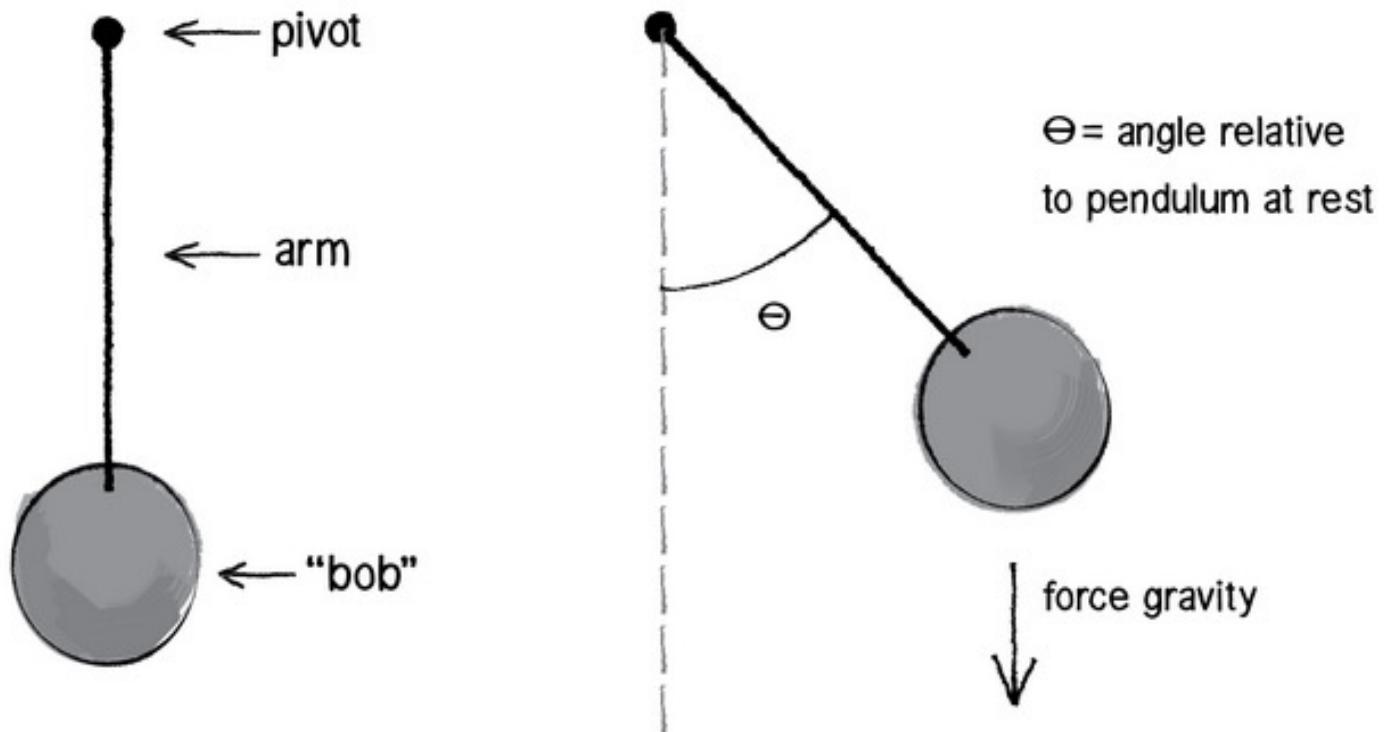


# Pendulum State Estimation using Kalman Filtering & State Space Analysis

Prepared by:  
Nishantkumar V Patel

# Introduction:



The above picture is of simple pendulum with its different components which later on used in the simulink physical model from Mathworks.

The angle theta is the angle which is ultimately the state which we need to estimate and also the velocity during the pendulum oscillations.

The angle theta is our case considered at initial condition.

# Parameter Values:

```
1
2
3 % gravity
4 g = 9.81; % (m/s^2)
5 % Pendulum mass
6 m = 2; % (kg)
7 % Pendulum length
8 l = 0.75; % (m)
9 % Process noise covariance
10 Q = 1e-3;
11 % Measurement noise covariance
12 R = 1e-4;
13 % Sampling time
14 Ts = 0.01; % (sec)
15
16
17
18
19 % State Space Matrices
20 % A
21 system_matrix= [0 1; -g/l 0];
22 % B
23 input_matrix = [0; 1/(m*l^2)];
24 % C
25 output_matrix = [1 0];
26 % D
27 direct_transmission_matrix = 0;
28
29
30
31
32
33
34
```

# State Space Analysis for a Pendulum system behaviour

→ The simple pendulum system with no friction can be replaced by below equation of motion.

$$\boxed{J \frac{d^2\theta}{dt^2} + mgl \sin\theta = \tau} \quad \textcircled{X}$$

Where,  $J = ml^2$   
moment of inertia

→ To make the equation  $\textcircled{X}$  into linear differential equation.

$$\frac{d^2\theta}{dt^2} + \frac{mgl \sin\theta}{ml^2} = \frac{\tau}{ml^2} \quad (\because \text{divided by } J = ml^2)$$

$$\boxed{\therefore \frac{d^2\theta}{dt^2} + \frac{g}{l} \sin\theta = \frac{\tau}{ml^2}} \quad \textcircled{Y}$$

