

Algorithm 1: Kernel Disturbance Observer algorithm using dynamic regression for parameter estimation

Input: $\mathbf{p}_M = [p_M(t_k - \delta) : p_M(t_k - \delta)]$, $\Theta_M[\Theta_M(t_k - \delta) : \Theta_M(t_k)]$

Output: $\hat{d}_p(t_k)$, $\hat{d}_\Theta(t_k)$

1 Set $\mathbf{w}_1^0 = [1, 1, 1, 1]$, $k = 1$

2 Obtain Full State Estimate

for $\zeta \in [x, y, z, \phi, \theta, \psi]$ **do**

3 **Parameter Estimate for Local Surrogate Model:** Compute

\mathbf{w}_{ζ_k}

$$\hat{\mathbf{w}}_{\zeta_k} = \underset{\mathbf{w}_{\zeta_k}}{\operatorname{argmin}} \{ \gamma \| \mathbf{w}_{\zeta_k} - \hat{\mathbf{w}}_{\zeta_{k-1}} \|_2^2 + \lambda \| \mathbf{w}_{\zeta_k} \|_2^2 + \| \zeta_M - \zeta_{est} \|_2^2 \}$$

4 **State Estimation:** Compute state estimate $\hat{\zeta}(t_k)$

$$\hat{\zeta}(t_k) = \frac{1}{((t_k) - (t_k - \delta))^4} \int_{t_k - \delta}^{t_k} K_{DS}(t_k, \tau) \zeta_M(\tau) d\tau$$

5 **Higher Order Derivatives:** Compute the higher order derivatives

$\hat{\zeta}(t_k)$ and $\hat{\dot{\zeta}}(t_k)$ using (??)-(??) in Section ?? of Appendix.

6 **end**

7 Compute External Disturbances $\hat{d}_p(t_k) = [d_x, d_y, d_z]$ and

$$\hat{d}_\Theta(t_k) = [d_\phi, d_\theta, d_\psi]:$$

$$\hat{d}_\phi = \hat{\ddot{\phi}} - (r_1 \hat{\theta} \hat{\psi} - r_2 \hat{\theta} w + q_1 U_2); \quad \hat{d}_x = \hat{\ddot{x}} - \left((C_{\hat{\phi}} S_{\hat{\theta}} C_{\hat{\psi}} + S_{\hat{\phi}} S_{\hat{\psi}}) \frac{1}{m} U_1 \right)$$

$$\hat{d}_\theta = \hat{\ddot{\theta}} - (r_3 \hat{\phi} \hat{\psi} + r_4 \hat{\phi} w + q_2 U_3); \quad \hat{d}_y = \hat{\ddot{y}} - \left((C_{\hat{\phi}} S_{\hat{\theta}} S_{\hat{\psi}} - S_{\hat{\phi}} C_{\hat{\psi}}) \frac{1}{m} U_1 \right)$$

$$\hat{d}_\psi = \hat{\ddot{\psi}} - (r_5 \hat{\theta} \hat{\phi} + q_3 U_4); \quad \hat{d}_z = \hat{\ddot{z}} - \left(-g + (C_{\hat{\phi}} C_{\hat{\theta}}) \frac{1}{m} U_1 \right)$$

8 Set $k \leftarrow k + 1$ and go to Step 2.
