



Short communication

A two-step EMG-and-optimization process to estimate muscle force during dynamic movement

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ABSTRACT

The present study proposed a two-step EMG-and-optimization method for muscle force estimation in dynamic condition. Considering the strengths and the limitations of existing methods, the proposed approach exploited the advantages of min/max optimization with constraints on the contributions of the flexor and extensor muscle groups to the net joint moment estimated through an EMG-to-moment approach. Our methodology was tested at the knee joint during dynamic half squats, and was compared with traditional min/max optimization. In general, results showed significant differences in muscle force estimates from EMG-and-optimization method when compared with those from traditional min/max optimization. Muscle forces were higher – especially in the antagonist muscles – and more consistent with EMG patterns because of the ability of the proposed approach to properly account for agonist/antagonist cocontraction. In addition, muscle forces agree with mechanical constraints regarding the net, the agonist, and the antagonist moments, thus greatly improving the confidence in muscle force estimates. The proposed two-step EMG-and-optimization method for muscle force estimation is easy to implement with relatively low computational requirements and, thus, could offer interesting advantages for various applications in many fields, including rehabilitation, clinical, and sports biomechanics.

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1. Introduction

The knowledge of the force developed by muscles during dynamic activity could be of primary interest especially in biomechanics. Due to the very slight possibility of *in vivo* measurements, muscle force estimation remains a major challenge despite numerous existing methods (see Erdemir et al., 2007). Static optimization methods can lead to biased muscle force, especially because those of antagonist muscles are erroneously set to zero (Challis, 1997). Forward dynamics methods (Spagele et al. 1999; Neptune et al., 2000; Seth and Pandey, 2007) increase confidence in muscle force estimates but at high complexity and computational costs, thereby reducing the potential for clinical applications (Erdemir et al., 2007). Pure optimization techniques with appropriate search strategies and algorithms have demonstrated their potential to improve the accuracy of muscle force estimates (e.g., Rasmussen et al., 2001). However, studies have highlighted the need to incorporate electromyography (EMG) in the optimization problem formulation to provide reliable redundant information on muscular activity (e.g., Amarantini and Martin, 2004; Dowling, 1997; Vigouroux et al., 2007). Additional information about

force–length and force–velocity relationships (Olney and Winter, 1985; Buchanan et al., 2004) is also required to gain more confidence in muscle force estimates in dynamic condition.

This study aimed at proposing a two-step EMG-and-optimization method to provide muscle force in dynamic condition while properly accounting for agonist–antagonist cocontraction. The first step was to estimate the contributions of the flexor and extensor muscle groups to the net joint moment using a clinically applicable EMG-to-moment approach (Centomo et al., 2007; Rao et al., 2009). The novelty of the proposed method lies in the second step, which exploited the advantages of min/max optimization with equality constraints imposed on agonist and antagonist muscle moments to estimate muscle force. The results obtained at the knee joint using this two-step EMG-and-optimization approach were compared with those from traditional min/max optimization during half squats. We hypothesized higher force in the antagonist muscles with the proposed method because the presence of cocontraction would be properly taken into account.

2. Methods

The proposed approach was tested using data from a single male participant (age: 28 years, height: 1.77 m, mass: 73 kg) performing seven consecutive half squats at self-selected speed, loaded with 20% body weight. This closed chain

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exercise was chosen as the experimental task because of possible applications of this study especially in rehabilitation (Rao et al., 2009).

The first step of the proposed methodology aimed at computing the net joint moment as well as contributions of the agonist and antagonist muscle groups to the net joint moment. The required input data were similar to those summarized in Amarantini and Martin (2004) (Fig. 1), but the isometric calibration was directly incorporated into the routine used for dynamic conditions for simplified and enhanced clinical applicability (Centomo et al., 2007; Rao et al., 2009). Briefly, inputs to the model were force plate data (AMTI, 240 Hz), sagittal kinematics of the ankle, knee, and hip joints (Vicon, 120 Hz), and EMG from the gastrocnemius medialis (GA), biceps femoris (BF), rectus femoris (RF), and vastus medialis (VM) muscles (BIOPAC MP150, 1000 Hz). Interestingly at this step, the proposed method allows for estimating agonist and antagonist muscle group moments using other models (e.g., Billot et al., 2009; Doorenbosch and Harlaar, 2003).

