

pendent noise ε_{b^2} with zero mean and variance b^2 :

$$(5.37) \quad \hat{\delta}_{\text{rot1}} = \delta_{\text{rot1}} - \varepsilon_{\alpha_1 \delta_{\text{rot1}}^2 + \alpha_2 \delta_{\text{trans}}^2}$$

$$(5.38) \quad \hat{\delta}_{\text{trans}} = \delta_{\text{trans}} - \varepsilon_{\alpha_3 \delta_{\text{trans}}^2 + \alpha_4 \delta_{\text{rot1}}^2 + \alpha_4 \delta_{\text{rot2}}^2}$$

$$(5.39) \quad \hat{\delta}_{\text{rot2}} = \delta_{\text{rot2}} - \varepsilon_{\alpha_1 \delta_{\text{rot2}}^2 + \alpha_2 \delta_{\text{trans}}^2}$$

As in the previous section, ε_{b^2} is a zero-mean noise variable with variance b^2 . The parameters α_1 to α_4 are robot-specific error parameters, which specify the error accrued with motion.

Consequently, the true position, x_t , is obtained from x_{t-1} by an initial rotation with angle $\hat{\delta}_{\text{rot1}}$, followed by a translation with distance $\hat{\delta}_{\text{trans}}$, followed by another rotation with angle $\hat{\delta}_{\text{rot2}}$. Thus,

$$(5.40) \quad \begin{pmatrix} x' \\ y' \\ \theta' \end{pmatrix} = \begin{pmatrix} x \\ y \\ \theta \end{pmatrix} + \begin{pmatrix} \hat{\delta}_{\text{trans}} \cos(\theta + \hat{\delta}_{\text{rot1}}) \\ \hat{\delta}_{\text{trans}} \sin(\theta + \hat{\delta}_{\text{rot1}}) \\ \hat{\delta}_{\text{rot1}} + \hat{\delta}_{\text{rot2}} \end{pmatrix}$$

Notice that algorithm `sample_motion_model_odometry` implements Equations (5.34) through (5.40).

The algorithm `motion_model_odometry` is obtained by noticing that lines 5-7 compute the motion parameters $\hat{\delta}_{\text{rot1}}$, $\hat{\delta}_{\text{trans}}$, and $\hat{\delta}_{\text{rot2}}$ for the hypothesized pose x_t , relative to the initial pose x_{t-1} . The difference of both,

$$(5.41) \quad \delta_{\text{rot1}} - \hat{\delta}_{\text{rot1}}$$

$$(5.42) \quad \delta_{\text{trans}} - \hat{\delta}_{\text{trans}}$$

$$(5.43) \quad \delta_{\text{rot2}} - \hat{\delta}_{\text{rot2}}$$

is the *error* in odometry, assuming of course that x_t is the true final pose. The error model (5.37) to (5.39) implies that the probability of these errors is given by

$$(5.44) \quad p_1 = \varepsilon_{\alpha_1 \delta_{\text{rot1}}^2 + \alpha_2 \delta_{\text{trans}}^2} (\delta_{\text{rot1}} - \hat{\delta}_{\text{rot1}})$$

$$(5.45) \quad p_2 = \varepsilon_{\alpha_3 \delta_{\text{trans}}^2 + \alpha_4 \delta_{\text{rot1}}^2 + \alpha_4 \delta_{\text{rot2}}^2} (\delta_{\text{trans}} - \hat{\delta}_{\text{trans}})$$

$$(5.46) \quad p_3 = \varepsilon_{\alpha_1 \delta_{\text{rot2}}^2 + \alpha_2 \delta_{\text{trans}}^2} (\delta_{\text{rot2}} - \hat{\delta}_{\text{rot2}})$$

with the distributions ε defined as above. These probabilities are computed in lines 8-10 of our algorithm `motion_model_odometry`, and since the errors are assumed to be independent, the joint error probability is the product $p_1 \cdot p_2 \cdot p_3$ (c.f., line 11).