

UNCLASSIFIED

A Self-adjusting Model for Maneuvering Targets

by

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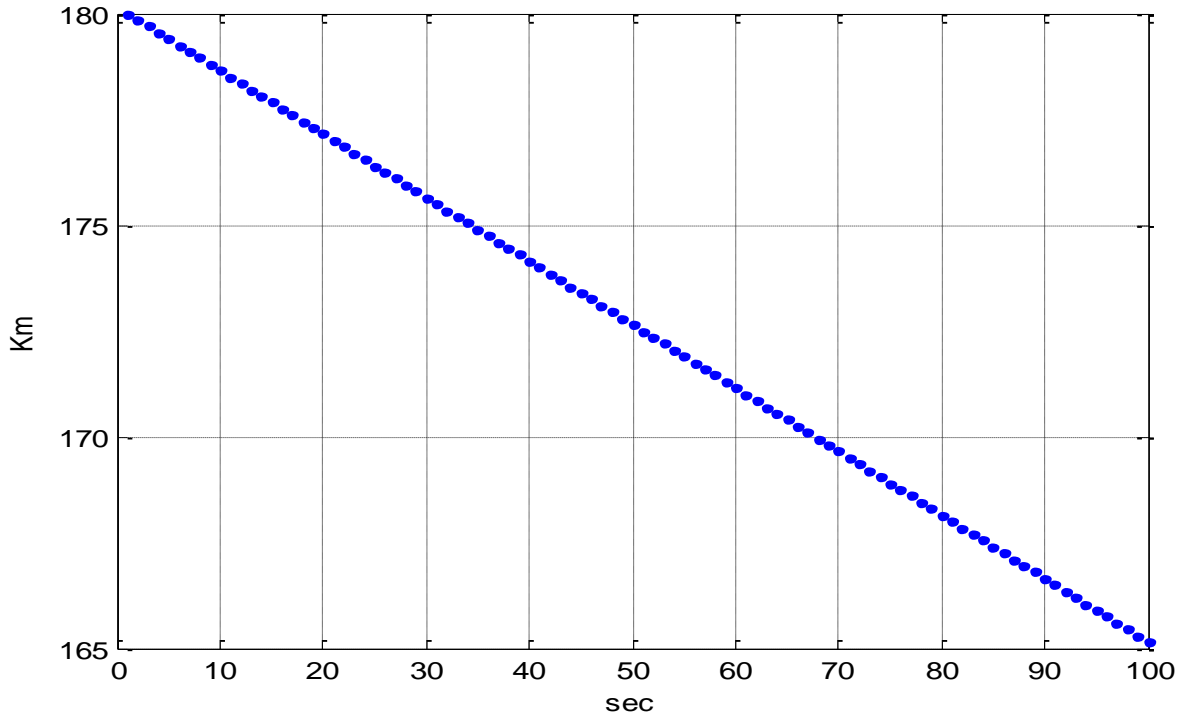
The Primary Task of Tracking

The **primary task** of tracking is prediction and filtering.

Target dynamics model and tracking algorithm work jointly.

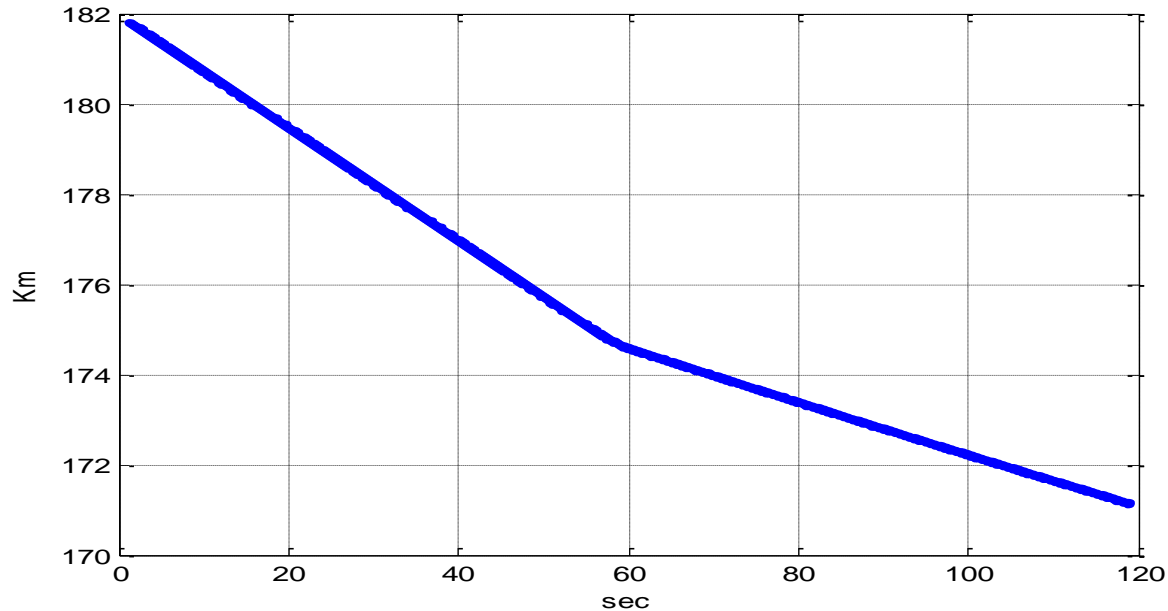
The closer the model matches the dynamics, the better the tracking performance.

