

Vehicle Control

Vehicle Dynamics를 이용한 차량 위치 추정,
Pure pursuit을 이용한 **Steering Control**

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1. CAR MAKER

CM CarMaker for Simulink - Test: hkhk - 'win-5q85262dfen' online

File Application Simulation Parameters Settings Help 

Car: DemoCar
Typical, unvalidated data for passenger car with Front Wheel Drive Select

Trailer: - Select


Tires: Ex...JRT_195_65R15 Ex...JRT_195_65R15 Select
Ex...JRT_195_65R15 Ex...JRT_195_65R15

Load: 0 kg Select

Maneuver

0 :

Simulation

Perf.:  realtime


Status: (1.0x)

Idle

Time: 39.7

Distance: 416.86

Storage of Results

Mode:  collect only

Buffer: 33.6 MB, 1118 s

Save Stop Abort

Start

Stop

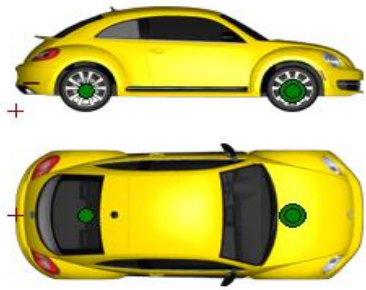
Road Sensors

List of Sensors

Sensor.Road. RD00

- RD00
- RD01
- RD02
- RD03
- RD04
- RD05
- RD06
- RD07

Add Delete



RD00

Preview distance [m]

Preview mode  Along Route Centerline

Consider bumps

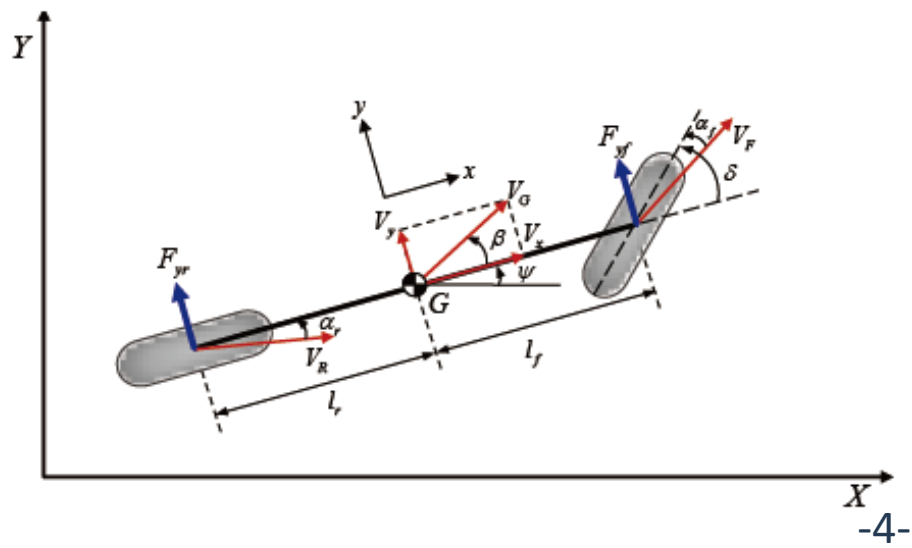
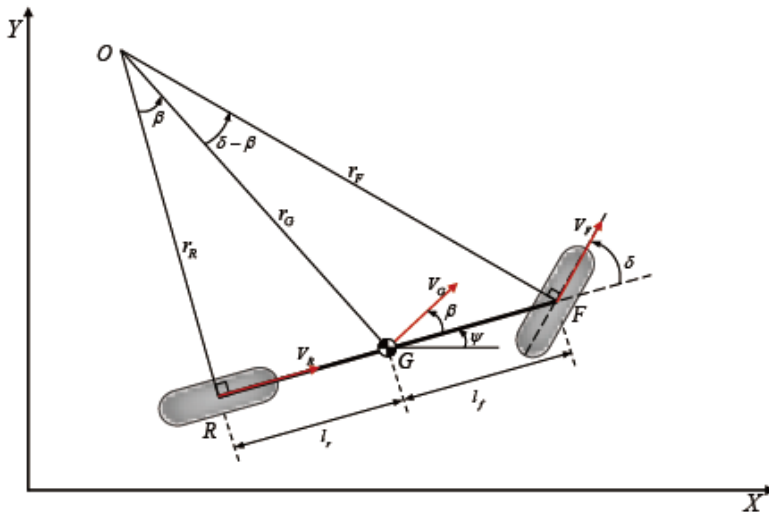
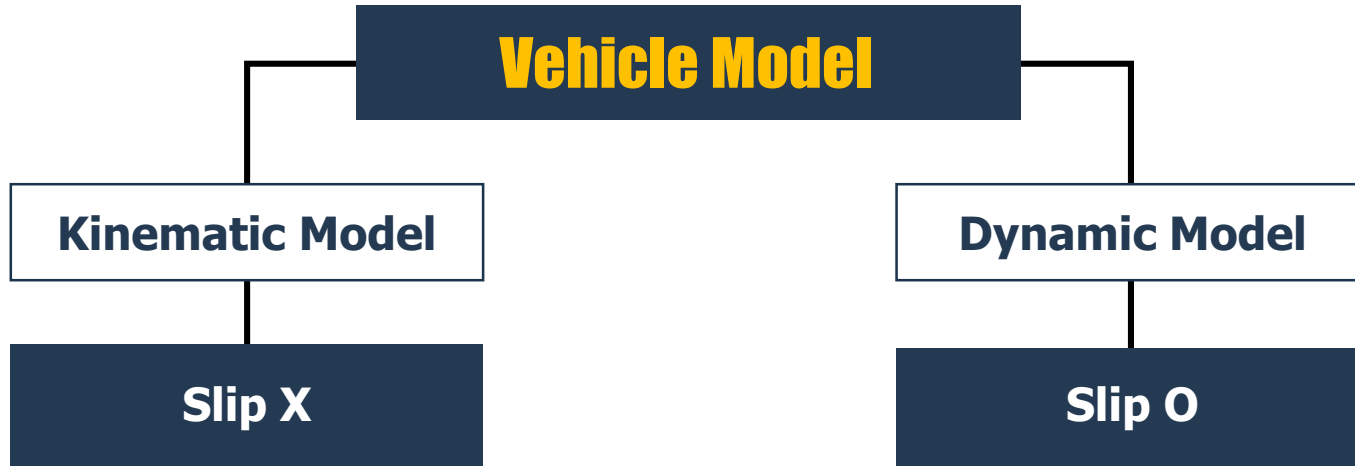


Rigid Vehicle Body

Override internally computed vehicle body proportioning

	x [m]	y [m]	z [m]	Mass [kg]	Ixx [kgm ²]	Iyy [kgm ²]	Izz [kgm ²]
Vehicle Body	2.43	0	0.60	1301	470	1500	1600
Vehicle Body B	2.15	0.0	0.58	650.5	180.0	900.0	900.0
Joint A - B	2.43	0	0.60				
Calculated vehicle overall mass [kg]				1463.00	Info		

2. Vehicle Localization



2-1) Kinematic Model

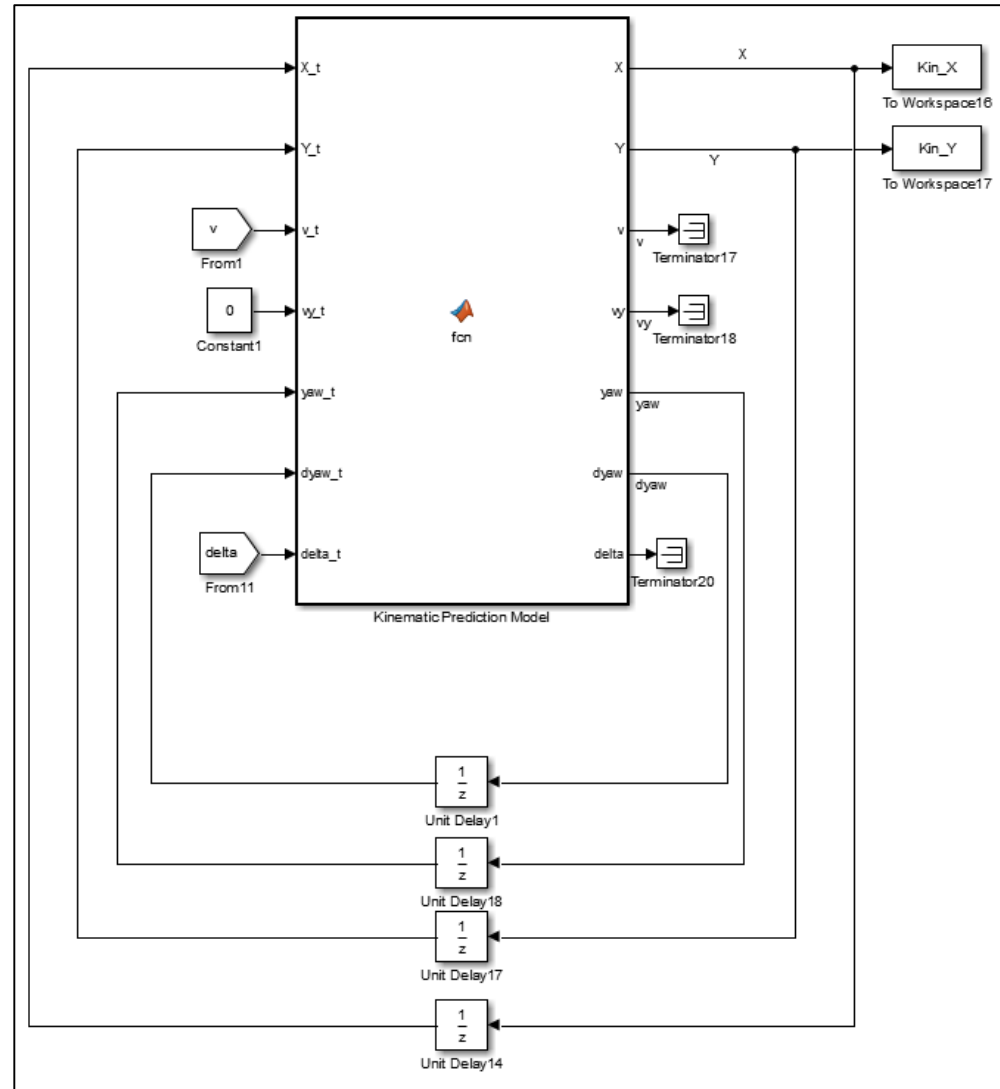
$$\begin{bmatrix} \dot{X} \\ \dot{Y} \\ \dot{\psi} \end{bmatrix} = \begin{bmatrix} V_G \cos(\psi + \beta) \\ V_G \sin(\psi + \beta) \\ V_G \cos(\beta) \tan(\delta) / (l_f + l_r) \end{bmatrix}$$

where

$$\beta = \tan^{-1} \left(\frac{l_r \tan(\delta)}{l_f + l_r} \right)$$

```

X = X_t + v_t * cos(beta + yaw_t) * dT;
Y = Y_t + v_t * sin(beta + yaw_t) * dT;
yaw = yaw_t + v_t * tan(delta_t)/(l_r+l_f) * dT;
dyaw = dyaw_t;
v = v_t;
vy = vy_t;
delta = delta_t;
    
```



