

Robotic Unicycle Loss Function

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The state of the system is given as

$$\mathbf{x} = [\dot{\theta} \quad \dot{\phi} \quad \dot{\psi}_w \quad \dot{\psi}_f \quad \dot{\psi}_t \quad \theta \quad \phi \quad \psi_w \quad \psi_f \quad \psi_t]^\top,$$

where θ is the roll angle, ϕ is yaw, ψ_w wheel angle, ψ_f pitch angle and ψ_t the disc angle. The *reward* R and *loss* L of being in state \mathbf{x} is

$$R(\mathbf{x}) = \exp\left(-\frac{\alpha}{2}d(\mathbf{x})^2 - \frac{\phi^2}{2(4\pi)^2}\right), \quad \text{and} \quad L(\mathbf{x}) = 1 - R(\mathbf{x}),$$

where α is a (positive) constant, and d^2 is the squared distance between the tip of the unicycle and its ideal location when upright in the centre of the coordinate system.

$$\begin{aligned} d(\mathbf{x})^2 &= d_x^2 + d_y^2 + d_z^2 \quad \text{with} \\ d_x &= x_c - r \sin \psi_f \\ d_y &= y_c - (r + r_w) \sin \theta \\ d_z &= (r_w + r - r_w \cos \theta - r \cos \theta \cos \psi_f)^2 \\ &= (r_w + r - r_w \cos \theta - \frac{r}{2} \cos(\theta - \psi_f) - \frac{r}{2} \cos(\theta + \psi_f))^2 \end{aligned}$$

Here, r_w is the radius of the wheel and r is the length of the fork; x_c and y_c are coordinates of the laboratory origin in the self-centred coordinate system. Apart from those two latter coordinates, the relevant parts of the state \mathbf{x} are the sine and cosine of θ , the sine of ψ_f , and the difference/sum of the angles θ and ψ_f . Therefore, we define $\alpha := \theta - \psi_f$ and $\beta := \theta + \psi_f$. We assume that the joint distribution of these three variables is approximately Gaussian

$$\mathbf{z} = \begin{bmatrix} x_c \\ \sin \psi_f \\ y_c \\ \sin \theta \\ \cos \theta \\ \cos \alpha \\ \cos \beta \\ \phi \end{bmatrix} \quad \text{and} \quad \mathcal{N}(\mathbf{m}, S).$$

The *expected reward* is

$$\mathbb{E}[R] = \langle R(\mathbf{x}) \rangle_{\mathbf{z}} = \left\langle \exp\left(-\frac{1}{2}(\mathbf{z} - \mathbf{z}_0)^\top \mathbf{T}^{-1}(\mathbf{z} - \mathbf{z}_0)\right) \right\rangle_{\mathbf{z}},$$

where we defined

$$\mathbf{z}_0 := [0 \ 0 \ 0 \ 0 \ 1 \ 1 \ 1 \ 0]^\top$$

and

$$\begin{aligned} \mathbf{T}^{-1} &= a \operatorname{diag} \left\{ \begin{bmatrix} 1 & -r \\ -r & r^2 \end{bmatrix}, \begin{bmatrix} 1 & -(r+r_w) \\ -(r+r_w) & (r+r_w)^2 \end{bmatrix}, \begin{bmatrix} \frac{r_w^2}{2} & \frac{r_w r}{2} & \frac{r_w r}{2} \\ \frac{r_w r}{2} & \frac{r^2}{4} & \frac{r^2}{4} \\ \frac{r_w r}{2} & \frac{r^2}{4} & \frac{r^2}{4} \end{bmatrix}, \left[\frac{1}{16\pi^2} \right] \right\} \\ &= a \operatorname{diag}(\mathbf{C}_x \mathbf{C}_x^\top, \mathbf{C}_y \mathbf{C}_y^\top, \mathbf{C}_z \mathbf{C}_z^\top) \end{aligned}$$

with

$$\mathbf{C}_x = \begin{bmatrix} 1 \\ -r \end{bmatrix} \quad \mathbf{C}_y = \begin{bmatrix} 1 \\ -(r+r_w) \end{bmatrix} \quad \mathbf{C}_z = \begin{bmatrix} r_w \\ r/2 \\ r/2 \end{bmatrix}$$