



Robotics 2

Robust Trajectory Control

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Problem formulation

- given the **real** robot, modeled by

$$B(q)\ddot{q} + n(q, \dot{q}) = u$$

- assuming an **estimated** feedback linearization control

$$u = \hat{B}(q)a + \hat{n}(q, \dot{q})$$

- we would like to design a so as to obtain

- asymptotic stability of the closed-loop system
- the best possible trajectory tracking performance

- the linear feedback choice is not enough...

$$a = \ddot{q}_d + K_D(\dot{q}_d - \dot{q}) + K_P(q_d - q)$$

- questions:

- which should be the conditions on the estimates?
- can we **guarantee** stability/performance, based on **known bounds on the uncertainties**?



Closed-loop equations - 1

- under uncertain conditions (estimated \neq real dynamic coefficients), feedback linearization is only approximate and the closed-loop equations are **still nonlinear**

$$\begin{aligned}\ddot{q} &= B^{-1}(q)(\hat{B}(q)a + \hat{n}(q, \dot{q}) - n(q, \dot{q})) \\ &= a + (B^{-1}(q)\hat{B}(q) - I)a \\ &\quad + B^{-1}(q)(\hat{n}(q, \dot{q}) - n(q, \dot{q})) \\ &= a + E(q)a + B^{-1}(q)\Delta n(q, \dot{q}) \\ &= a + \eta(a, q, \dot{q})\end{aligned}$$

where η depends on the **amount of uncertainty**

$$\begin{aligned}E(q) &= B^{-1}(q)\Delta B(q) = B^{-1}(q)(\hat{B}(q) - B(q)) \\ \Delta n(q, \dot{q}) &= \hat{n}(q, \dot{q}) - n(q, \dot{q})\end{aligned}$$



Closed-loop equations - 2

- closed-loop **state equations** are written as

$$\dot{x} = Ax + B(a + \eta)$$

$$A = \begin{bmatrix} 0 & I \\ 0 & 0 \end{bmatrix} \quad B = \begin{bmatrix} 0 \\ I \end{bmatrix} \quad x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} q \\ \dot{q} \end{bmatrix}$$

- closed-loop **error equations** with respect to a desired $q_d(t)$ are rewritten as

$$e_1 = x_1 - x_{1d} = q - q_d \quad e_2 = x_2 - x_{2d} = \dot{q} - \dot{q}_d$$

$$\dot{e}_1 = e_2$$

$$\dot{e}_2 = \ddot{q} - \ddot{q}_d = a + \eta(a, e_1, e_2, q_d, \dot{q}_d) - \ddot{q}_d$$

$$\Rightarrow \boxed{\dot{e} = Ae + B(a + \eta - \ddot{q}_d)}$$

note that errors are defined here with **opposite signs** w.r.t. usual



Solution approach

- add an external **robust control** term/loop
 - based on computable bounds on the uncertainties
- based on the theory of **guaranteed stability for nonlinear uncertain system**
- Lyapunov-based analysis
- a **discontinuous** control law will result
 - difficult to implement because of **chattering** effects
 - **smoothed** version with only uniformly ultimate boundedness (**u.u.b. stability**) of the tracking error



Working assumptions

1. bound on the desired **trajectory**

$$\sup_{t \geq 0} \|\ddot{q}_d\| < Q_{\max} < \infty$$

2. bound on the estimate of the robot **inertia** matrix

$$\|E(q)\| = \|B^{-1}(q)\hat{B}(q) - I\| \leq \alpha < 1$$

with $\alpha \geq 0$, holding for all configurations q

3. bound on the estimate of **nonlinear dynamic** terms

$$\|\Delta n(q, \dot{q})\| \leq \phi(e, t)$$

with a known function ϕ , bounded for all t

- as a **general rule**, exploiting the model structure (e.g., its linear parameterization) may lead to more stringent bounds



Bound on the inertia matrix

- assumption 2. can always be satisfied, knowing some upper and lower bounds (that always exist due to the positive definiteness) on the **inverse** of the inertia matrix

$$0 < m \leq \|B^{-1}(q)\| \leq M < \infty$$

- it is then **sufficient** to choose as estimate

$$\hat{B} = \frac{1}{c} I \quad \text{with} \quad c = \frac{M + m}{2}$$

- in fact, using the SVD factorization of the inverse inertia matrix, it can be shown that (see Appendix A)

