SQRT(P(2,2))
28             XHERR=X-XH
29             XDHERR=XD-XDH
30             WRITE(9,*)T,X,XH,XD,XDH
31             WRITE(1,*)T,X,XH,XD,XDH
32             WRITE(2,*)T,XHERR,SP11,-SP11,XDHERR,SP22,-SP22
33
34             ENDIF
35
36             CONTINUE
37             PAUSE
38             END

```

C  
C  
C

## Batch Processing Yields Large Hangoff Error in Position When Measurement is Just Averaged

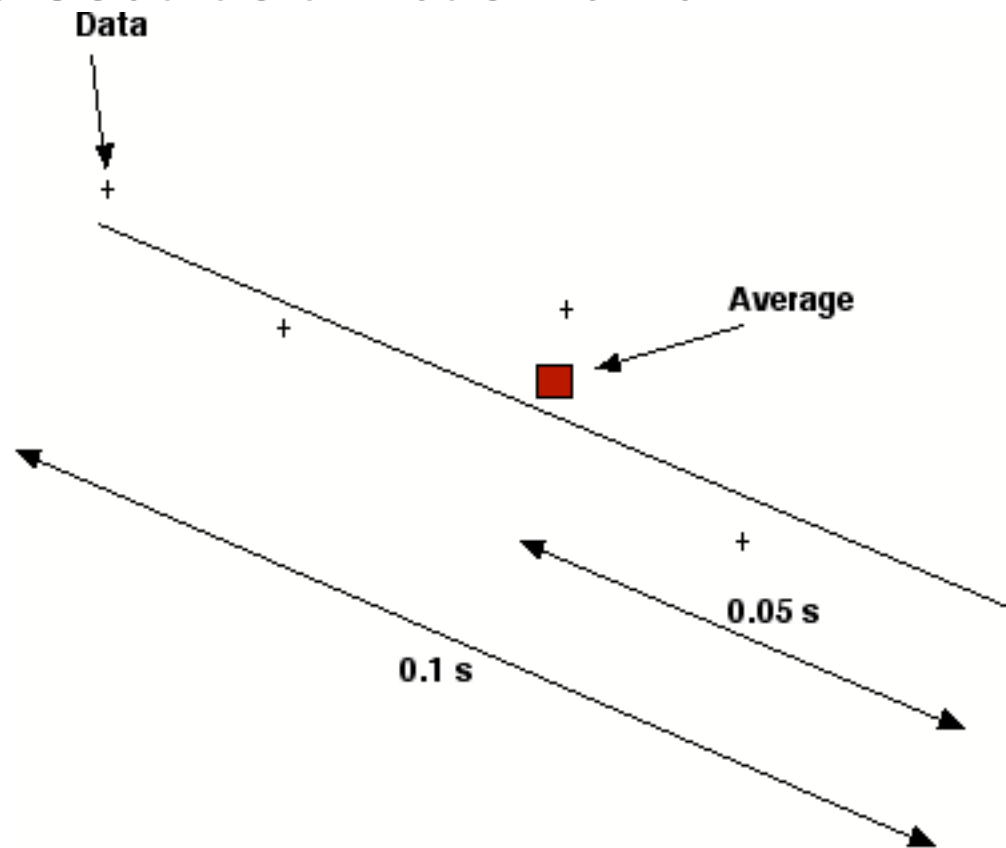


**FORTRAN**

XS=EAV

