

Torque Minimization for Redundant Manipulators

Amr Aly

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Abstract

Introduction

Dynamic Resolution

- Minimum Torque Norm

- Peak Torque Reduction

- Stable Torque Minimization Using Short Preview

Simulations

Conclusion

We study torque minimization as a criterion for resolving kinematic redundancy in manipulators. The minimization is studied, locally, at different time instants (current and/or future) in order to address the torque instability problem.

Preliminaries:

- ▶ Solving inverse kinematics for redundant robots yields infinite solutions
- ▶ Secondary objective: minimum joint torques
- ▶ Take into account manipulator dynamics

Different methods for torque optimization

- ▶ Hollerbach – Instantaneous torque minimization
 - ▶ Inclusion of torque bounds
- ▶ Li – Use current torque to minimize torque at next instant
- ▶ Al Khudir – Minimize torque over two successive instants
 - ▶ Inclusion of damping in the null space

Dynamic Model:

$$\boldsymbol{\tau} = \boldsymbol{M}(\boldsymbol{q})\ddot{\boldsymbol{q}} + \boldsymbol{c}(\boldsymbol{q}, \dot{\boldsymbol{q}}) + \boldsymbol{g}(\boldsymbol{q})$$

$$\boldsymbol{n}(\boldsymbol{q}, \dot{\boldsymbol{q}}) = \boldsymbol{c}(\boldsymbol{q}, \dot{\boldsymbol{q}}) + \boldsymbol{g}(\boldsymbol{q})$$

Velocity relationship:

$$\dot{\boldsymbol{x}} = \boldsymbol{J}\dot{\boldsymbol{q}}$$

Acceleration relationship:

$$\ddot{\boldsymbol{x}} = \boldsymbol{J}\ddot{\boldsymbol{q}} + \dot{\boldsymbol{J}}\dot{\boldsymbol{q}} = \boldsymbol{J}\ddot{\boldsymbol{q}} + \boldsymbol{h}(\boldsymbol{q}, \dot{\boldsymbol{q}})$$

Hollerbach minimizes the instantaneous motor torques by resolving the redundancy at the acceleration level.

$$\begin{aligned} \min_{\ddot{\mathbf{q}}_k} \quad & H_1 = \frac{1}{2} \|\boldsymbol{\tau}_k\|^2 \\ \text{s.t.} \quad & \boldsymbol{\tau}_k = \mathbf{M}_k \ddot{\mathbf{q}}_k + \mathbf{n}_k, \\ & \ddot{\mathbf{x}}_k = \mathbf{J}_k \ddot{\mathbf{q}}_k + \mathbf{h}_k \end{aligned}$$

Extension:

Consider the upper and lower joint torque limits are $\boldsymbol{\tau}^+$ and $\boldsymbol{\tau}^-$, respectively. The goal is to place the joint torques closest to the midpoints of the limits $\frac{1}{2}(\boldsymbol{\tau}^+ + \boldsymbol{\tau}^-)$.

$$\begin{aligned} \min_{\ddot{\mathbf{q}}_k} \quad & H_2 = \frac{1}{2} \left(\boldsymbol{\tau}_k - \frac{\boldsymbol{\tau}^+ + \boldsymbol{\tau}^-}{2} \right)^T \mathbf{W} \left(\boldsymbol{\tau}_k - \frac{\boldsymbol{\tau}^+ + \boldsymbol{\tau}^-}{2} \right) \\ \text{s.t.} \quad & \boldsymbol{\tau}_k = \mathbf{M}_k \ddot{\mathbf{q}}_k + \mathbf{n}_k, \\ & \ddot{\mathbf{x}}_k = \mathbf{J}_k \ddot{\mathbf{q}}_k + \mathbf{h}_k \end{aligned}$$

Consider the interval $t_k \leq t \leq t_{k+1}$, where t_k, t_{k+1} are two immediate time instants. The goal is to use the future state $k+1$ as an objective and use current torque τ_k to accelerate/decelerate the joint velocity from $\dot{\mathbf{q}}_k$ to $\dot{\mathbf{q}}_{k+1}^{opt}$ (thus, minimizing τ_{k+1}), while keeping τ_k within limits.