$$\|B^{-1}\hat{B} - I\| \leq \frac{M - m}{M + m} = \alpha < 1$$



Control design – step 1

- linear control law with an **added robust** term

$$a = \ddot{q}_d - K_P e_1 - K_D e_2 + \Delta a$$

where the PD gains are **diagonal** and **positive** matrices

- we obtain

$$\dot{e} = \bar{A}e + B(\Delta a + \bar{\eta})$$

being

$$\bar{A} = A - BK \quad K = [K_P \quad K_D]$$

where \bar{A} has all eigenvalues with negative real part,
and

$$\bar{\eta} = E(\ddot{q}_d - Ke + \Delta a) + B^{-1}\Delta n$$



Control design – step 2

- (same) **bound** on **nonlinear** terms and **added robust** term

$$\|\bar{\eta}\| < \rho(e, t) \quad \|\Delta a\| < \rho(e, t)$$

- we can use the previous data and implicitly define the bound $\rho(e, t)$ from

$$\begin{aligned} \|\bar{\eta}\| &= \|E\Delta a + E(\ddot{q}_d - \mathcal{K}e) + B^{-1}\Delta n\| \\ &\leq \alpha\rho(e, t) + \alpha(Q_{\max} + \|\mathcal{K}\| \cdot \|e\|) + M\phi(e, t) \\ &=: \rho(e, t) \end{aligned}$$

yielding the well-defined (since $0 < \alpha < 1$), limited and possibly time-varying function

$$\boxed{\rho(e, t)} = \frac{1}{1 - \alpha} [\alpha(Q_{\max} + \|\mathcal{K}\| \cdot \|e\|) + M\phi(e, t)]$$



Control design – step 3

- solve an associated (linear) **Lyapunov equation**, for any given symmetric $Q > 0$ matrix

$$\bar{A}^T P + P \bar{A} + Q = 0$$

finding the unique (symmetric) solution matrix $P > 0$

- finally, define the **discontinuous** robust term as

$$\Delta a = \begin{cases} -\rho(e, t) \frac{\mathcal{B}^T P e}{\|\mathcal{B}^T P e\|} & \text{if } \|\mathcal{B}^T P e\| \neq 0 \\ 0 & \text{if } \|\mathcal{B}^T P e\| = 0 \end{cases}$$

that also satisfies, by its own structure, $\|\Delta a\| < \rho(e, t)$



Solving the Lyapunov equation

- in general, using `lyap` in Matlab (only once, in advance)
- closed-form solution in an interesting scalar case (one robot joint/link), to get a “feeling”...

$$A = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \quad B = \begin{bmatrix} 0 \\ 1 \end{bmatrix} \quad K = [k_P \quad k_D] \quad \bar{A} = A - BK = \begin{bmatrix} 0 & 1 \\ -k_P & -k_D \end{bmatrix} \quad k_P > 0, k_D > 0$$

choose, e.g., $Q = q \cdot I_{2 \times 2} > 0$ \Rightarrow find $P = \begin{bmatrix} p_{11} & p_{12} \\ p_{12} & p_{22} \end{bmatrix} \underbrace{> 0}_{\text{test!}}$

$$\begin{aligned} \bar{A}^T P + P \bar{A} + q \cdot I &= \\ &= \begin{bmatrix} -2p_{12}k_P + q & p_{11} - p_{12}k_D - p_{22}k_P \\ \text{symm} & 2(p_{12} - k_D p_{22}) + q \end{bmatrix} = 0 \end{aligned}$$

$$p_{12} = \frac{q}{2k_P} \quad p_{22} = \frac{q}{2} \left(\frac{1 + k_P}{k_P k_D} \right) > 0$$

$$p_{11} = \frac{q}{2} \left(\frac{k_D}{k_P} + \frac{1 + k_P}{k_D} \right) > 0$$

$$\Rightarrow p_{11}p_{22} - p_{12}^2 > 0$$

so that (also in the n-dof case) $\boxed{B^T P e} = \text{block} \left\{ \frac{q}{2} \left(\frac{e_1}{k_P} + \frac{1 + k_P}{k_P} \frac{e_2}{k_D} \right) \right\}$



Stability analysis

Theorem 1

Defining $V(e) = e^T P e$, the presented robust control law with the discontinuous term is such that $\dot{V}(e) < 0$ along the trajectories of the closed-loop error system

Proof

$$\begin{aligned}\dot{V}(e) &= \dot{e}^T P e + e^T P \dot{e} \\ &= e^T (\bar{A}^T P + P \bar{A}) e + 2e^T P B (\Delta a + \bar{\eta}) \\ &= -e^T Q e + 2e^T P B (\Delta a + \bar{\eta}) \\ &= -e^T Q e + 2w^T (\Delta a + \bar{\eta})\end{aligned}$$

if $w = 0 \Rightarrow \dot{V} = -e^T Q e < 0$

if $w \neq 0 \Rightarrow \Delta a = -\rho w / \|w\| \Rightarrow$

$$\begin{aligned}w^T \left(-\rho \frac{w}{\|w\|} + \bar{\eta}\right) &= -\rho \frac{w^T w}{\|w\|} + w^T \bar{\eta} \\ &\leq -\rho \|w\| + \|w\| \cdot \|\bar{\eta}\| \\ &= \|w\| (-\rho + \|\bar{\eta}\|) \leq 0\end{aligned}$$

note: because of the discontinuity we cannot directly conclude on the (global) asymptotic stability of $e=0$

