

Driver assistance system design A

Lane keeping - part 1

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Outline

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1 Introduction

2 Dynamic single-track model

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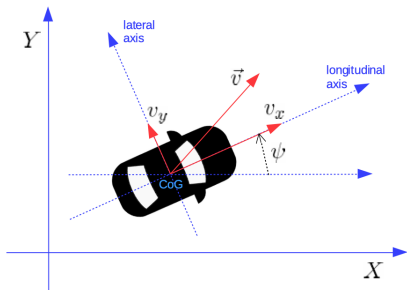
Introduction

- The goal of lane keeping systems is to **maintain the vehicle within the lane** through a control action on the steer.
- Indeed, lateral dynamics is unstable and control is necessary to keep the vehicle in the lane (in manual driving, controller = driver).
- The control system is not intended to replace the driver.
- The control system is aimed to improve safety:
 - ▶ help the driver in emergency situations, e.g. in the case of tiredness, lack of attention, critical road conditions, etc.
 - ▶ reduce the driver's tiredness/drowsiness.



Dynamic single-track model

- To investigate the lane keeping problem and design suitable controllers, we recall the dynamic single-track (DST) model.



► Vehicle variables:

X, Y : coordinates of the vehicle CoG in an inertial reference frame

ψ : yaw angle

$\omega_\psi \doteq \dot{\psi}$: yaw rate

$\vec{v} \equiv V$: velocity vector in the inertial frame

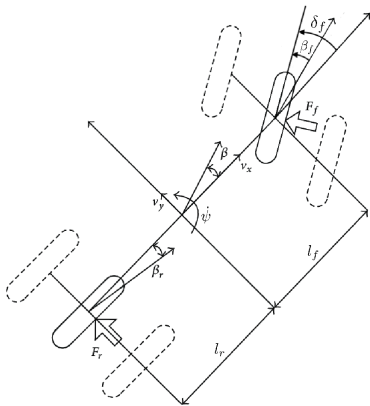
v_x : longitudinal speed = \vec{v} component along the longitudinal axis

v_y : lateral speed = \vec{v} component along the lateral (transverse) axis

a_x : longitudinal acceleration in the inertial frame.

Dynamic single-track model

- The DST model (or bicycle model) is characterized by a single-track (two wheels: one rear wheel, one front wheel), which is equivalent to a vehicle with four wheels, where the left part is equal to the right part.



- Vehicle variables:
 - δ_f : steering angle
 - β : vehicle slip angle = angle between the vehicle longitudinal axis and velocity
 - β_f, β_r : tire slip angles = angles between the tire longitudinal axis and velocity.
- Vehicle parameters:
 - CoG: center of gravity
 - m, J : mass and moment of inertia
 - l_f : distance CoG - front wheel center
 - l_r : distance CoG - rear wheel center
 - c_f, c_r : front/rear cornering stiffnesses.

Dynamic single-track model

- The state equations of the DST model are

$$\dot{X} = v_x \cos \psi - v_y \sin \psi$$

$$\dot{Y} = v_x \sin \psi + v_y \cos \psi$$

$$\dot{\psi} = \omega_\psi$$

$$\dot{v}_x = v_y \omega_\psi + a_x$$

$$\dot{v}_y = -v_x \omega_\psi + \frac{2}{m} (F_{yf} + F_{yr})$$

$$\dot{\omega}_\psi = \frac{2}{J} (l_f F_{yf} - l_r F_{yr})$$

where F_{yf} and F_{yr} are the lateral forces exchanged between tire and road. Different tire models can be considered (next slide).

- State: $\zeta = (X, Y, \psi, v_x, v_y, \omega_\psi)$, input: $u = (a_x, \delta_f)$.

Dynamic single-track model

- Linear (for $v_x = \text{const}$) tire model:

$$\begin{aligned} F_{yf} &= -c_f \beta_f, & F_{yr} &= -c_r \beta_r \\ \beta_f &= \frac{v_y + l_f \omega_\psi}{v_x} - \delta_f, & \beta_r &= \frac{v_y - l_r \omega_\psi}{v_x}. \end{aligned} \quad (1)$$

- Nonlinear simplified tire model:

$$\begin{aligned} F_{yf} &= -c_f \beta_f \cos \delta_f, & F_{yr} &= -c_r \beta_r \\ \beta_f &= \text{atan} \left(\frac{v_y + l_f \omega_\psi}{v_x} \right) - \delta_f, & \beta_r &= \text{atan} \left(\frac{v_y - l_r \omega_\psi}{v_x} \right). \end{aligned} \quad (2)$$

- Nonlinear Pacejka's tire model:

$$F_{yf} = -f_P(\beta_f) \cos \delta_f, \quad F_{yr} = -f_P(\beta_r) \quad (3)$$

where β_f and β_r are those in (2) and $f_P(\beta)$ is given by the Pacejka's magic formula (next slide).