Data available every 0.02 s, **averaged** and sent to KF every 0.1 s

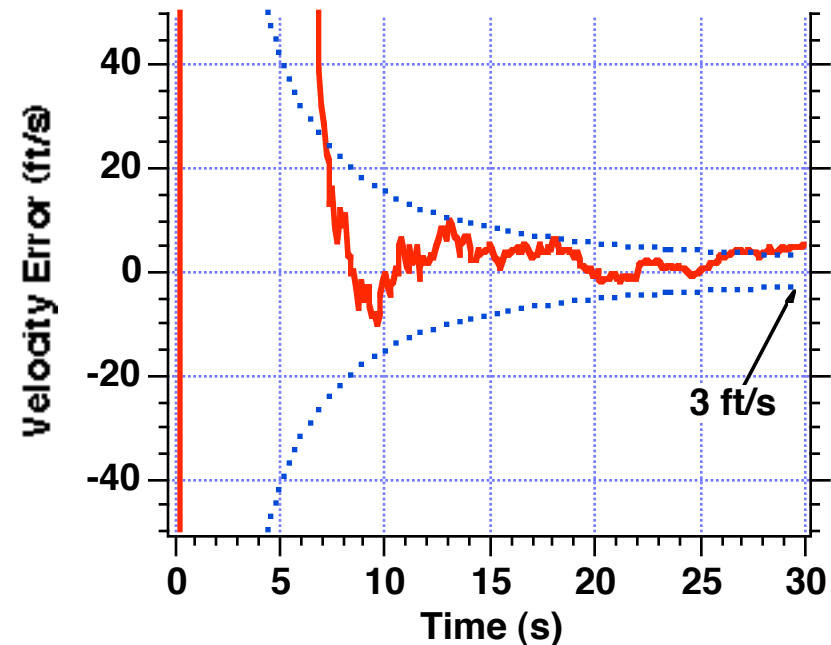
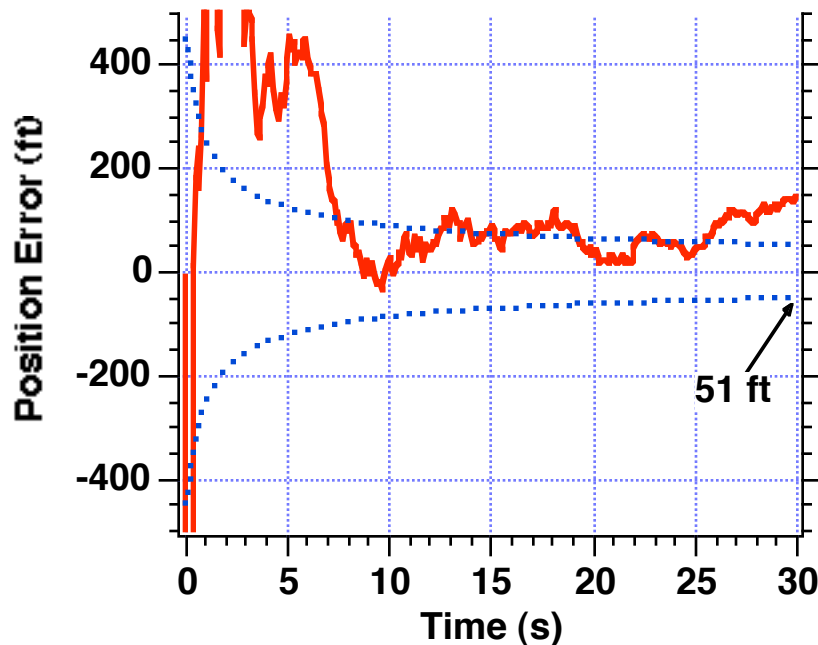
# We Must Propagate Average Forward In Order To Be Used as a Measurement



```
FORTTRAN  
XS=EAV+.05*XDH-16,1*.05*.05
```

$$\text{Measurement} = \text{Average} + .05 \hat{x} - 16.1 \cdot .05 \cdot .05$$

## Hangoff Error is Mostly Removed and 0.02 s Performance Achieved When Measurement is Averaged and Propagated



**FORTRAN**

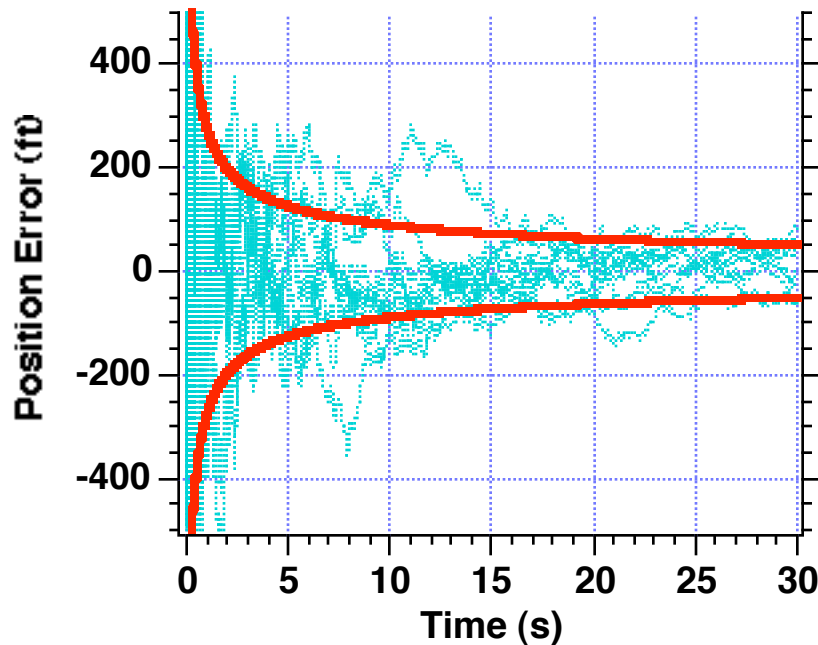
