

AA 274: Principles of Robotic Autonomy

Algos for Lecture 11

Data: $\{p_{k,t-1}\}, u_t, z_t$
Result: $\{p_{k,t}\}$
 $\eta \leftarrow 0;$
foreach k **do**
 $\bar{p}_{k,t} = \sum_i \tilde{\eta} |x_{k,t}| p(\hat{x}_k | u_t, \hat{x}_i) p_{i,t-1};$
end

foreach k **do**
 $p_{k,t} = \eta p(z_t | \hat{x}_k, m) \bar{p}_{k,t};$
 $\eta \leftarrow \eta + p_{k,t};$
end

foreach k **do**
 $p_{k,t} = \eta^{-1} p_{k,t};$
end

Return $\{p_{k,t}\}$

Data: $\mathcal{X}_{t-1}, u_t, z_t$

Result: \mathcal{X}_t

$\bar{\mathcal{X}}_t = \mathcal{X}_t = \emptyset;$

for $i = 1$ **to** M **do**

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

$w_t^{[m]} = p(z_t | x_t^{[m]});$

$\bar{\mathcal{X}}_t = \bar{\mathcal{X}}_t \cup (x_t^{[m]}, w_t^{[m]});$

end

for $m = 1$ **to** M **do**

 Draw i with probability $\propto w_t^{[i]};$

 Add $x_t^{[i]}$ to $\mathcal{X}_t;$

end

Return \mathcal{X}_t

Data: $bel(x_{t-1}), u_t, z_t, \textcolor{red}{m}$
Result: $bel(x_t)$
foreach x_t **do**
 $\overline{bel}(x_t) = \int p(x_t | u_t, x_{t-1}, \textcolor{red}{m}) bel(x_{t-1}) dx_{t-1};$
 $bel(x_t) = \eta p(z_t | x_t, \textcolor{red}{m}) \overline{bel}(x_t);$
end
Return $bel(x_t)$

Data: $\mathcal{X}_{t-1}, u_t, z_t$

Result: \mathcal{X}_t

$\bar{\mathcal{X}}_t = \mathcal{X}_t = \emptyset;$

for $i = 1$ **to** M **do**

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

$w_t^{[m]} = p(z_t | x_t^{[m]});$

$\bar{\mathcal{X}}_t = \bar{\mathcal{X}}_t \cup (x_t^{[m]}, w_t^{[m]});$

end

for $i = 1$ **to** M **do**

 Draw i with probability $\propto w_t^{[i]};$

 Add $x_t^{[i]}$ to $\mathcal{X}_t;$

end

Return \mathcal{X}_t

Data: $\{p_{k,t-1}\}, u_t, z_t$
Result: $\{p_{k,t}\}$
foreach k **do**
 $\bar{p}_{k,t} = \sum_i p(X_t = x_k | u_t, X_{t-1} = x_i) p_{i,t-1};$
 $p_{k,t} = \eta p(z_t | X_t = x_k) \bar{p}_{k,t};$
end
Return $\{p_{k,t}\}$

Data: $(\mu_{t-1}, \Sigma_{t-1}), u_t, z_t$
Result: (μ_t, Σ_t)
 $\mathcal{X}_{t-1} = (\mu_{t-1} \quad \mu_{t-1} + \gamma\sqrt{\Sigma_{t-1}} \quad \mu_{t-1} - \gamma\sqrt{\Sigma_{t-1}});$
 $\mathcal{X}_t^* = g(u_t, \mathcal{X}_{t-1});$
 $\bar{\mu}_t = \sum_{i=0}^{2n} w_m^{[i]} \bar{\mathcal{X}}_t^{*[i]};$
 $\bar{\Sigma}_t = \sum_{i=0}^{2n} w_c^{[i]} (\bar{\mathcal{X}}_t^{*[i]} - \bar{\mu}_t) (\bar{\mathcal{X}}_t^{*[i]} - \bar{\mu}_t)^T + R_t;$
 $\bar{\mathcal{X}}_t = (\bar{\mu}_t \quad \bar{\mu}_t + \gamma\sqrt{\bar{\Sigma}_t} \quad \bar{\mu}_t - \gamma\sqrt{\bar{\Sigma}_t});$
 $\bar{\mathcal{Z}}_t = h(\bar{\mathcal{X}}_t);$
 $\hat{z}_t = \sum_{i=0}^{2n} w_m^{[i]} \bar{\mathcal{Z}}_t^{[i]};$
 $S_t = \sum_{i=0}^{2n} w_c^{[i]} (\bar{\mathcal{Z}}_t^{[i]} - \hat{z}_t) (\bar{\mathcal{Z}}_t^{[i]} - \hat{z}_t)^T + Q_t;$
 $\bar{\Sigma}_t^{x,z} = \sum_{i=0}^{2n} w_c^{[i]} (\bar{\mathcal{X}}_t^{[i]} - \bar{\mu}_t) (\bar{\mathcal{Z}}_t^{[i]} - \hat{z}_t)^T;$
 $K_t = \bar{\Sigma}_t^{x,z} S_t^{-1};$
 $\mu_t = \bar{\mu}_t + K_t(z_t - \hat{z}_t);$
 $\Sigma_t = \bar{\Sigma}_t - K_t S_t K_t^T;$
Return (μ_t, Σ_t)

Data: $(\mu_{t-1}, \Sigma_{t-1}), u_t, z_t$
Result: (μ_t, Σ_t)
 $\bar{\mu}_t = g(u_t, \mu_{t-1});$
 $\bar{\Sigma}_t = G_t \Sigma_{t-1} G_t^T + R_t;$
 $K_t = \bar{\Sigma}_t H_t^T (H_t \bar{\Sigma}_t H_t^T + Q_t)^{-1};$
 $\mu_t = \bar{\mu}_t + K_t(z_t - h(\bar{\mu}_t));$
 $\Sigma_t = (I - K_t H_t) \bar{\Sigma}_t;$
Return (μ_t, Σ_t)

Data: $(\mu_{t-1}, \Sigma_{t-1}), u_t, z_t$
Result: (μ_t, Σ_t)
 $\bar{\mu}_t = A_t \mu_{t-1} + B_t u_t;$
 $\bar{\Sigma}_t = A_t \Sigma_{t-1} A_t^T + R_t;$
 $K_t = \bar{\Sigma}_t C_t^T (C_t \bar{\Sigma}_t C_t^T + Q_t)^{-1};$
 $\mu_t = \bar{\mu}_t + K_t(z_t - C_t \bar{\mu}_t);$
 $\Sigma_t = (I - K_t C_t) \bar{\Sigma}_t;$
Return (μ_t, Σ_t)

Data: $bel(x_{t-1}), u_t, z_t$
Result: $bel(x_t)$
foreach x_t **do**
 $\bar{bel}(x_t) = \int p(x_t | u_t, x_{t-1}) bel(x_{t-1}) dx_{t-1};$
 $bel(x_t) = \eta p(z_t | x_t) \bar{bel}(x_t);$
end
Return $bel(x_t)$

Data: Set S containing N points

Result: Line fitting the points in S

Initialize $n_\alpha \times n_r$ accumulator H with zeros;

foreach $(x_i, y_i) \in S$ **do**

foreach $\alpha \in \{\alpha_1, \dots, \alpha_{n_\alpha}\}$ **do**

 compute $r = x_i \cos \alpha + y_i \sin \alpha$;

$H[\alpha, r] \leftarrow H[\alpha, r] + 1$;

end

end

Choose (α^*, r^*) that corresponds to largest count in H ;

Return line defined by (α^*, r^*)