Non maneuvering target (straight motion)



$$\mathbf{X}(k + 1) = \mathbf{F}(k) \mathbf{X}(k) + \mathbf{G}(k) \mathbf{W}(k)$$

Maneuvering target



$$\mathbf{X}(k+1) = \mathbf{F}(k)\mathbf{X}(k) + \mathbf{C}(k)\mathbf{U}(k) + \mathbf{G}(k)\mathbf{W}(k)$$



Acceleration (completely unknown)

Dilemma

The closer the model matches the dynamics, the better the tracking performance

Once a model is chosen, the prediction process is determined and does not reflect the reality in case of dynamics deviations

Major Problem

The **major problem** of tracking is prediction of position of maneuvering targets

Standard solution:

Detect the target maneuver

Estimate the strength of the maneuver

Estimate the duration of the maneuver

Correct the model

Novel approach to tracking maneuvering targets

Construct the **model** to be **compatible** with a **broad range of accelerations** which automatically handles both non maneuvering and maneuvering cases

Approach

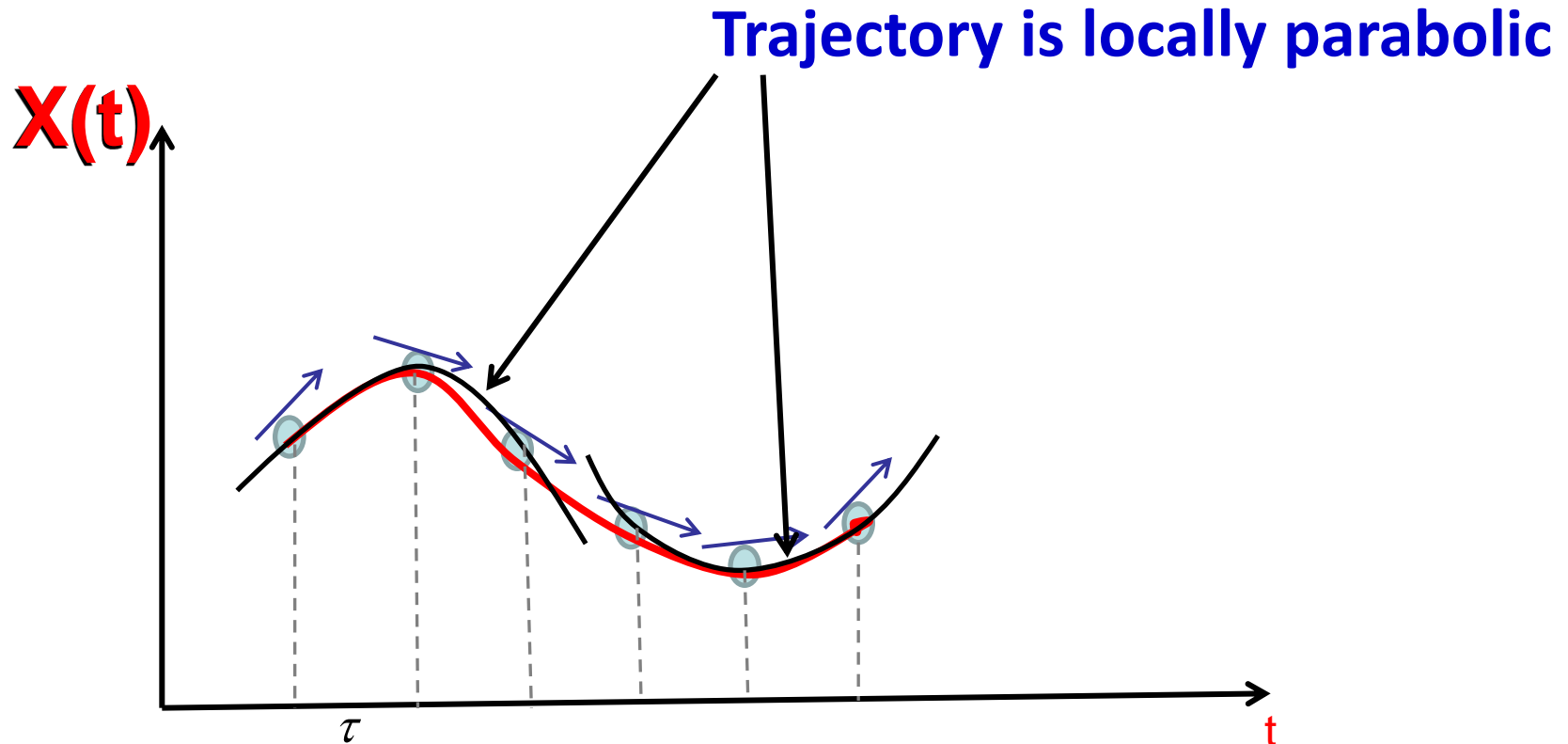
Physical basis:

- Real trajectories are smooth – the first derivatives are continuous

Assumption for our model

- Second derivative for physical systems can be viewed as constant for any three adjacent time samples

Approach (ctd)



Approach (ctd)

$\mathbf{X}(k)$ is the four component vector -

$$\mathbf{X}(k) = \begin{bmatrix} x(k) \\ \dot{x}(k) \\ x(k-1) \\ \dot{x}(k-1) \end{bmatrix} \quad \mathbf{X}(k+1) = \begin{bmatrix} a_1(\tau) & a_2(\tau) & a_3(\tau) & a_4(\tau) \\ b_1(\tau) & b_2(\tau) & b_3(\tau) & b_4(\tau) \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \end{bmatrix} \begin{bmatrix} x(k) \\ \dot{x}(k) \\ x(k-1) \\ \dot{x}(k-1) \end{bmatrix}$$

Approach (ctd)

If the following equations are satisfied

$$\begin{cases} a_1 + a_3 = 1 \\ -a_1\tau + a_2 - 2a_3\tau + a_4 = 0 \\ a_1\frac{\tau}{2} - a_2 + 2a_3\tau - 2a_4 = 0 \end{cases}$$

Then $x(k+1) = a_1(\tau)x(k) + a_2(\tau)\dot{x}(k) + a_3(\tau)x(k-1) + a_4(\tau)\dot{x}(k-1)$

Is correct for any acceleration

Approach (ctd)

Same for the $\dot{x}(k + 1)$

If the following equations are satisfied

$$\begin{cases} b_1 + b_3 = 0 \\ -b_1\tau + b_2 - 2b_3\tau + b_4 = 1 \\ b_1\frac{\tau}{2} - b_2 + 2b_3\tau - 2b_4 = 0 \end{cases}$$

Then $\dot{x}(k + 1) = b_1(\tau)x(k) + b_2(\tau)\dot{x}(k) + b_3(\tau)x(k - 1) + b_4(\tau)\dot{x}(k - 1)$

Is correct for any acceleration

Approach (ctd)

Deterministic relations

$$\begin{cases} X(1, k+1) = A^T X(1, k) \\ X(2, k+1) = B^T X(1, k) \end{cases}$$

$$A = \begin{array}{c|c|c|c|c} -1 & 0 & 1 & & \\ \tau/2 & 3/2 & 0 & & \\ 1 & 0 & 0 & & \\ \tau/2 & -1/2 & 0 & & \end{array} \begin{array}{c} a_3 + \\ \\ \tau + \\ \end{array}$$

$$B = \begin{array}{c|c|c|c} -1 & 0 & & \\ \tau/2 & 2 & & \\ 1 & 0 & & \\ \tau/2 & -1 & & \end{array} \begin{array}{c} b_3 + \\ \\ \\ \end{array}$$

