

Application of PDE control to problems in drilling

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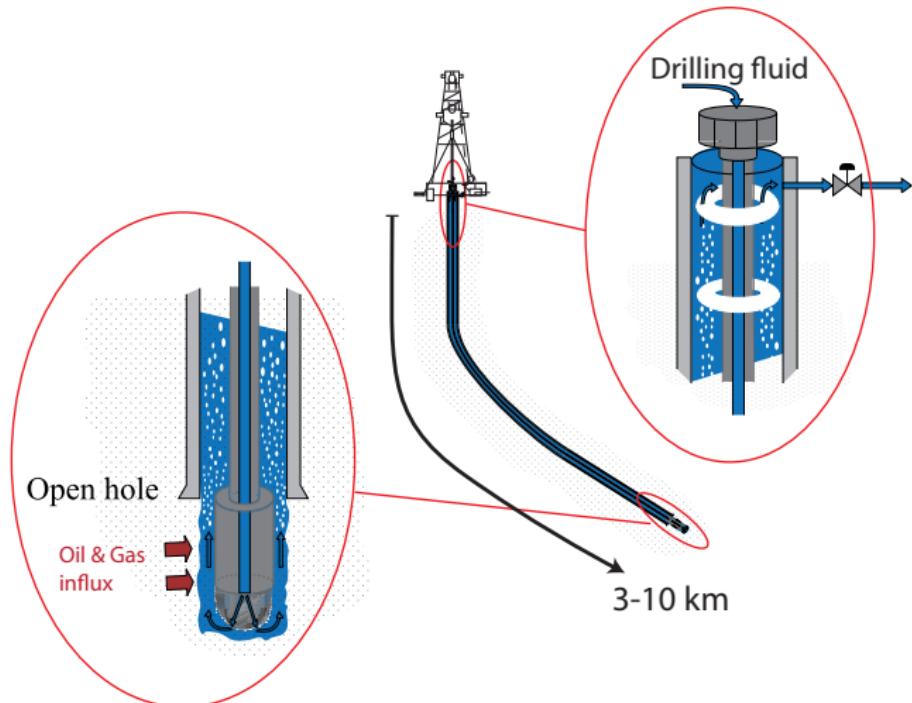
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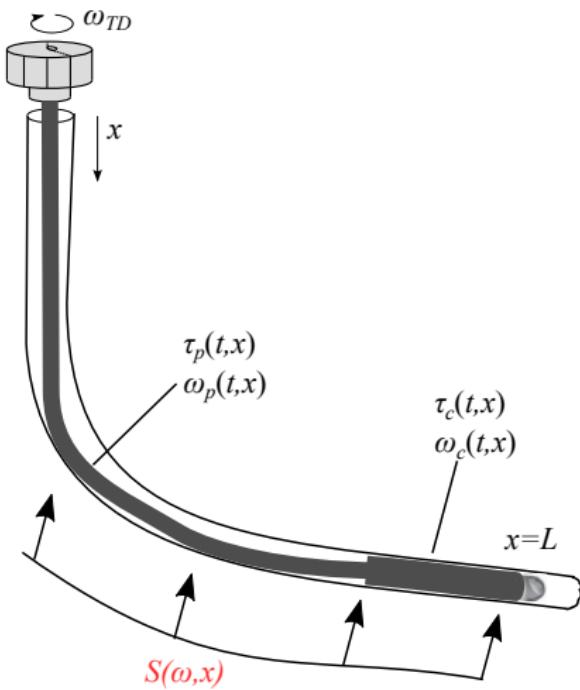
Research goals

- ▶ Develop application oriented results for control and estimation of PDEs
 - ▶ For systems with important distributed effects
 - ▶ Taking into account delays / wave propagation
- ▶ Apply results to problems in drilling

