Four Wheel Steering (4WS) Control

Name: Rakshith Vishwanatha ASU ID: 1217756241 Mobile: 480-781-6116

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A. Bike Dynamics Equation

$$(I_{bz} + I_{fz})\ddot{\phi}_b = (I_{fx} - I_{fy})\{\Omega^2 cos(\phi_w)^2 sin(\phi_b)cos(\phi_b) - \Omega cos(\phi_w)(cos(2\phi_b) + I_{fz})\dot{\phi}_w\} + (m_b h_b + m_f h_f)g sin(\phi_b)$$
(1)

$$X = \begin{bmatrix} \phi_b \\ \dot{\phi}_b \\ \phi_w \end{bmatrix}$$

$$u = \dot{\phi}_w$$
(2)

$$\dot{X} = f(X) + g(X)u \tag{3}$$

$$\begin{bmatrix}
\dot{\phi_{b}} \\
\dot{\phi_{b}} \\
\dot{\phi_{w}}
\end{bmatrix} = \begin{bmatrix}
\frac{(I_{fx} - I_{fy})}{(I_{bz} + I_{fz})} \{\Omega^{2} cos(\phi_{w})^{2} sin(\phi_{b}) cos(\phi_{b})\} + \\
\frac{1}{(I_{bz} + I_{fz})} (m_{b}h_{b} + m_{f}h_{f}) g sin(\phi_{b})
\end{bmatrix} + \\
0 \\
\begin{bmatrix}
0 \\
\frac{(I_{fx} - I_{fy})}{(I_{bz} + I_{fz})} \{-\Omega cos(\phi_{w}) (cos(2\phi_{b}) + I_{fz}) \dot{\phi_{w}}\} \\
1
\end{bmatrix} \dot{\phi_{w}}$$
(4)

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B. Control Barrier Equations

Requirement is:

$$|\phi_b| \le \phi_{bMax}$$

$$|\dot{\phi}_b| \le \dot{\phi}_{bMax}$$

$$|\phi_w| \le \phi_{wMax}$$
(5)

So CBFs are defined as:

$$|\phi_b| \le \phi_{bMax}$$

$$|\dot{\phi}_b| \le \dot{\phi}_{bMax}$$

$$|\phi_w| \le \phi_{wMax}$$
(6)

This leads to siz CBF equations:

$$h_{1}(X): \quad 0 \leq \phi_{bMax} - \phi_{b}$$

$$h_{2}(X): \quad 0 \leq \phi_{bMax} + \phi_{b}$$

$$h_{3}(X): \quad 0 \leq \dot{\phi}_{bMax} - \dot{\phi}_{b}$$

$$h_{4}(X): \quad 0 \leq \dot{\phi}_{bMax} + \dot{\phi}_{b}$$

$$h_{5}(X): \quad 0 \leq \phi_{wMax} - \phi_{w}$$

$$h_{6}(X): \quad 0 \leq \phi_{wMax} + \phi_{w}$$

$$(7)$$

C. QP Solution

$$u^* = \arg\min_{u \in U} \frac{1}{2} ||u_{des} - u_{asif}||^2$$
s.t. $L_f h_i + L_g h_i u + \alpha h_i \ge 0$
(8)

The Lie derivative of a scalar function such as h (as opposed to a vector function) is given by:

$$Lfh = \nabla h f(X)$$

$$= \left\langle \left[\frac{\delta}{\delta x_1}, \frac{\delta}{\delta x_2}, \frac{\delta}{\delta x_3} \right], h \right\rangle f(X)$$

$$= \left\langle \left[\frac{\delta}{\delta \phi_b}, \frac{\delta}{\delta \dot{\phi}_b}, \frac{\delta}{\delta \phi_w} \right], h \right\rangle f(X)$$
(9)

Thus, the Lie derivative inequalities provided above in the squared error optimizer are:

 $L_f h_1 + L_g h_1 u + \alpha h_1 \ge 0$

$$\left\langle [-1, \ 0, \ 0], h_1(X) \right\rangle f(X) + \left\langle [-1, \ 0, \ 0], h_1(X) \right\rangle g(X)u + \alpha h_1 \ge 0
- f_1(X) + 0 + \alpha h_6 \ge 0
- \dot{\phi}_b + 0 + \alpha (\phi_{bMax} - \phi_b) \ge 0
L_f h_2 + L_g h_2 u + \alpha h_2 \ge 0
\left\langle [+1, \ 0, \ 0], h_2(X) \right\rangle f(X) + \left\langle [+1, \ 0, \ 0], h_2(X) \right\rangle g(X)u + \alpha h_2 \ge 0
+ f_1(X) + 0 + \alpha h_6 \ge 0
+ \dot{\phi}_b + 0 + \alpha (\phi_{bMax} + \phi_b) \ge 0$$
(10)

$$\begin{split} & L_f h_3 + L_g h_3 u + \alpha h_3 \geq 0 \\ & \Big\langle [0, -1, 0], h_3(X) \Big\rangle f(X) + \Big\langle [0, -1, 0], h_3(X) \Big\rangle g(X) u + \alpha h_3 \geq 0 \\ & - f_2(X) - g_2(X) u + \alpha h_3 \geq 0 \end{split}$$