2-2) Dynamic Model

$$\begin{bmatrix} \dot{X} \\ \dot{Y} \\ \dot{\psi} \\ \dot{V}_y \end{bmatrix} = \begin{bmatrix} V_x \cos(\psi) - V_y \sin(\psi) \\ V_y \cos(\psi) + V_x \sin(\psi) \\ -\left(\frac{2C_f l_f^2 + 2C_r l_r^2}{I_z V_x}\right) \dot{\psi} - \left(\frac{2C_f l_f - 2C_r l_r}{I_z V_x}\right) V_y + \frac{2C_f l_f}{I_z} \\ -\left(V_x + \frac{2C_f l_f - 2C_r l_r}{V_x m}\right) \dot{\psi} - \left(\frac{2C_f + 2C_r}{V_x m}\right) V_y + \frac{2C_f \delta}{m} \end{bmatrix}$$

```

if v_t >= 0 && v_t <= 1
    v_tt = 1;
elseif v_t < 0 && v_t >= -1
    v_tt = -1;
else
    v_tt = v_t;
end

```

```
X = X_t + (v_t * cos(yaw_t) - vy_t * sin(yaw_t)) * dT;
```

```
Y = Y_t + (vy_t * cos(yaw_t) + v_t * sin(yaw_t)) * dT;
```

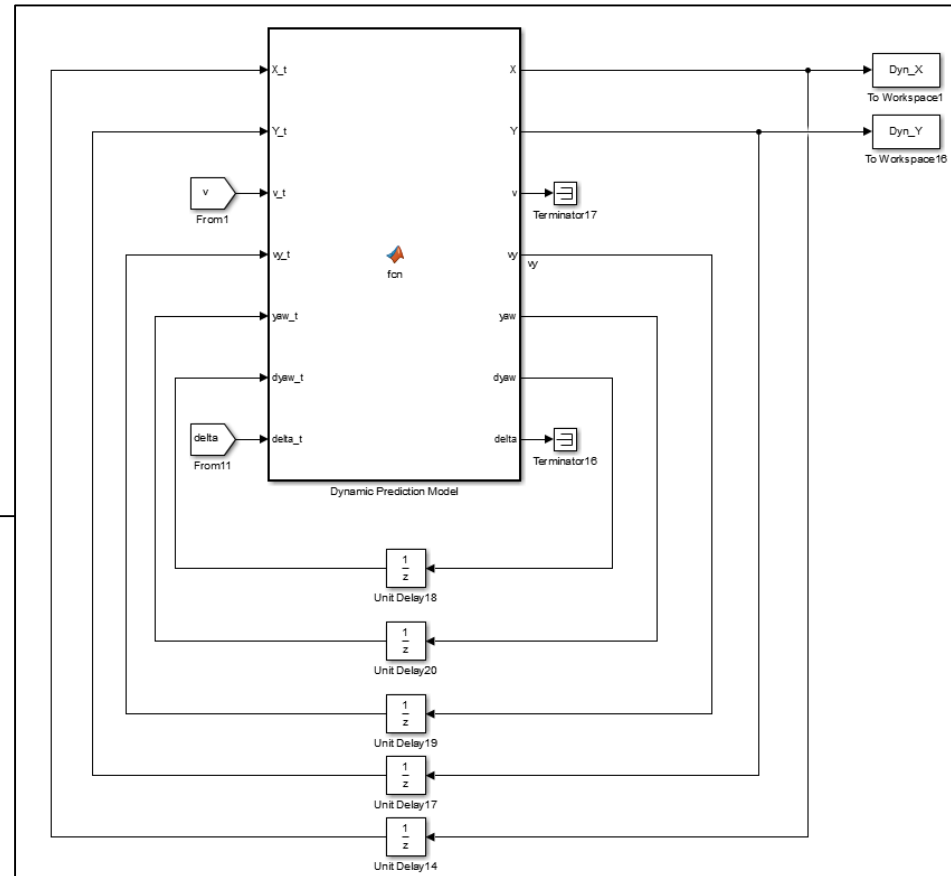
```
v = v_t;
```

```
vy = vy_t + (-(v_t + 2*(Cf*lf - Cr*lr)/(v_tt*m))*dyaw_t - 2*(Cf + Cr)/(v_tt*m)*vy_t + 2*Cf*delta_t/m) * dT;
```

```
yaw = yaw_t + dyaw_t * dT;
```

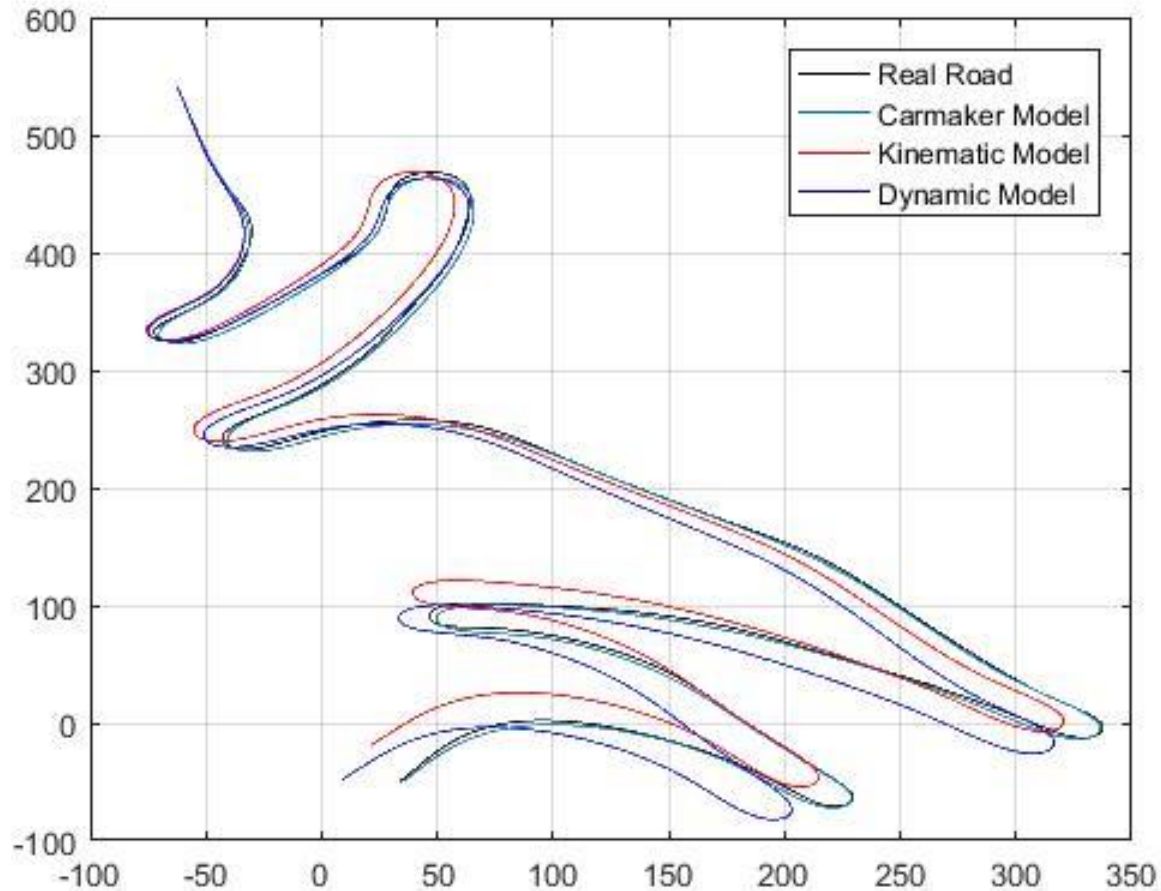
```
dyaw = dyaw_t + (-(2*(Cf*lf^2 + Cr*lr^2)/(Iz*v_tt))*dyaw_t - 2*((Cf*lf - Cr*lr)/(Iz*v_tt))*vy_t + 2*Cf*lf*delta_t/Iz) * dT;
```

```
delta = delta_t;
```



2-3) Analysis

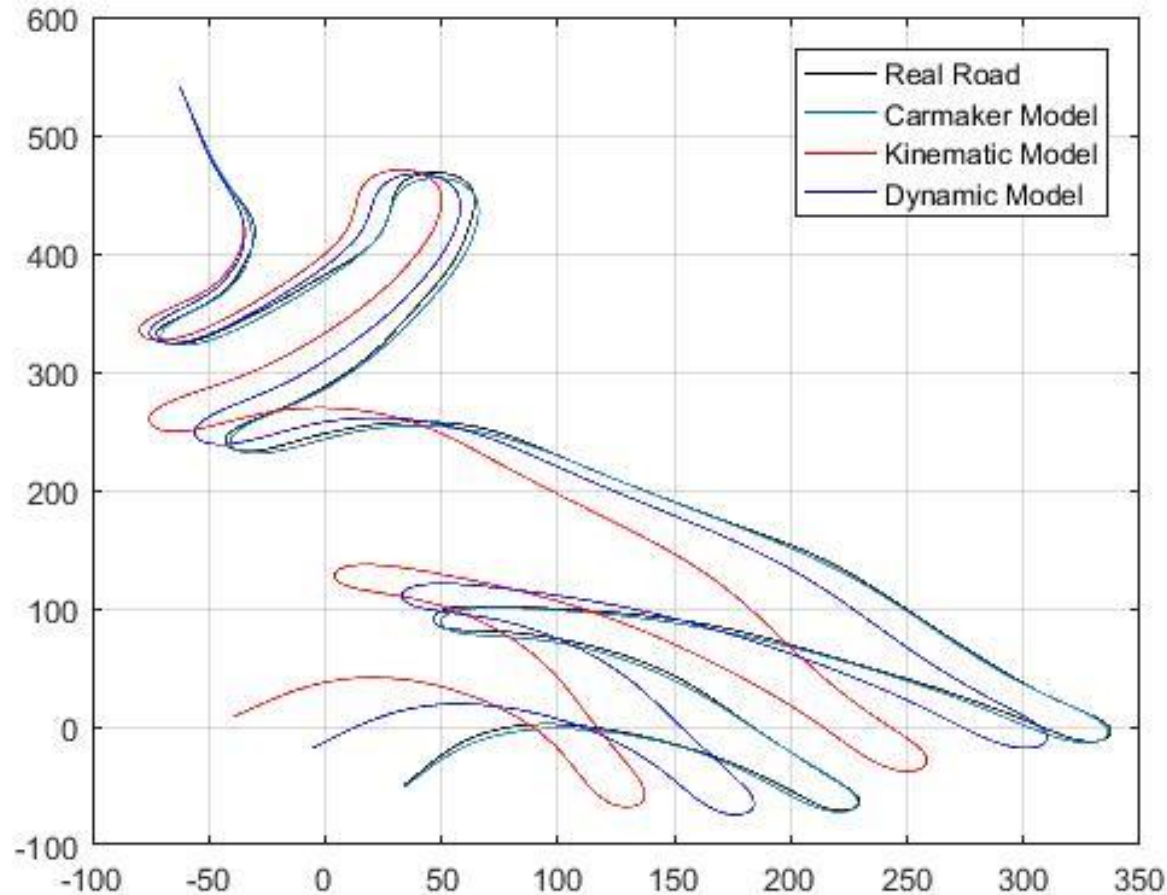
Low speed (20km/h)



Kinematic Model is reasonable for low-speed vehicle motion.

