An INTRODUCTION to SWITCHING ADAPTIVE CONTROL

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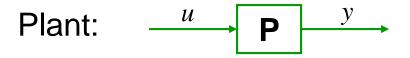


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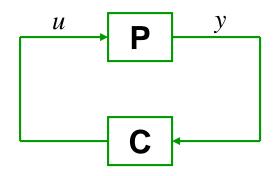
Based on joint work with J.P. Hespanha (UCSB) and A.S. Morse (Yale)

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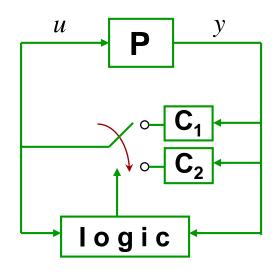
SWITCHING CONTROL



Classical continuous feedback paradigm:



But logical decisions are often necessary:



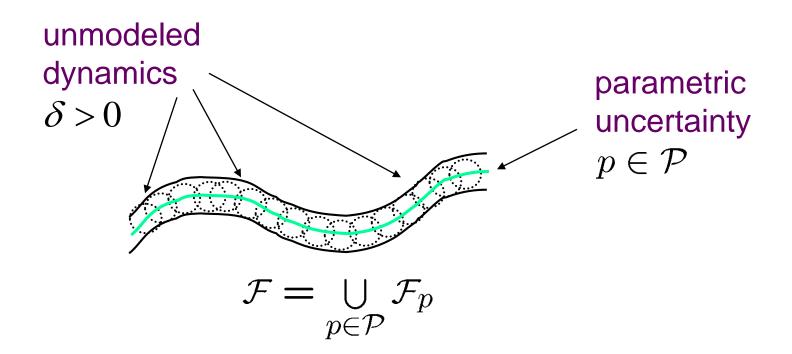
REASONS for SWITCHING

- Nature of the control problem
- Sensor or actuator limitations
- Large modeling uncertainty
- Combinations of the above

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MODELING UNCERTAINTY



Also, noise n and disturbance d

Adaptive control (continuous tuning) vs. supervisory control (switching)

EXAMPLE

Scalar system:

$$\dot{y} = y^2 + p^* u$$

 $\mathcal{P} = [-10, -0.1] \cup [0.1, 10]$

 $p^{*} \in \mathcal{P}$, otherwise unknown

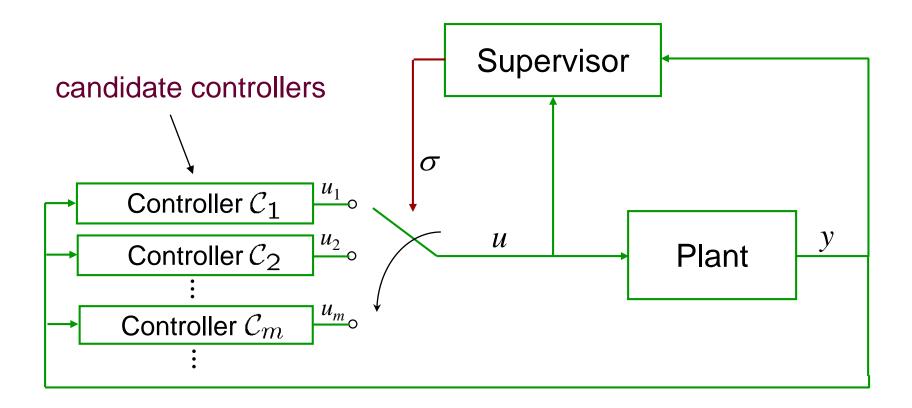
(purely parametric uncertainty)

$$u = -\frac{1}{p^*}(y^2 + y) \Rightarrow \dot{y} = -y$$
 stable \checkmark not implementable

Controller family: $u_q = -\frac{1}{q}(y^2 + y), \ q \in \mathcal{P}$

Could also take $u_q, q \in \mathcal{Q} = \{-1, 1\}$ controller index set

SUPERVISORY CONTROL ARCHITECTURE



 σ – switching signal, takes values in ${\cal Q}$

 C_{σ} – switching controller

TYPES of **SUPERVISION**

Prescheduled (prerouted)