Since the  $\theta$  is very small  
 $\therefore \sin\theta \approx \theta$

→ The state space representation,

Assume,

$\theta = \text{output}$  - we want  
 $\tau = \text{input}$  we applied.

$$\Rightarrow \theta = y = x_1$$

$$\Rightarrow x_2 = \dot{x}_1 = \dot{\theta} = (\frac{d\theta}{dt})$$

$$\Rightarrow \dot{x}_2 = \ddot{x}_1 = \ddot{\theta} = (\frac{d^2\theta}{dt^2})$$

$$\Rightarrow u = \tau$$

$$\Rightarrow \frac{\tau}{ml^2} - \frac{g}{l} \theta = \frac{d^2\theta}{dt^2}$$

$$\dot{x}_2 = \frac{1}{ml^2}(u) - \frac{g}{l}x_1$$

→ State-Space Equation

$$\boxed{\begin{aligned} \therefore \dot{x} &= Ax + Bu \\ \therefore y &= Cx + Du \end{aligned}} \quad (\text{where } D=0)$$

$$\rightarrow \begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ -\frac{g}{l} & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \begin{bmatrix} 0 \\ \frac{1}{ml^2} \end{bmatrix} \tau$$

$$[y] = [1 \ 0] \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + 0$$

∴ So, we get the A, B, C and D matrices.

$$\therefore A = \begin{bmatrix} 0 & 1 \\ -\frac{g}{m} & 0 \end{bmatrix}, \quad \therefore B = \begin{bmatrix} 0 \\ \frac{1}{mL^2} \end{bmatrix}$$

$$\therefore C = [1 \ 0], \quad \therefore D = 0.$$

The state space matrices A, B, C and D are later on used for the block parameters of Kalman filter by selecting the State Space analysis method.

**Block Parameters: Linear Kalman Filter**

**Kalman Filter**

Estimate the states of a discrete-time or continuous-time linear system. Time-varying systems are supported.

**Filter Settings**

Time domain: **Continuous-Time**

**Model Parameters**    **Options**

**System Model**

Model source: Individual A, B, C, D matrices

A: `system_matrix`    B: `input_matrix`

C: `output_matrix`    D: `direct_transmission_matrix`

**Initial Estimates**

Source: Dialog

Initial states  $x(0)$ : `[pi/16 0]`

**Noise Characteristics**

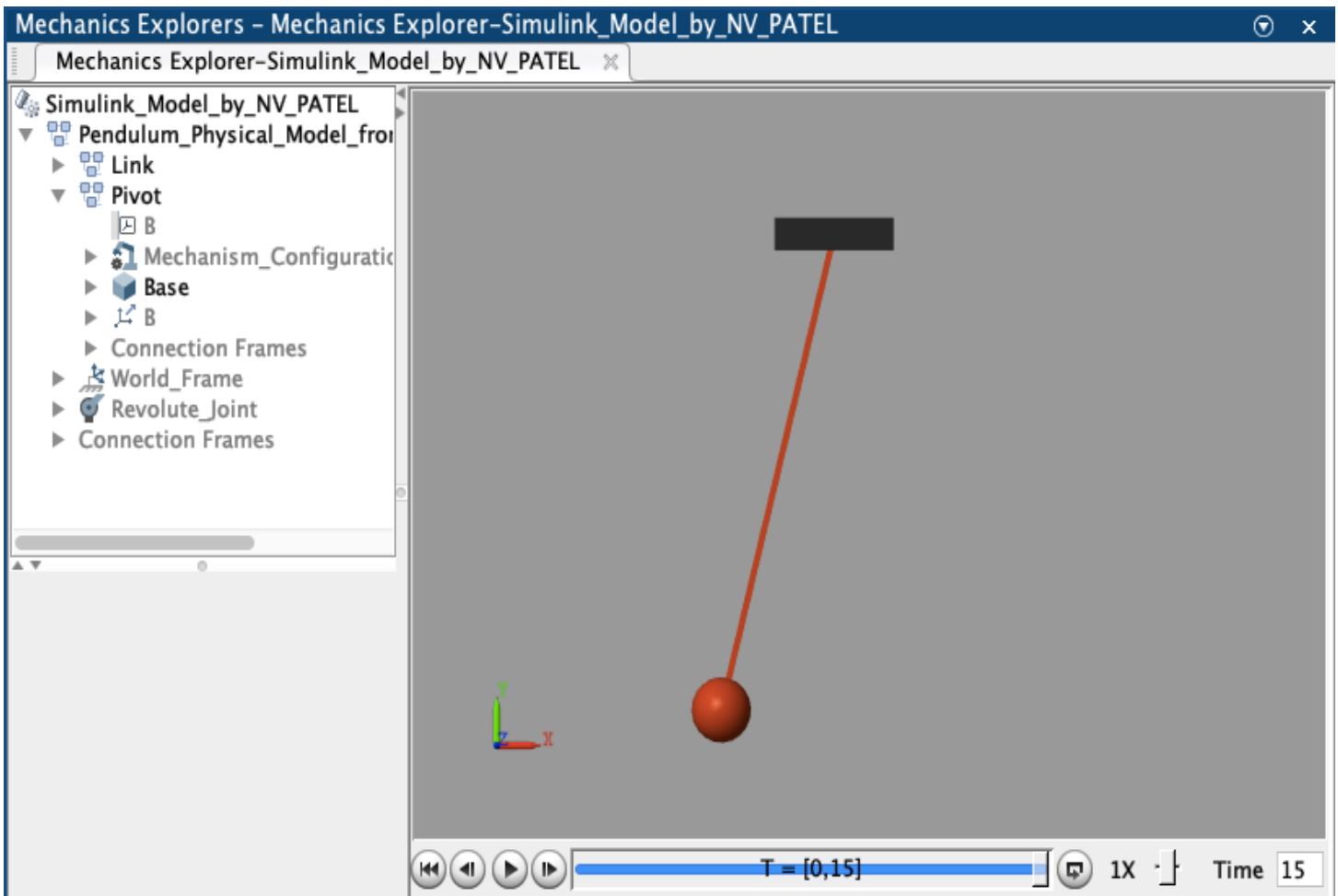
Use G and H matrices (default  $G=I$  and  $H=0$ )