$\Rightarrow \dot{V} < 0$





A smoother robust controller

- for any given (small) $\epsilon > 0$, define the **continuous** robust term as

$$\Delta a = \begin{cases} -\rho(e, t) \frac{\mathcal{B}^T P e}{\|\mathcal{B}^T P e\|} & \text{if } \|\mathcal{B}^T P e\| \geq \epsilon \\ -\frac{\rho(e, t)}{\epsilon} \mathcal{B}^T P e & \text{if } \|\mathcal{B}^T P e\| < \epsilon \end{cases}$$

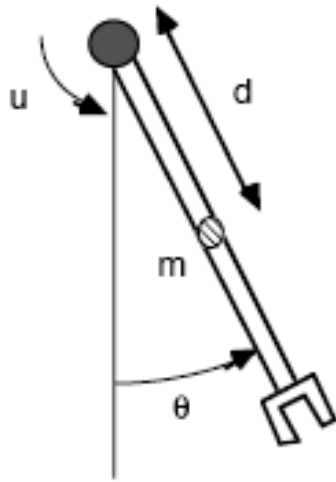
Theorem 2

With the above continuous robust control law, any solution $e(t)$, with $e(0) = e_0$, of the closed-loop error system is **uniformly ultimately bounded** with respect to a suitable set S (a neighborhood of the origin)

Proof in Appendix B



Case study: Single-link under gravity



model $I\ddot{\theta} + mgd \sin \theta = u$ (no friction)

error equations

$$\dot{e}_1 = e_2$$

$$\dot{e}_2 = \frac{1}{I} u - \frac{mgd}{I} \sin \theta - \ddot{\theta}_d$$

$$= \frac{1}{I} [\hat{I}(a + \Delta a) + \widehat{mgd} \sin \theta] - \frac{mgd}{I} \sin \theta - \ddot{\theta}_d$$

$$e_1 = \theta - \theta_d$$

$$e_2 = \dot{\theta} - \dot{\theta}_d$$

$$= a + \Delta a + \left(\frac{\hat{I}}{I} - 1 \right) (a + \Delta a) + \frac{\Delta mgd}{I} \sin \theta - \ddot{\theta}_d$$

known bounds for control design

$$5 \leq I \leq 10 \quad 5 \leq mgd \leq 7$$



Calculations for robust control

```
% real robot
I=5; mgd=7;
% initial robot state
th0=0; thp0=0;
% range of uncertainties
I_min=5; I_max=10;
mgd_min=5; mgd_max=7;
% linear tracking stabilizer gains
kp=25; kd=10; % two poles in -5

% robust control part
% Lyapunov matrix P and b^T P term
A=[0 1; -kp -kd];
q=1; Q=q*eye(2);
P=lyap(A',Q); % solve A'*P+P*A+Q=0
b=[0 1];
bP=b*P; % =[0.02 0.052]
```

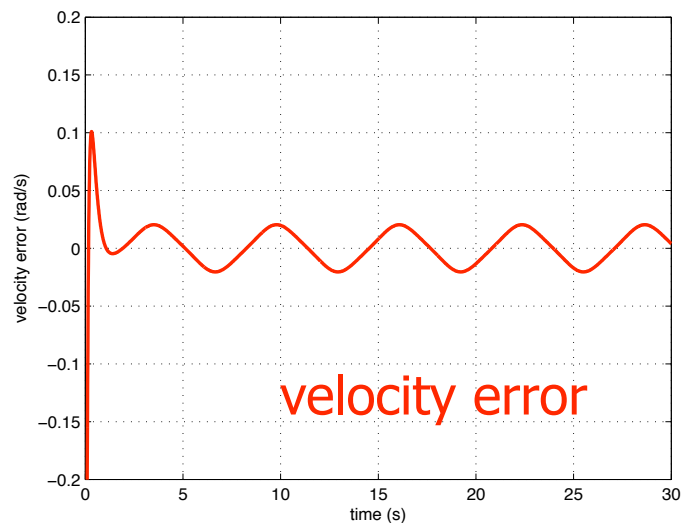
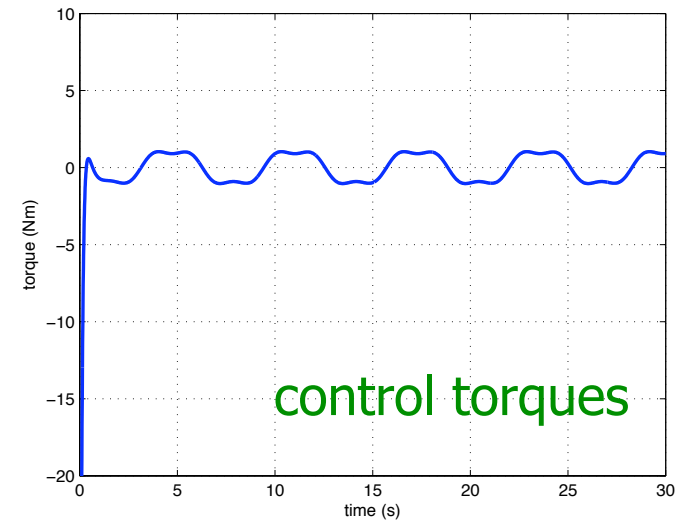
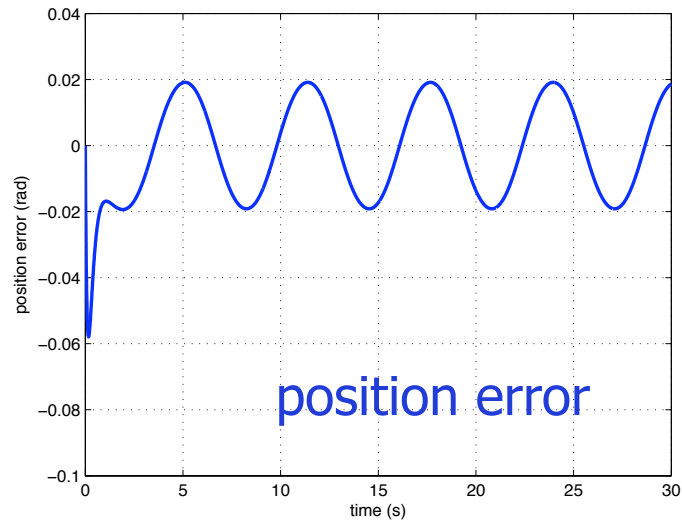
```
% bounding dynamic terms
% inertia
m=1/I_max; M=1/I_min;
c=(M+m)/2;
alpha=(M-m)/(M+m);
Ihat=1/c; % =6.6667
% nonlinear terms (only gravity)
Mphi=M*(mgd_max-mgd_min);
mgdhat=5;
% overall bounding
rho0=Mphi/(1-alpha) % =0.6
rho1=alpha/(1-alpha) % =0.5
% smoothed version
epsilon=5*10^-4;

red values are used in Simulink
```