2-3) Analysis

High speed (80km/h)



Dynamic Model is reasonable for high-speed vehicle motion.

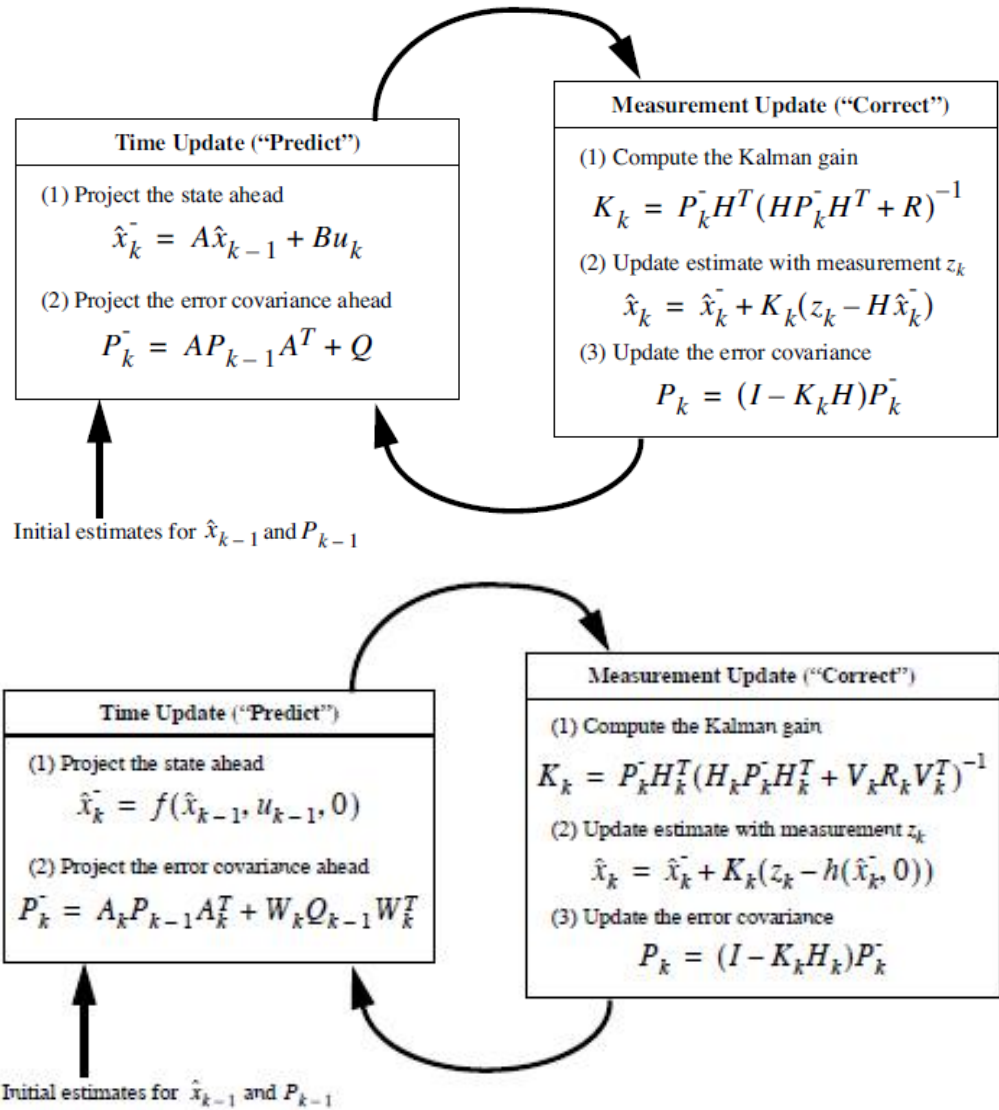
3. Extended Kalman Filter

Linear System

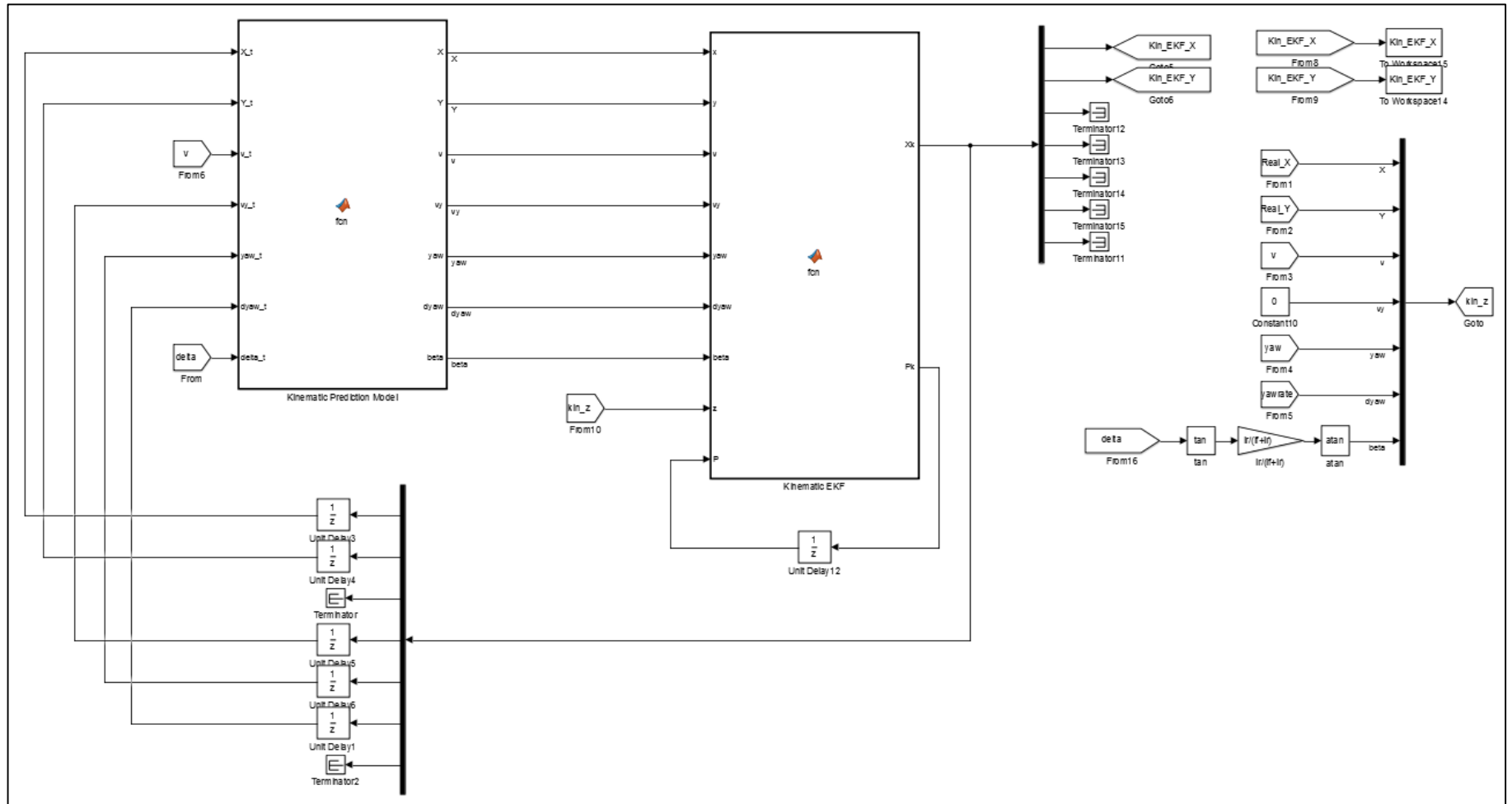
Kalman Filter

Extended Kalman Filter

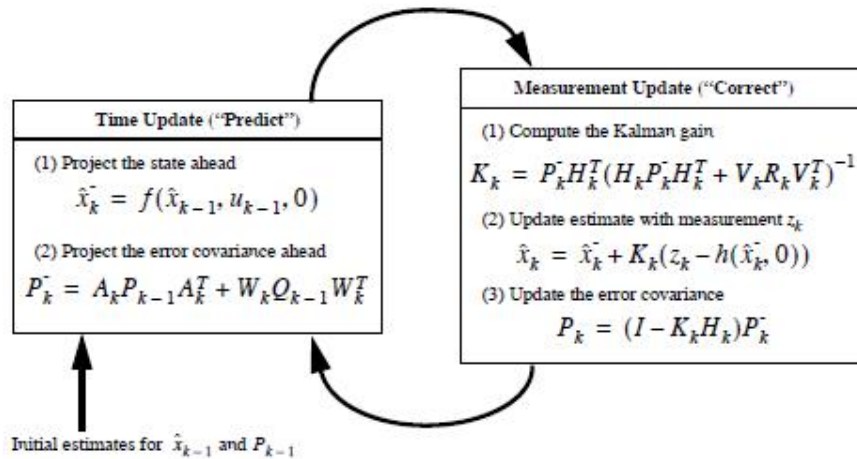
Nonlinear System



3-1) Kinematic Model with EKF



3-1) Kinematic Model with EKF



$$\begin{bmatrix} \dot{X} \\ \dot{Y} \\ \dot{\psi} \end{bmatrix} = \begin{bmatrix} V_G \cos(\psi + \beta) \\ V_G \sin(\psi + \beta) \\ V_G \cos(\beta) \tan(\delta) / (l_f + l_r) \end{bmatrix}$$