• Performance-based (direct)

• Estimator-based (indirect)

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OUTLINE

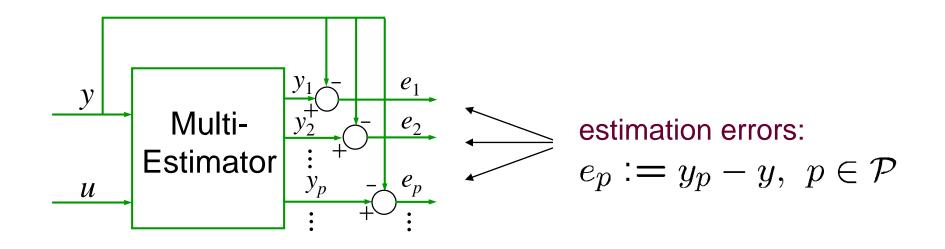
- Basic components of supervisor
- Design objectives and general analysis
- Achieving the design objectives (highlights)

OUTLINE

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SUPERVISOR



Want e_{p^*} to be small Then e_p small indicates $p = p^*$ likely

EXAMPLE

$$\dot{y} = y^2 + p^* u$$

Multi-estimator:

$$\dot{y}_p = -(y_p - y) + y^2 + pu, \ p \in \mathcal{P}$$

EXAMPLE

$$\dot{y} = y^2 + p^*u - d$$

disturbance

Multi-estimator:

$$\dot{y}_p = -(y_p - y) + y^2 + pu, \ p \in \mathcal{P}$$

STATE SHARING

$$\dot{y}_p = -(y_p - y) + y^2 + pu, \ p \in \mathcal{P}$$

Bad! Not implementable if \mathcal{P} is infinite

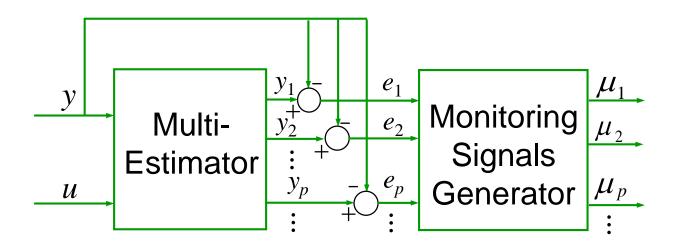
The system

$$\dot{z}_1 = -z_1 + y + y^2$$
$$\dot{z}_2 = -z_2 + u$$
$$y_p = z_1 + pz_2, \ p \in \mathcal{P}$$

produces the same signals

$$\dot{y}_p = \dot{z}_1 + p\dot{z}_2 = -z_1 + y + y^2 - pz_2 + pu = -y_p + y + y^2 + pu$$
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SUPERVISOR



Examples:

$$\mu_p(t) = \int_0^t |e_p(\tau)|^2 d\tau \Leftrightarrow \dot{\mu}_p = |e_p|^2, \ \mu_p(0) = 0$$

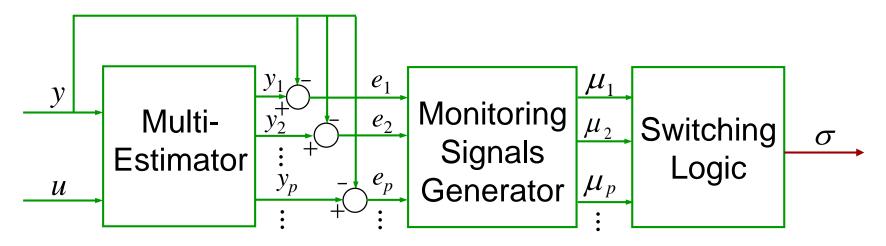
$$\mu_p(t) = \int_0^t e^{-\lambda(t-\tau)} |e_p(\tau)|^2 d\tau \Leftrightarrow \dot{\mu}_p = -\lambda\mu_p + |e_p|^2, \ \mu_p(0) = 0$$