Q: `diag([0,Q])`     Time-Invariant Q

R: `R`     Time-Invariant R

N: `0`     Time-Invariant N

**OK**    **Cancel**    **Help**    **Apply**



# 1. With Linear Kalman Filter:

Using linear kalman filter we can only estimate the position of pendulum with small value of angle at initial condition.

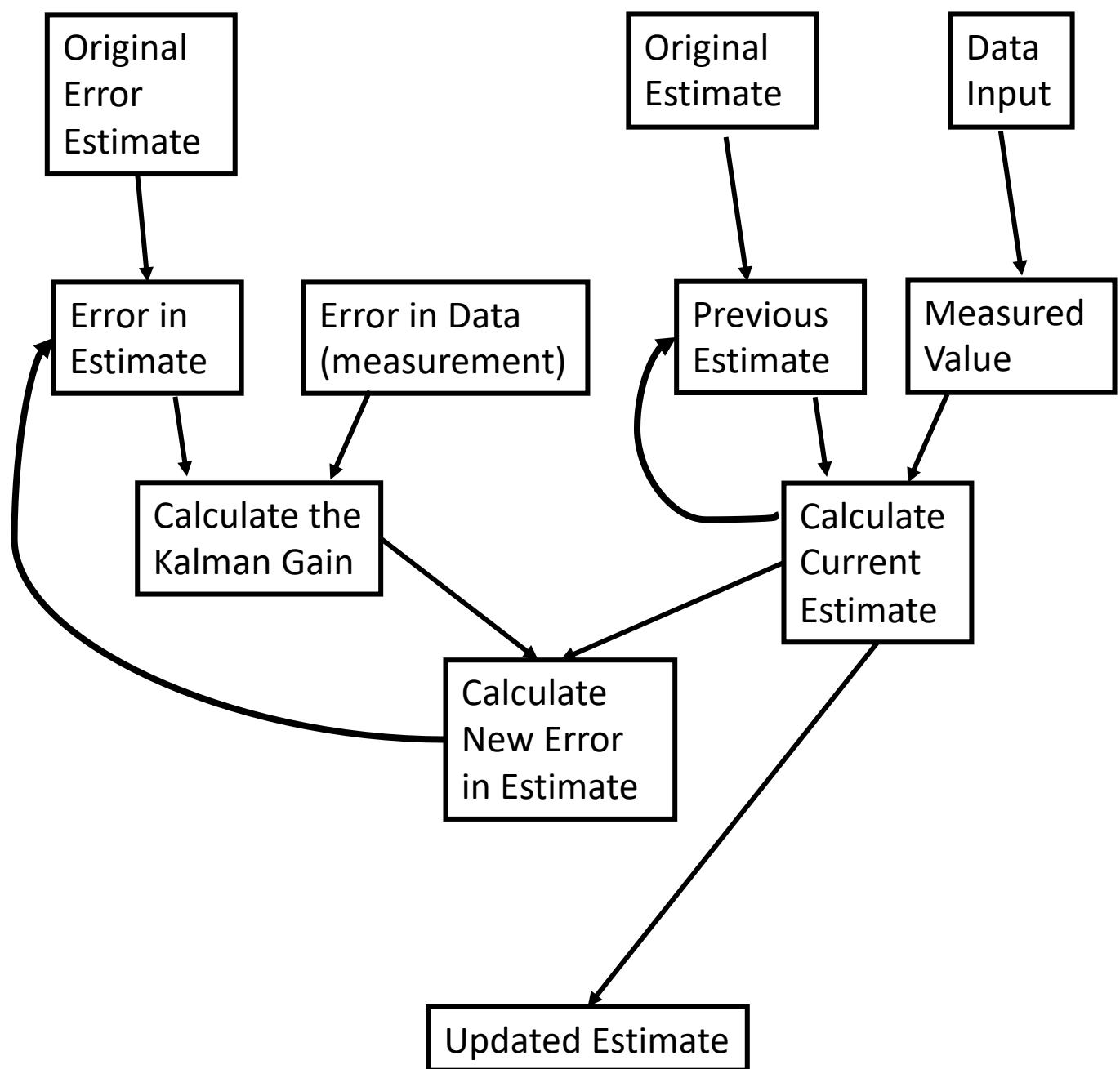
As in this case I used the initial condition of angle theta equal to 100.