$$\begin{aligned}
 \min_{\dot{\mathbf{q}}} \quad & H'_3 = \frac{1}{2} \left(\tau_{k+1}(\mathbf{q}_{k+1}, \dot{\mathbf{q}}_{k+1}) - \frac{\tau^+ + \tau^-}{2} \right)^T \\
 & \times \mathbf{W} \left(\tau_{k+1}(\mathbf{q}_{k+1}, \dot{\mathbf{q}}_{k+1}) - \frac{\tau^+ + \tau^-}{2} \right) \\
 \text{s.t.} \quad & \dot{\mathbf{x}}_{k+1} = \mathbf{J}(\mathbf{q}_{k+1})\dot{\mathbf{q}}_{k+1}, \\
 & \tau_k^- \leq \tau_k^i \leq \tau_k^+
 \end{aligned}$$

Al Khudir uses two successive discrete-time samples to minimize the joint torque norm.

$$\begin{aligned} \min_{\ddot{\mathbf{q}}_k, \ddot{\mathbf{q}}_{k+1}} \quad & H_4 = \frac{1}{2} \left(\omega_k \|\boldsymbol{\tau}_k\|^2 + \omega_{k+1} \|\boldsymbol{\tau}_{k+1}\|^2 \right) \\ \text{s.t.} \quad & \boldsymbol{\tau}_k = \mathbf{M}_k \ddot{\mathbf{q}}_k + \mathbf{n}_k, \\ & \ddot{\mathbf{x}}_k = \mathbf{J}_k \ddot{\mathbf{q}}_k + \mathbf{h}_k, \\ & \boldsymbol{\tau}_{k+1} = \mathbf{M}_{k+1} \ddot{\mathbf{q}}_{k+1} + \mathbf{n}_{k+1}, \\ & \ddot{\mathbf{x}}_{k+1} = \mathbf{J}_{k+1} \ddot{\mathbf{q}}_{k+1} + \mathbf{h}_{k+1} \end{aligned}$$

But the problem loses the original Linear-Quadratic formulation...

By applying dynamic approximations to the future state (namely, $\ddot{\mathbf{x}}_{k+1}$ and $\boldsymbol{\tau}_{k+1}$), we can obtain a linear dependence of the constraints and a quadratic dependence of the objective on $\ddot{\mathbf{q}}_k$. Thus...

$$\begin{aligned}
 \min_{\ddot{\mathbf{q}}_k, \ddot{\mathbf{q}}_{k+1}} \quad & H_4 = \frac{1}{2} \left(\omega_k \|\boldsymbol{\tau}_k\|^2 + \omega_{k+1} \|\boldsymbol{\tau}_{k+1}\|^2 \right) \\
 \text{s.t.} \quad & \boldsymbol{\tau}_k = \mathbf{M}_k \ddot{\mathbf{q}}_k + \mathbf{n}_k, \\
 & \ddot{\mathbf{x}}_k = \mathbf{J}_k \ddot{\mathbf{q}}_k + \mathbf{h}_k, \\
 & \boldsymbol{\tau}_{k+1} = \mathbf{M}_{k+} \ddot{\mathbf{q}}_{k+1} + \mathbf{S}_{k+} (\dot{\mathbf{q}}_k + \ddot{\mathbf{q}}_k^T) + \mathbf{g}_{k+}, \\
 & \ddot{\mathbf{x}}_{k+1} = \mathbf{J}_{k+} \ddot{\mathbf{q}}_{k+1} + (\mathbf{J}_{k+} - \mathbf{J}_k) \ddot{\mathbf{q}}_k + \mathbf{h}_{k+}
 \end{aligned}$$

Minimize the difference in norm with respect to a suitable desired target torque. Define

$$\boldsymbol{\tau}_{D_k} = -\boldsymbol{D}_k \boldsymbol{M}_k \dot{\boldsymbol{q}}_k$$

When $\boldsymbol{\tau}_k = \boldsymbol{\tau}_{D_k}$ and the damping matrix is in the form $\boldsymbol{D}_k = d\boldsymbol{I} > 0$, the joint acceleration $\ddot{\boldsymbol{q}}_k$ becomes

$$\ddot{\boldsymbol{q}}_k = -\boldsymbol{M}_k^{-1}(\boldsymbol{S}_k \dot{\boldsymbol{q}}_k + \boldsymbol{g}_k) - d\dot{\boldsymbol{q}}_k$$

Effect of $\boldsymbol{\tau}_{D_k}$ is opposite to the current joint velocity, thus acting as a damper on the joint motion.