In the second step, muscle force estimation was formulated as a new constrained min/max optimization problem designed to produce a solution that “distributes the collaborative muscle forces in such a way that the maximum relative muscle force is as small as possible” (Rasmussen et al., 2001) while accounting for agonist–antagonist cocontraction. The input parameters included moment–arms and physiological cross-sectional area (PCSA) while the equality constraints were imposed on the agonist and antagonist muscle moments obtained from first step: find t_i

$$\text{that minimizes } C(t_i) = \max \left(\frac{t_i}{PCSA_i} \right), \quad i \in N^*, \quad i = 1, \dots, p \quad (1)$$

$$\text{subject at each time instant to } \begin{cases} t_i > 0 \\ t_i < \sigma_{\max} PCSA_i \\ \sum_{x=1}^m (r_x t_x) = M_{K\text{flex}} \\ \sum_{y=1}^n (r_y t_y) = M_{K\text{ext}} \end{cases} \quad (2)$$

where t_i and $PCSA_i$ are, respectively, the force and the cross-sectional area (Visser et al., 1990) of muscle i , σ_{\max} is the maximum muscle stress, set constant for all muscles to 40 N/cm² (Prilutsky and Gregor, 1997); p , n , and m are, respectively, the total number of muscles considered (9), the number of knee flexors (biceps femoris short and long heads, semi-membranous, semi-tendinous, gastrocnemius), and knee extensors (vastus lateralis, medialis, and intermedius and rectus femoris). $M_{K\text{flex}}$ and $M_{K\text{ext}}$ are the contributions of the flexor and extensor muscle group, respectively, to the knee net joint moment estimated in the first step of the proposed method and r_i is the moment–arm of muscle i (Visser et al., 1990).

The results by this EMG-and-optimization approach were compared to those obtained by using a traditional min/max optimization method (1) with the constraints on the knee net joint moment (M_K) only:

$$\begin{cases} 0 < t_i < \sigma_{\max} PCSA_i \\ \sum_{i=1}^p (r_i t_i) = M_K \end{cases}, \quad i \in N^*, \quad i = 1, \dots, p \quad (3)$$

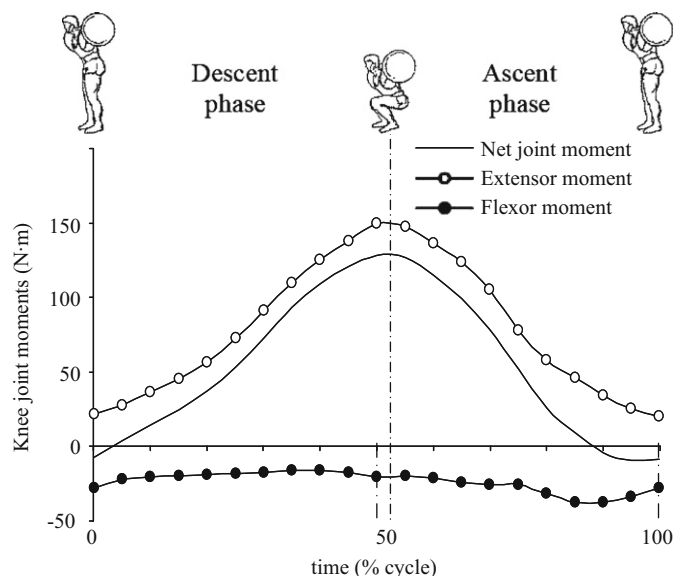


Fig. 1. From top to bottom: schematic representation of movement phases (descent/ascent); patterns of the net, flexor, and extensor moments acting at the knee joint during dynamic half squat.

3. Results

The squat cycle phases and the net, flexor, and extensor moments acting at the knee joint are presented in Fig. 1.

Both the traditional and the EMG-and-optimization approaches provided force estimates below the maximum force of the investigated muscles acting across the knee.

Using the traditional min/max optimization (Figs. 2a–b), force estimates for the muscles agonist to the net joint moment (knee extensors; Fig. 2a) were consistent with observed EMG from RF and VM (not presented here; see Isear et al. (1997) for typical EMG traces during squat), whereas the forces of the antagonist muscles (knee flexors) were predicted to be null during almost the entire squat cycle (Fig. 2b).

The patterns of the agonist muscle (knee extensors) forces were qualitatively similar between EMG-and-optimization and traditional min/max optimization. However, the normalized increase of force was $13.69 \pm 11.19\%$ (overall mean value; Fig. 2c vs. Fig. 2a), with most notable increase for the VL ($28.45 \pm 12.44\%$) compared with the RF, VM, and VI ($8.75 \pm 4.38\%$, $8.84 \pm 4.41\%$, and $8.71 \pm 4.35\%$, respectively). Regarding antagonist muscles (knee flexors), EMG-and-optimization solutions were considerably different from those by traditional min/max optimization (Fig. 2d vs. Fig. 2b), with muscle forces by the EMG-and-optimization approach being very consistent with observed EMG from GA and BF muscles. Whatever the phase of movement, no antagonist muscle force was incorrectly set to zero. As expected from Eq. (2), the sum of individual muscle moments equaled simultaneously the knee net joint moment and the muscle group moments, thus being capable of accounting for agonist/antagonist cocontraction in muscle force estimation.