$XS=EAV+.05*XDH-16,1*.05*.05$

Data available every 0.02 s, **averaged** and sent to KF every 0.1 s

## Monte Carlo Position Error (10 Runs)

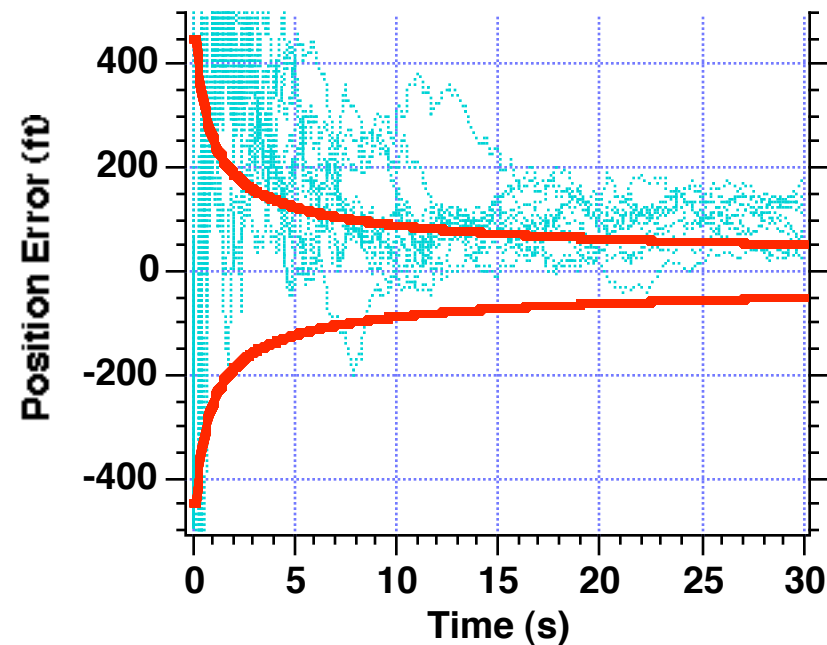
Data available every 0.02 s

Sent to KF every 0.02 s



Data available every 0.02 s

Averaged and sent to KF every 0.1 s



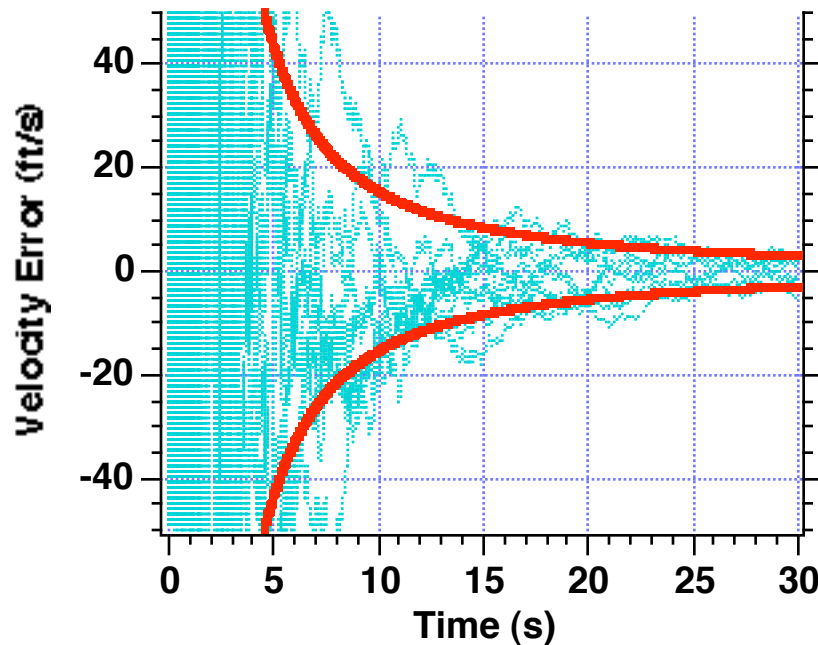
**There is still a slight bias in position error when measurement has been averaged and propagated**



## Monte Carlo Velocity Error (10 Runs)

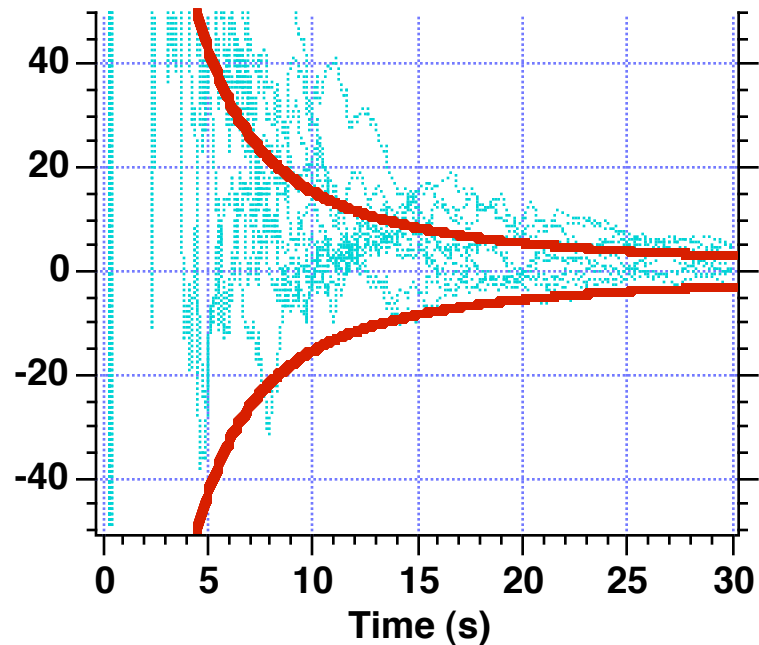