MTN – with damping (MTND)

$$\begin{aligned} H'_1 &= \frac{1}{2} \|\boldsymbol{\tau}_k - \boldsymbol{\tau}_{D_k}\|^2 \\ &= \frac{1}{2} \|\mathbf{M}_k \ddot{\mathbf{q}}_k + (\mathbf{S}_k + \mathbf{D}_k \mathbf{M}_k) \dot{\mathbf{q}}_k + \mathbf{g}_k\|^2 \end{aligned}$$

MBP – with damping (MBPD)

$$H'_4 = \frac{1}{2} \left(\omega_k \|\boldsymbol{\tau}_k - \boldsymbol{\tau}_{D_k}\|^2 + \omega_{k+1} \|\boldsymbol{\tau}_{k+1} - \boldsymbol{\tau}_{D_{k+1}}\|^2 \right)$$

To obtain new solutions replace \mathbf{S}_k with $(\mathbf{S}_k + \mathbf{D}_k \mathbf{M}_k)$

3R planar manipulator on horizontal plane

- ▶ 3 uniform links, each of length $l = 1$ [m], mass $m_l = 10$ [kg], and moment of inertia $I_l = m_l l^2 / 12$
- ▶ upper and lower torque limits for joints 1-3 are $\pm 54, \pm 24$ and ± 6 [N.m]

Other parameters:

- ▶ Integration step $T_s = 0.001$ [s]
- ▶ Cartesian task error gains $\mathbf{K}_P = 10\mathbf{I}$ and $\mathbf{K}_D = \mathbf{I}$
- ▶ Damping matrix $\mathbf{D}_k = 10\mathbf{I}$

The simulated end-effector motion is a rest-to-rest straight-line Cartesian trajectory, with bang-bang acceleration profile.

1. Short path: path length $\|L_1\| = 0.2828$ [m] and acceleration bound $\|A_1\| = 2.8284$ [m/s^2]
2. Long path: path length $\|L_2\| = 1.1738$ [m] and acceleration bound $\|A_2\| = 1.1412$ [m/s^2]

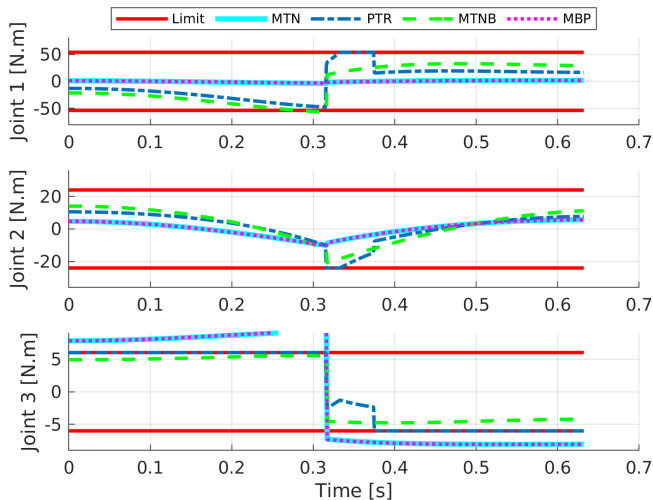


Figure: Torque profiles

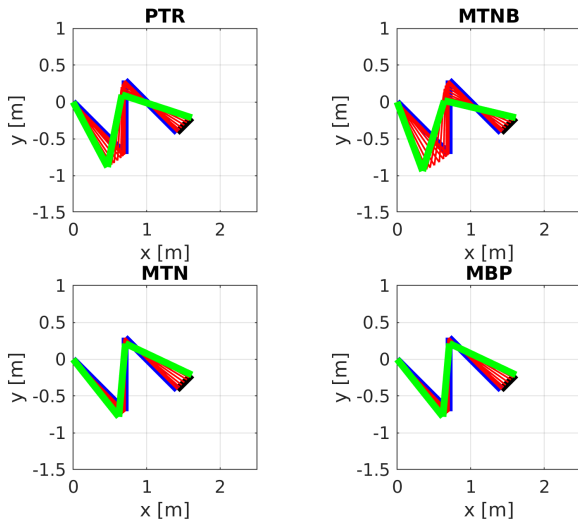


Figure: Continual arm motions (initial in blue, final in green)

