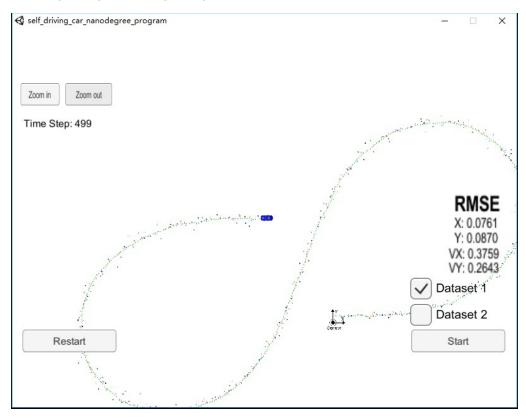
# **Unscented Kalman Filter Project**

Self-Driving Car Engineer Nanodegree Program



# **Overview**

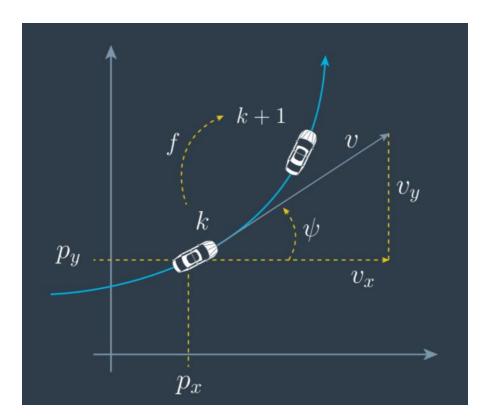
In this project utilize an Unscented Kalman Filter to estimate the state of a moving object of interest with noisy lidar and radar measurements. Passing the project requires obtaining RMSE values [px,py,vx,vy] that are less than or equal to the values [0.09,0.10,0.40,0.30].

This project involves the Term 2 Simulator which can be downloaded here

# Write Up

# **Motion Model**

In this project,we will work with the CTRV (Constant Turn Rate and Velocity magnitude model) model.



CTRV Model State Vector : 
$$x_k = \begin{bmatrix} p_x \\ p_y \\ v \\ \psi \\ \dot{\psi} \end{bmatrix}$$

### CTRV Model Equation:

$$\dot{p_x} = v_x = cos(\psi) \cdot v$$

$$\dot{p_y} = v_y = sin(\psi) \cdot v$$

$$x_{k+1} = x_k + \int_{t_k}^{t_{k+1}} egin{bmatrix} \dot{p}_x(t) \ \dot{p}_y(t) \ \dot{v}(t) \ \dot{\psi}(t) \ \ddot{\psi}(t) \ \ddot{\psi}(t) \end{bmatrix} dt = x_k + egin{bmatrix} \int_{t_k}^{t_{k+1}} \cos(\psi(t)) \cdot v(t) dt \ \int_{t_k}^{t_{k+1}} \sin(\psi(t)) \cdot v(t) dt \ 0 \ \dot{\psi}_k \Delta t \ 0 \end{bmatrix} = x_k + egin{bmatrix} v_k \int_{t_k}^{t_{k+1}} \cos(\psi_k + \dot{\psi}_k(t-t_k)) dt \ v_k \int_{t_k}^{t_{k+1}} \sin(\psi_k + \dot{\psi}_k(t-t_k)) dt \ 0 \ \dot{\psi}_k \Delta t \ 0 \end{bmatrix}$$

*Note:* We assume that velocity (v) and turn rate ( $\dot{\psi}$  ) are constant.

If 
$$\dot{\psi}_{\scriptscriptstyle k} 
eq 0$$
 ,then

$$x_{k+1} = x_k + egin{bmatrix} rac{v_k}{\dot{\psi}_k} \left( sin(\psi_k + \dot{\psi}_k \Delta t) - sin(\psi_k) 
ight) \ rac{v_k}{\dot{\psi}_k} \left( -cos(\psi_k + \dot{\psi}_k \Delta t) + cos(\psi_k) 
ight) \ 0 \ \dot{\psi}_k \Delta t \ 0 \end{bmatrix}$$

If 
$$\dot{\psi}_k=0$$
 , then

$$x_{k+1} = x_k + egin{bmatrix} v_k cos(\psi_k) \Delta t \ v_k sin(\psi_k) \Delta t \ 0 \ 0 \ 0 \end{bmatrix}$$

Process Noise Vector : 
$$\nu_k = \begin{bmatrix} \nu_{a,k} \\ \nu_{\vec{\psi},k} \end{bmatrix}$$