Results

first trajectory – feedback linearization, no robust loop



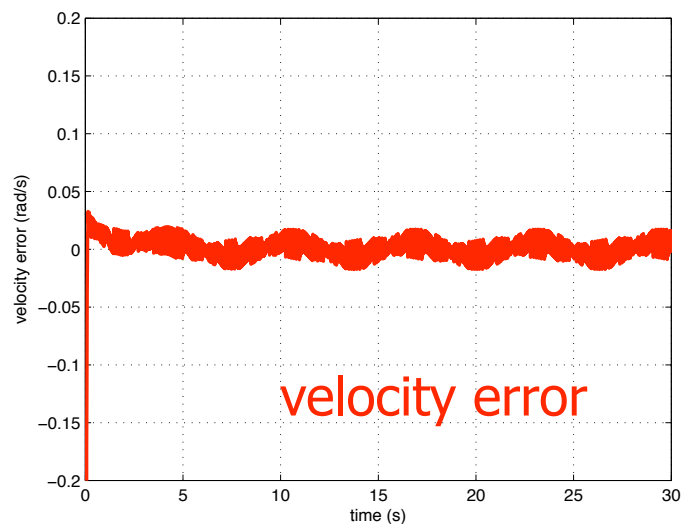
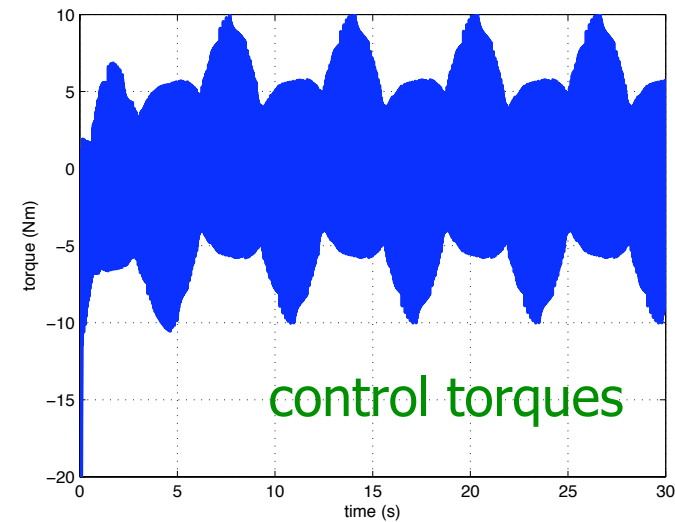
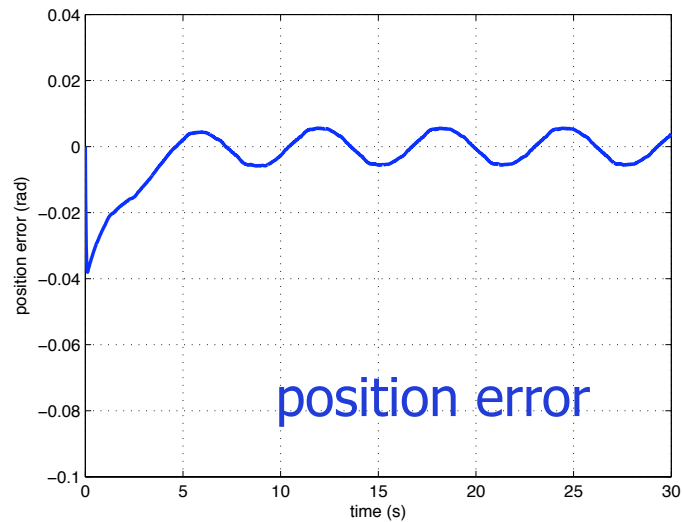
$$\theta_d(t) = -\sin t$$
$$\theta(0) = 0, \dot{\theta}(0) = 0$$