Approach (ctd)

Prediction accuracies

$$J_A = A^T \mathbf{P} A = \begin{vmatrix} -1 & 0 & 1 \\ \tau/2 & 3/2 & 0 \\ 1 & 0 & 0 \\ \tau/2 & -1/2 & 0 \end{vmatrix}^{T} \mathbf{P} \begin{vmatrix} -1 & 0 & 1 \\ \tau/2 & 3/2 & 0 \\ 1 & 0 & 0 \\ \tau/2 & -1/2 & 0 \end{vmatrix}$$

$$J_B = B^T \mathbf{P} B = \begin{vmatrix} -1 & 0 \\ \tau/2 & 2 \\ 1 & 0 \\ \tau/2 & -1 \end{vmatrix}^{T} \mathbf{P} \begin{vmatrix} -1 & 0 \\ \tau/2 & 2 \\ 1 & 0 \\ \tau/2 & -1 \end{vmatrix}$$

Approach (ctd)

Finding the minimum with respect to a_3, b_3

$$\frac{\partial J}{\partial a_3} = \begin{bmatrix} -1 \\ \tau/2 \\ 1 \\ \tau/2 \end{bmatrix}^T \mathbf{P} \begin{bmatrix} -1 \\ \tau/2 \\ 1 \\ \tau/2 \end{bmatrix} a_3 + \begin{bmatrix} 0 \\ 3/2 \\ 0 \\ -1/2 \end{bmatrix} \tau + \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix} = 0$$

$$\frac{\partial J}{\partial b_3} = \begin{bmatrix} -1 \\ \tau/2 \\ 1 \\ \tau/2 \end{bmatrix}^T \mathbf{P} \begin{bmatrix} -1 \\ \tau/2 \\ 1 \\ \tau/2 \end{bmatrix} b_3 + \begin{bmatrix} 0 \\ 2 \\ 0 \\ -1 \end{bmatrix} = 0$$

Approach (ctd)

Test for minimum

$$\frac{\partial^2 J}{\partial (a_3)^2} = \begin{vmatrix} -1 \\ \tau/2 \\ 1 \\ \tau/2 \end{vmatrix}^T \mathbf{P} \begin{vmatrix} -1 \\ \tau/2 \\ 1 \\ \tau/2 \end{vmatrix} \geq 0$$

$$\frac{\partial^2 J}{\partial (b_3)^2} = \begin{vmatrix} -1 \\ \tau/2 \\ 1 \\ \tau/2 \end{vmatrix}^T \mathbf{P} \begin{vmatrix} -1 \\ \tau/2 \\ 1 \\ \tau/2 \end{vmatrix} \geq 0$$