Application domain: drilling



Torsional Model: ODE-PDE-PDE



► Topside BC

$$\dot{\omega}_{TD} = \frac{1}{I_{TD}}(\tau_m - \tau(0, t))$$

► Distributed wave eq.: $i \in \{p, c\}$

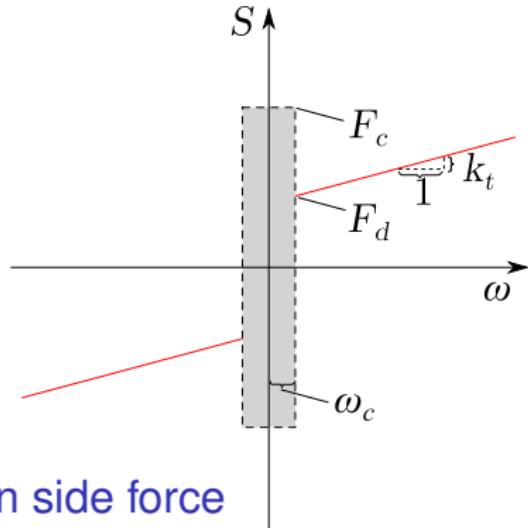
$$\frac{\partial \tau_i(t, x)}{\partial t} + J_i G \frac{\partial \omega_i(t, x)}{\partial x} = 0$$
$$J_i \rho \frac{\partial \omega_i(t, x)}{\partial t} + \frac{\partial \tau_i(t, x)}{\partial x} = -S(\omega_i, x),$$

► Coupling

$$\omega_p(L_p, t) = \omega_c(0, t)$$

$$\tau_p(L_p, t) = \tau_c(0, t)$$

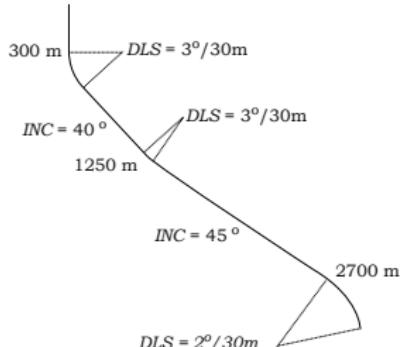
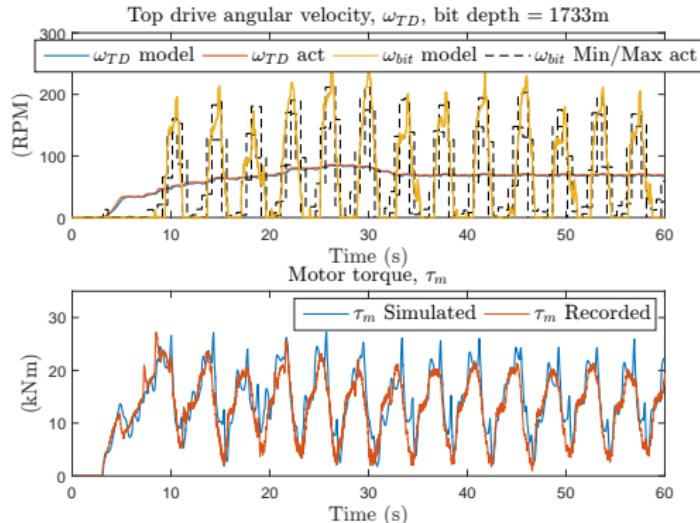
Model: Side force



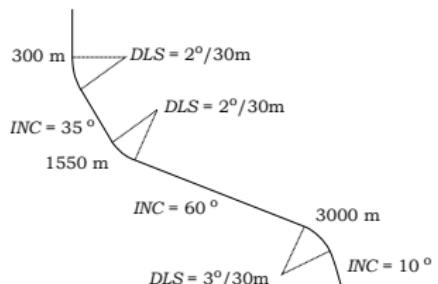
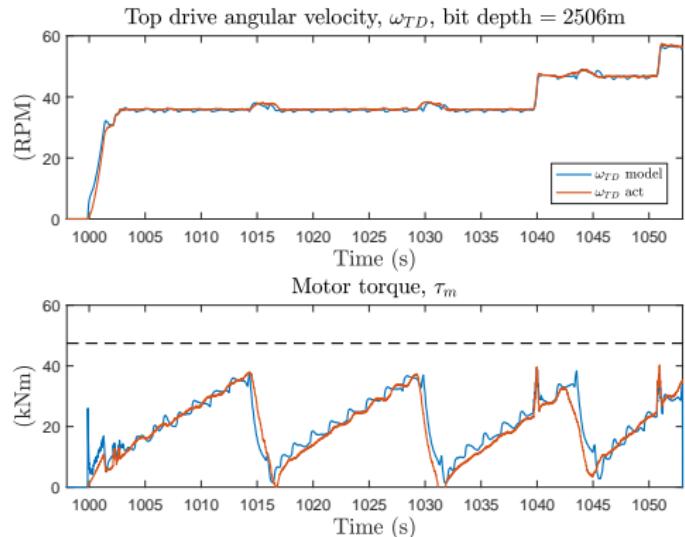
Coulomb friction side force

$$\begin{cases} S(\omega, x) = F_d(x), & \omega \geq \omega_c \\ S(\omega, x) \in [-F_c(x), F_c(x)] & |\omega| < \omega_c \\ S(\omega, x) = F_d(x), & \omega \leq -\omega_c \end{cases}$$