non-zero initial error
on velocity



Results

first trajectory – discontinuous robust control



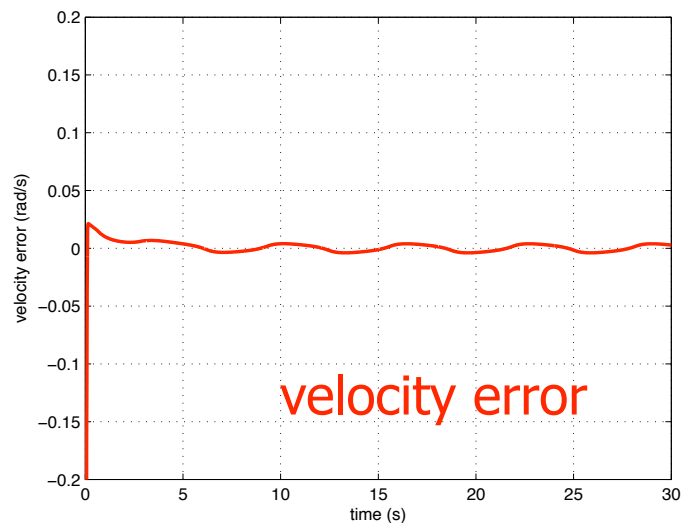
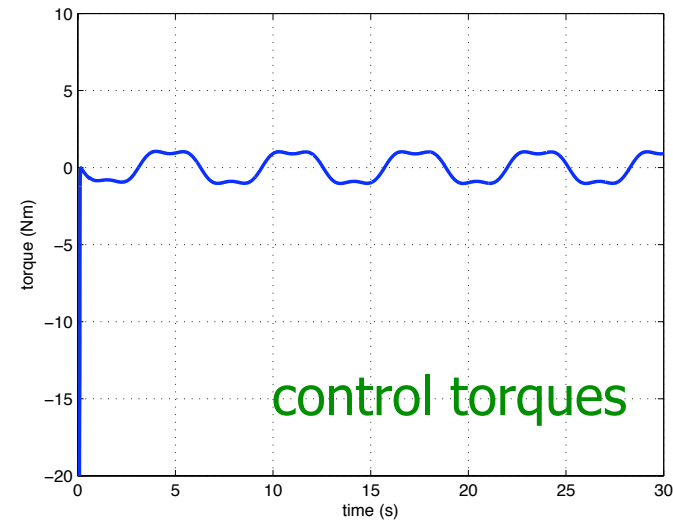
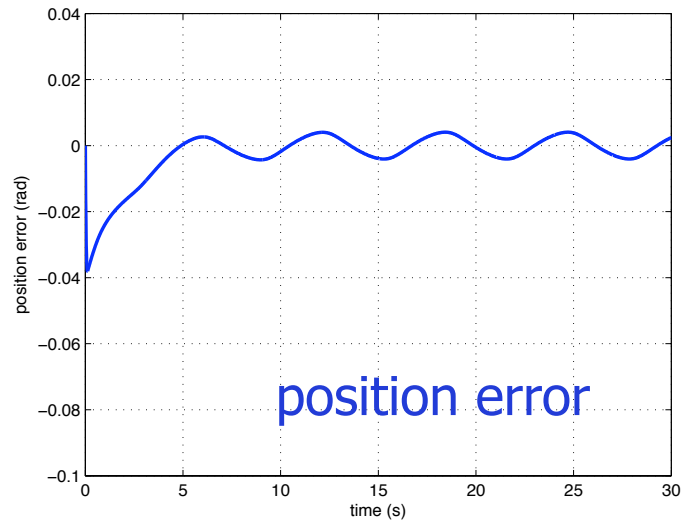
$$\theta_d(t) = -\sin t$$

position and velocity errors
are largely **reduced**,
but control **chattering**
at high frequency
(when error is close to zero)