Simulation Results

Proposed Algorithm Response

Target with maneuver of 6 sec duration

Maximum range acceleration 2g

Constant sample rate 0.5 Hz

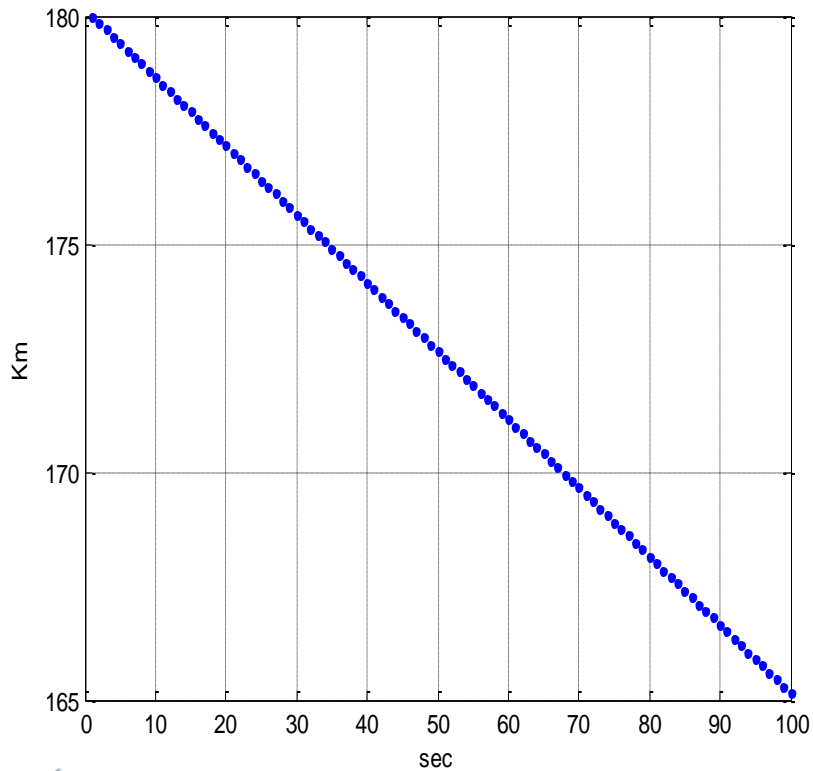
Measurement range accuracy 16 m

Track rms accuracy computed from 100 Monte-Carlo runs

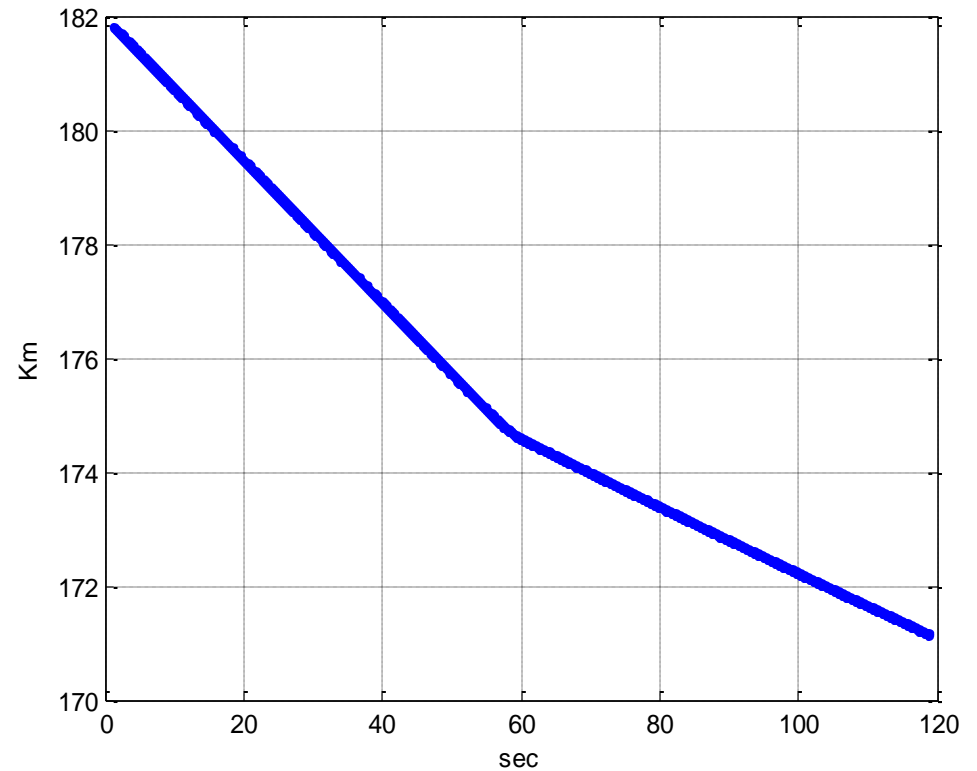
Simulation Results