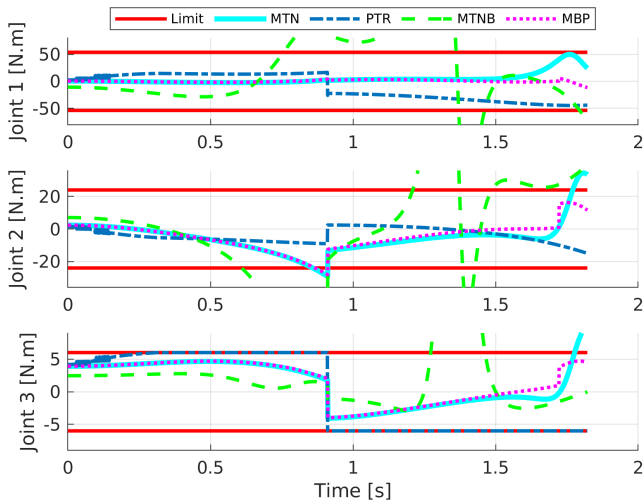


Figure: Torque profiles

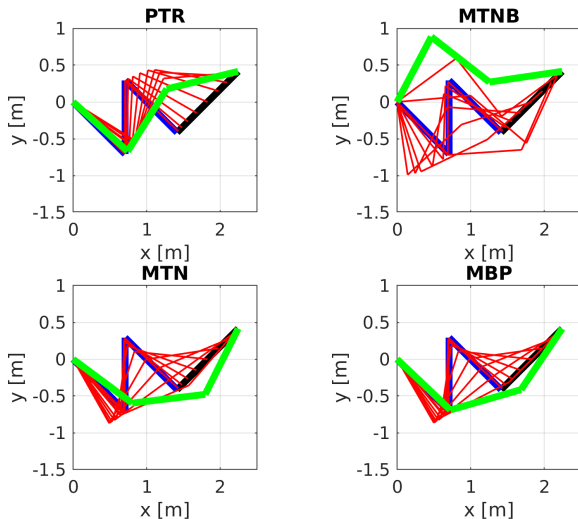


Figure: Continual arm motions (initial in blue, final in green)