Results

first trajectory – smoothed robust control



$$\theta_d(t) = -\sin t$$

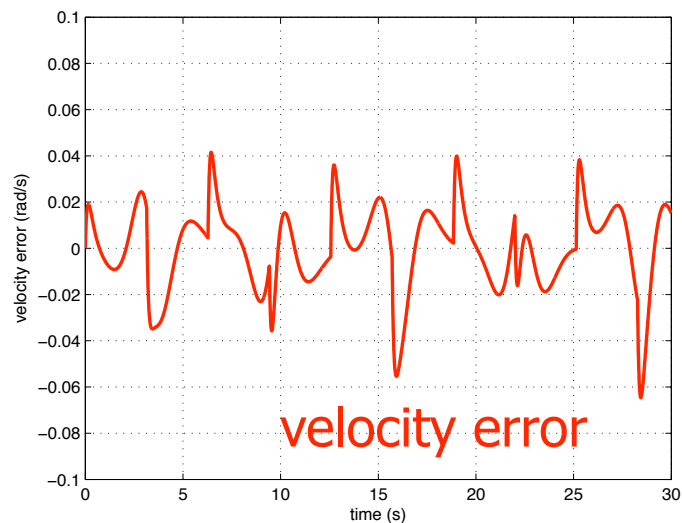
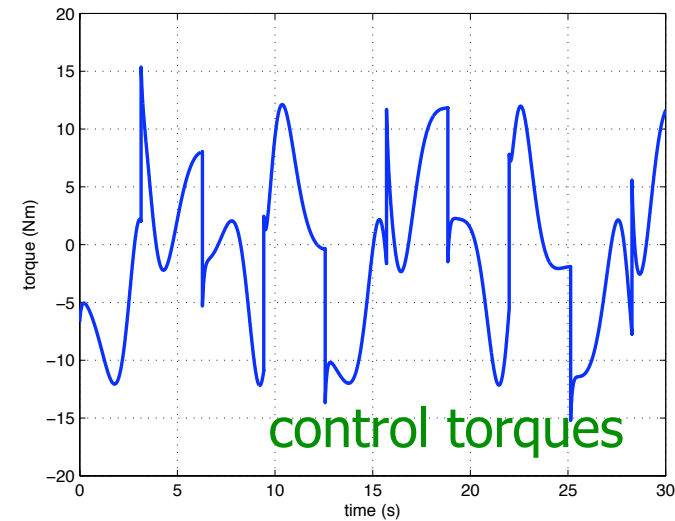
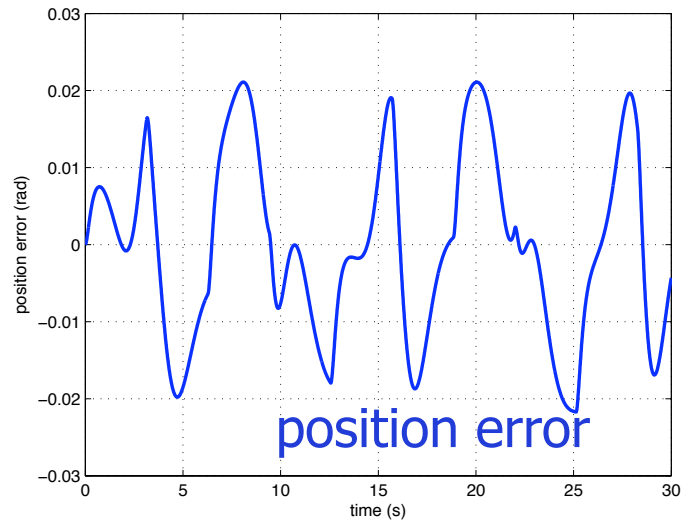
position and velocity errors
are similarly **reduced**,
without control chattering

(using here $\epsilon > 0$)