Proposed Algorithm Response

Non-maneuvering target range



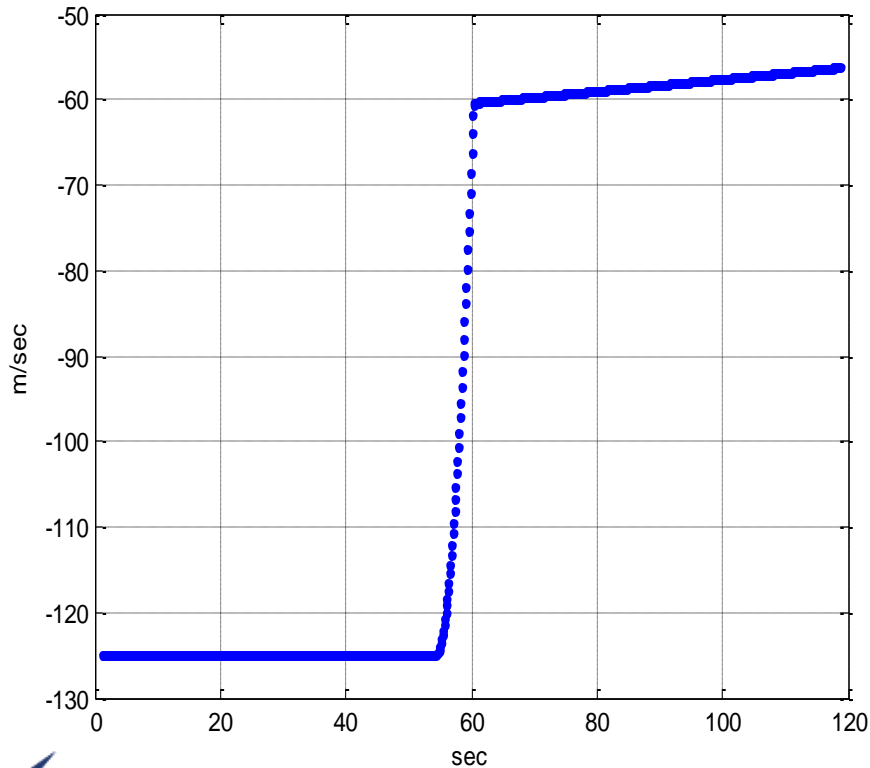
Maneuvering target range



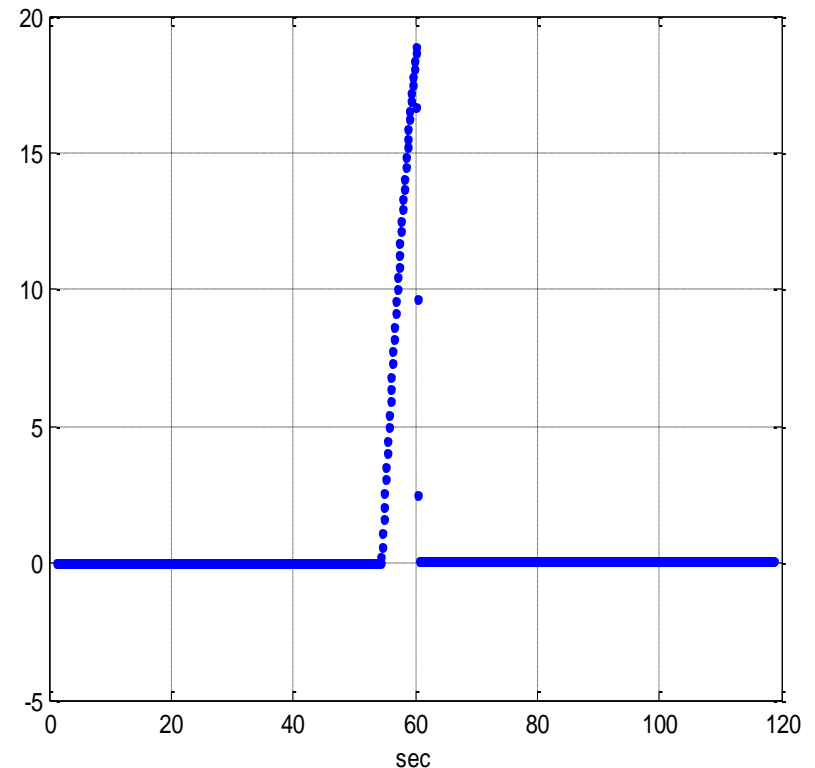
Simulation Results

Proposed Algorithm Response

Maneuvering target range rate



Maneuvering target range acceleration

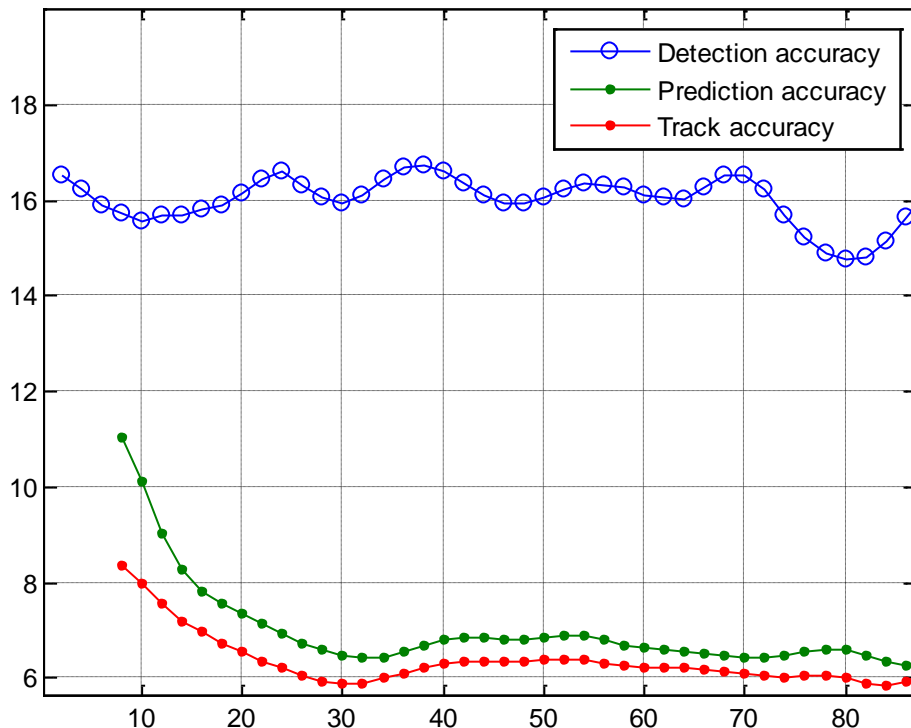