Field data ex 1. 1,733 meter [Aarsnes, UJF and Shor, RJ 2018]



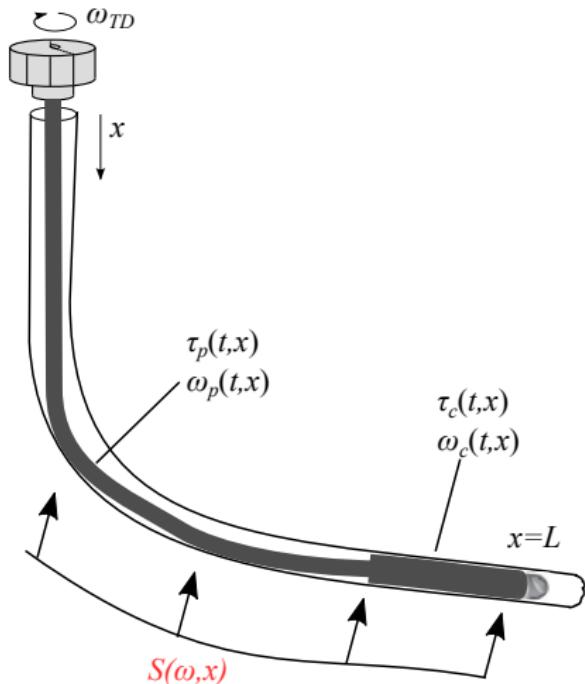
Field data ex 2: 2,2506 meter [Aarsnes, UJF and Shor, RJ 2018]



Control and estimation objectives

1. Control: Avoid “stick-slip” oscillations
2. Estimate: Downhole state and side forces
3. Will use approximate *ODE-PDE-ODE* model

Approximate Model: ODE-PDE-ODE



► Topside BC

$$\dot{\omega}_{TD} = \frac{1}{I_{TD}} (\tau_m - \tau(0, t))$$

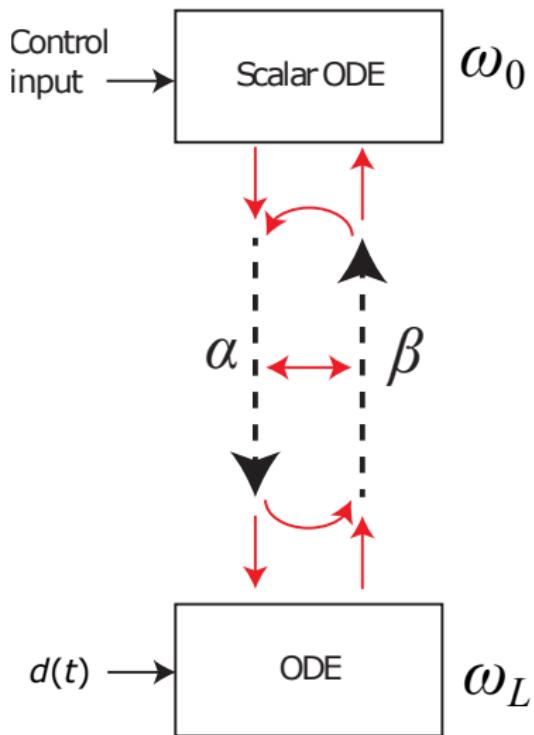
► Drillstring

$$\begin{aligned}\frac{\partial \tau_p(t, x)}{\partial t} + J_p G \frac{\partial \omega_i(t, x)}{\partial x} &= 0 \\ J_p \rho \frac{\partial \omega_p(t, x)}{\partial t} + \frac{\partial \tau_p(t, x)}{\partial x} &= 0,\end{aligned}$$

► Lumped collar

$$\begin{aligned}\dot{\omega}_L &= \frac{1}{I_{BHA}} (\tau_p(L_p, t) - \textcolor{red}{d}(t)), \\ \textcolor{red}{d}(t) &\approx \int_0^L S(\omega, x)\end{aligned}$$

Model structure



Riemann invariants

$$\alpha = \omega + \frac{c_t}{JG}\tau, \quad \beta = \omega - \frac{c_t}{JG}\tau,$$

Structural property: flatness

Parametrize measured state, ω_0 , and control, τ_m , using desired output $z(t) = \omega_L(t)$.