Data available every 0.02 s

Sent to KF every 0.02 s



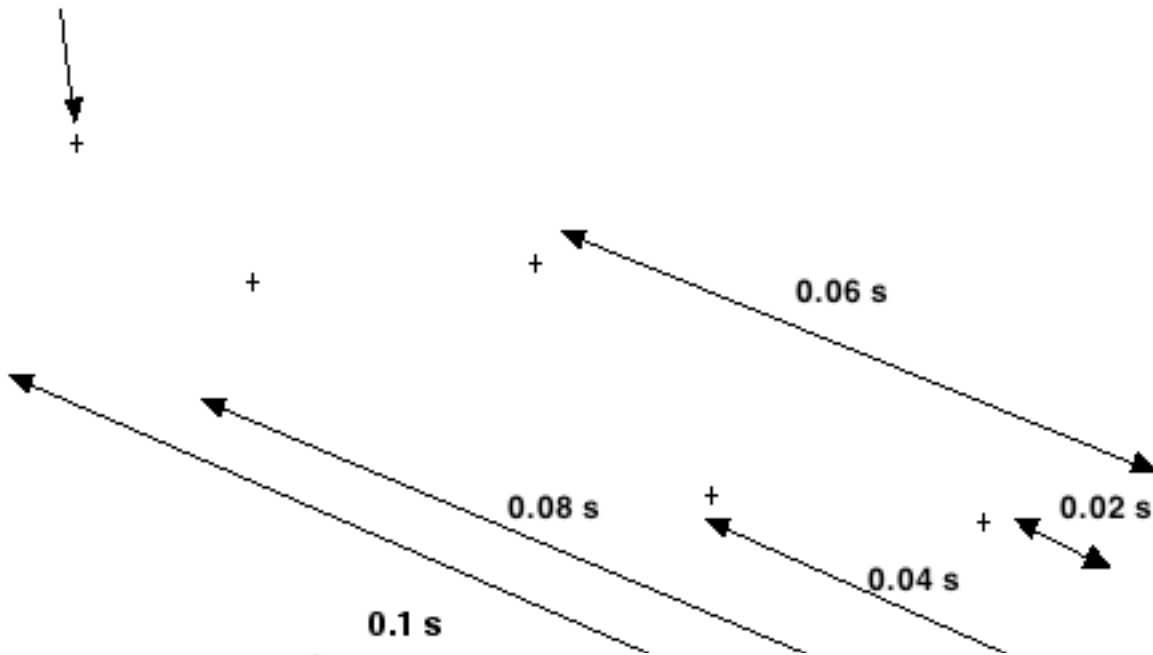
Data available every 0.02 s

Averaged and sent to KF every 0.1 s



**There is virtually no bias in velocity error when measurement has been averaged and propagated**

Data



$$\text{Proj1} = \text{Measurement} + 0.1 \hat{x} - 16.1 * 0.1 * 0.1$$

$$\text{Proj2} = \text{Measurement} + 0.08 \hat{x} - 16.1 * 0.08 * 0.08$$

$$\text{Proj3} = \text{Measurement} + 0.06 \hat{x} - 16.1 * 0.06 * 0.06$$