For larger amount of angle theta linear kalman filter is not able to track the precise position of pendulum over the period of time.

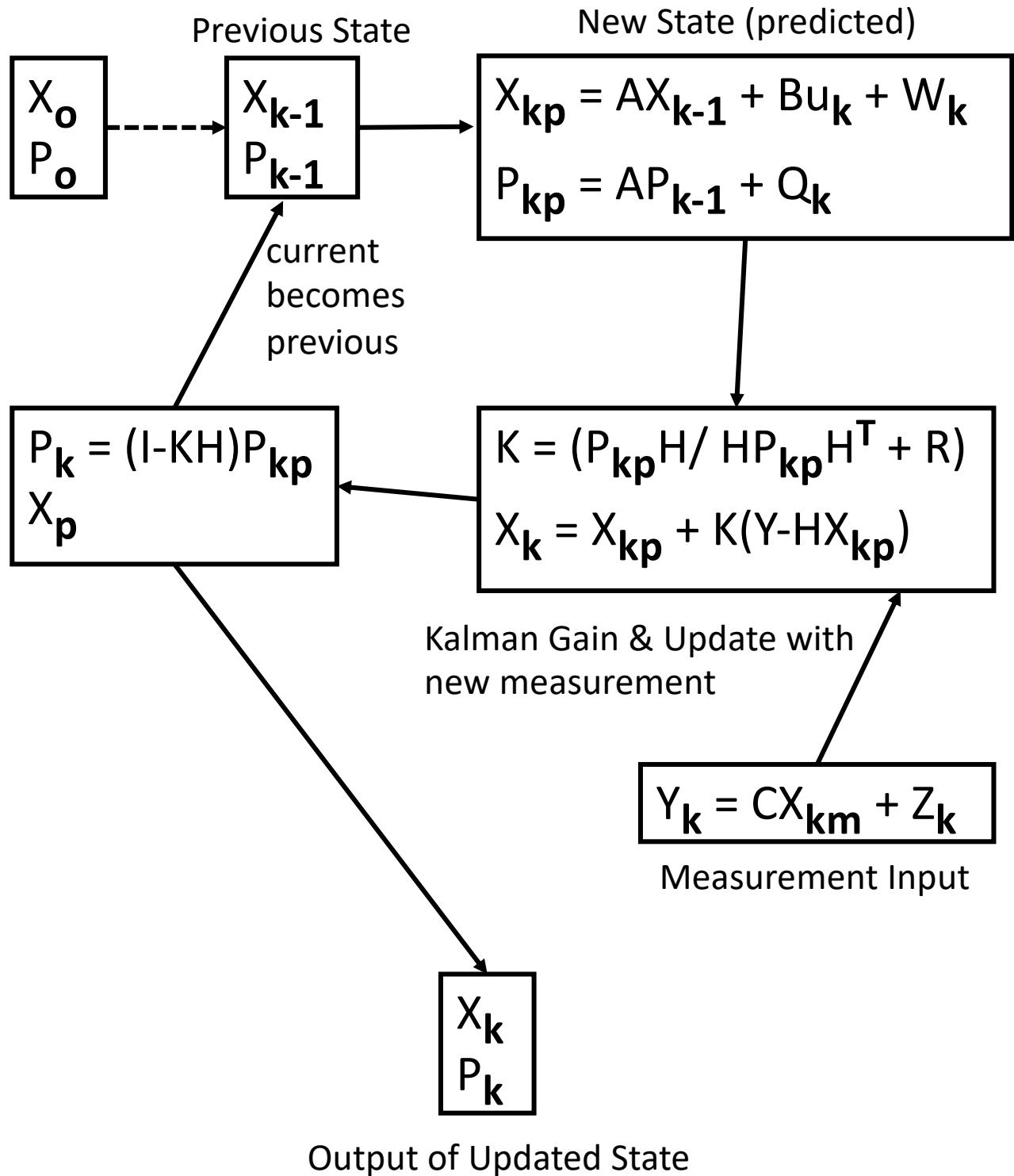
Suppose, I would take the larger angle for example theta of value 300.

I will compare the both results with small and large theta using linear kalman filter.

Basically how the kalman filter actually works are described as below using process flow chart.

