

Lab 09 – Some Tips

Autonomous Navigation of Mobile Robots

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Key Steps of FastSLAM

- Extend the path posterior by sampling a new pose for each sample

$$x_t^{[k]} \sim p(x_t | x_{t-1}^{[k]}, u_t)$$

- Compute particle weight

$$w^{[k]} = |2\pi Q|^{-\frac{1}{2}} \exp \left\{ -\frac{1}{2} (z_t - \hat{z}^{[k]})^T Q^{-1} (z_t - \hat{z}^{[k]}) \right\}$$

$\hat{z}^{[k]}$ – expected observation

Q – measurement covariance

- Update belief of observed landmarks (EKF update rule)
- Resample

FastSLAM

Part 1

```
1:  FastSLAM1.0_known_correspondence( $z_t, c_t, u_t, \mathcal{X}_{t-1}$ ):
2:      for  $k = 1$  to  $N$  do                                // loop over all particles
3:          Let  $\langle x_{t-1}^{[k]}, \langle \mu_{1,t-1}^{[k]}, \Sigma_{1,t-1}^{[k]} \rangle, \dots \rangle$  be particle  $k$  in  $\mathcal{X}_{t-1}$ 
4:           $x_t^{[k]} \sim p(x_t \mid x_{t-1}^{[k]}, u_t)$           // sample pose
5:           $j = c_t$                                           // observed feature
6:          if feature  $j$  never seen before
7:               $\mu_{j,t}^{[k]} = h^{-1}(z_t, x_t^{[k]})$         // initialize mean
8:               $H = h'(\mu_{j,t}^{[k]}, x_t^{[k]})$               // calculate Jacobian
9:               $\Sigma_{j,t}^{[k]} = H^{-1} Q_t (H^{-1})^T$     // initialize covariance
10:              $w^{[k]} = p_0$                                 // default importance weight
11:         else
```

FastSLAM

Part 2

```
11:         else
12:              $\langle \mu_{j,t}^{[k]}, \Sigma_{j,t}^{[k]} \rangle = \text{EKF-Update}(\dots)$     // update landmark
13:              $w^{[k]} = |2\pi Q|^{-\frac{1}{2}} \exp \left\{ -\frac{1}{2} (z_t - \hat{z}^{[k]})^T Q^{-1} (z_t - \hat{z}^{[k]}) \right\}$ 
                ↑                               ↑
    measurement cov.  $Q = H \Sigma_{j,t-1}^{[k]} H^T + Q_t$     exp. observation
14:         endif
15:         for all unobserved features  $j'$  do
16:              $\langle \mu_{j',t}^{[k]}, \Sigma_{j',t}^{[k]} \rangle = \langle \mu_{j',t-1}^{[k]}, \Sigma_{j',t-1}^{[k]} \rangle$     // leave unchanged
17:         endfor
18:     endfor
19:      $\mathcal{X}_t = \text{resample} \left( \left\langle x_t^{[k]}, \langle \mu_{1,t}^{[k]}, \Sigma_{1,t}^{[k]} \rangle, \dots, w^{[k]} \right\rangle_{k=1, \dots, N} \right)$ 
20:     return  $\mathcal{X}_t$ 
```

FastSLAM

Part 2 Detailed

```
11:         else
12:              $\hat{z}^{[k]} = h(\mu_{j,t-1}^{[k]}, x_t^{[k]})$  // measurement prediction
13:              $H = h'(\mu_{j,t-1}^{[k]}, x_t^{[k]})$  // calculate Jacobian
14:              $Q = H \Sigma_{j,t-1}^{[k]} H^T + Q_t$  // measurement covariance
15:              $K = \Sigma_{j,t-1}^{[k]} H^T Q^{-1}$  // calculate Kalman gain
16:              $\mu_{j,t}^{[k]} = \mu_{j,t-1}^{[k]} + K(z_t - \hat{z}^{[k]})$  // update mean
17:              $\Sigma_{j,t}^{[k]} = (I - K H) \Sigma_{j,t-1}^{[k]}$  // update covariance
18:              $w^{[k]} = |\pi Q|^{-\frac{1}{2}} \exp \left\{ -\frac{1}{2} (z_t - \hat{z}^{[k]})^T \right.$   

                $\left. Q^{-1} (z_t - \hat{z}^{[k]}) \right\}$  // importance factor
19:         endif
20:         for all unobserved features  $j'$  do
21:              $\langle \mu_{j',t}^{[k]}, \Sigma_{j',t}^{[k]} \rangle = \langle \mu_{j',t-1}^{[k]}, \Sigma_{j',t-1}^{[k]} \rangle$  // leave unchanged
22:         endfor
23:     endfor
24:
25:      $\mathcal{X}_t = \text{resample} \left( \left\langle x_t^{[k]}, \left\langle \mu_{1,t}^{[k]}, \Sigma_{1,t}^{[k]} \right\rangle, \dots, w^{[k]} \right\rangle_{k=1, \dots, N} \right)$ 
26:     return  $\mathcal{X}_t$ 
```