4. Discussion

This study was aimed at developing a two-step EMG-and-optimization process to estimate muscle force in dynamic condition. The novelty and originality of the method was the introduction of equality constraints on agonist and antagonist muscle moments estimated through an EMG-to-moment optimization procedure (Centomo et al., 2007; Rao et al., 2009). A comparison of our results was made with those from traditional min/max optimization to evaluate the ability of the proposed method to estimate muscle force and to fulfill the specific purpose of taking properly into account agonist/antagonist cocontraction.

The two different approaches provided force estimates for all the 9 knee muscles considered in the present study, even for those from which EMG was not recorded. Qualitatively similar patterns of agonist muscle forces were found, with higher magnitudes obtained from the EMG-and-optimization approach. Such a difference could be explained by significant improvements in the estimation of antagonist muscle forces, and thus by the ability of the EMG-and-optimization approach to adequately account for agonist/antagonist cocontraction. As expected from Eq. (3), the sum of individual muscle moments from traditional min/max optimization equaled the knee net joint moment but differed from the muscle group moments, with forces inadequately set to zero without consistency with EMG of the knee flexors. On the contrary, antagonist muscle forces from EMG-and-optimization were most consistent with EMG patterns, without EMG being directly enforced as a constraint in the min/max optimization problem. EMG was taken only as an indicator of muscle activity to provide input data in the first step dedicated to the estimation of muscle group moments (see Eqs. (1) and (2); Amarantini and Martin, 2004; Centomo et al., 2007; Rao et al., 2009). Even if the EMG-and-optimization approach incorporates