where

$$\beta = \tan^{-1} \left(\frac{l_r \tan(\delta)}{l_f + l_r} \right)$$

```
X = zeros(7,1);

X(1) = x;
X(2) = y;
X(3) = v;
X(4) = vy;
X(5) = yaw;
X(6) = dyaw;
X(7) = beta;

F = [0, 0, cos(yaw+beta), 0, -v*sin(yaw+beta), 0, -v*sin(yaw+beta); ...
     0, 0, sin(yaw+beta), 0, v*cos(yaw+beta), 0, v*cos(yaw+beta); ...
     0, 0, 0, 0, 0, 0, 0; ...
     0, 0, 0, 0, 0, 0, 0; ...
     0, 0, 1/lr*sin(beta), 0, 0, 0, v/lr*cos(beta); ...
     0, 0, 0, 0, 0, 0, 0; ...
     0, 0, 0, 0, 0, 0, 0];

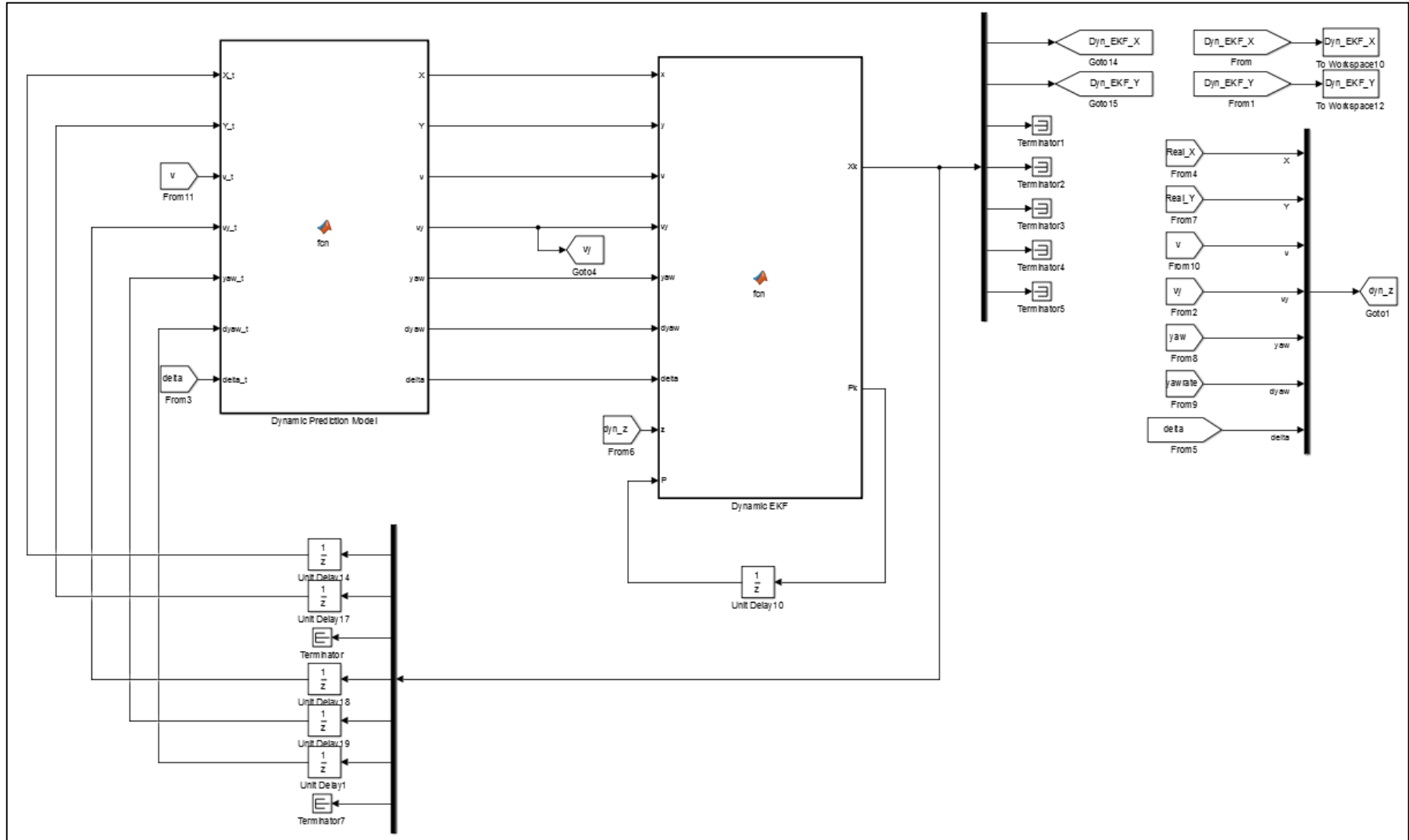
Pi = eye(7) + F*dT;

Q = diag([0, 0, 0.1^2, 0, (0.05*pi/180*dT)^2, (0.05*pi/180)^2, (0.1*pi/180)^2]);

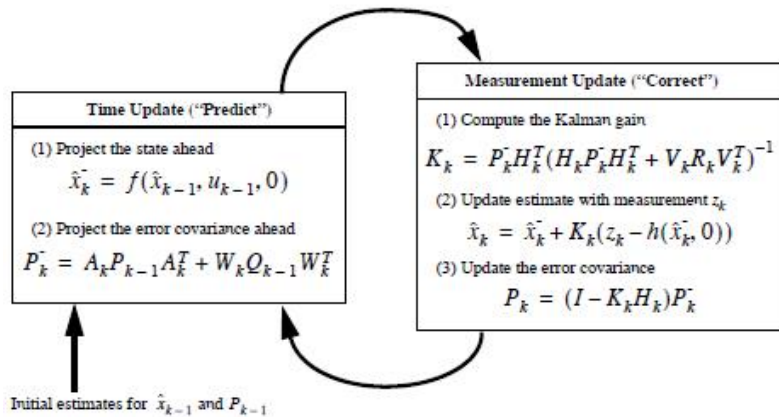
P = Pi * P + Pi' + Q;
R = diag([2^2, 2^2, 0.1^2, 0.1^2, 0, 0, (0.1*pi/180)^2]);
H = eye(7);
S = H*P*H' + R;
K = P*H'/S;
Xk = X + K*(z - H*X);

Pk = (eye(7) - K*H)*P;
```

3-2) Dynamic Model with EKF



3-2) Dynamic Model with EKF



$$\begin{bmatrix} \dot{X} \\ \dot{Y} \\ \dot{\psi} \\ \dot{V}_y \end{bmatrix} = \begin{bmatrix} V_x \cos(\psi) - V_y \sin(\psi) \\ V_y \cos(\psi) + V_x \sin(\psi) \\ -\left(\frac{2C_f l_f^2 + 2C_r l_r^2}{I_z V_x}\right) \dot{\psi} - \left(\frac{2C_f l_f - 2C_r l_r}{I_z V_x}\right) V_y + \frac{2C_f l_f}{I_z} \\ -\left(V_x + \frac{2C_f l_f - 2C_r l_r}{V_x m}\right) \dot{\psi} - \left(\frac{2C_f + 2C_r}{V_x m}\right) V_y + \frac{2C_f \delta}{m} \end{bmatrix}$$

```
X = zeros(7,1);

X(1) = x;
X(2) = y;
X(3) = v;
X(4) = vy;
X(5) = yaw;
X(6) = dyaw;
X(7) = delta;

if v >= 0 && v <= 1
    vt = 1;
elseif v < 0 && v >= -1
    vt = -1;
else
    vt = v;
end

F = [0, 0, cos(yaw), -sin(yaw), -v*sin(yaw)-vy*cos(yaw), 0, 0; ...
     0, 0, sin(yaw), cos(yaw), v*cos(yaw)-vy*sin(yaw), 0, 0; ...
     0, 0, 0, 0, 0, 0, 0; ...
     0, 0, (-1+2*(Cf*lf-Cr*lr)/(m*vt^2))+dyaw+2*(Cf+Cr)/(m*vt^2)+vy, ...
     -2*(Cf+Cr)/(m*vt), 0, -(v+2*(Cf*lf-Cr*lr)/(m*vt)), 2*Cf/m, 0, 0, 0, 0, 1, 0; ...
     0, 0, 2*(Cf*lf^2+Cr*lr^2)/(Iz*vt^2)+dyaw+2*(Cf*lf-Cr*lr)/(Iz*vt^2)+vy, -2*(Cf*lf-Cr*lr)/(Iz*vt), 0, ...
     -2*(Cf*lf^2+Cr*lr^2)/(Iz*vt), 2*Cf*lf/Iz, 0, 0, 0, 0, 0, 0, 0];

Pi = eye(7) + F*dT;

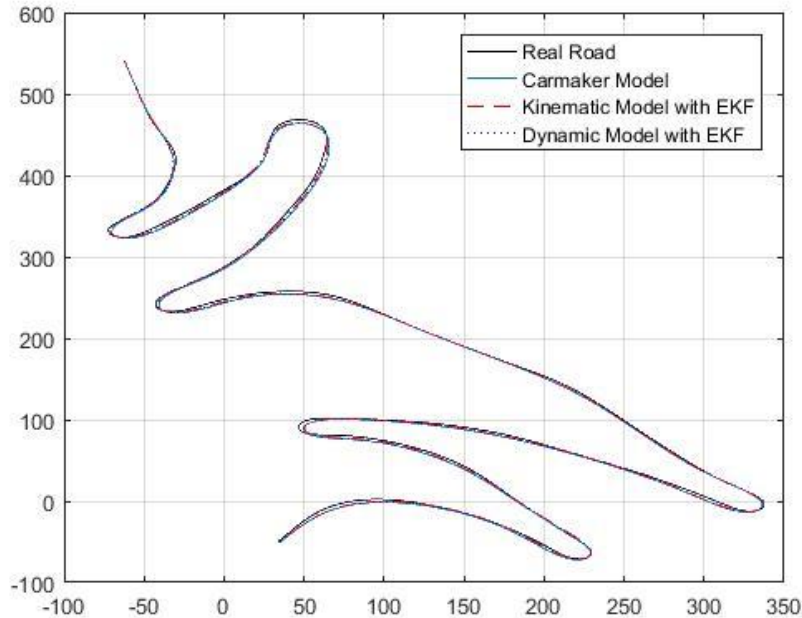
Q = diag([0, 0, 0.1^2, 0, (0.05*pi/180*dT)^2, (0.05*pi/180)^2, (0.1*pi/180)^2]);

P = Pi * P + Pi' + Q;
R = diag([2^2, 2^2, 0.1^2, 0.1^2, 0, 0, 0]);
H = eye(7);
S = H*P*H' + R;
K = P*H'/S;
Xk = X + K*(z - H*X);

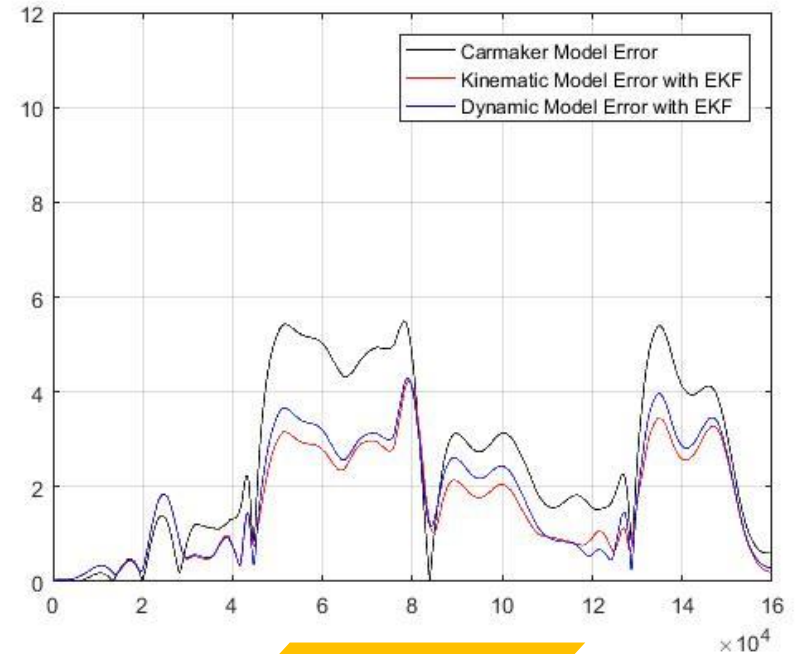
Pk = (eye(7) - K*H)*P;
```