EXAMPLE

Multi-estimator:

$$\begin{split} \dot{z}_1 &= -z_1 + y + y^2 \\ \dot{z}_2 &= -z_2 + u \\ y_p &= z_1 + pz_2, \ p \in \mathcal{P} \\ \\ \hline \mu_p &= e_p^2 \end{bmatrix} - \text{can use state sharing} \\ e_p^2 &= (z_1 + pz_2 - y)^2 = (z_1 - y)^2 + 2pz_2(z_1 - y) + p^2 z_2^2 \\ \dot{\eta}_1 &= (z_1 - y)^2 \\ \dot{\eta}_2 &= 2z_2(z_1 - y) \\ \dot{\eta}_3 &= z_2^2 \\ \mu_p &= \eta_1 + p\eta_2 + p^2 \eta_3, \ p \in \mathcal{P} \end{split}$$

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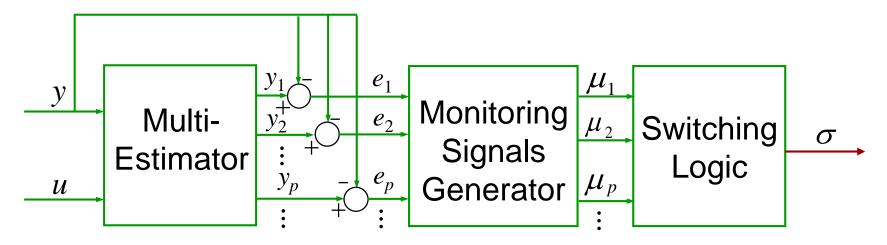
Basic idea:
$$\sigma(t) = \arg\min_{p \in \mathcal{P}} \mu_p(t)$$

Justification? Plant $\in \mathcal{F} = \bigcup_{p \in \mathcal{P}} \mathcal{F}_p$, controllers: $\mathcal{C}_p, p \in \mathcal{P}$

 $\mu_p \text{ small } => e_p \text{ small } => \text{ plant likely in } \mathcal{F}_p => \mathcal{C}_p \text{ gives stable }$ closed-loop system

("certainty equivalence")

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Basic idea:
$$\sigma(t) = \arg\min_{p \in \mathcal{P}} \mu_p(t)$$

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 $\begin{array}{ll} \mu_p \text{ small } \Longrightarrow e_p \text{ small } \swarrow p \text{ lant likely in } \mathcal{F}_p \implies \mathcal{C}_p \text{ gives stable} \\ & \text{only know converse!} & \text{closed-loop system} \end{array}$

Need: e_p small $\Rightarrow C_p$ gives stable closed-loop system

This is detectability w.r.t. e_p

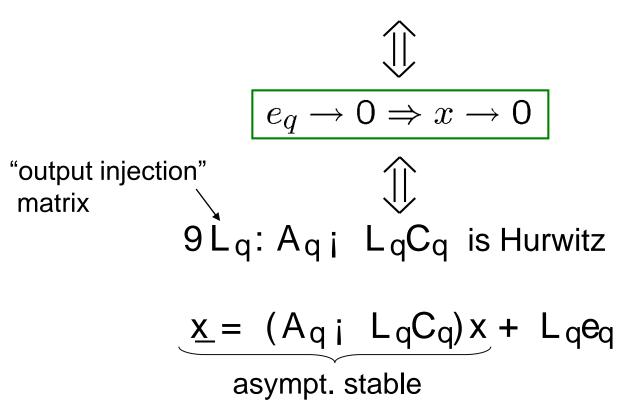
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DETECTABILITY

Linear case:

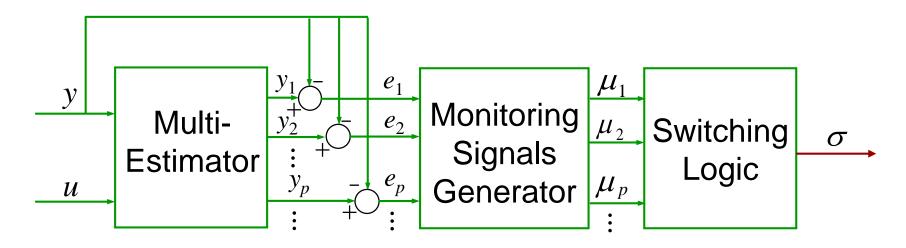
 $\dot{x} = A_q x \qquad \longleftarrow \begin{array}{c} \text{plant in closed} \\ \text{loop with } \mathcal{C}_q \end{array}$ $e_q = C_q x \qquad \longleftarrow \begin{array}{c} \text{view as output} \end{array}$