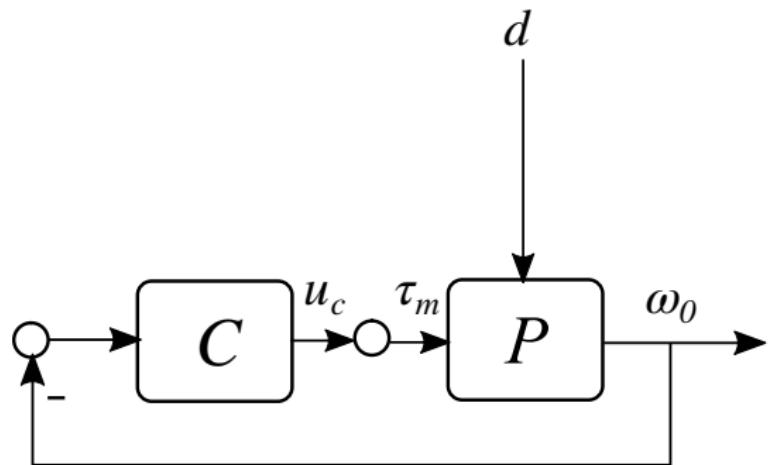
$$\omega_0 = \frac{z(t + t_D) + z(t - t_D)}{2} + \frac{\dot{z}(t + t_D) - \dot{z}(t - t_D)}{2a_L}, \quad (1)$$

$$\frac{2}{I_{TD}}\tau_m = \dot{\omega}_0 + a_0\omega_0 + \frac{a_0}{a_L}\dot{z}(t - t_D) - a_0z(t - t_D). \quad (2)$$

Trajectory planning for $z(t)$ is easy....