3-3) Analysis

Low speed (20km/h)



Road Plot

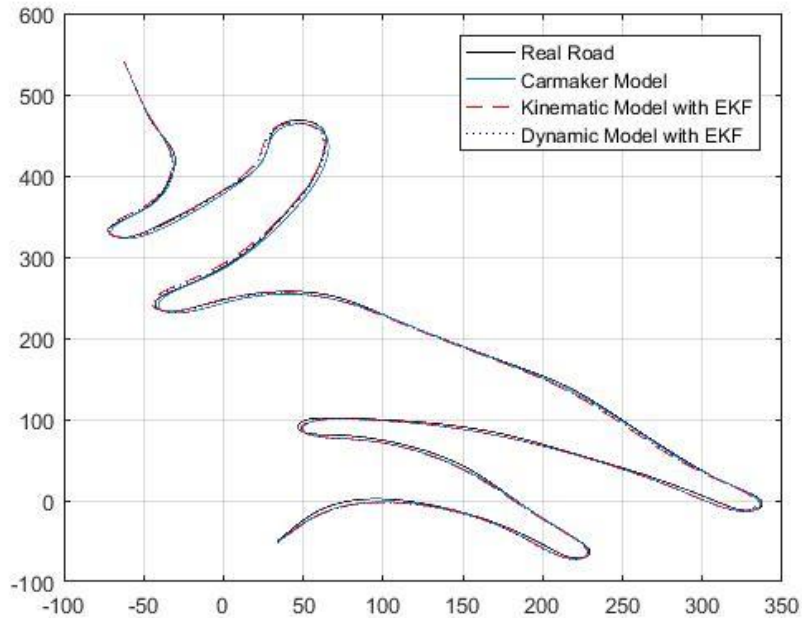


Error Plot

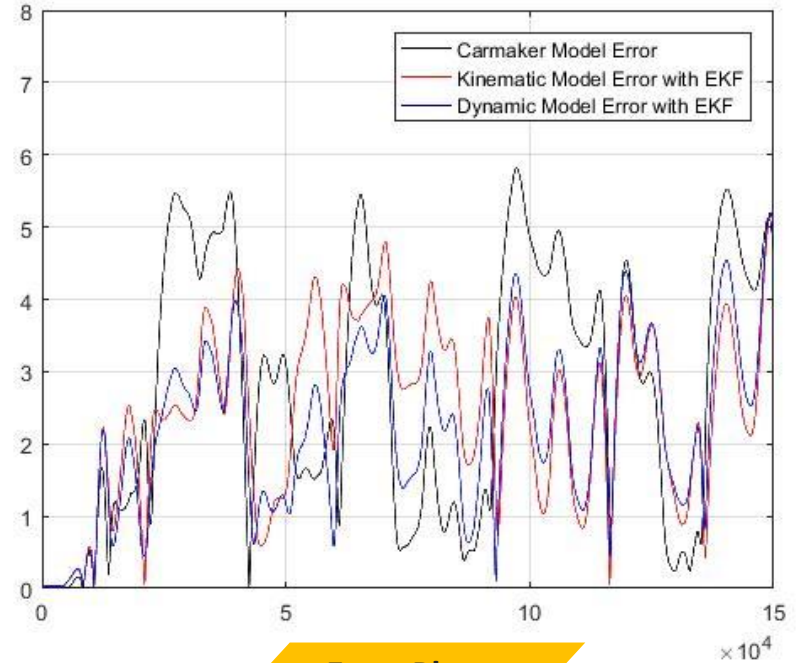
The error is noticeably reduced.
But Kinematic Model is reasonable for low-speed vehicle motion.

3-3) Analysis

High speed (80km/h)



Road Plot

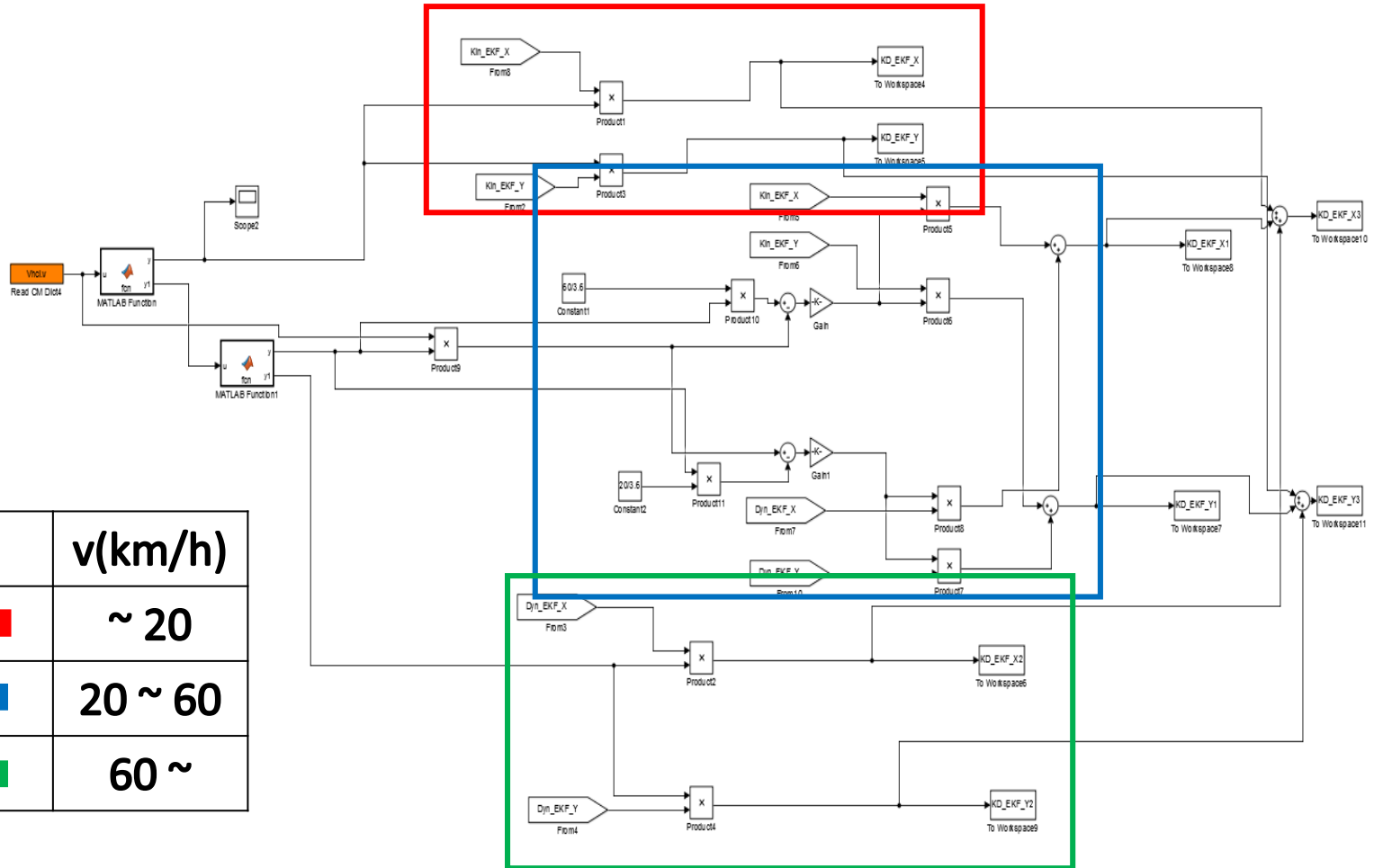





Error Plot

The error is noticeably reduced.
But Dynamic Model is reasonable for high-speed vehicle motion.

3-3) Integration Model

Kinematic + Dynamic (Velocity)

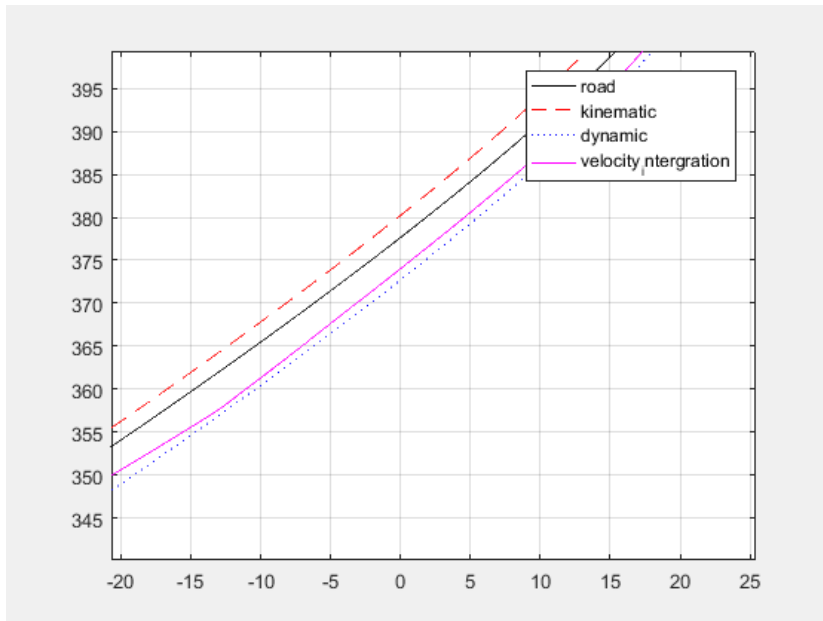


	v(km/h)
	~ 20
	20 ~ 60
	60 ~

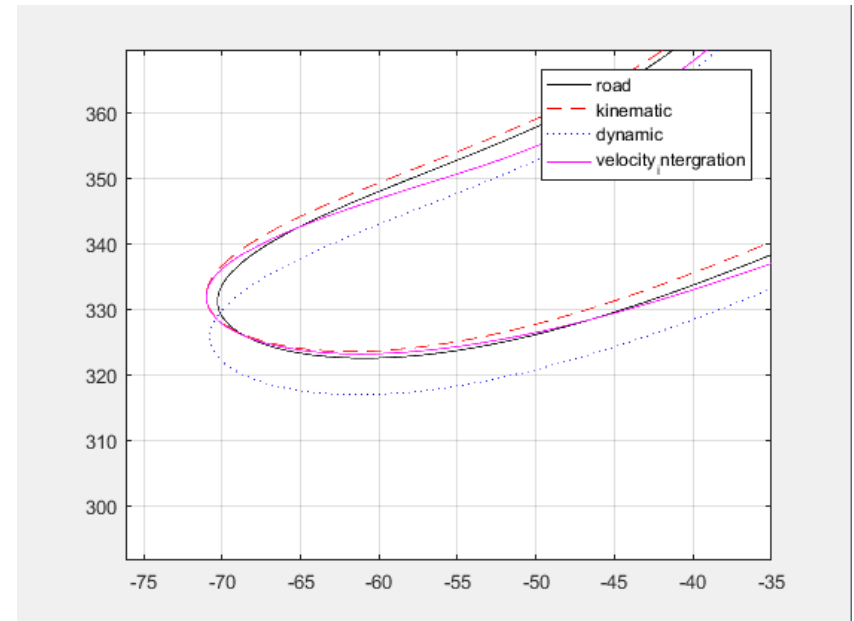
3-3) Integration Model

Kinematic + Dynamic (Velocity)