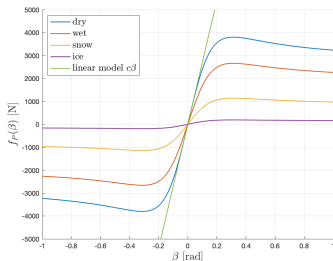
Dynamic single-track model

- Pacejka's magic formula:

$$f_P(\beta) \doteq p_1 \sin(p_2 \operatorname{atan}(p_3 \beta - p_4 (p_3 \beta - \operatorname{atan}(p_3 \beta))))$$

p_1 : peak value, p_2 : shape factor, p_3 : stiffness factor, p_4 : curvature factor. Linearizing this formula, we find $p_1 p_2 p_3 = c_f$ (or c_r).

- ▶ In real applications, these parameters are difficult to measure/estimate.
- ▶ They change in function of the road conditions.



1 Introduction

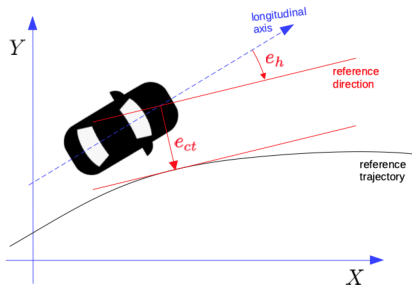
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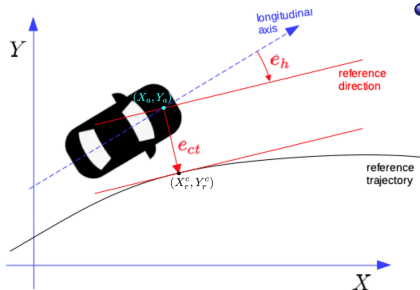
Lane keeping

- The goal of lane keeping is to maintain the vehicle within the lane or, more precisely, to have the vehicle track a given reference trajectory.
- The lateral dynamics is unstable. For constant v_x , the linearized DST model have unstable eigenvalues.
- The following tracking errors can be suitably used for lane keeping control.



- **Heading error e_h** : angle between the vehicle longitudinal axis and the reference direction.
- **Cross-track error e_{ct}** : displacement from the center of the vehicle front axle to the closest point on the trajectory.

Lane keeping



- To compute the two errors, we define the following quantities:

$p_a \doteq (X_a, Y_a, \psi)$: vehicle front axle pose

$P_r \doteq \{p_r^1, \dots, p_r^N\}$: reference trajectory

$p_r^i \doteq (X_r^i, Y_r^i, \psi_r^i) \in P_r$: reference pose

(X_r^c, Y_r^c) : trajectory point closest to the vehicle: $c = \arg \min_i \|(X_r^i, Y_r^i) - (X_a, Y_a)\|$

ψ_r^c : corresponding reference yaw angle.

- Heading error: $e_h \doteq \psi_r^c - \psi$.

- Cross-track error: $e_{ct} \doteq (Y_r^c - Y_a) \cos \psi_r^c - (X_r^c - X_a) \sin \psi_r^c$.

e_{ct} is the signed distance between the vehicle and the reference trajectory (see next slide).

Lane keeping

- Define the 3D vectors

$$\rho \doteq \begin{bmatrix} X_r^c - X_a \\ Y_r^c - Y_a \\ 0 \end{bmatrix} \quad \text{vector from } (X_a, Y_a) \text{ to } (X_r^c, Y_r^c)$$

$$\xi \doteq \begin{bmatrix} \cos \psi_r^c \\ \sin \psi_r^c \\ 0 \end{bmatrix} \quad \text{reference direction of motion.}$$