Want this system to be detectable



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We know: e_{p^*} is small

Switching logic (roughly): $\sigma(t) = \arg\min_{p \in \mathcal{P}} \mu_p(t)$

This (hopefully) guarantees that e_{σ} is small

Need: e_{σ} small => stable closed-loop switched system

This is switched detectability

DETECTABILITY under SWITCHING

Switched system: $\dot{x} = A_{\sigma} x \quad \longleftarrow$ plant in closed loop with C_{σ} $e_{\sigma} = C_{\sigma} x \quad \longleftarrow$ view as output

Want this system to be detectable: $e_{\sigma} \rightarrow 0 \Rightarrow x \rightarrow 0$

Assumed detectable for each frozen value of σ

Output injection:

$$\dot{x} = (A_{\sigma} - L_{\sigma}C_{\sigma})x + L_{\sigma}e_{\sigma}$$

need this to be asympt. stable

Thus σ needs to be "non-destabilizing":

- switching stops in finite time
- slow switching (on the average)

SUMMARY of BASIC PROPERTIES

Multi-estimator:

- 1. At least one estimation error (e_p *) is small
- $e_{p^*} \rightarrow 0 \ \forall u$ when $n = 0, d = 0, \delta = 0$
- e_{p^*} is bounded for bounded n & d

Candidate controllers:

2. For each C_q , closed-loop system is detectable w.r.t. e_q

Switching logic:

- $3 e_{\sigma}$ is bounded in terms of the smallest e_p
- 4 Switched closed-loop system is detectable w.r.t. e_{σ} provided this is true for every frozen value of σ

- conflicting: for 3, want to switch to $\arg \min_p \mu_p(t)$ for 4, want to switch slowly or stop

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Switching logic:

- 3. e_{σ} is bounded in terms of the smallest e_p
- 4. Switched closed-loop system is detectable w.r.t. e_{σ} provided this is true for every frozen value of σ

Analysis: $1 + 3 => e_{\sigma}$ is small 2 + 4 => detectability w.r.t. e_{σ} => state is small \checkmark

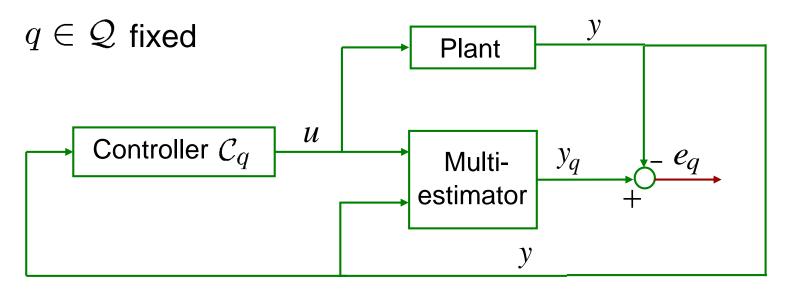
OUTLINE

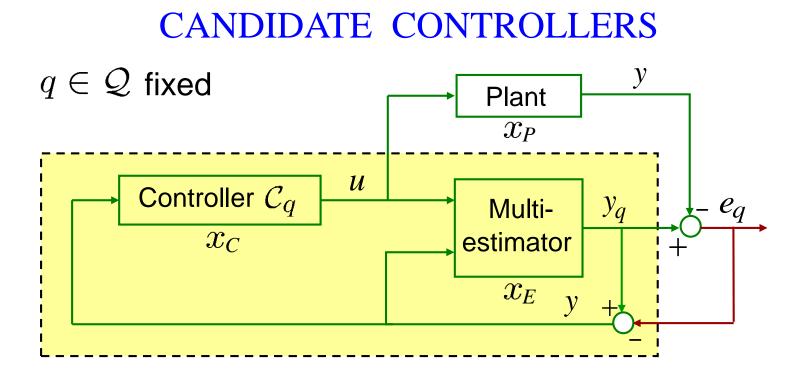
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CANDIDATE CONTROLLERS