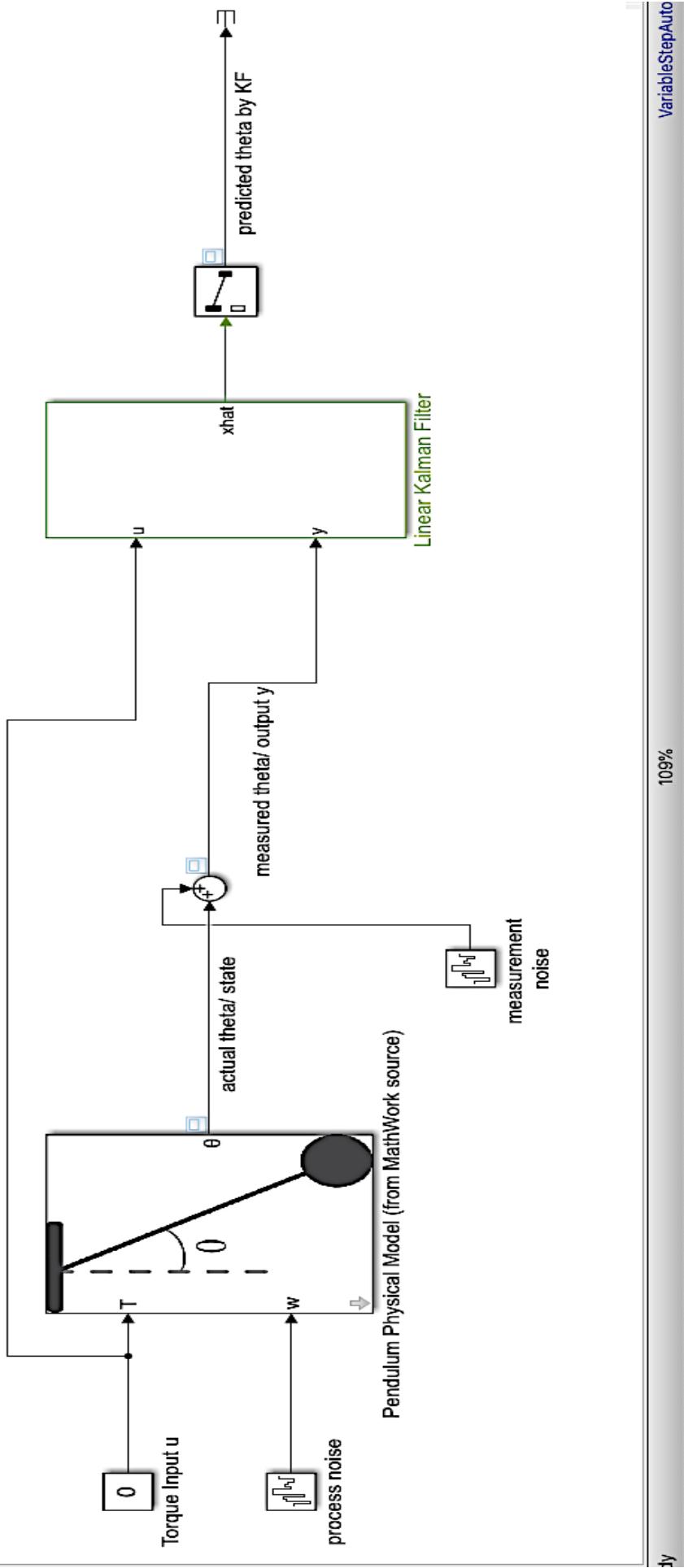


# Kalman Filtering Equations



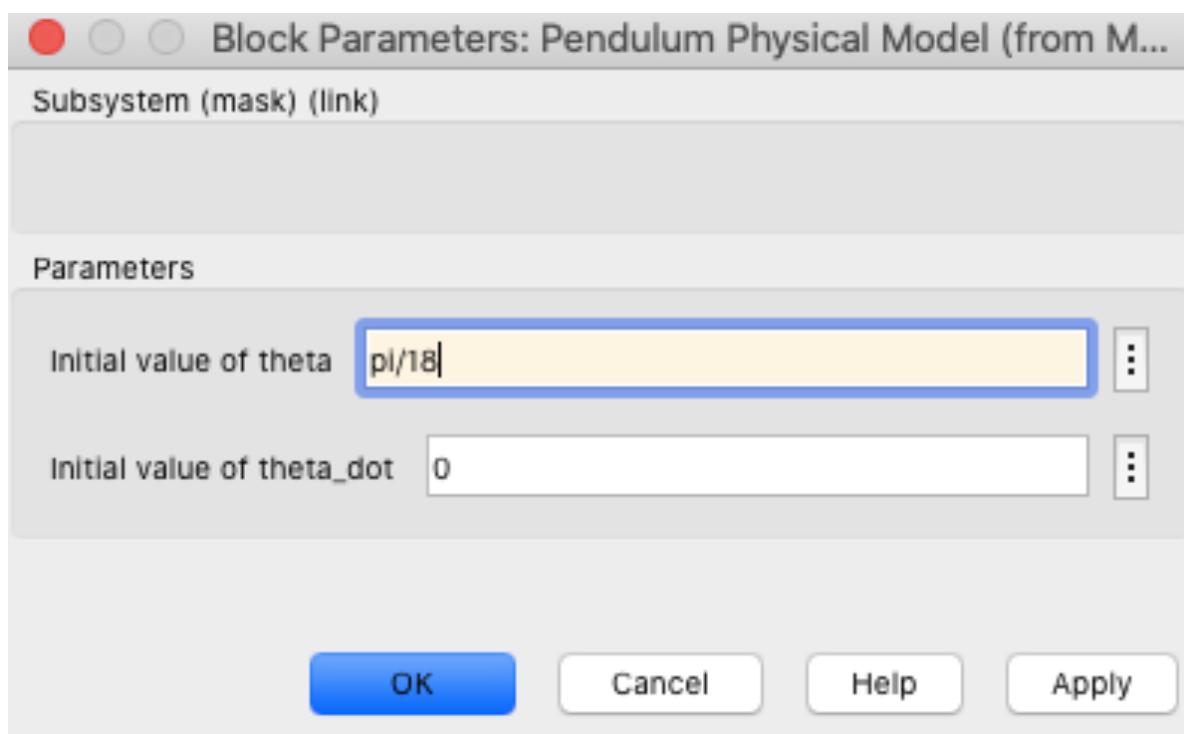
# Simulink Model (linear KF):