Analysis in variable speed



Straight Road

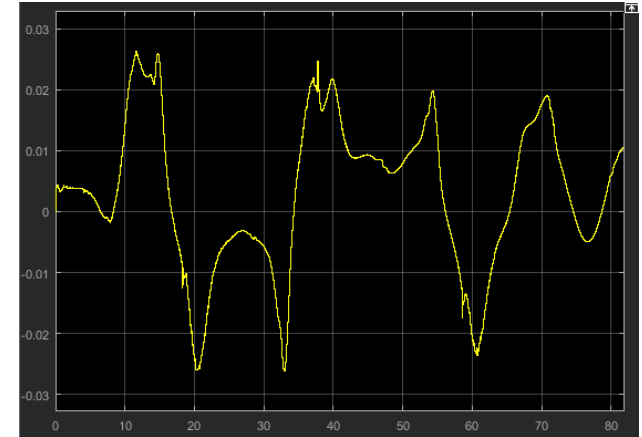
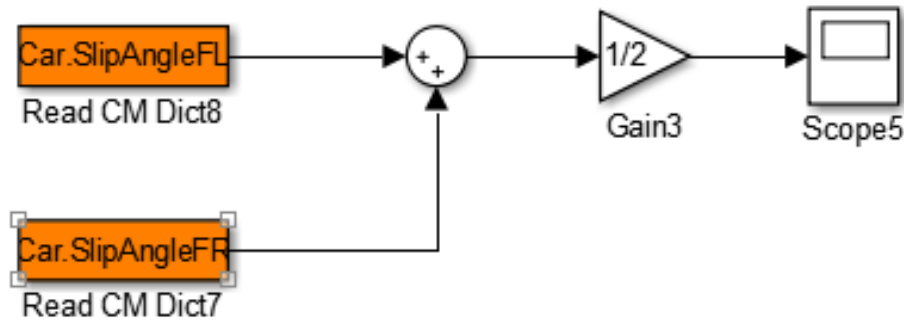


Curve

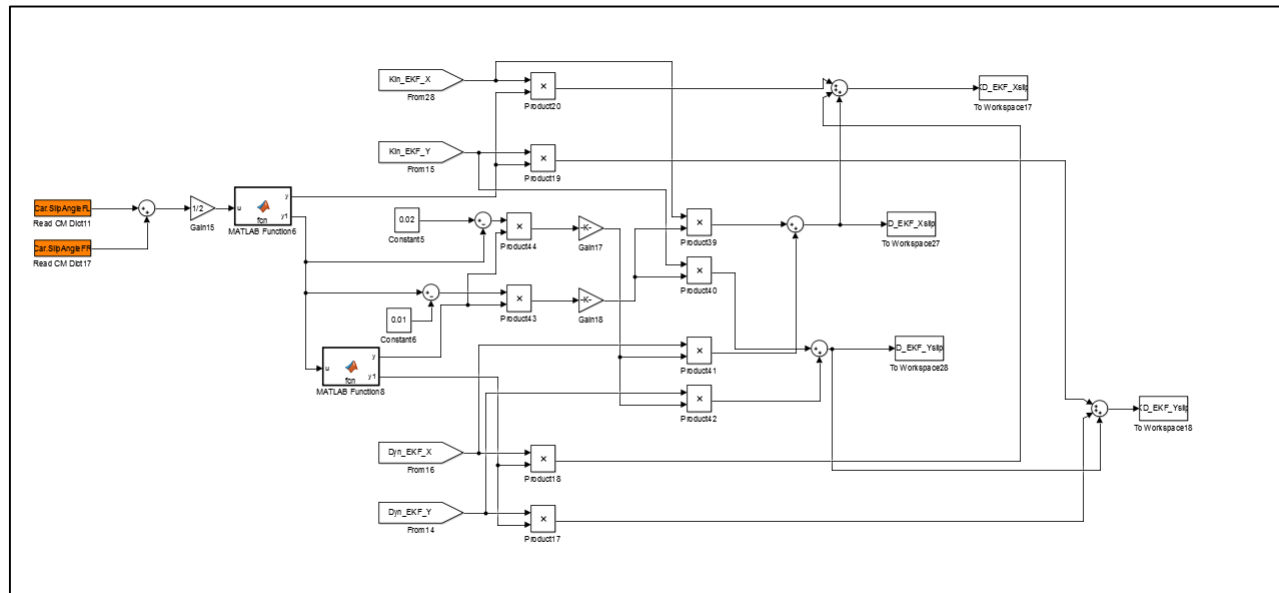
Integration Model according to velocity is reasonable for curve.

3-3) Integration Model

Kinematic + Dynamic (Slip)



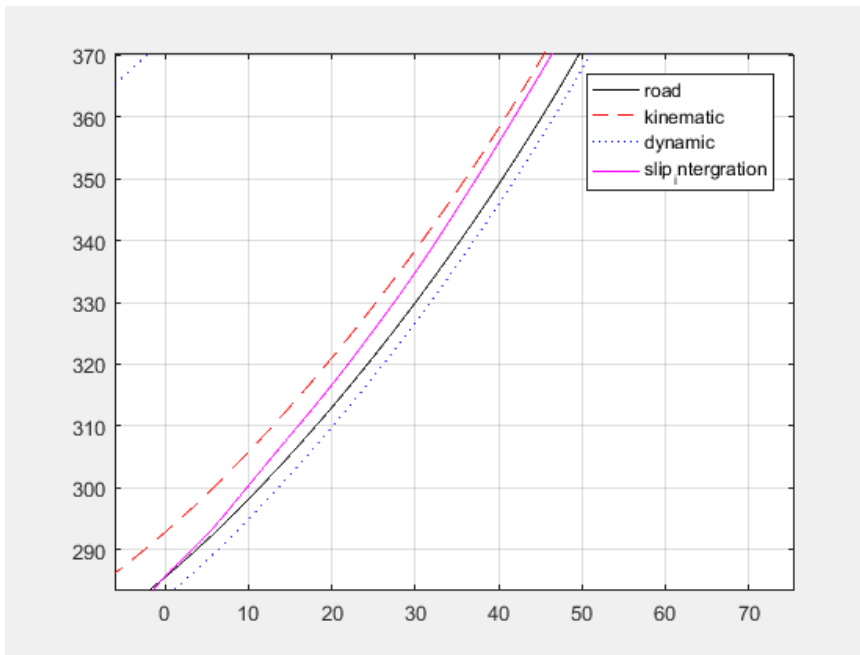
-0.03 ~ 0.03



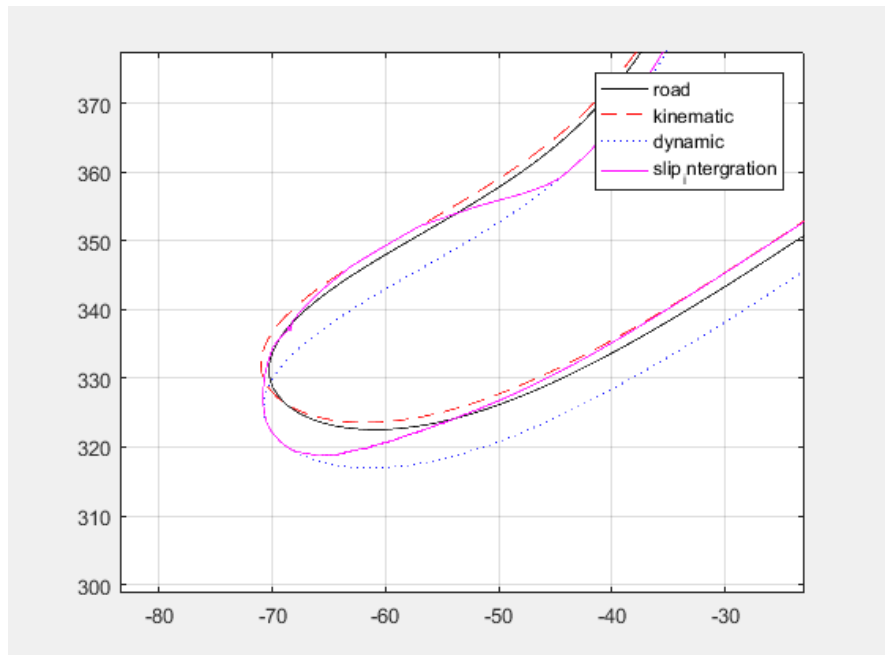
3-3) Integration Model

Kinematic + Dynamic (Slip)