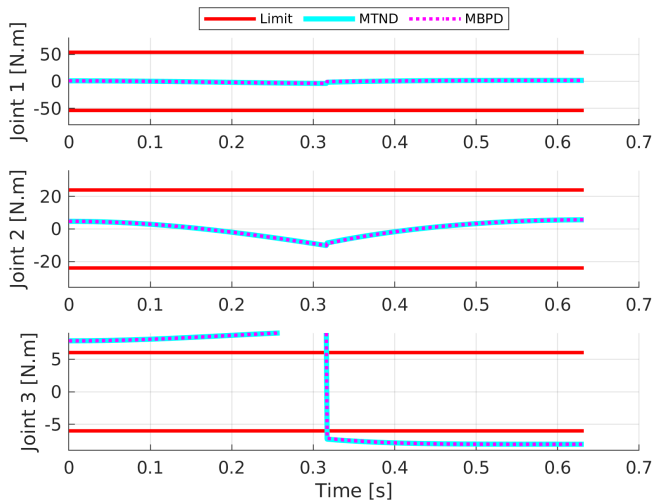


Figure: Torque profiles for short path

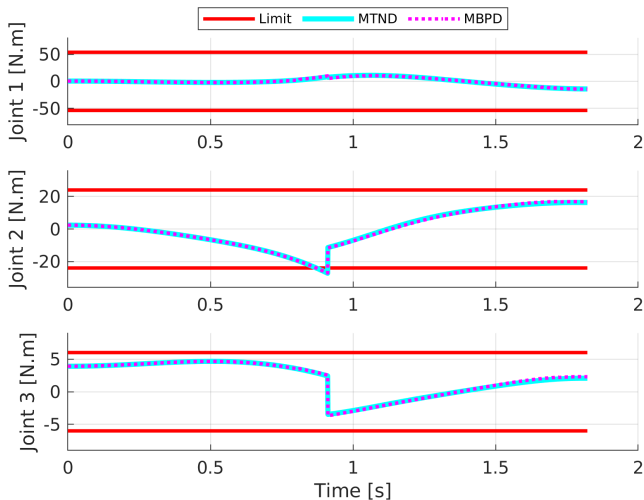


Figure: Torque profiles for long path

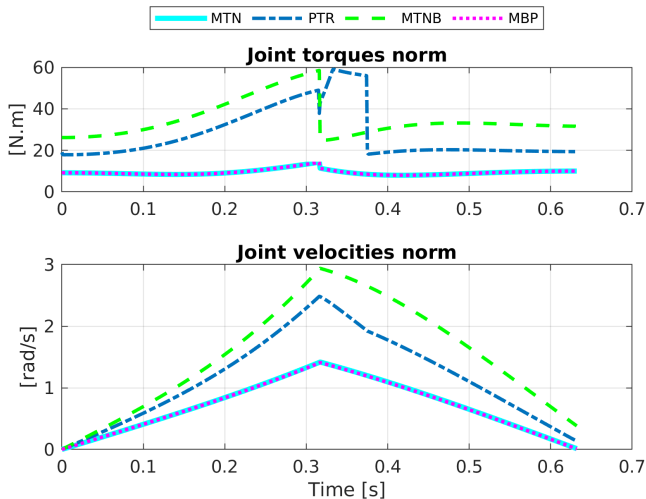


Figure: Joint torque and velocity norms for short path

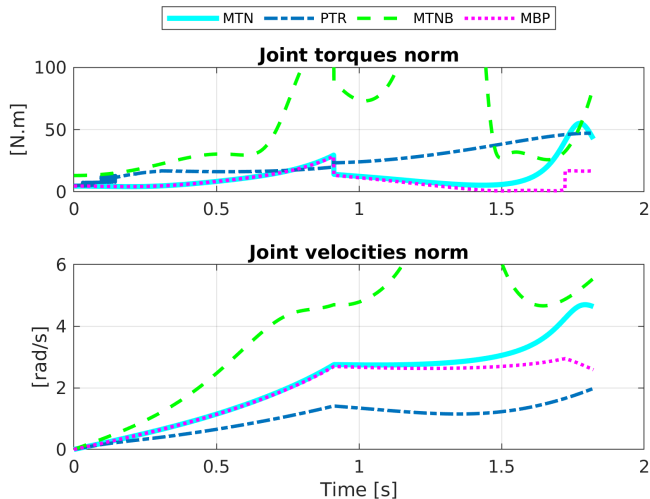


Figure: Joint torque and velocity norms for long path

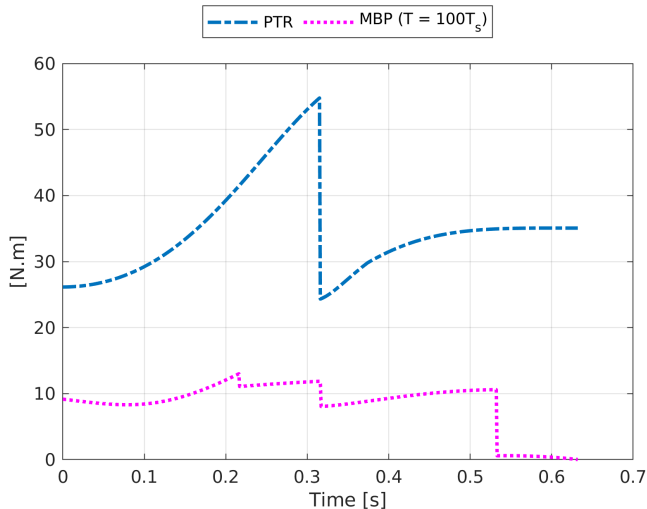


Figure: Future torque estimated by each method (short path)

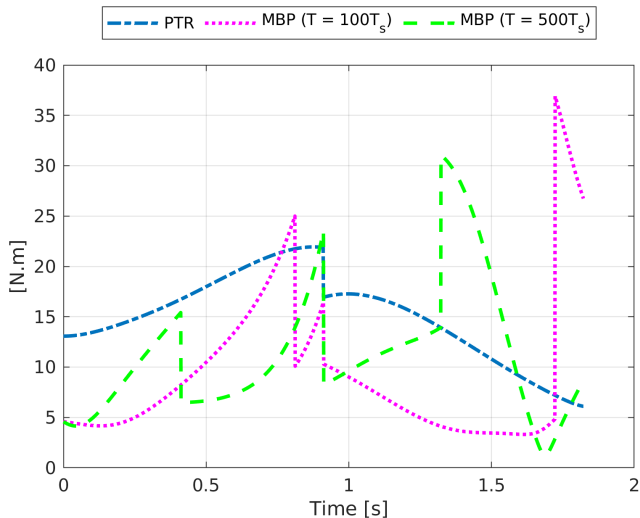


Figure: Future torque estimated by each method (long path)

- ▶ PTR can't be used for real-time control
 - ▶ PTR always respects torque bounds, while MTNB has unstable behaviour for longer paths
 - ▶ PTR has smoother future torque prediction, while MBP has lower norm
 - ▶ No method guarantees stability and respecting the bounds at the same time.
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