

# AE3M33MKR

## Kalman Filter

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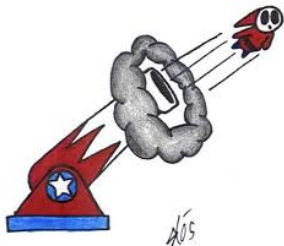
# Assignment

We will use a common physics problem.

This assignment will involve firing a ball from a cannon at an unknown angle and at an unknown velocity.

However, we will have a camera that will record the ball's position from the side. The positions measured from the camera have significant a measurement error.

We want to know the position of the ball as precisely as possible.



## Mathematical Description

The position  $\mathbf{x} = (x, y)$  of the ball is computed from its initial position,  $\mathbf{x} = (x_0, y_0)$ , its initial velocity  $\mathbf{v}_0 = (v_{0x}, v_{0y})$  and the gravitational acceleration  $g$ :

$$\begin{aligned}x(t) &= x_0 + v_{0x}t \\y(t) &= y_0 - v_{0y}t - \frac{1}{2}gt^2 \\v_x(t) &= v_{0x} \\v_y(t) &= v_{0y} - gt\end{aligned}\tag{1}$$



# Mathematical Description

In the discrete case the system can be described by:

$$\begin{aligned}x_n &= x_{n-1} + v_{n-1_x} \Delta t \\y_n &= y_{n-1} - v_{n-1_y} \Delta t - \frac{1}{2} g \Delta t^2 \\v_{n_x} &= v_{n-1_x} \\v_{n_y} &= v_{n-1_y} - g \Delta t\end{aligned}\tag{2}$$



# State description

$$x_n = \mathbf{A}x_{n-1} + \mathbf{B}u \quad (3)$$

$u$  is the control input, in this case the control input is the influence of the gravitational acceleration.



# Kalman filter

used constants:

**A:** State transition matrix. Basically, multiply state by this and add control factors, and you get a prediction of the state for the next time step.

**B:** Control matrix. This is used to define the linear relation between control and state.

**H:** Observation matrix. Multiply a state vector by  $H$  to translate it into a measurement vector.

**Q:** Covariance of the estimated state error.

**R:** Covariance of the estimated measurement error.



# Kalman filter

## Inputs:

$u$ : Control vector. Constant in this case.

$z_n$ : Measurement vector. Position of the cannon ball measured in this time step. Measured values contain noise.

## Outputs:

$x_n$ : Newest estimate of the current "true" state.

$P_n$ : Newest covariance of the estimate of the average error for each parts of the state.



## Kalman filter - How to compute it

Predict where the cannon ball is going to be:

$$x_{predicted} = \mathbf{A}x_{n-1} + \mathbf{B}u \quad (4)$$

Predict the covariance of the error:

$$P_{predicted} = \mathbf{A}P_{n-1}\mathbf{A}^T + \mathbf{Q} \quad (5)$$

Compare the reality against the prediction, we call it innovation:

$$\tilde{y} = z_n - \mathbf{H}x_{predicted} \quad (6)$$

Compute the covariance of the innovation:

$$S = \mathbf{H}P_{predicted}\mathbf{H}^T + \mathbf{R} \quad (7)$$





## Kalman filter - How to compute it, cont.

Now the key part, the Kalman Gain:

$$K = \mathbf{P}_{predicted} \mathbf{H}^T \mathbf{S}^{-1} \quad (8)$$

and finally we update the state:

$$x_n = x_{predicted} + \mathbf{K} \tilde{y} \quad (9)$$

and the covariance of the state:

$$P_n = (\mathbf{I} - \mathbf{K}\mathbf{H})\mathbf{P}_{predicted}, \text{ with } \mathbf{I} \text{ the identity matrix} \quad (10)$$



## code explanation

Initialization of the matrices, the size of the state vector is 4 and the size of measurement vector is 2.

```
Matrix4f A,B,P,Q;  
Matrix2f R;  
Matrix<float,n,m> H;  
Vector4f x;  
float dt = 0.1;
```

The matrices can be filled like this:

```
P << 1, 0, 0, 0,  
     0, 1, 0, 0,  
     0, 0, 1, 0,  
     0, 0, 0, 1;
```



## Code explanation, cont.

Initialization of the GUI and the system simulator:

```
Gui gui;  
System system;
```

A point is defined by two co-ordinates  $x$  and  $y$ .

Used variables:

**measurement** holds the measured position of the ball

**truth** holds the true position (used for painting)

**kfPosition** holds the position estimated by the Kalman filter



## Code explanation, cont.

The main loop simulates one step of the system, stores the true and measured positions in the appropriate variables

```
for(int i=1; i<nSteps; i++) {  
    system.makeStep();  
    truth = system.getTruthPosition();  
    measurement = system.getMeasurement();
```

estimates the position by Kalman filter

```
    kfPosition = KalmanFilter(measurement);
```

and plots all the positions by calling

```
    gui.setPoints(truth, measurement, kfPosition);  
}  
gui.startInteractor();
```



# Assignment

Your task is to implement a Kalman filter in the KalmanFilter function:

```
Point KalmanFilter(const Point measuredPosition )  
{  
  
}
```

which returns the estimated position as a variable of type Point.