Simulation Results

Proposed Algorithm Response

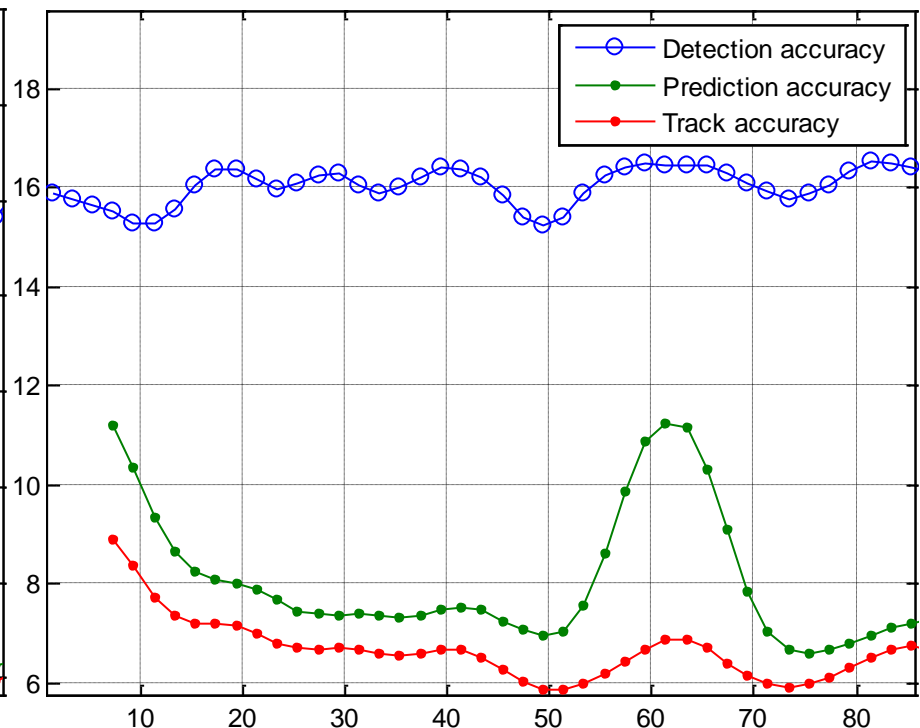
Non-maneuvering target

Range accuracy (m) vs time (sec)



Maneuvering target

Range accuracy (m) vs time (sec)



How to proceed ?

Possible research directions:

Tracking with 6-component state vector

Target jerk estimation

Dynamic sample rate

Thank you !!