#### Process Model:

$$x_{k+1} = x_k + egin{bmatrix} rac{v_k}{\dot{\psi}_k} \left( sin(\psi_k + \dot{\psi}_k \Delta t) - sin(\psi_k) 
ight) \\ rac{v_k}{\dot{\psi}_k} \left( -cos(\psi_k + \dot{\psi}_k \Delta t) + cos(\psi_k) 
ight) \\ 0 \\ \dot{\psi}_k \Delta t \\ 0 \end{bmatrix} + egin{bmatrix} rac{rac{1}{2} \left( \Delta t 
ight)^2 cos(\psi_k) \cdot 
u_{a,k}}{rac{1}{2} \left( \Delta t 
ight)^2 sin(\psi_k) \cdot 
u_{a,k}}{\Delta t \cdot 
u_{a,k}} \\ rac{1}{2} \left( \Delta t 
ight)^2 \cdot 
u_{\ddot{\psi},k} \\ \Delta t \cdot 
u_{\ddot{\psi},k} \end{bmatrix}$$

## **UKF Roadmap**

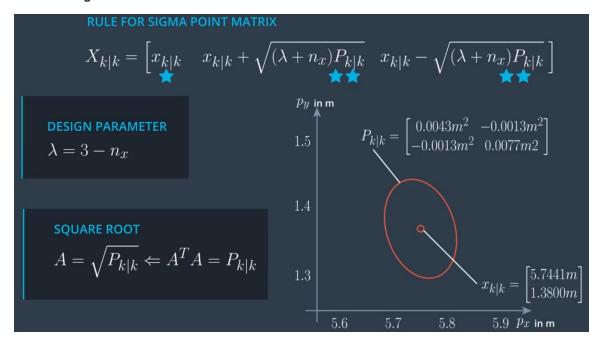
#### Prediction Step:

- · Generate Sigma Points
- · Predict Sigma Points
- · Predict Mean and Covariance

#### **Update Step:**

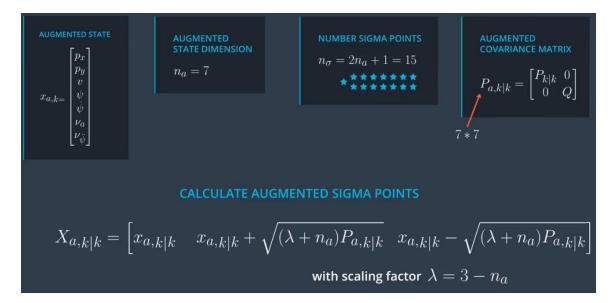
- Predict Measurement
- · Update State

## **Generate Sigma Points**



- Posterior State :  $x_{k|k}$
- Posterior Covariance Matrix : $P_{k|k}$
- State Vector :  $x = \begin{bmatrix} p_x \\ p_y \end{bmatrix}$
- State Dimension :  $n_x=2$
- Number Sigma Points :  $n_{\sigma}=2n_{x}+1=7$

#### **UKF Augmention**:

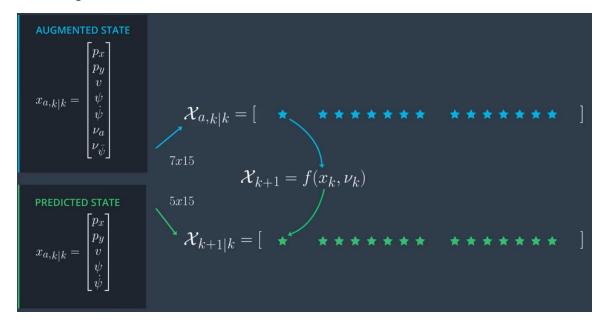


- Posterior Distribution :  $x_{k|k}, P_{k|k}$ 

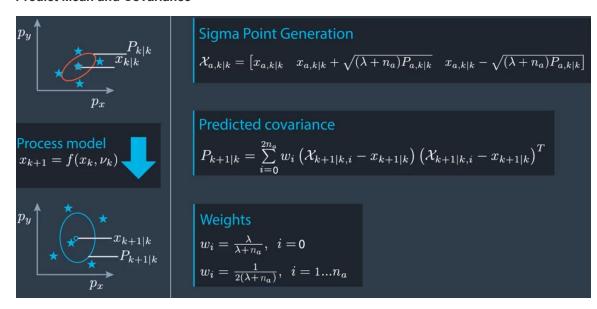
• Process Noise : 
$$u_k = \begin{bmatrix} \nu_{a,k} \\ \nu_{\ddot{\psi},k} \end{bmatrix}$$
• Process Model :  $x_{k+1} = f(x_k, \nu_k)$ 

• Process Nosie Covariance Matrix : 
$$Q=E\left\{\,
u_k\cdot 
u_k^T\,
ight\}=\left[egin{array}{cc} \sigma_a^2 & 0 \\ 0 & \sigma_{\Vec\psi}^2 \end{array}
ight]$$