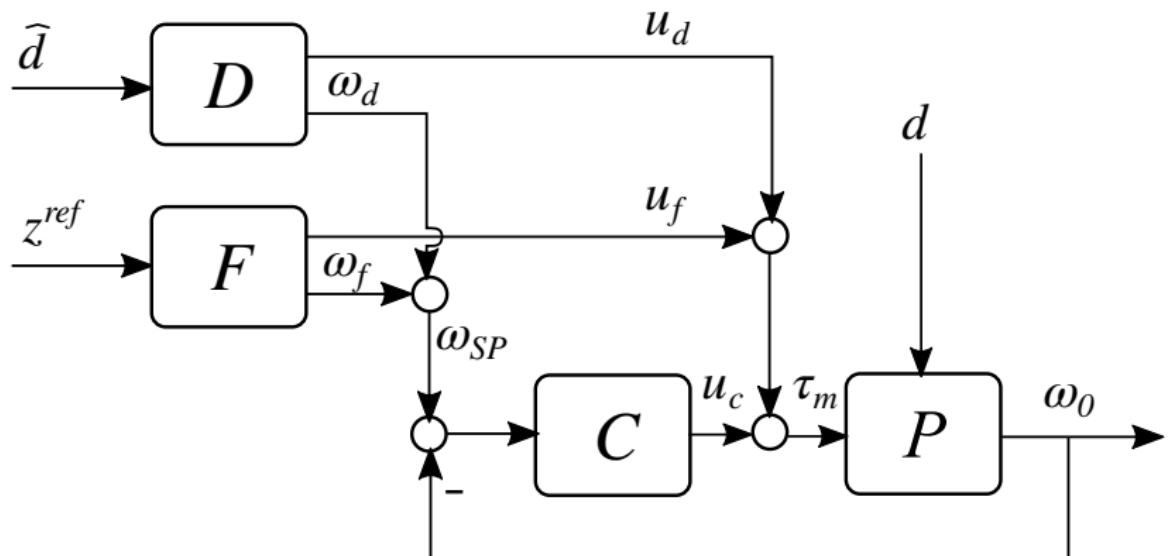
Feedforward control at startup

Disturbance cancellation + trajectory planning

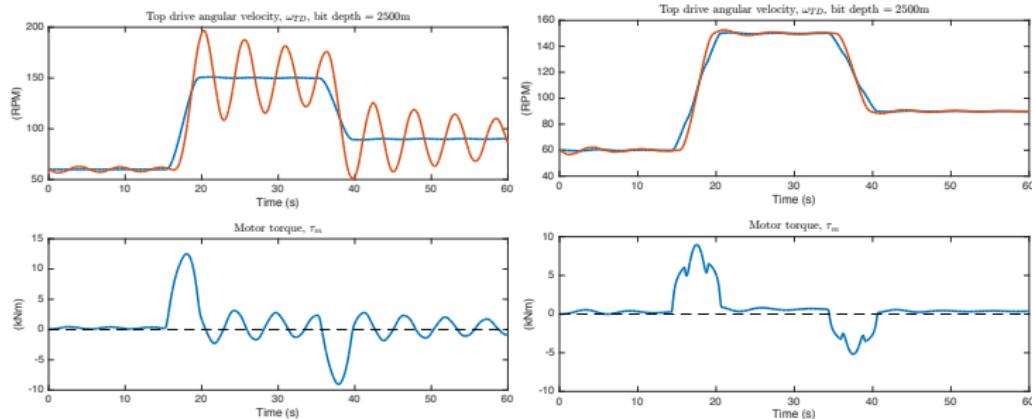


Feedforward control at startup

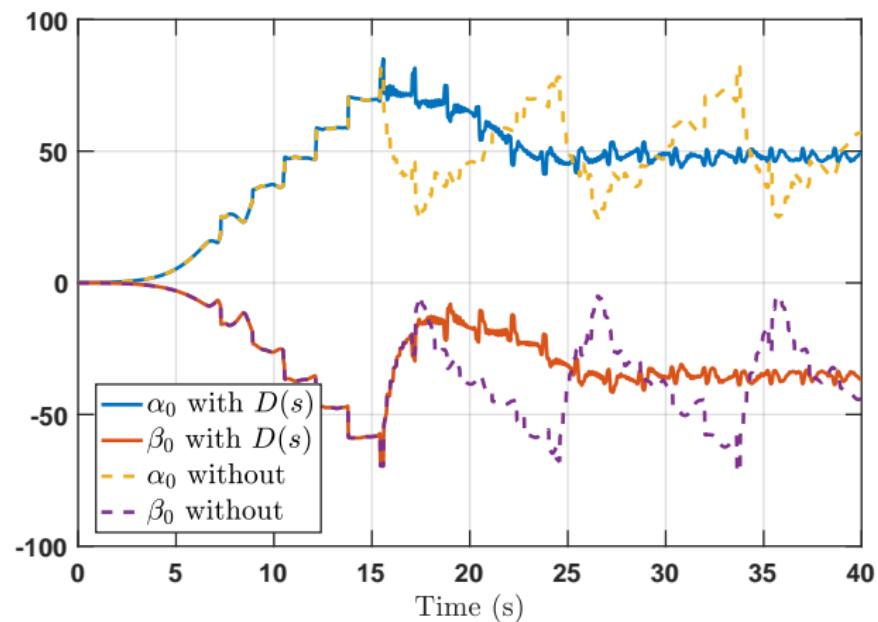
Disturbance cancellation + trajectory planning



Avoid exciting vibration modes



Feedforward control at startup: Canceling $d(t)$ term



Feedforward control at startup

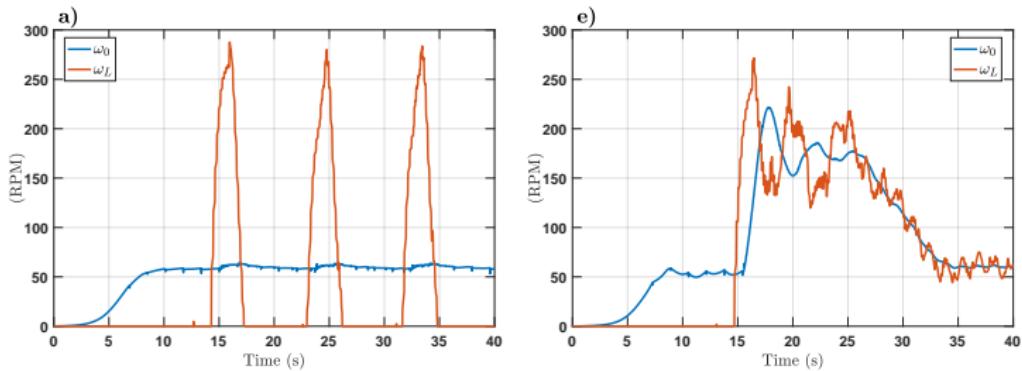


Figure: Nominal case (**left**) with feedforward disturbance cancellation (**right**).