$$\text{Proj4} = \text{Measurement} + 0.04 \hat{x} - 16.1 * 0.04 * 0.04$$

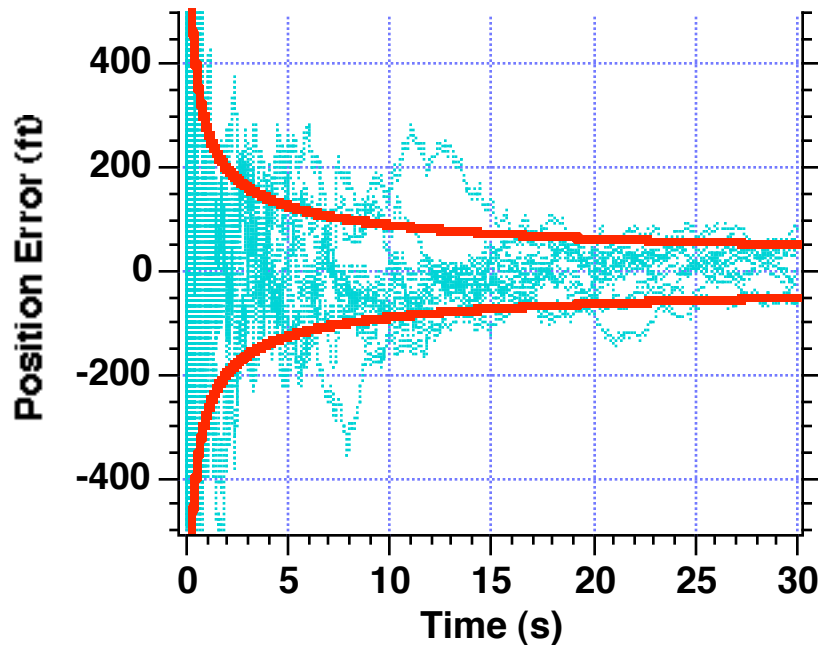
$$\text{Proj5} = \text{Measurement} + 0.02 \hat{x} - 16.1 * 0.02 * 0.02$$

$$\text{Average Measurement} \quad \frac{\text{Proj1} + \text{Proj2} + \text{Proj3} + \text{Proj4} + \text{Proj5}}{5}$$

## New Monte Carlo Position Error (10 Runs)

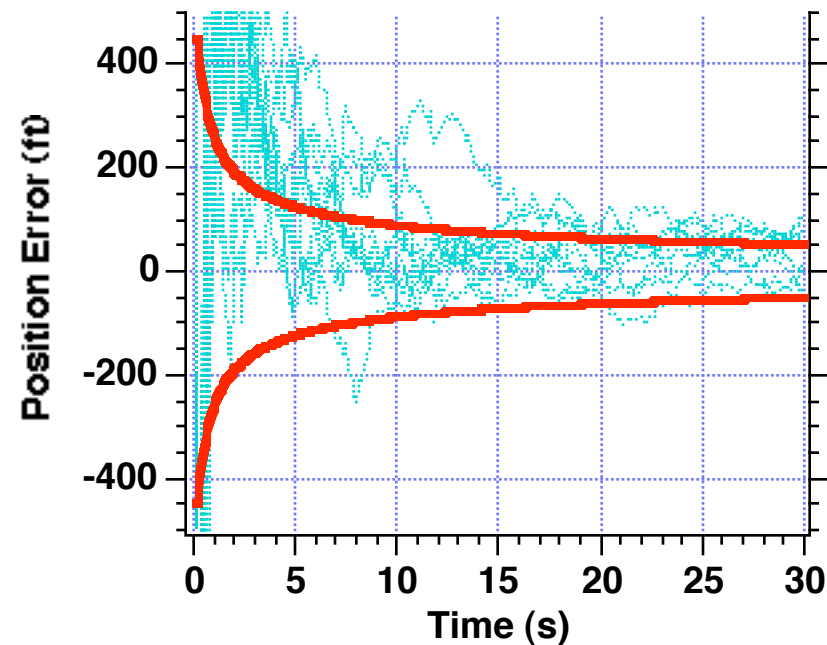
Data available every 0.02 s

Sent to KF every 0.02 s



Data available every 0.02 s