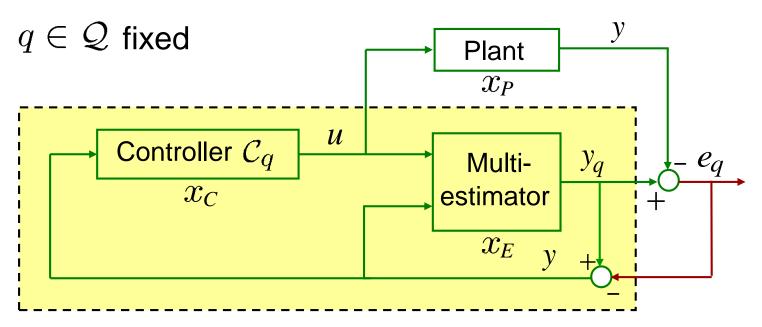
Linear: overall system is detectable w.r.t. e_q if i. system inside the box is stable ii. plant is detectable

ii. plant is detectable

Need to show: $e_q \rightarrow 0 \implies x_P$, x_C , $x_E \rightarrow 0$

 $e_q \rightarrow 0 \Longrightarrow x_C, x_E \rightarrow 0 \Longrightarrow u, y_q \rightarrow 0 \Longrightarrow y = y_q - e_q \rightarrow 0 \Longrightarrow x_P \xrightarrow{} 0$

CANDIDATE CONTROLLERS



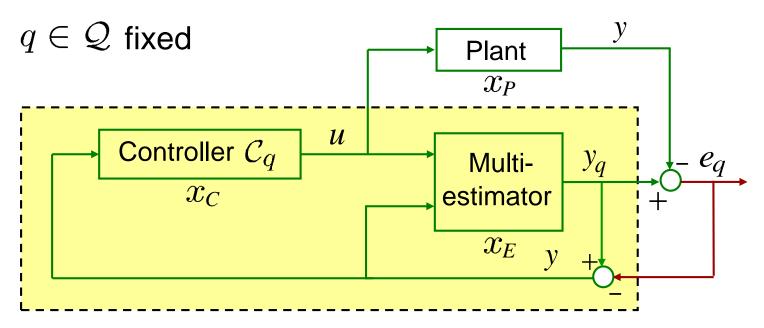
Linear: overall system is detectable w.r.t. e_q if

- i. system inside the box is stable
- ii. plant is detectable

Nonlinear: same result holds if stability and detectability are interpreted in the ISS/OSS sense:

$$|x(t)| \leq \beta(|x(0)|, t) + \gamma\left(\|\dot{v}\|_{[0,t]}\right)$$

CANDIDATE CONTROLLERS



Linear: overall system is detectable w.r.t. e_q if

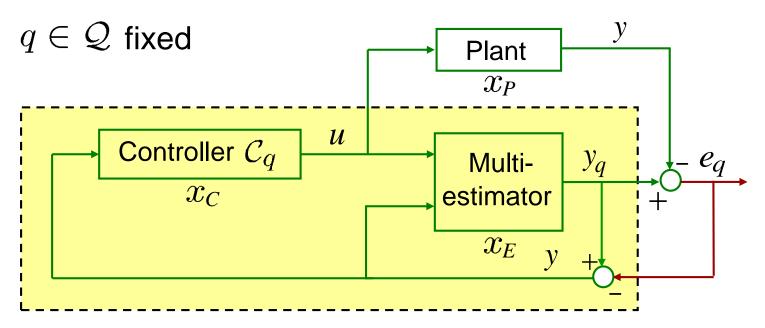
- i. system inside the box is stable
- ii. plant is detectable

Nonlinear: same result holds if stability and detectability are interpreted in the integral-ISS/OSS sense:

$$|x(t)| \leq \beta(|x(0)|,t) + \int_0^t \gamma(|v(\tau)|) d\tau$$

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CANDIDATE CONTROLLERS



Linear: overall system is detectable w.r.t. e_q if i. system inside the box is stable

ii. plant is detectable

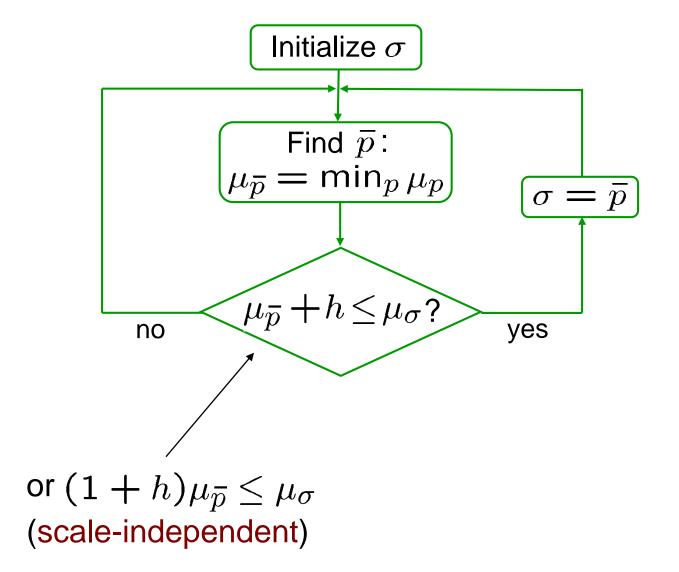
For minimum-phase plants, it is enough to ask that the system inside the box be output-stabilized

SWITCHING LOGIC: DWELL-TIME $\mu_p, \ p \in \mathcal{P}$ – monitoring signals $\tau_d > 0$ – dwell time Initialize σ Wait τ_d time units Find \overline{p} : $\mu_{\overline{p}} = \min_p \mu_p$ $\sigma = \bar{p}$ $\mu_{\overline{p}} < \mu_{\sigma}$? no yes Detectability is preserved if τ_d is large enough \checkmark Obtaining a bound on e_{σ} in terms of e_{p^*} is harder

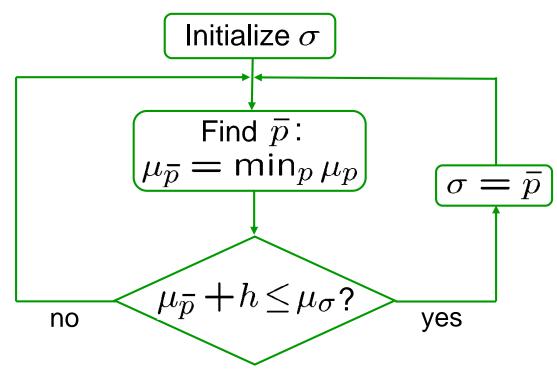
Not suitable for nonlinear systems (finite escape)

SWITCHING LOGIC: HYSTERESIS

 $\mu_p, p \in \mathcal{P}$ – monitoring signals h > 0 – hysteresis constant

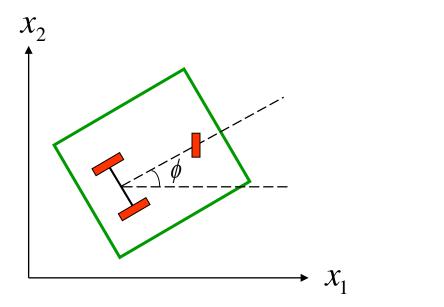


SWITCHING LOGIC: HYSTERESIS



 $\mathcal{P} \text{ finite, } \mu_p\uparrow, \mu_{p^*} \text{ bounded } => \text{ switching stops in finite time}$ This applies to $\delta, n, d = 0, e_{p^*} \rightarrow 0 \text{ exp fast, } \mu_p = \int |e_p|^2$ Linear, $\delta = 0, n, d$ bounded $=> \text{ average dwell time } \tau_a(h)$ $\int |e_\sigma|^2 \leq |\mathcal{P}|(1+h) \int |e_p|^2$ _{33 of 34}

TOY EXAMPLE: PARKING PROBLEM



 $\dot{x}_1 = p_1 w_1 \cos \phi$ $\dot{x}_2 = p_1 w_1 \sin \phi$ $\dot{\phi} = p_2 w_2$

Unknown parameters p_1, p_2 correspond to the radius of rear wheels and distance between them