Boundary observer design [Auriol et al. 2018]

- ▶ Backstepping transformation, following [Bou Saba et al., 2017]:

$$u(x, t) \equiv \alpha(x, t) - \int_x^L K^{uu}(x, y)\alpha(y, t)dy - \int_x^L K^{uv}(x, y)\beta(y, t)dy - G_u(x)^\top \omega_L(t)$$
$$v(x, t) \equiv \beta(x, t) - \int_x^L K^{vu}(x, y)\alpha(y, t)dy - \int_x^L K^{vv}(x, y)\beta(y, t)dy - G_v(x)^\top \omega_L(t),$$

with well posed kernels K^{uu} , K^{uv} , K^{vu} , K^{vv} and G_u , G_v to place poles of the distal ODE.

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with well posed kernels K^{uu} , K^{uv} , K^{vu} , K^{vv} and G_u , G_v to place poles of the distal ODE.

- ▶ Slightly modified variable change for ω_0 , c.f. [Bou Saba et al., 2017] no reflection cancellation term:

$$\sigma_0(t) \equiv \omega_0(t) - c_0^{-1} \int_0^L [\tilde{K}_\alpha(y)\alpha(y) + \tilde{K}_\beta(y)\beta(y)] dy - c_0^{-1} \tilde{G}^\top \omega_L(t)$$

with \tilde{K}_α , \tilde{K}_β , \tilde{G} functions of the kernels.

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with \tilde{K}_α , \tilde{K}_β , \tilde{G} functions of the kernels.

- ▶ Chose control law such that

$$\dot{\sigma}_0(t) = \check{p}\sigma_0(t) + d_0 \frac{\check{p}}{c_0} \beta(0, t)$$

$$u(0, t) = c_0 \sigma_0(t) + d_0 v(0, t)$$

with tuning variable $\check{p} > 0$.

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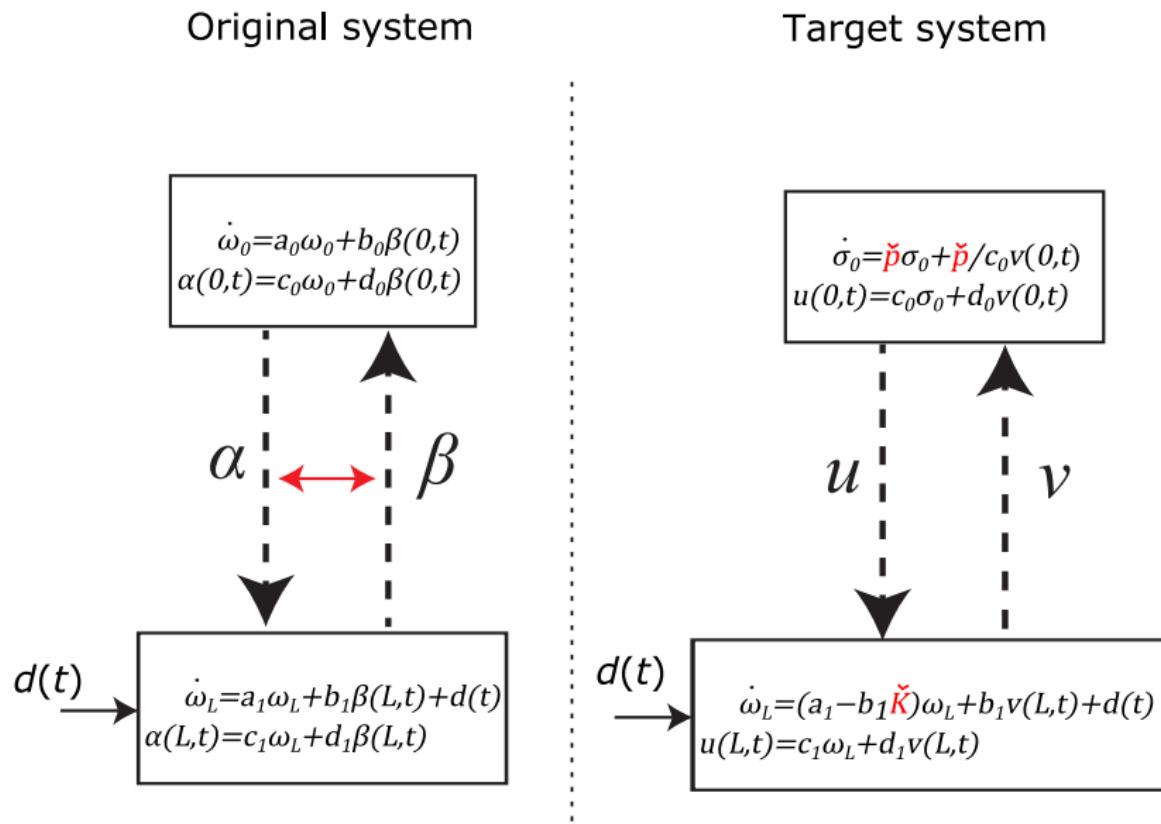
$$\begin{aligned} \dot{\sigma}_0(t) &= \check{p}\sigma_0(t) + d_0 \frac{\check{p}}{c_0} \beta(0, t) \\ u(0, t) &= c_0 \sigma_0(t) + d_0 v(0, t) \end{aligned}$$

