

The function  $g(u_t, \mu_{t-1})$  is simply obtained by replacing the exact state  $x_{t-1}$ —which we do not know—by our expectation  $\mu_{t-1}$ —which we know. The Jacobian  $G_t$  is the derivative of the function  $g$  with respect to  $x_{t-1}$  evaluated at  $u_t$  and  $\mu_{t-1}$ :

$$(7.8) \quad G_t = \frac{\partial g(u_t, \mu_{t-1})}{\partial x_{t-1}} = \begin{pmatrix} \frac{\partial x'}{\partial \mu_{t-1,x}} & \frac{\partial x'}{\partial \mu_{t-1,y}} & \frac{\partial x'}{\partial \mu_{t-1,\theta}} \\ \frac{\partial y'}{\partial \mu_{t-1,x}} & \frac{\partial y'}{\partial \mu_{t-1,y}} & \frac{\partial y'}{\partial \mu_{t-1,\theta}} \\ \frac{\partial \theta'}{\partial \mu_{t-1,x}} & \frac{\partial \theta'}{\partial \mu_{t-1,y}} & \frac{\partial \theta'}{\partial \mu_{t-1,\theta}} \end{pmatrix}$$

Here  $\mu_{t-1} = (\mu_{t-1,x} \ \mu_{t-1,y} \ \mu_{t-1,\theta})^T$  denotes the mean estimate factored into its individual three values, and  $\frac{\partial x'}{\partial \mu_{t-1,x}}$  is short for the derivative of  $g$  along the  $x'$  dimension, taken with respect to  $x$  at  $\mu_{t-1}$ . Calculating these derivatives from Equation (7.6) gives us the following matrix:

$$(7.9) \quad G_t = \begin{pmatrix} 1 & 0 & \frac{v_t}{\omega_t}(-\cos \mu_{t-1,\theta} + \cos(\mu_{t-1,\theta} + \omega_t \Delta t)) \\ 0 & 1 & \frac{v_t}{\omega_t}(-\sin \mu_{t-1,\theta} + \sin(\mu_{t-1,\theta} + \omega_t \Delta t)) \\ 0 & 0 & 1 \end{pmatrix}$$

To derive the covariance of the additional motion noise,  $\mathcal{N}(0, R_t)$ , we first determine the covariance matrix  $M_t$  of the noise in *control space*. This follows directly from the motion model in Equation (7.5):

$$(7.10) \quad M_t = \begin{pmatrix} \alpha_1 v_t^2 + \alpha_2 \omega_t^2 & 0 \\ 0 & \alpha_3 v_t^2 + \alpha_4 \omega_t^2 \end{pmatrix}$$

The motion model in (7.6) requires this motion noise to be mapped into *state space*. The transformation from control space to state space is performed by another linear approximation. The Jacobian needed for this approximation, denoted  $V_t$ , is the derivative of the motion function  $g$  with respect to the motion parameters, evaluated at  $u_t$  and  $\mu_{t-1}$ :

$$(7.11) \quad V_t = \frac{\partial g(u_t, \mu_{t-1})}{\partial u_t} = \begin{pmatrix} \frac{\partial x'}{\partial v_t} & \frac{\partial x'}{\partial \omega_t} \\ \frac{\partial y'}{\partial v_t} & \frac{\partial y'}{\partial \omega_t} \\ \frac{\partial \theta'}{\partial v_t} & \frac{\partial \theta'}{\partial \omega_t} \end{pmatrix} = \begin{pmatrix} \frac{-\sin \theta + \sin(\theta + \omega_t \Delta t)}{\omega_t} & \frac{v_t(\sin \theta - \sin(\theta + \omega_t \Delta t))}{\omega_t^2} + \frac{v_t \cos(\theta + \omega_t \Delta t) \Delta t}{\omega_t} \\ \frac{\cos \theta - \cos(\theta + \omega_t \Delta t)}{\omega_t} & -\frac{v_t(\cos \theta - \cos(\theta + \omega_t \Delta t))}{\omega_t^2} + \frac{v_t \sin(\theta + \omega_t \Delta t) \Delta t}{\omega_t} \\ 0 & \Delta t \end{pmatrix}$$

The multiplication  $V_t M_t V_t^T$  then provides an approximate mapping between the motion noise in control space to the motion noise in state space.