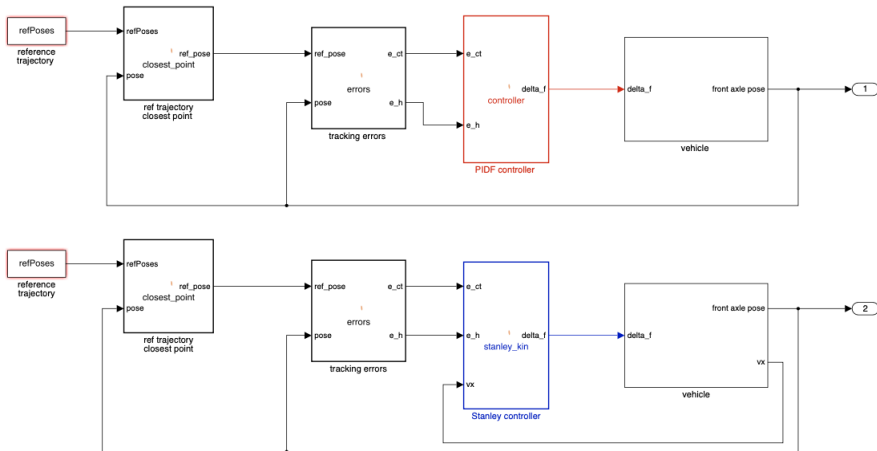
- Their cross-product is $\xi \times \rho = \begin{bmatrix} 0 \\ 0 \\ e_{ct} \end{bmatrix}$.
- Since ξ and ρ are orthogonal, $\|\xi \times \rho\| = \|\xi\| \|\rho\| = \|\rho\|$. Hence,

$$|e_{ct}| = \|\rho\| = \|(X_r^c, Y_r^c) - (X_a, Y_a)\|.$$

- The sign of e_{ct} is as follows:
 - ▶ vehicle on the left of reference trajectory $\iff e_{ct} < 0$
 - ▶ vehicle on the right of reference trajectory $\iff e_{ct} > 0$.

Lane keeping

Closed-loop schemes

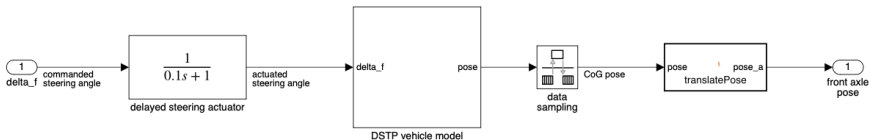


- Two controller types considered: PIDF, Stanley.

Lane keeping

Vehicle model

- DSTP: dynamic single-track vehicle model with Pacejka's tire formula.
 - ▶ Vehicle parameters: $l_f = 1.2$ m, $l_r = 1.6$ m, $m = 1575$ kg, $J = 4000$ kg m², $c_f = c_r = 27e3$ N/rad.
 - ▶ Tire parameters: $p_1 = 3863$ N, $p_2 = 1.5$, $p_4 = -0.5$, and $p_3 = c_f/p_1/p_2$ (front tire) or $p_3 = c_r/p_1/p_2$ (rear tire).
 - ▶ A constant speed v_x is directly imposed in "DSTP vehicle model".
- Other blocks:
 - ▶ steering actuator dynamics
 - ▶ pose translation from CoG to front axel
 - ▶ data sampling blocks (sampling time = 0.05 s).



Lane keeping

Designed controllers

- **PIDF1** (discrete-time):
 - ▶ Uses the cross-track error as the feedback variable.
 - ▶ Designed using the Simulink PID tuner (sampling time $T_s = 0.05$ s).
- **PIDF2** (discrete-time):
 - ▶ Uses the cross-track error and heading error as feedback variables.
 - ▶ Designed using the Simulink PID tuner (sampling time $T_s = 0.05$ s).
- **Stanley kinematic controller**:
 - ▶ “Classical” geometric path tracking controller, consisting of a simple nonlinear feedback law (winner of a DARPA challenge).
 - ▶ The variables used for feedback are the cross-track error and the heading error.

Lane keeping

PID controllers

- Discrete-time PIDF controller general form:

$$C(z) = K_P + K_I \frac{T_s}{z-1} + K_D \frac{z-1}{T_f(z-1)+T_s}.$$

- Consider the following controllers:
 - ▶ $C_{ct}(z)$: PIDF controller using e_{ct} as the feedback variable.
 - ▶ $C_h(z)$: PIDF controller using e_h as the feedback variable.
- PIDF1 control law: $\delta_f = C_{ct}(z) e_{ct}$.
- PIDF2 control law: $\delta_f = C_{ct}(z) e_{ct} + C_h(z) e_h$.
- Controller design: PIDF controllers' parameters designed using the Simulink PID tuner, assuming a constant speed $v_x = 40$ km/h.