with tuning variable $\check{p} > 0$.

- ▶ Yields “low-pass filtered” reflection (in Laplace domain):

$$u(s, 0) = d_0 \frac{s}{s - \check{p}} v(s, 0). \tag{3}$$

Boundary observer design [Auriol et al. 2018]



Results

Simplified case: $S(\omega, x)$ linear, constant disturbance d

$$\dot{\hat{d}}(t) = \gamma(\hat{\omega}_{TD} - y(t)) \quad (4)$$

- ▶ theorem guarantees convergence and robustness to delay mismatch
- ▶ degrees of freedom to calibrate the observer
- ▶ convergence speed vs. noise sensitivity

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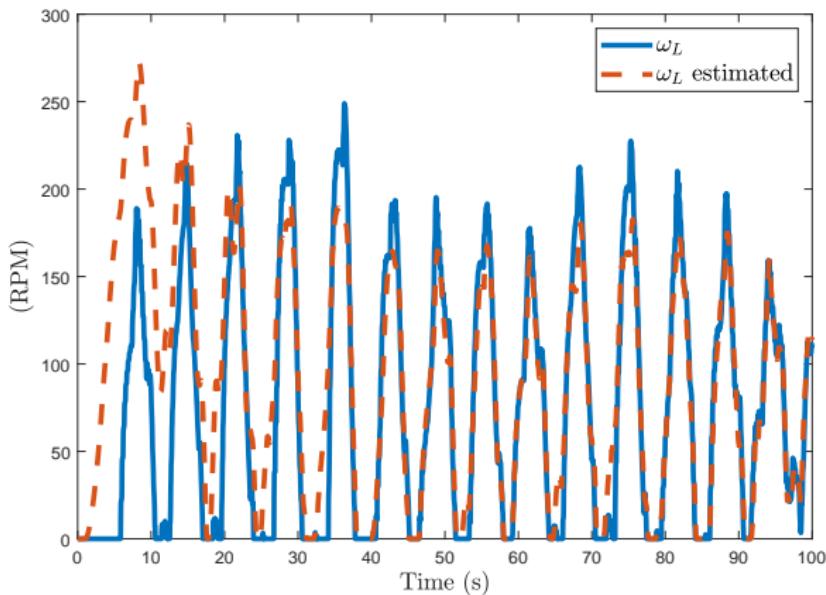
- ▶ theorem guarantees convergence and robustness to delay mismatch
- ▶ degrees of freedom to calibrate the observer
- ▶ convergence speed vs. noise sensitivity

HACK: Friction estimation (*work in progress*)

$$\begin{cases} \hat{d}(t) = \hat{d}_d & \text{if } |\hat{\omega}_L| > \omega_c \\ \hat{d}(t) \in \pm \hat{d}_s & \text{if } |\hat{\omega}_L| < \omega_c \end{cases}$$

Results

Use gains from ‘ODE-PDE-ODE’ backstepping result on ‘ODE-PDE-PDE’ model. Estimate disturbance $d(t)$ to update Coulomb friction model.



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