Analysis in variable speed



Straight Road

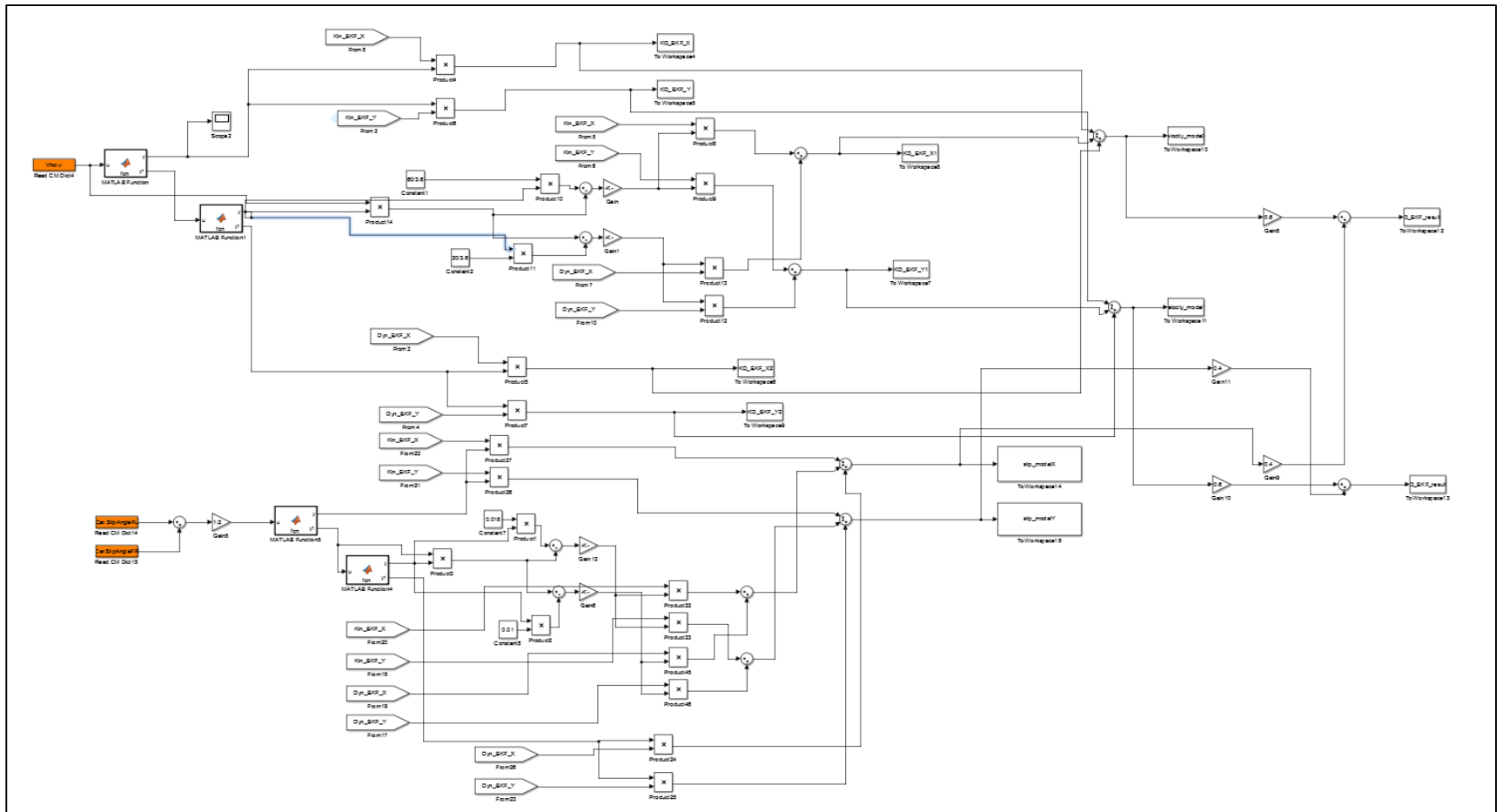


Curve

Integration Model according to slip is reasonable for straight road.

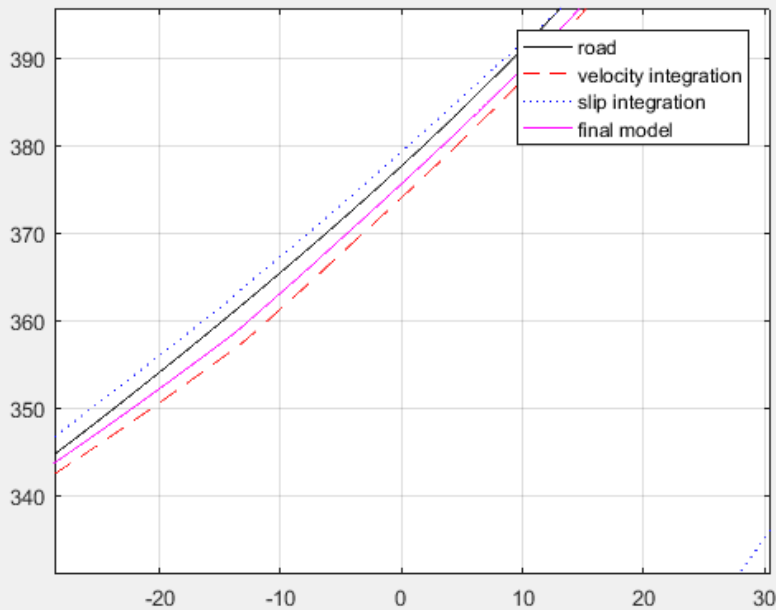
3-3) Integration Model

Kinematic + Dynamic (Velocity + Slip)

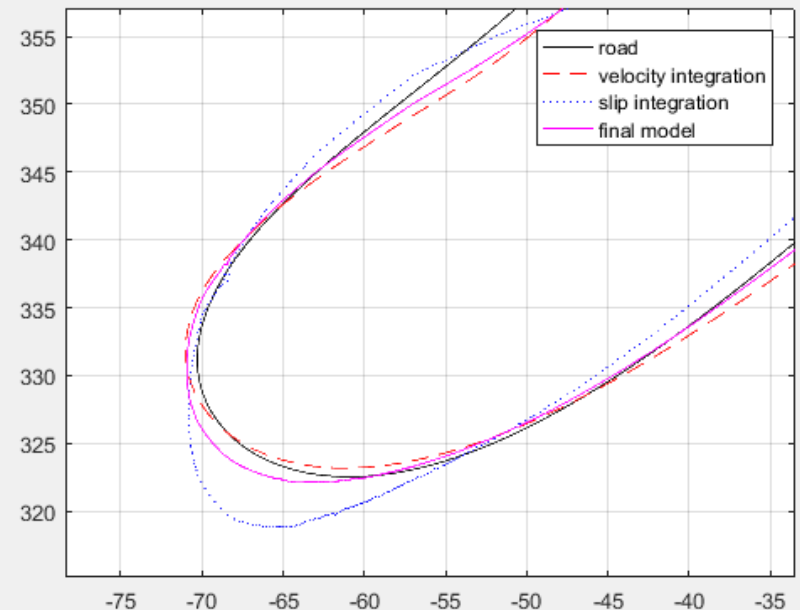


3-3) Integration Model

Kinematic + Dynamic (Velocity 7 + Slip 3) Analysis in variable speed



Straight Road

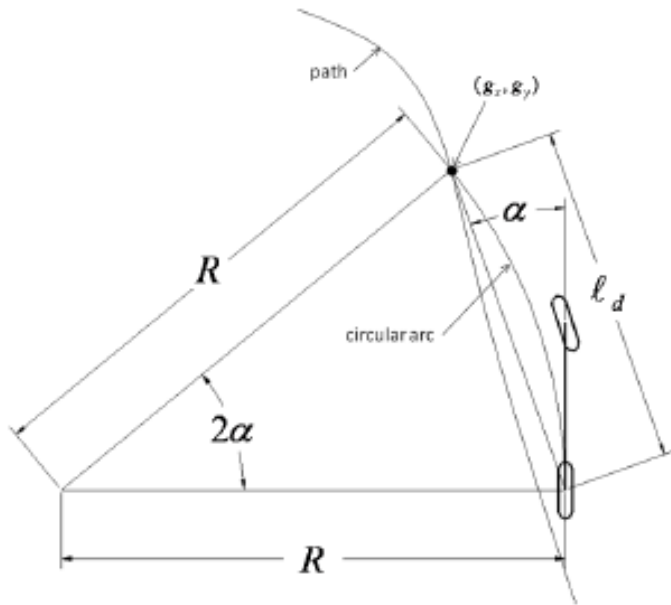


Curve

As a result of the tuning, the integration model with the ratio of 7: 3 was the best along the road.

