

扩展卡尔曼滤波

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1 预测更新

设系统在 $t, t+1$ 时刻的状态值有函数关系:

$$x_{t+1} = f(x_t) + w_t, \quad w_t \sim \mathcal{N}(0, Q_t) \quad (1)$$

自然地, 有:

$$\hat{x}_{t+1} = f(\hat{x}_t) \quad (2)$$

其误差

$$\begin{aligned} \varepsilon_{t+1} &= x_{t+1} - \hat{x}_{t+1} \\ &= f(x_t) + w_t - f(\hat{x}_t) \\ &= f(\hat{x}_t) + f'(\hat{x}_t)(x_t - \hat{x}_t) + O((x_t - \hat{x}_t)^2) + w_t - f(\hat{x}_t) \\ &= f'(\hat{x}_t)(x_t - \hat{x}_t) + w_t + O((x_t - \hat{x}_t)^2) \end{aligned} \quad (3)$$

忽略高阶无穷小, 求协方差矩阵得:

$$\begin{aligned} \hat{P}_{t+1} &= E(\varepsilon_{t+1}\varepsilon_{t+1}^T) \\ &= E((f'(\hat{x}_t)(x_t - \hat{x}_t) + O((x_t - \hat{x}_t)^2) + w_t - f(\hat{x}_t))(f'(\hat{x}_t)(x_t - \hat{x}_t) + O((x_t - \hat{x}_t)^2) + w_t - f(\hat{x}_t))^T) \\ &\approx f'(\hat{x}_t)E((x_t - \hat{x}_t)(x_t - \hat{x}_t)^T)f'(\hat{x}_t)^T + E(w_t w_t^T) \\ &= F_t \hat{P}_t F_t^T + Q_t \end{aligned} \quad (4)$$

其中 $F_t = f'(\hat{x}_t)$, Q_t 为随机误差 w_t 的方差。

2 测量更新

设系统状态 x 与观测值 z 之前有函数关系:

$$z = h(x) + v, \quad x \sim \mathcal{N}(\mu, P), \quad v \sim \mathcal{N}(0, R) \quad (5)$$

现有此刻系统状态的估计值 x_0 , 则对观测值的估计为:

$$z_0 = h(x_0) \quad (6)$$

又有此刻的观测值 z , 欲求 x 。通过泰勒展开线性化式(5):

$$z = z_0 + h'(x_0)(x - x_0) + O((x - x_0)^2) + v \quad (7)$$

令 $\delta z = z - z_0$, $\delta x = x - x_0$, 且令 $H = h'(x_0) = \frac{dz}{dx}|_{x=x_0}$ 为 $h: x \rightarrow z$ 的雅可比矩阵, 则有:

$$\delta z \approx H\delta x + v, \quad \delta x \sim \mathcal{N}(0, P), \quad v \sim \mathcal{N}(0, R) \quad (8)$$

解出Kalman增益 $K = PH^T(HPH^T + R)^{-1}$, 则 δx 的最优后验估计为:

$$\hat{\delta x} = 0 + K(\delta z - 0) = K\delta z \quad (9)$$

相应地:

$$\hat{x} = x_0 + \hat{\delta x} = x_0 + K(z - h(x_0)) \quad (10)$$

由于 x 与 δx 的方差相同, 更新 x 的方差估计为:

$$\hat{P} = (I - KH)P \quad (11)$$