 $-f_2(X) - g_2(X)u + \alpha h_3 \ge 0$

$$-\left[\frac{(I_{fx} - I_{fy})}{(I_{bz} + I_{fz})} \{\Omega^{2} cos(\phi_{w})^{2} sin(\phi_{b}) cos(\phi_{b})\} + \frac{1}{(I_{bz} + I_{fz})} (m_{b}h_{b} + m_{f}h_{f}) g sin(\phi_{b})\right]$$

$$-\left[\frac{(I_{fx} - I_{fy})}{(I_{bz} + I_{fz})} \{-\Omega cos(\phi_{w}) (cos(2\phi_{b}) + I_{fz}) \dot{\phi_{w}}\}\right]$$

$$+\alpha (\dot{\phi}_{bMax} - \dot{\phi}_{b}) \geq 0$$

$$L_{f}h_{3} + L_{g}h_{3}u + \alpha h_{3} \geq 0$$

$$\langle [0, -1, 0], h_{3}(X) \rangle f(X) + \langle [0, -1, 0], h_{3}(X) \rangle g(X)u + \alpha h_{3} \geq 0$$

$$(12)$$

$$+\left[\frac{(I_{fx}-I_{fy})}{(I_{bz}+I_{fz})}\{\Omega^{2}cos(\phi_{w})^{2}sin(\phi_{b})cos(\phi_{b})\}+\right]$$

$$\frac{1}{(I_{bz}+I_{fz})}(m_{b}h_{b}+m_{f}h_{f})g\sin(\phi_{b})$$

$$+\left[\frac{(I_{fx}-I_{fy})}{(I_{bz}+I_{fz})}\{-\Omega\cos(\phi_{w})(\cos(2\phi_{b})+I_{fz})\dot{\phi_{w}}\}\right]$$

$$+\alpha(\dot{\phi}_{bMax}+\dot{\phi}_{b})>0$$
(13)

$$L_{f}h_{5} + L_{g}h_{5}u + \alpha h_{5} \geq 0$$

$$\left\langle [0, 0, -1], h_{5}(X) \right\rangle f(X) + \left\langle [0, 0, -1], h_{5}(X) \right\rangle g(X)u + \alpha h_{5} \geq 0$$

$$0 - g_{3}(X)u + \alpha h_{6} \geq 0$$

$$0 - \dot{\phi_{w}} + \alpha(\phi_{wMax} - \phi_{w}) \geq 0$$
(14)

$$L_{f}h_{6} + L_{g}h_{6}u + \alpha h_{6} \geq 0$$

$$\left\langle [0, 0, -1], h_{6}(X) \right\rangle f(X) + \left\langle [0, 0, -1], h_{6}(X) \right\rangle g(X)u + \alpha h_{6} \geq 0$$

$$0 + g_{3}(X)u + \alpha h_{6} \geq 0$$

$$0 + \dot{\phi}_{w} + \alpha(\phi_{wMax} + \phi_{w}) \geq 0$$
(15)

D. Calculation of α in QP Equations

 α is a class Kappa function defined as:

$$\alpha(r) \triangleq -\min_{\substack{x \in A(r) \\ u \in U}} \left(\min_{i \in [i,n]} L_f h_i + L_g h_i u \right)$$
where,
$$A(r) \triangleq \{ x \in S \mid \exists i \in [1,n], \ 0 \leq h_i(X) \leq r \} \subseteq S$$
where, S is not just the safety set, but is the VIABLE safety set.

In its expanded form this means: α is a class Kappa function defined as:

$$\alpha(r) \triangleq - \min_{\substack{x \in A(r) \\ u \in U}} \left(\min \begin{bmatrix} L_f h_1 + L_g h_1 u \\ L_f h_2 + L_g h_2 u \\ L_f h_3 + L_g h_3 u \\ L_f h_4 + L_g h_4 u \\ L_f h_5 + L_g h_5 u \\ L_f h_6 + L_g h_6 u \end{bmatrix} \right)$$

with the constraints placed by the A(r) definition/inequality

$$0 \le h_1 \le r$$

$$0 \le h_2 \le r$$

$$0 \le h_3 \le r$$

$$0 \le h_4 \le r$$

$$0 \le h_5 \le r$$

$$0 \le h_6 \le r$$

$$(17)$$

The inequalities $0 \le h_i \le r$ place the additional constraint/relation between the α Kappa function's r and the CBF state constraints of (obtained by summing up similar CBF equations and simplifying):

$$-\frac{r}{2} \le \phi_b \le +\frac{r}{2}$$

$$-\frac{r}{2} \le \dot{\phi}_b \le +\frac{r}{2}$$

$$-\frac{r}{2} \le \phi_w \le +\frac{r}{2}$$
(18)

So CBF max state values chosen should be within the overall r/2 bounds placed by the class Kappa function.

When we put the $\alpha(r)$ equation in words, it effectively means:

- Create an empty 2D table of size $x \times u$ to represent all combos of states and controls
- The number of states itself that can be created/checked is defined by all states within the viable safety set which satisfy $0 \le h_i \le r$ given in the A(r) equation, and expanded in (17)'s inequalities.
- For each of the entries in the 2D matrix, calculate a $n \times 1$ vector corresponding to each of the CBF equations.
- Now for each of the vectors in the 2D matrix, calculate and store in-place the minimum of that vector
- Then, calculate the min over the whole 2D matrix
- Negate this value to obtain the value of α

Ofcourse, all these steps can be vectorized and handled more efficiently using an optimization toolbox like fmincon() in MATLAB.