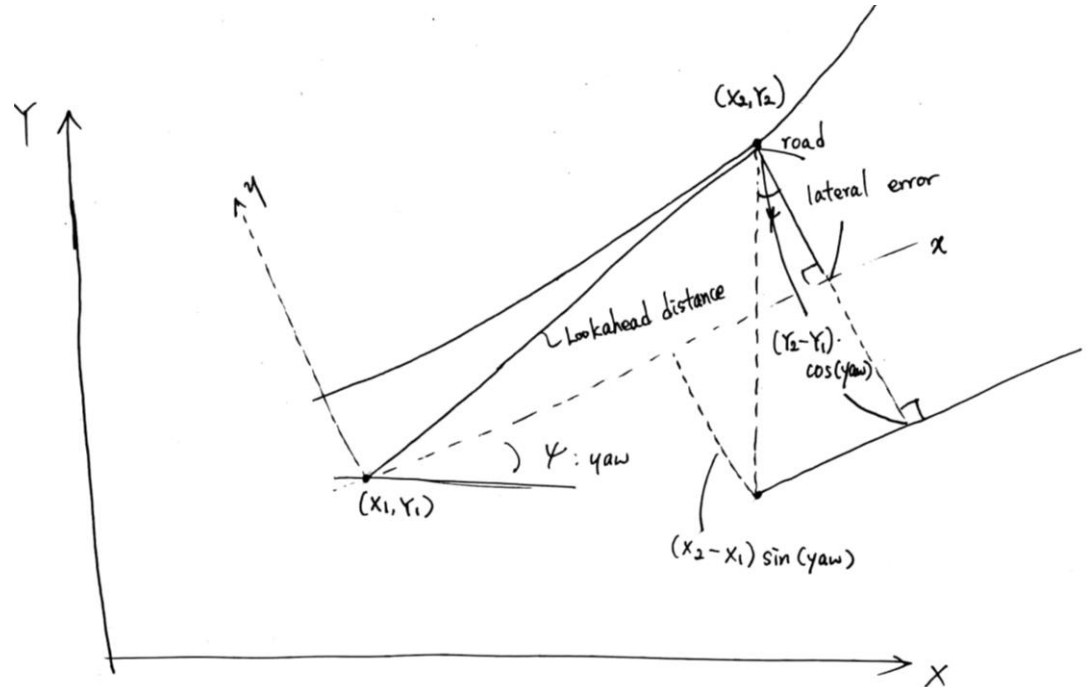
4. Steering Control

Pure pursuit



$$\delta(t) = \tan^{-1} \left(\frac{2L \sin(\alpha(t))}{l_d} \right)$$

$$\sin(\alpha) = \frac{e l_d}{l_d}$$



x, y : Global Coordinate

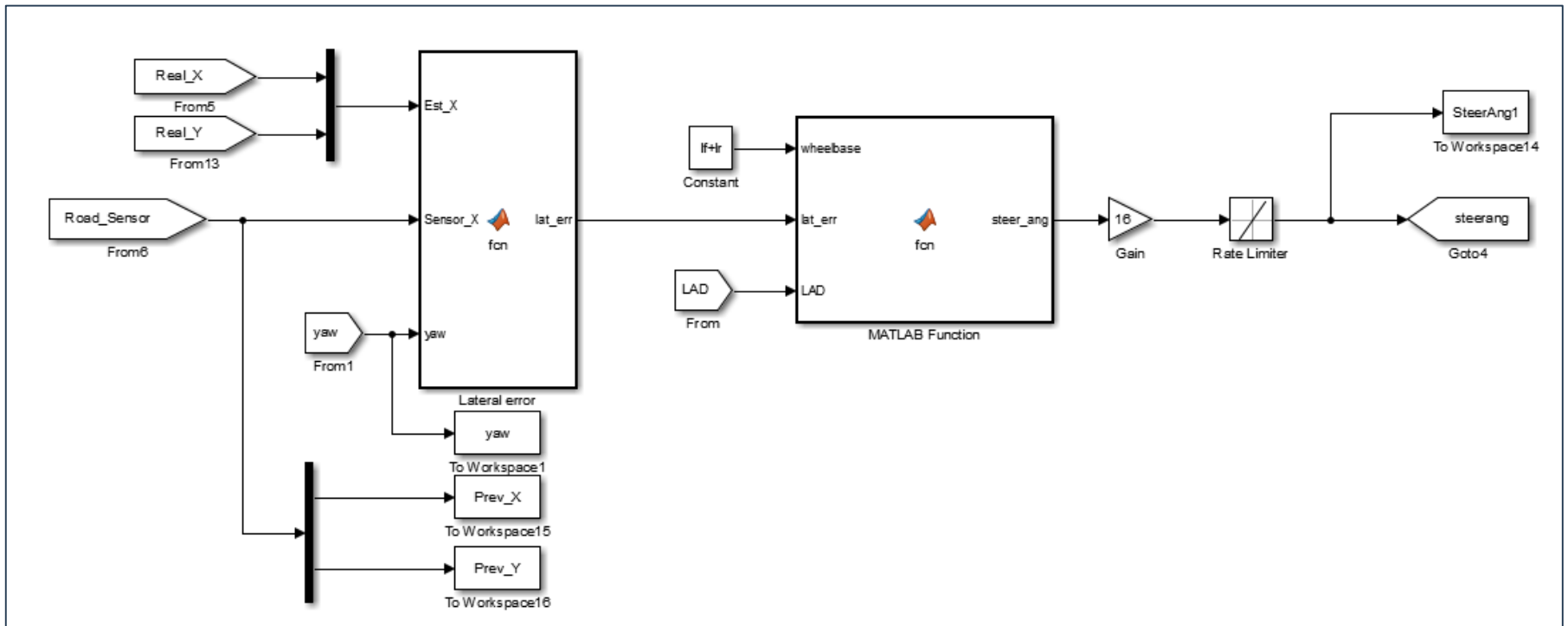
α, ψ : local coordinate

$$\therefore \text{lateral error} = (y_2 - y_1) \cos(\text{yaw}) - (x_2 - x_1) \sin(\text{yaw})$$

$$e_{l_d} = (Y_2 - Y_1) \cos(\text{yaw}) - (X_2 - X_1) \cos(\text{yaw})$$

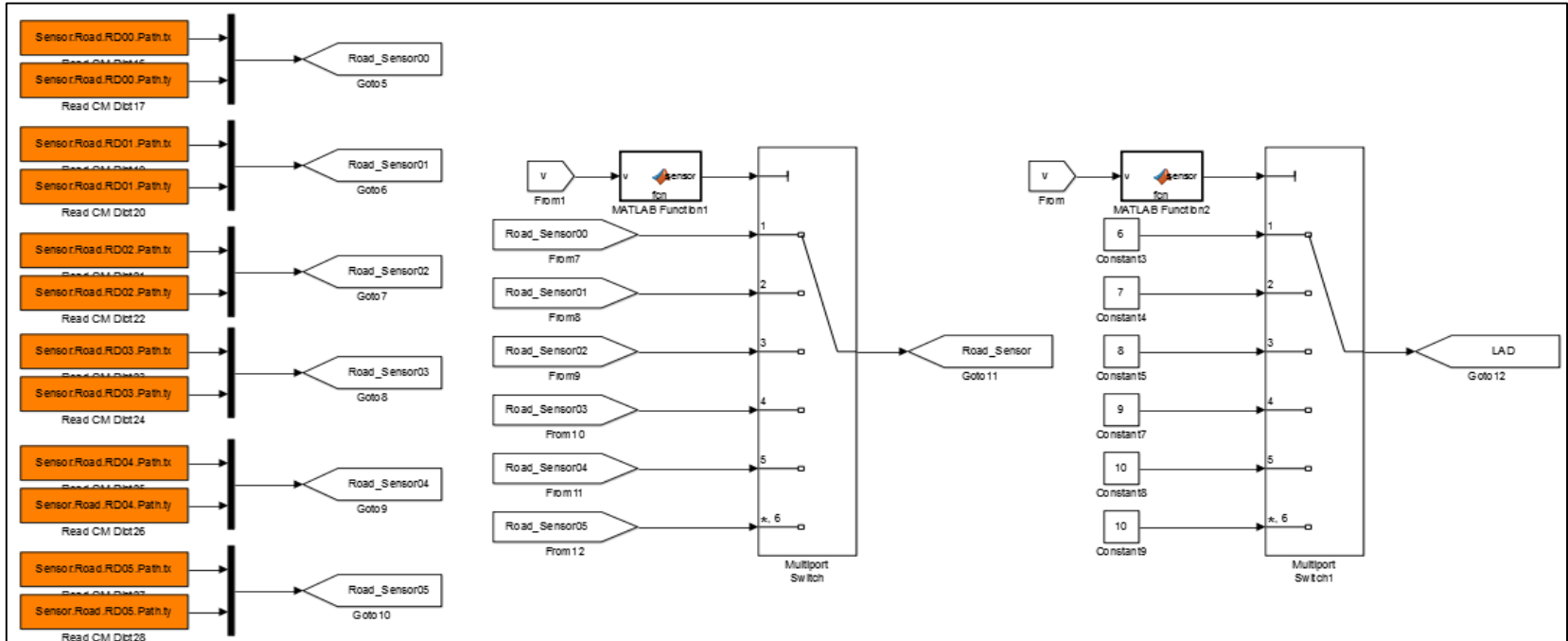
4. Steering Control

Pure pursuit



4. Steering Control

Look-ahead distance control



```
if v <= 30/3.6
    sensor = 1;
elseif v <= 40/3.6
    sensor = 2;
elseif v <= 60/3.6
    sensor = 3;
elseif v <= 80/3.6
    sensor = 4;
elseif v <= 100/3.6
    sensor = 5;
else
    sensor = 6;
end
```



4-1) Video

Carmaker model



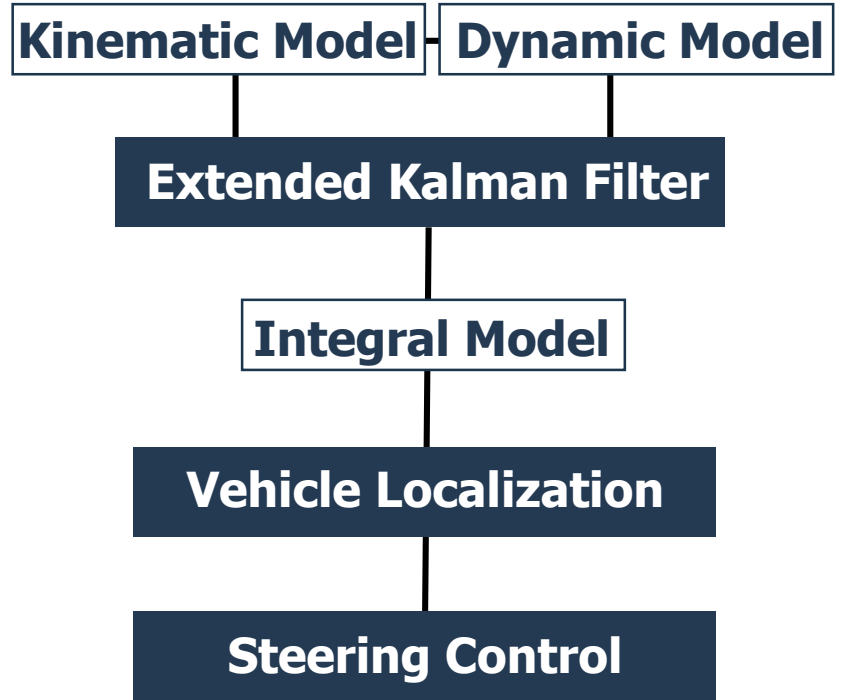
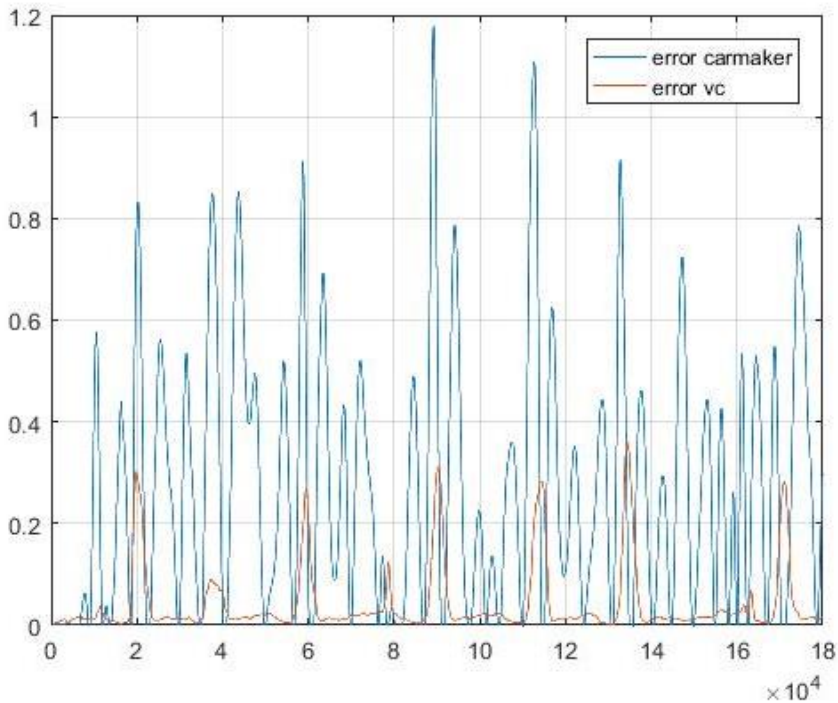
Integration Model



START

4-2) Analysis & Conclusion

Error Analysis



Better steering control than the original model

Reference

Jarrod M. Snider. Automatic Steering Methods for Autonomous Automobile Path Tracking. Carnegie Mellon University, 2009.

Kichun Jo. Integration of Multiple Vehicle Models with an IMM Filter for Vehicle Localization. IEEE Intelligent Vehicles Symposium, 2010.

https://en.wikipedia.org/wiki/Kalman_filter

THANK YOU