Averaged and sent to KF every 0.1 s

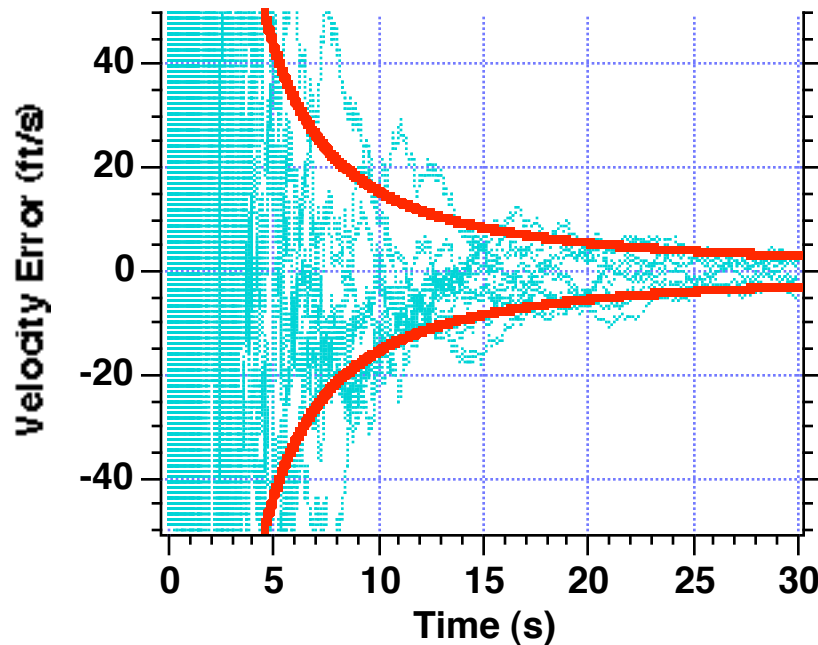


Now there is **no** bias in position error when measurement has been propagated and averaged

## New Monte Carlo Velocity Error (10 Runs)

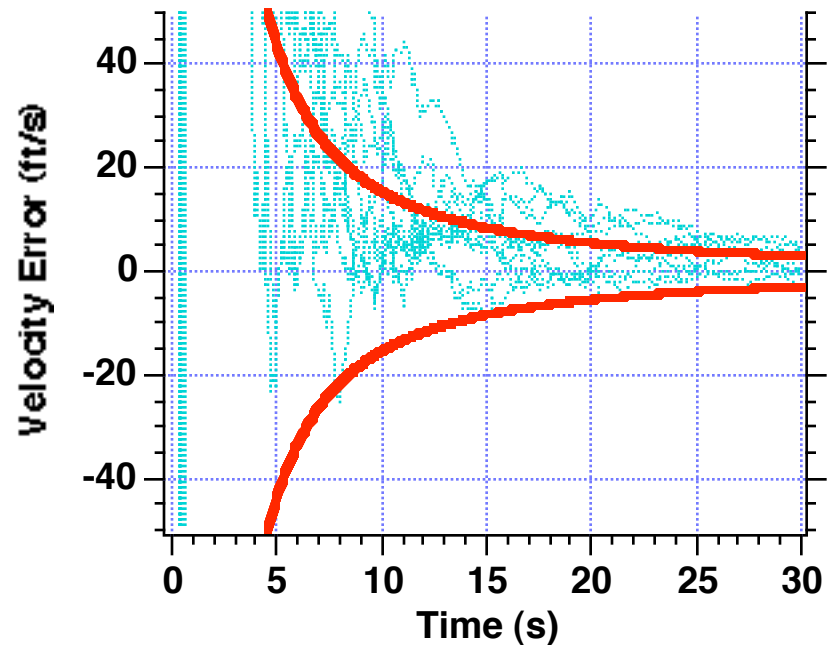
Data available every 0.02 s

Sent to KF every 0.02 s



Data available every 0.02 s

Averaged and sent to KF every 0.1 s



**There is still virtually no bias in velocity error when measurement has been propagated and averaged**