## **Predict Sigma Points**



#### **Predict Mean and Covariance**



#### Weights:

$$\omega_i = rac{\lambda}{\lambda + n_a}, i = 1$$
  $\omega_i = rac{1}{2(\lambda + n_a)}, i = 2...n_a$ 

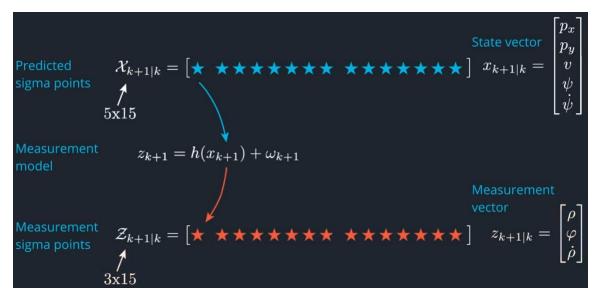
#### Predicted Mean :

$$x_{k+1|k} = \sum_{i=1}^{n_\sigma} \omega_i X_{k+1|k,i}$$

## Predicted Covariance:

$$P_{k+1|k} = \sum_{i=1}^{n_{\sigma}} \omega_i (X_{k+1|k,i} - x_{k+1|k}) (X_{k+1|k,i} - x_{k+1|k})^T$$

#### **Measurement Prediction**

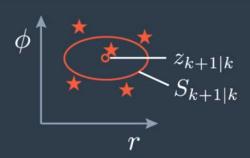


Measurement Nosie :  $\omega_{k+1}$ 

Measurement Model :  $z_{k+1|k} = h(x_{k+1}) + \omega_{k+1}$ 

$$ho = \sqrt{p_x^2 + p_y^2} \ arphi = arctan(rac{p_y}{p_x})$$

 $\dot{\rho} = \frac{p_x cos(\psi)v + p_y sin(\psi)v}{\sqrt{p_x^2 + p_y^2}}$  Predict Measurement Mean And Covariance :



Predicted measurement mean

$$z_{k+1|k} = \sum_{i=1}^{n_{\sigma}} w_i \mathcal{Z}_{k+1|k,i}$$

$$z_{k+1} = h(x_{k+1}) + \omega_{k+1}$$

$$S_{k+1|k} = \sum_{i=0}^{2n_{\sigma}} w_i \left( \mathcal{Z}_{k+1|k,i} - z_{k+1|k} \right) \left( \mathcal{Z}_{k+1|k,i} - z_{k+1|k} \right)^T + R$$

$$R = E\{\omega_k \cdot \omega_k^T\}$$

#### **UKF Update**

Kalman Gain

$$K_{k+1|k} = T_{k+1|k} S_{k+1|k}^{-1}$$

$$x_{k+1|k+1} = x_{k+1|k} + K_{k+1|k}(z_{k+1} - z_{k+1|k})$$

Covariance matrix update

$$P_{k+1|k+1} = P_{k+1|k} - K_{k+1|k} S_{k+1|k} K_{k+1|k}^T$$

New here: Cross-correlation between sigma points in state space and

$$T_{k+1|k} = \sum_{i=0}^{2n_{\sigma}} w_i \left( \mathcal{X}_{k+1|k,i} - x_{k+1|k} \right) \left( \mathcal{Z}_{k+1|k,i} - z_{k+1|k} \right)^T$$

# **Run Project**

Once the install for uWebSocketIO is complete, the main program can be built and ran by doing the following from the project top directory.

- 1. mkdir build
- 2. cd build

- 3. cmake ..
- 4. make
- 5. ./UnscentedKF

Here is the main protcol that main.cpp uses for uWebSocketIO in communicating with the simulator.

INPUT: values provided by the simulator to the c++ program

["sensor\_measurement"] => the measurment that the simulator observed (either lidar or radar)

OUTPUT: values provided by the c++ program to the simulator

["estimate\_x"] <= kalman filter estimated position x

["estimate\_y"] <= kalman filter estimated position y

["rmse\_x"]

["rmse\_y"]

["rmse\_vx"]

["rmse\_vy"]

# **Other Important Dependencies**

- cmake >= 3.5
- All OSes: click here for installation instructions
- make >= 4.1 (Linux, Mac), 3.81 (Windows)
- Linux: make is installed by default on most Linux distros
- Mac: install Xcode command line tools to get make
- Windows: Click here for installation instructions
- gcc/g++>=5.4
- Linux: gcc / g++ is installed by default on most Linux distros
- Mac: same deal as make install Xcode command line tools
- Windows: recommend using MinGW

# **Basic Build Instructions**

- 1. Clone this repo.
- 2. Make a build directory: mkdir build && cd build
- 3. Compile: cmake .. && make
- 4. Run it: ./UnscentedKF Previous versions use i/o from text files. The current state uses i/o from the simulator.