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Stanley's kinematic controller

- Define

$$\delta_o = e_h + \text{atan} \left(\frac{K_S e_{ct}}{v_b + v_x} \right)$$

where v_x is the vehicle long. speed, v_b is a “small” term introduced to avoid a null denominator, and K_S is a gain to be tuned.

- Stanley's kinematic control law:

$$\delta_f = \text{sat}(\delta_o, \delta_m) \doteq \begin{cases} \delta_m, & \delta_o \geq \delta_m \\ \delta_o, & -\delta_m < \delta_o < \delta_m \\ -\delta_m, & \delta_o \leq -\delta_m \end{cases}$$

where δ_m is the maximum steering angle.

- When applied to the kinematic single-track model, this control law guarantees convergence to zero of e_{ct} and e_h .
- Parameters: $K_S = 2$, $v_b = 1$, $\delta_m = 35^\circ$ (from a Matlab tutorial).

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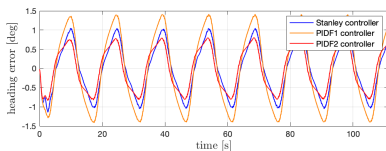
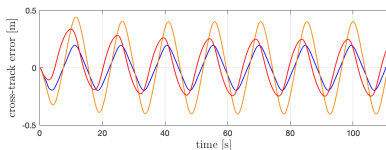
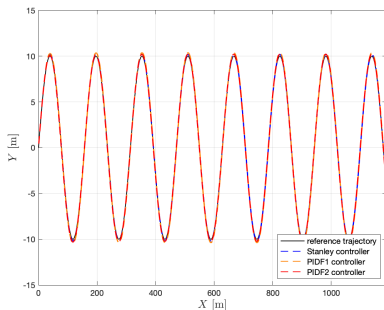
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Simulation results for $v_x = 40$ km/h

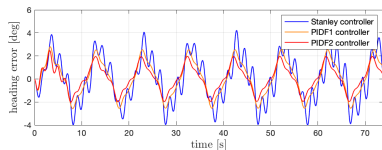
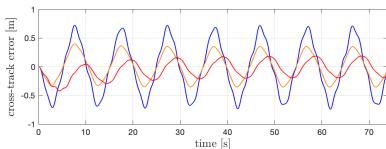
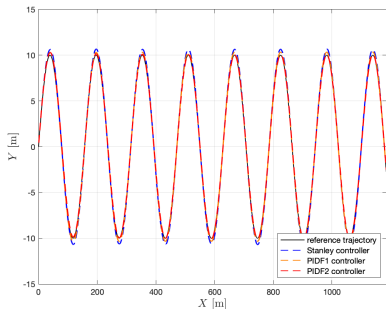
- Relatively challenging reference trajectory: $Y_r = 10 \sin(0.04X_r)$.



- For all the controllers $|e_{ct}| < 0.5$ m.
- Stanley provides the best performance: $|e_{ct}| < 0.2$ m.
- PIDF2 shows improvements wrt PIDF1.

Lane keeping

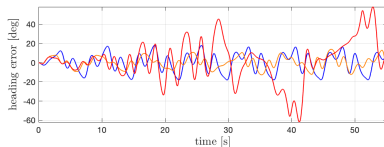
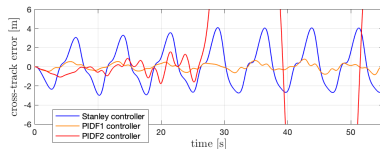
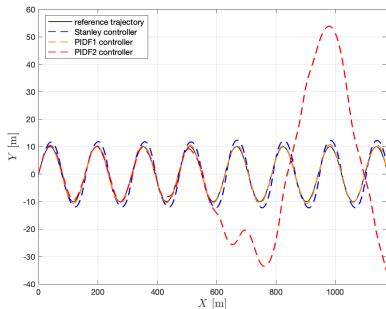
Simulation results for $v_x = 60$ km/h



- The performance of all controllers worsen.
- Stanley shows the worst degradation.
- PDF2 shows improvements wrt PDF1.
- In (more realistic) scenarios with smaller curvatures, smaller errors are obtained.

Lane keeping

Simulation results for $v_x = 80$ km/h



- PDF2 does not work anymore (failure depends on the tuning).
- PDF1 provides the best performance $|e_{ct}| < 1$ m. However, values $|e_{ct}| \cong 1$ m are not acceptable in real scenarios.
- **Remark:** LTI controllers may only work in certain speed ranges and fail in other ranges \rightarrow gain-scheduling/nonlinear control.