Results

second trajectory – fbk linearization, no robust loop



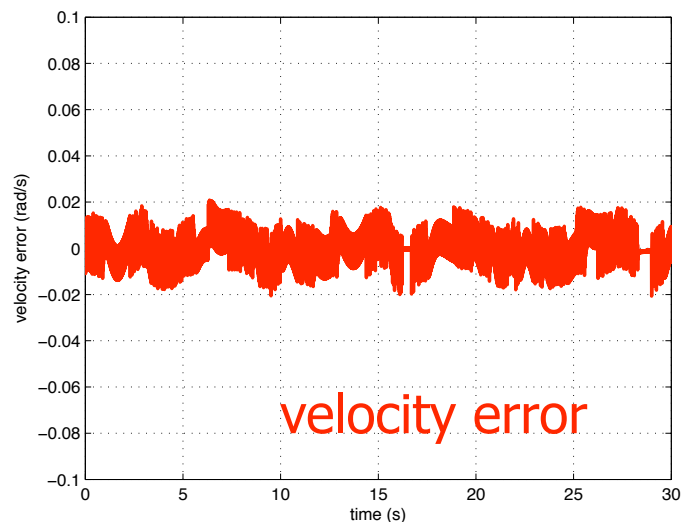
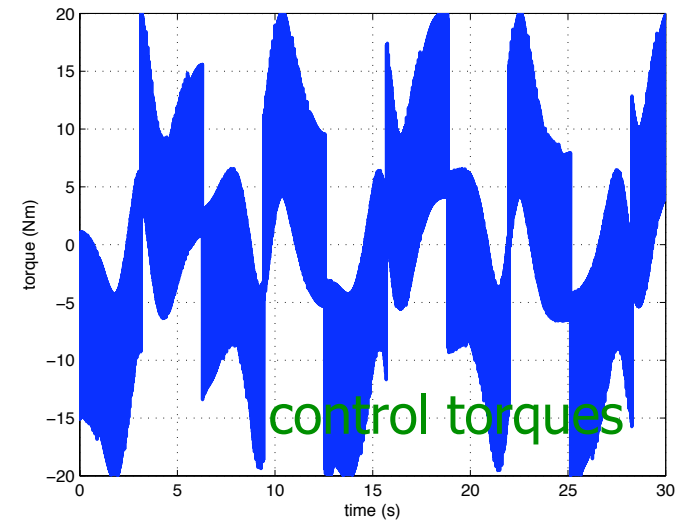
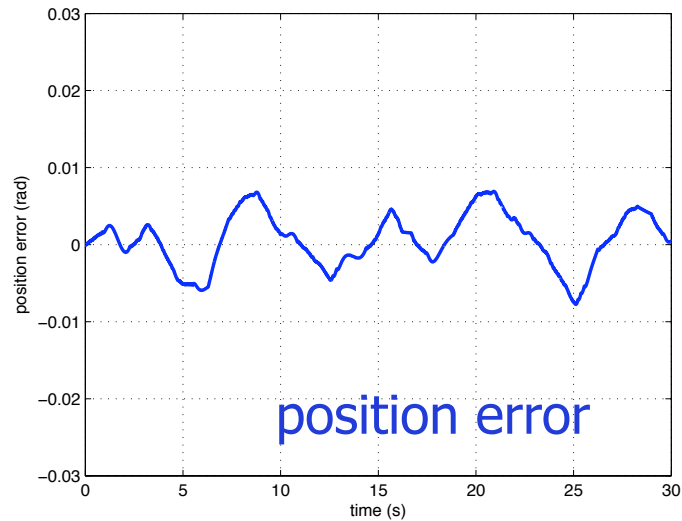
bang-bang acceleration profile
at 1 rad/s frequency and
with $Q_{max} = 1 \text{ rad/s}^2$

zero initial tracking error
(matching state conditions)



Results

second trajectory – discontinuous robust control



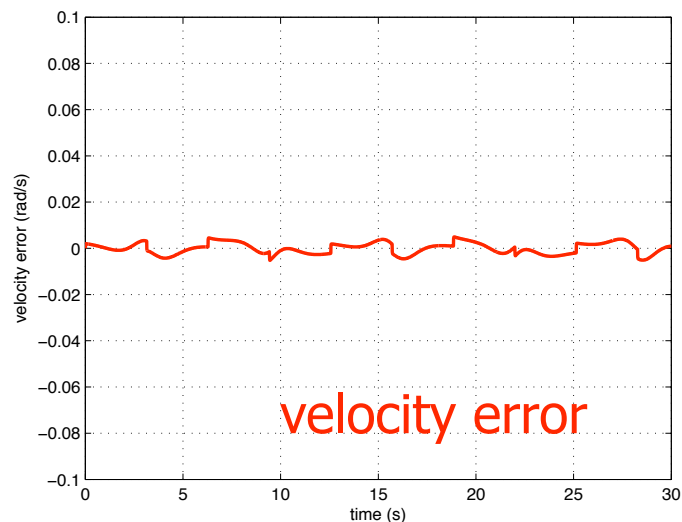
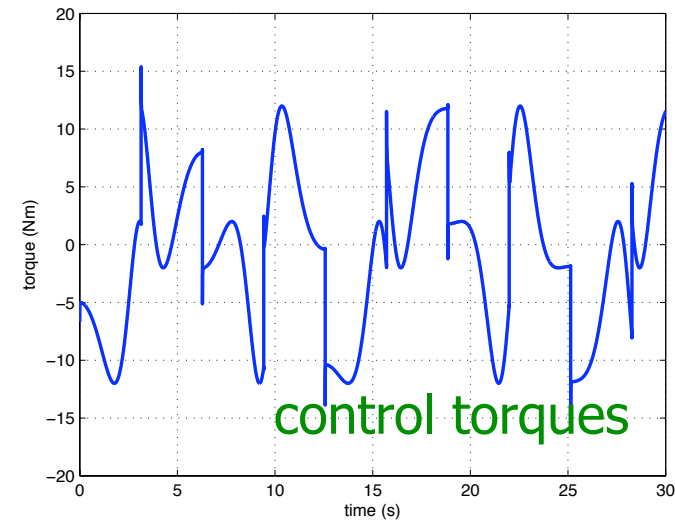
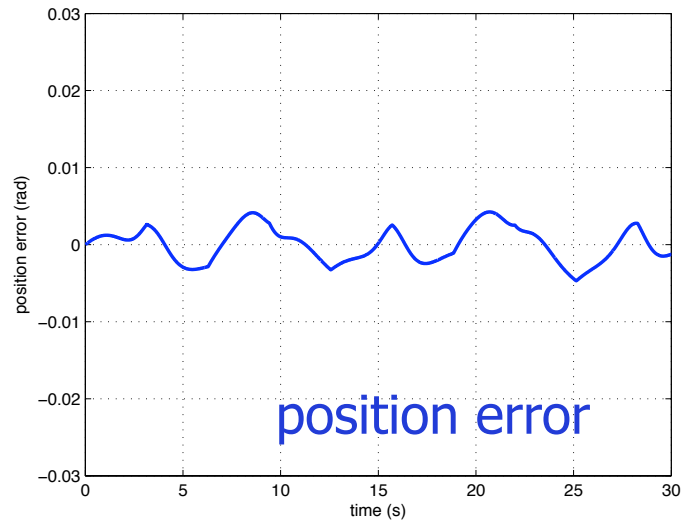
bang-bang acceleration profile