Pendulum State Estimation by Kalman Filtering and State Space Analysis  
Modelling by Nishantkumar V Patel

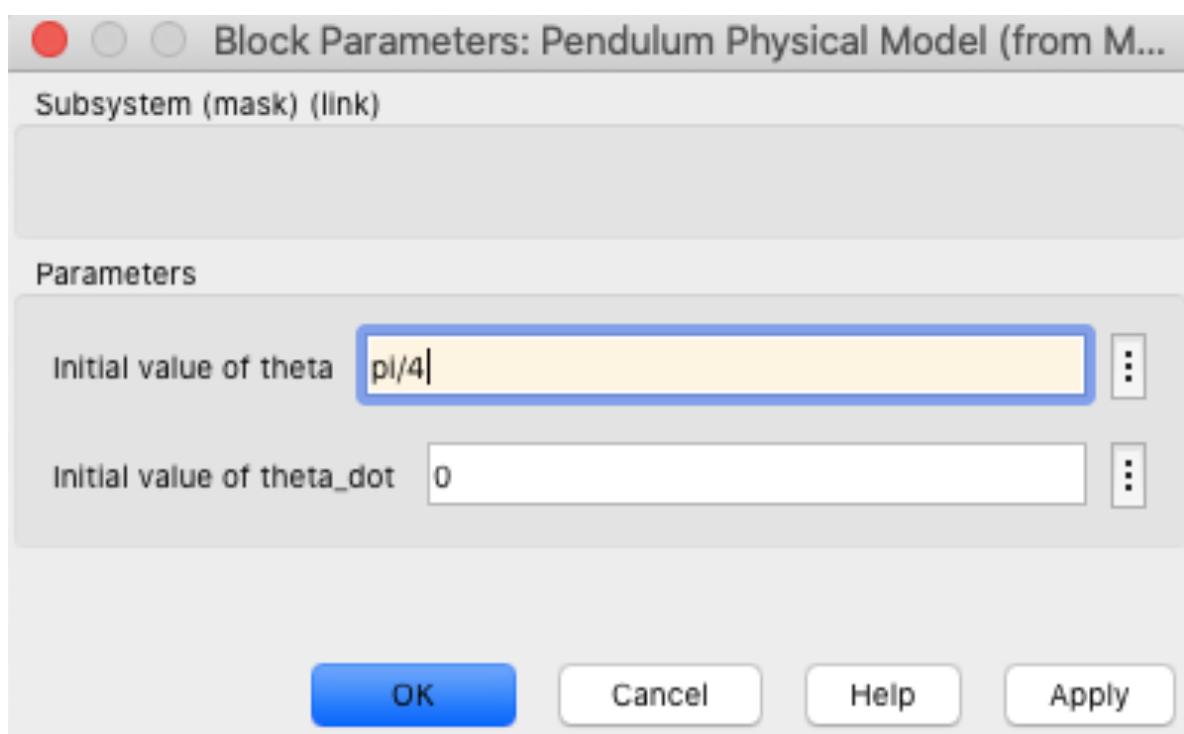


# Pendulum Block Parameters