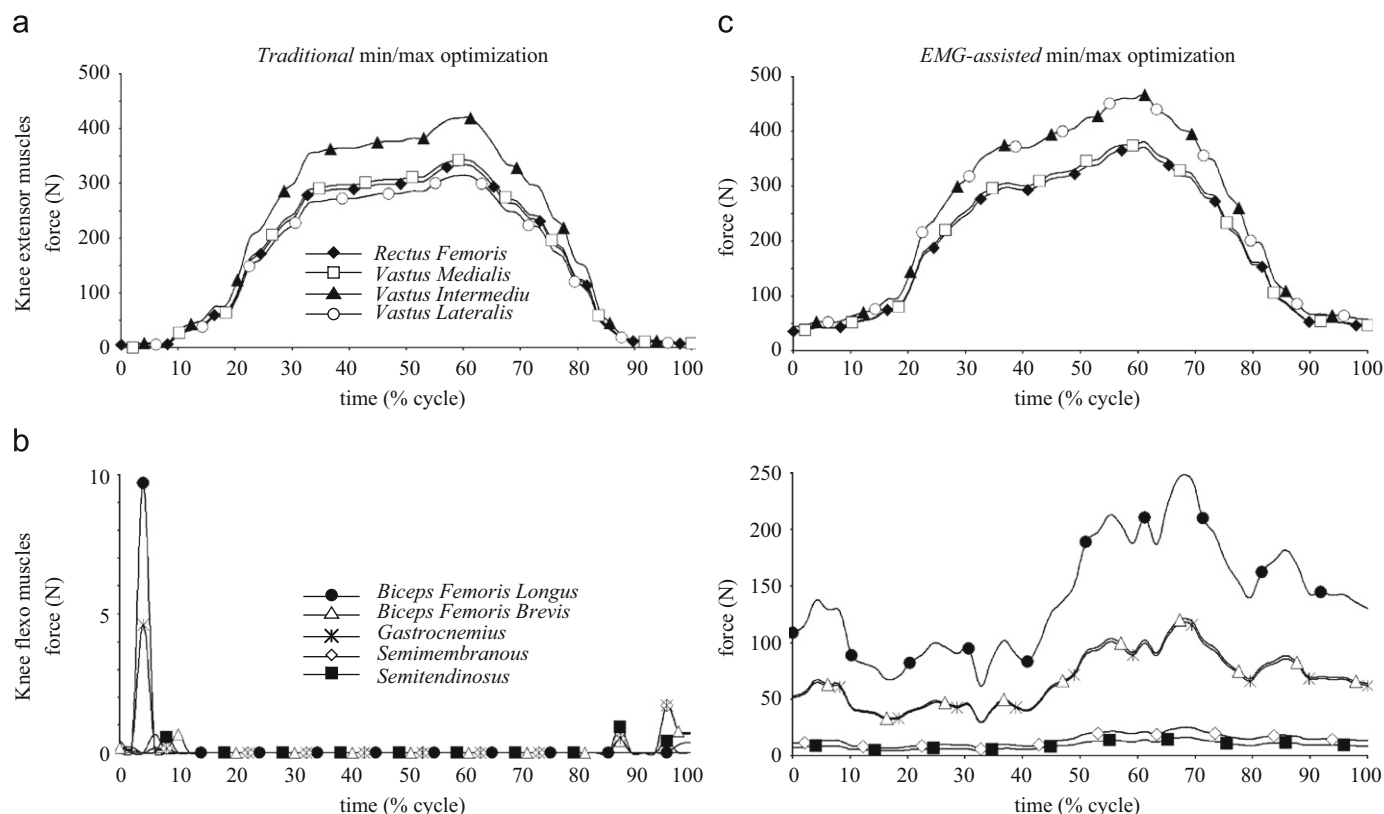


Fig. 2. Force estimates of knee extensor muscles (a)–(c) and knee flexor muscles (b)–(d) obtained from traditional min/max optimization (a, b) and from the proposed EMG-assisted min/max optimization (c, d) during dynamic half squat.

the strength of min/max optimization in force-sharing (Rasmussen et al., 2001) and provides muscle forces compatible with VL muscle dominance (Pincivero et al., 2006; 2008), our results do not definitively resolve the question of appropriateness of force distribution within muscles. In the first step, the proposed method may benefit from more precise modeling of the force-generating potential of muscles. In the second step, it could incorporate PCSA values depending on joint angles (Narici et al., 1996) or pennation angles (Manal et al., 2006), different for each muscle (Buchanan, 1995), to enhance the accuracy of force-sharing solutions. However, an interesting feature of the proposed two-step approach is to reduce the sensitivity of muscle group moments to muscle parameters (Raikova and Prilutsky, 2001), thus providing more reliable equality constraints on min/max optimization in the second step.

The results emphasize that pure traditional min/max optimization can produce misleading muscle force estimates because cocontraction remains hidden in the net joint moment. On the contrary, pure min/max optimization with equality constraints on the agonist and antagonist moments computed in a first step using EMG-to-moment optimization (Centomo et al., 2007; Rao et al., 2009) greatly improves our confidence in muscle force estimation. The solutions by the two-step EMG-and-optimization approach indicate remarkable correspondence with both observed and reported (Isear et al., 1997) EMG patterns. The formulation of the proposed method adequately exploited min/max optimization to properly account for cocontraction.

The proposed method extends the EMG-to-moment model originally developed by Amarantini and Martin (2004) and enables estimation of muscle force during dynamic movement treated as 2D motion. Such reduction of movement dimensionality is reasonable during half squats (Toutoungi et al., 2000) but our force-sharing solutions might not match those of the “real”

3D musculoskeletal system because the key point is the mapping between the solution-space of the optimization problem and the number of degrees of freedom of the musculoskeletal system (Jinha et al., 2006). Nevertheless, the proposed approach is not restricted to 2D analyses. Extension to 3D can be performed by incorporating additional EMG (e.g., Lloyd and Besier, 2003) and moment equilibrium constraints (Cholewicki and McGill, 1994, 1995) to provide mechanically valid optimal solutions for 3D models.

Other improvements can be expected to enhance the accuracy of muscle force estimates using the proposed method. Another criterion than minimization of muscle stress (Eq. (1)) can be used to best represent the neural command and dynamic optimization (Chao and Rim, 1973) can be incorporated – to the detriment of computational cost – to obtain muscle force that best reproduces the measured motion. Even if static and dynamic optimization may produce comparable results (Anderson and Pandey, 2001), a direction for future research could be a comparison of muscle forces from EMG-and-optimization with those from forward dynamics to address questions of validity and sensitivity of force-sharing solutions. An interesting complement would also be to compare the proposed method with other EMG-assisted processes (e.g., Cholewicki and McGill, 1994) to help determine the best way to implement EMG into optimization problems.

In conclusion, the proposed two-step EMG-and-optimization process has the advantages of being noninvasive, of requiring a single experimental session, and of using static optimization, thus making the proposed method easy to implement with acceptably low computational requirements. We think that EMG-and-optimization may offer a convenient way to estimate the muscle force for applications in various fields, including rehabilitation and sports biomechanics. In considering the observations by Erdemir et al. (2007), the generalization of the proposed method

might be of particular interest in clinical biomechanics, e.g., to investigate neural control and to improve diagnosis or surgery planning.

Conflict of interest

We wish to confirm that there are no known conflicts of interest associated with this publication and there has been no significant financial support for this work that could have influenced its outcome.

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