position and velocity errors
again largely **reduced**,
but control **chattering**
and larger effort



Results

second trajectory – smoothed robust control



bang-bang acceleration profile

position and velocity errors
are further **reduced**,
without control chattering
and **same control effort** as
without robustifying term

(using here $\epsilon > 0$)



Appendix A

Proof of bounds on the inertia matrix

- the SVD factorization of the (symmetric) inverse inertia matrix is

$$B^{-1} = U \Sigma^{-1} U^T = U \operatorname{diag}\left\{\frac{1}{\sigma_1}, \dots, \frac{1}{\sigma_n}\right\} U^T \quad m \leq \frac{1}{\sigma_n} \leq \dots \leq \frac{1}{\sigma_1} \leq M$$

so that, with the choice made for its estimate, it follows that

$$\begin{aligned} \|B^{-1} \hat{B} - I\| &= \|U \Sigma^{-1} U^T \hat{B} - I\| \\ &= \|U \Sigma^{-1} U^T \cdot \left(\frac{1}{c} I\right) - I\| \\ &= \|U (\Sigma^{-1} \cdot \frac{1}{c} - I) U^T\| \\ &\leq \|U\| \cdot \|\Sigma^{-1} \cdot \frac{1}{c} - I\| \cdot \|U^T\| \\ &= \|\Sigma^{-1} \cdot \frac{1}{c} - I\| \leq \frac{M}{c} - 1 \\ &= \frac{M - c}{c} = \frac{M - m}{M + m} = \alpha < 1 \end{aligned}$$



Appendix B

Proof of Theorem 2

- setting $w = \mathcal{B}^T P e$, note that for the robust term in the control law it is

$$\|\Delta a\| = \begin{cases} \rho & \text{if } \|w\| \geq \epsilon \\ (\rho/\epsilon)\|w\| < \rho & \text{if } \|w\| < \epsilon \end{cases}$$

- defining as before $V(e) = e^T P e$, we have

$$\begin{aligned} \dot{V}(e) &= -e^T Q e + 2w^T (\Delta a + \bar{\eta}) \\ &\leq -e^T Q e + 2w^T \left(\Delta a + \rho \frac{w}{\|w\|} \right) \end{aligned}$$

having used the chain of inequalities

$$w^T \bar{\eta} \leq \|w^T \bar{\eta}\| \leq \|w\| \cdot \|\bar{\eta}\| \leq \|w\| \rho = w^T \rho \frac{w}{\|w\|}$$

- if $\|w\| \geq \epsilon$, the rest of the proof is the same as in Theorem 1

%



Appendix B

Proof of Theorem 2 (cont)

- if $\|w\| < \epsilon$, the second term in the derivative of V is

$$2w^T \left(-\frac{\rho}{\epsilon} w + \rho \frac{w}{\|w\|} \right) = 2\rho \left(-\frac{\|w\|^2}{\epsilon} + \|w\| \right)$$

with a maximum value $\rho \frac{\epsilon}{2}$ attained for $\|w\| = \frac{\epsilon}{2}$

- therefore, it is

$$\dot{V}(e) \leq -e^T Q e + \rho \frac{\epsilon}{2} \leq -\lambda_{\min}(Q) \|e\|^2 + \rho \frac{\epsilon}{2} < 0$$

provided that

$$\|e\| \geq \left[\frac{\rho \epsilon}{2\lambda_{\min}(Q)} \right]^{1/2} := \omega$$

- if S is the *smallest* level set of $V = e^T P e$ (an ellipsoid) containing the hyper-sphere of radius ω , then

$$e \notin S \implies \dot{V}(e) < 0 \quad \text{and u.u.b. is obtained for } S$$

(an upper bound for the time needed to reach S can be given)