with Small & Large Angle at Initial Condition:

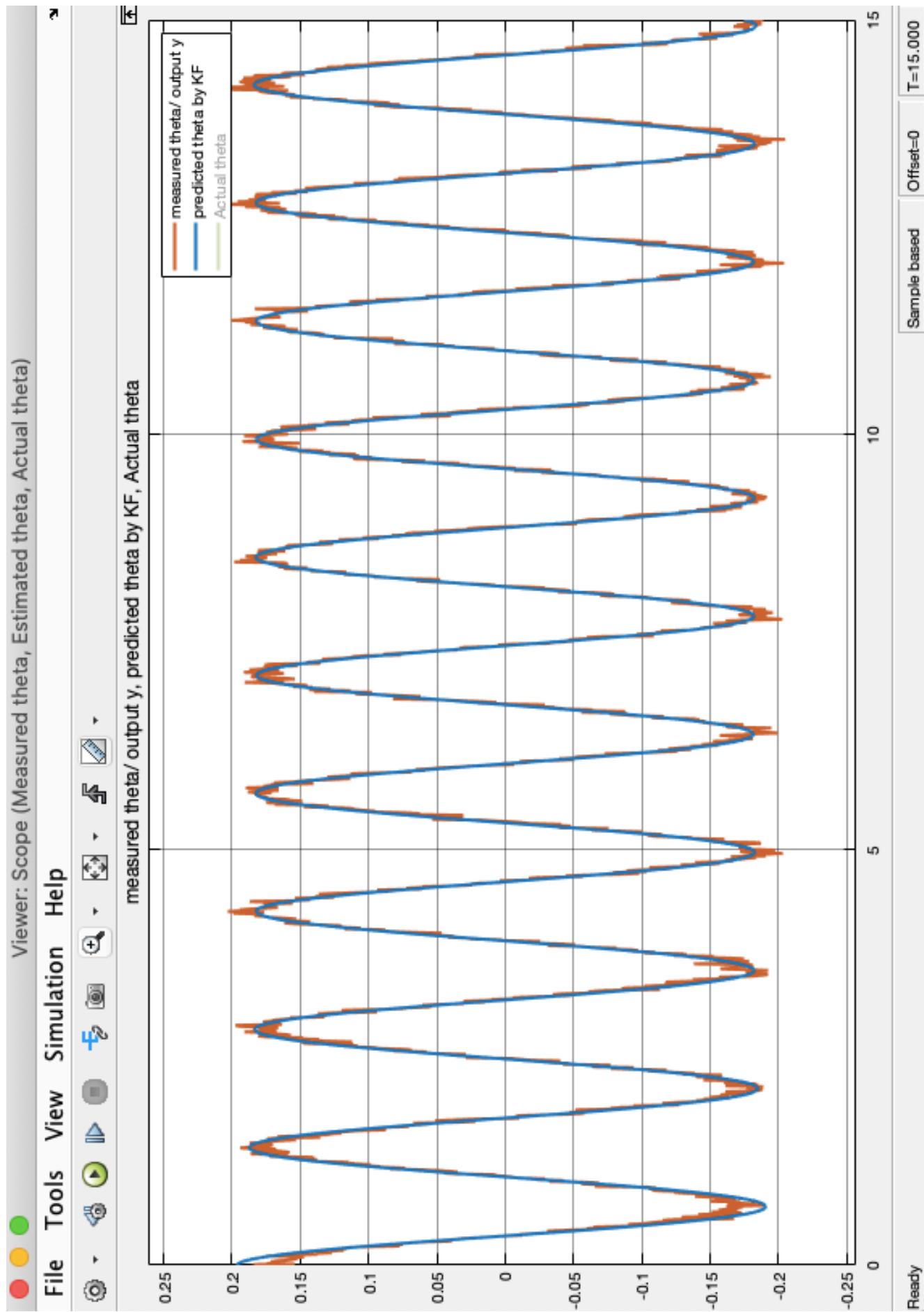
With small angle 10°:



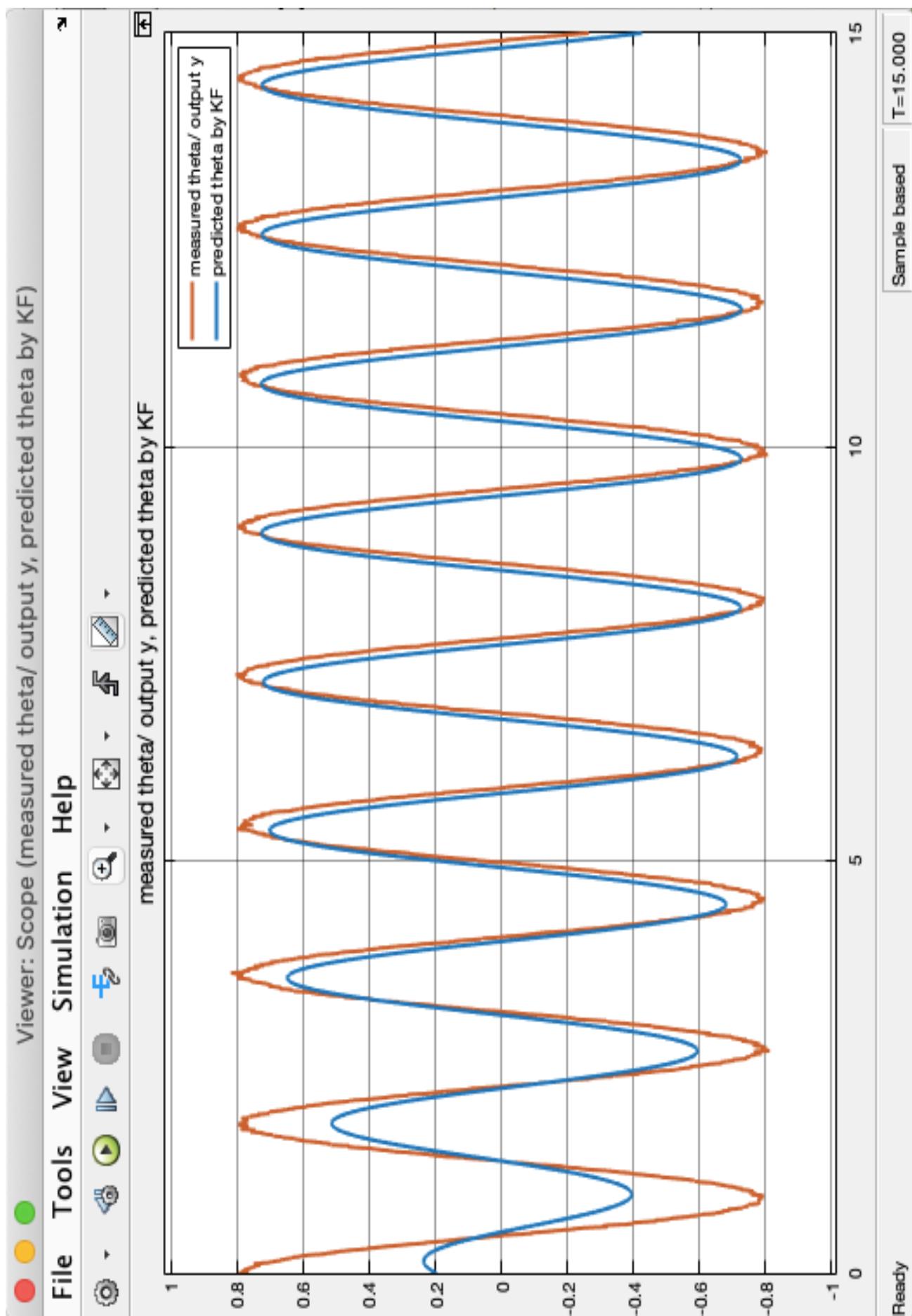
With large angle 45°:



# Results Comparison (with small angle):



# Results Comparison (with large angle):



## 2. With Extended Kalman Filter:

### **Why use?**

The extended kalman filter eliminates the inefficiency of linear kalman filter and it can be used for the large angle values.

Extended kalman filter is able to track the non linear behaviour of the system quite well.

I will show the same value of angle 300 using extended kalman filter and we will compare the scope graph that how fairly it can predict the states of pendulum.

# Block & Pendulum Parameters

Block Parameters: Extended Kalman Filter.

Extended Kalman Filter

Discrete-time extended Kalman filter. Estimate states of a nonlinear plant model. Use Simulink Function blocks or .m MATLAB Functions to specify state transition and measurement functions.

See block help for function syntaxes, which depend on if noise is additive or nonadditive.

System Model Multirate

State Transition

Function: myStateTransitionFcn  Jacobian

Process noise: Additive Covariance:  $\text{diag}([0 \ Q])$   Time-varying

Initialization

Initial state:  $[\pi/2 \ 0]$  Initial covariance:  $1e-6$

Measurement 1

Function: myMeasurementFcn  Jacobian  Add Enable port

Measurement noise: Additive Covariance: R  Time-varying

Has measurement wrapping

Add Measurement Remove Measurement

Settings

Use the current measurements to improve state estimates  
 Output state estimation error covariance

Data type: double

Sample time: Ts

OK Cancel Help Apply

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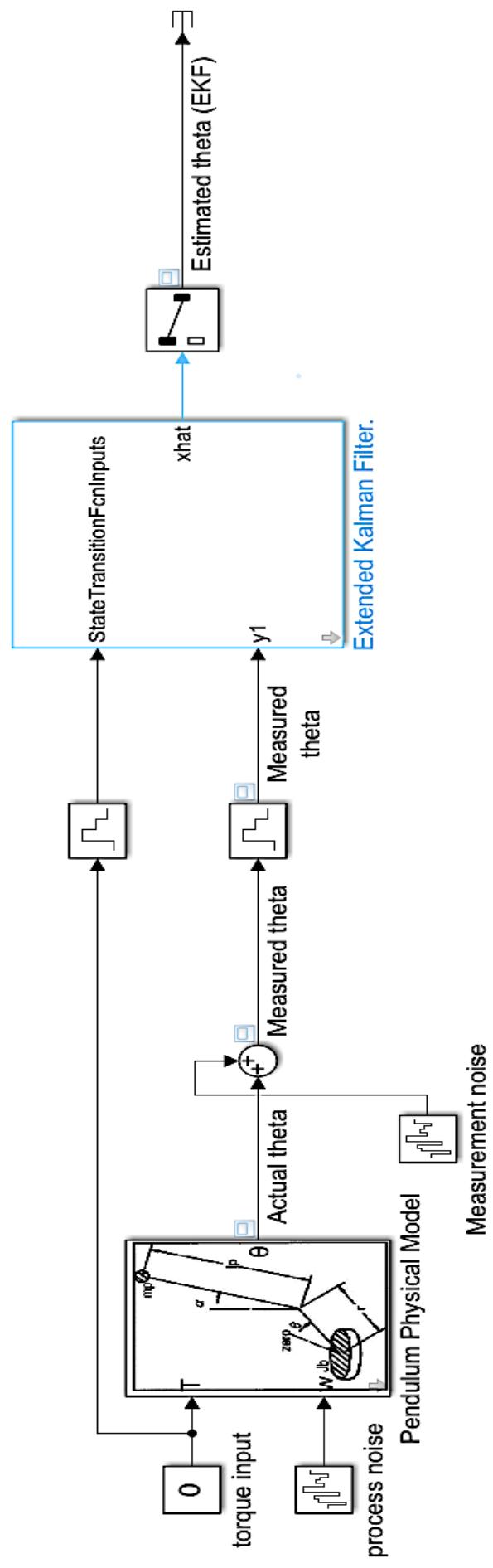
myStateTransitionFcn.m \* myMeasurementFcn.m +

```
1
2 function x = myStateTransitionFcn(x,u)
3 % Sample time [s]
4 dt = 0.01;
5
6 x = x + [x(2); -9.81/0.5*sin(x(1)) + u]*dt;
7 end
8
9
```

# Simulink Model (extended KF):

Simulink\_Model\_EKF\_by\_NV\_PATEL

Pendulum State Estimation (with large angle) using Extended Kalman Filter  
Modelling by  
Nishantkumar V Patel



# Result Comparison:

