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Review

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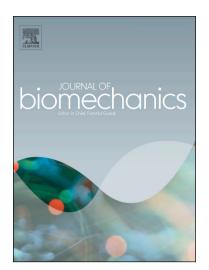
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Alterations of musculoskeletal models for a more accurate estimation of lower limb joint contact forces during normal gait: a systematic review

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Abstract

Musculoskeletal modelling is a methodology used to investigate joint contact forces during a movement. High accuracy in the estimation of the hip or knee joint contact forces can be obtained with subject-specific models. However, construction of subject-specific models remains time consuming and expensive. The purpose of this systematic review of the literature was to identify what alterations can be made on generic (*i.e.* literature-based, without any subject-specific measurement other than body size and weight) musculoskeletal models to obtain a better estimation of the joint contact forces. The impact of these alterations on the accuracy of the estimated joint contact forces were appraised.

The systematic search yielded to 141 articles and 24 papers were included in the review. Different strategies of alterations were found: skeletal and joint model (*e.g.* number of degrees of freedom, knee alignment), muscle model (*e.g.* Hill-type muscle parameters, level of muscular redundancy), and optimisation problem (*e.g.* objective function, design variables, constraints). All these alterations had an impact on joint contact force accuracy but it was not possible to highlight any trend defining which alteration had the largest impact.

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Keywords: survey, joint reaction force, skeletal and joint model, muscle model, optimisation problem, validation

1. Introduction

Musculoskeletal modelling is a computational methodology used to investigate musculotendon forces and joint contact forces during a movement, which cannot be easily measured
(Chèze et al., 2015; Erdemir et al., 2007; Pandy and Andriacchi, 2010). In particular, recent
literature has demonstrated that a high accuracy in the estimation of the hip or tibiofemoral
joint contact forces can be obtained, but induced the use of an extensive personalisation of the
model parameters (Ding et al., 2016; Gerus et al., 2013; Jung et al., 2016; Kia et al., 2014;
Marra et al., 2014). However, the construction of subject-specific models remains a time
consuming procedure that requires not always available data, such as medical imaging.
Conversely, several components of generic musculoskeletal models may be altered that have
noticeable influence on the estimated joint contact forces. Generic models, in this context,
refer to original models established only from literature data and not subject-specific
measurements. The present review focused on possible alterations of generic musculoskeletal
models by assuming that, through a more accurate estimation of joint contact forces, they
could reduce the need of model personalisation, and thus facilitate a broader use of
musculoskeletal modelling.

Several data repositories have been made available and allow to evaluate model accuracy during different motion tasks. Bergmann et al. (Bergmann et al., 2001) have disseminated *in vivo* measurements of hip contact forces, obtained through the use of instrumented hip implants, as well as kinematics and ground reaction forces and moments, during several daily activities (*e.g.* walking at different speeds, up and down stairs, sitting down, standing).

Similarly, Fregly et al. (Fregly et al., 2012) have disseminated a dataset concerning tibiofemoral medial and lateral contact forces during several variant of walking (*e.g.* normal, bouncy, medial trust, turn, acceleration). This data repository was completed with medical imaging data allowing the personalisation of a musculoskeletal model. These repositories have already been used in several studies and allowed to evaluate the accuracy of models with different levels of personalisation.

In this study, a systematic literature review was performed in order to identify what are the alterations that can be made on a generic musculoskeletal model to obtain a more accurate estimation of joint contact forces. The impact of these alterations on the accuracy of the estimated forces was appraised. More specifically, this study focused on lower limb musculoskeletal models and on hip and tibiofemoral joint contact forces during normal gait, because these forces have been now extensively validated against the previously mentioned *in vivo* measurements from instrumented implants.

2. Material & Methods

2.1. Search strategy

An electronic search was performed in Medline, Scopus, and Academic Search Premier databases. The logical (nested) expressions for the search were: musc* and contact and ((hip or knee or tibio*femoral) and (force* or load or reaction)) and (((force-measuring or instrumented or force-instrumented) and (replacement or implant or prosthesis)) or (valid* or accura* or in vivo measurement*)) not cadaver*. The search was based on the title, keywords

and abstract. References cited by the articles remaining after applying the exclusion criteria (see below) were also cross-referenced as well as the articles citing them.

2.2. Exclusion criteria

The articles retrieved from the search strategy were reviewed according to the following exclusion criteria. Studies were omitted if they 1) were published only as conference proceedings, 2) were written in a language other than English, 3) did not refer to a generic musculoskeletal model, 4) did not alter a generic musculoskeletal model, 5) did not include validation against instrumented prosthesis measurements, and 6) did not focus on normal gait. Note that the term generic musculoskeletal model refers in this systematic review to a model established only from literature data. It is the baseline situation in all included studies of this systematic review. However, body size and weight may have been personalised for scaling purposes. Moreover, depending on the study, the related generic musculoskeletal model can be different in terms of modelling approach and level of details.

2.3. Quality assessment

A customised checklist was developed on the basis of previous reviews in the field of biomechanics addressing connected topics (Kainz et al., 2015; Peters et al., 2010) to assess the methodological quality of the selected studies. Each question was rated two (satisfying description or justification), one (limited details) or zero (no information). The 16-item quality checklist used in this review was: Q1: Are the research objectives clearly stated? Q2: Is the study design clearly described? Q3: Is the scientific context clearly explained? Q4: Is the musculoskeletal model adequately described? Q5: Were the model alterations clearly

described? Q6: Is the model for joint contact force estimation adequately described? Q7: Were participant characteristics adequately described? Q8: Were movement tasks, equipment design, and set up clearly defined? Q9: Were the evaluation strategy appropriately justified? Q10: Were the analytical methods clearly described? Q11: Were the statistical methods justified and appropriately described (other than descriptive statistics)? Q12: Were the direct results easily interpretable? Q13: Were the main outcomes clearly stated and supported by the results? Q14: Were the limitations of the study clearly described? Q15: Were key findings supported by other literature? Q16: Were conclusions drawn from the study clearly stated? Each study was evaluated independently by the three authors (FM, LM, and RD) for this assessment. In case of discrepancy, the original article was checked to ensure the correct coding, and a consensus was found between authors.

2.4. Data extraction

A customised data extraction form was developed to extract key details from each selected study. One author (FM) performed the data extraction and the two other authors (LM and RD) checked the final form to ensure reliability. The retained information consisted in: 1) the investigated joint (*i.e.* hip joint or tibiofemoral joint), 2) the definition of the generic musculoskeletal model (*i.e.* software used in the simulations, original model, number of degrees of freedom (DoFs) and of muscular lines of action, muscle paths (*i.e.* straight lines, via points, wrapping surfaces), method used to solve the muscle redundancy problem and joint contact model), 3) the validation dataset used to assess model accuracy, 4) the alterations applied to the model, and 5) their impact on the joint contact force accuracy. These themes were chosen to provide an overview of the methods of each selected study together with the presented alterations and associated results. When the methods were only partially described

in the original article, comprehensive information was retrieved from references and author's previous works to provide comparable data across the selected studies.

3. Results

3.1. Search strategy yield

The search strategy allowed to identify 99 articles in Medline, 65 in Scopus, and 77 in Academic Search Premier, yielding to 141 articles without any duplicate (Fig. 1). According to the exclusion criteria, 20 articles were retained. Most of the excluded articles did not include validation against *in vivo* measurements from instrumented prosthesis. Others presented either a fully subject-specific musculoskeletal model (Ding et al., 2016; Gerus et al., 2013; Jung et al., 2016; Kia et al., 2014) or a generic musculoskeletal model without alteration (Heller et al., 2001; Kim et al., 2009; Modenese and Phillips, 2012; Purevsuren et al., 2016; Stansfield et al., 2003; Trepczynski et al., 2012). Four more articles were obtained by cross-referencing (Dumas et al., 2012; Lund et al., 2015; Manal and Buchanan, 2013; Steele et al., 2012), yielding the final set of 24 articles. Note that this final set was only composed of rigid body musculoskeletal models (some of them including deformable joints), while both rigid body and finite element musculoskeletal models were initially included. Quality assessment and data extraction results are reported below. Details can be found in Table S1 (available as supplementary material) and Table 1, respectively.

3.2. Quality assessment

The overall score of each article was thus calculated by the sum of rated questions divided by the sum of applicable questions. The selected studies were all of high quality, with a scoring ranged between 80% and 100%, and a mean score of 92% (Table S1, available as supplementary material). However, several questions were only partially answered. For example, validation (or comparison with validation dataset) was not always explicitly mentioned as an objective (or a method), elements of the methods (*i.e.* the musculoskeletal model, computation method for joint contact forces, or the validation dataset) were described just through a reference to previous work, or limitations of the study were not clearly stated or omitted. Only 3 of the 24 selected studies (see Table 1) reported a statistical method other than descriptive statistics (*i.e.* mean, standard deviation, root mean square error, correlation).

3.3. Generic musculoskeletal models

More than half of the studies (13 out of 24) were based on specialised biomechanical software such as Opensim (Delp et al., 2007) and Anybody (Damsgaard et al., 2006). Six studies used Matlab (The Mathworks, USA) to perform the computations. Finally, SIMM (Motion Analysis Corporation, Musculographics, USA), ADAMS (MSC Software, USA) or custom-made software were used in the remaining studies.

More than half of the selected studies (13 out of 24) used models based on the generic musculoskeletal model of Delp et al. (Delp, 1990). The original or a variant of the model developed by Klein Horsman et al. (Klein Horsman et al., 2007) or by Arnold et al. (Arnold et al., 2010) were also used. Two further models were also developed by other authors (Heller et al., 2005, 2001; Lin et al., 2010).

The number of DoFs was related to the generic musculoskeletal model used in each study. Briefly, the primary differences were observed at the hip, tibiofemoral, and patellofemoral joints were the number of DoFs were 3 or 6, 1 or 6, and 0, 1 or 6, respectively.

The number lines of action were also related to the generic musculoskeletal model used in each study and ranged between 11 (Lin et al., 2010) and 163 (Chen et al., 2014; Modenese et al., 2013, 2011; Moissenet et al., 2016; Zhang et al., 2015). In all the selected studies, the muscle paths were enhanced by via points only (15) or with (9) the use of wrapping surfaces. A majority of studies (17 out of 24) solved the muscle redundancy problem (also called the force distribution problem (Crowninshield and Brand, 1981)) through an inverse dynamicsbased optimisation, while some others used a forward dynamics-based optimisation (Guess et al., 2014), an EMG-driven method (Manal and Buchanan, 2013), a reduction method (Lundberg et al., 2013, 2012), computed muscle control method (Hast and Piazza, 2013; Thelen et al., 2014), or a mixed approach (Walter et al., 2014). Proposed objective functions included the sum of muscle activations, or musculo-tendon forces, at different power, both weighted and not. Some studies included additional terms to the objective function, such as reserve actuators (Modenese et al., 2013; Serrancoli et al., 2016) or joint loads (Demers et al., 2014; Lin et al., 2010; Moissenet et al., 2016, 2014). In one study, a min-max method was applied on musculo-tendon forces (Chen et al., 2014). Other objective functions (Manal and Buchanan, 2013; Walter et al., 2014) were mainly based on the tracking of a set of variables (e.g. joint moments).

The computation of joint contact forces was mainly based on 1-point or 2-point rigid contact models (respectively 10 and 5 out of 24 studies). However, several studies introduced deformable structures by using a force dependent kinematics method (Chen et al., 2014; Zhang et al., 2015), a deformable contact with viscous damping (Guess et al., 2014), a rigid body spring model (Hast and Piazza, 2013), or a surrogate contact modelling (Lin et al.,

2010). Some other studies are based on regression equations converting the varus-valgus moment to medial and lateral tibiofemoral contact forces (Lundberg et al., 2013, 2012), or on prosthesis calibration (Serrancoli et al., 2016; Walter et al., 2014).

3.4. Validation datasets

Among the 24 selected studies, 19 were interested in particular in the tibiofemoral joint, and 5 in the hip joint. The validation datasets used in these studies are directly related to the joint of interest.

In most of the selected studies, *in vivo* measurements of hip contact forces were obtained from the HIP98 dataset (Bergmann et al., 2001), while tibiofemoral joint from the six editions of the Grand Challenge Competition to Predict In Vivo Knee Loads (Fregly et al., 2012). In one study, the HIP98 dataset was pooled into a "typical patient" (Heller et al., 2005). In two studies (Lin et al., 2010; Lundberg et al., 2012), the datasets associated with the studies published by Zhao et al. (Zhao et al., 2007) and Mündermann et al. (Mündermann et al., 2008) were used.

3.5. Alterations of the generic musculoskeletal model

Three types of alterations of the generic musculoskeletal model have been proposed in the selected studies: 1) skeletal and joint models, 2) muscle model, and 3) optimisation problem. Details are given below.

3.5.1. Skeletal and joint models

Alterations of the skeletal and joint models concern kinematics and/or dynamics, and can be divided in nine sub-group: markers' placement (Lund et al., 2015; Navacchia et al., 2016), number of DoFs (Dumas et al., 2012; Lundberg et al., 2013), knee alignment (Lerner et al., 2015; Navacchia et al., 2016; Thelen et al., 2014), femoral anteversion (Heller et al., 2001), scaling strategy (Chen et al., 2014; Lund et al., 2015), medial-lateral tibiofemoral contact forces ratio (Lundberg et al., 2013), inertial parameters (Navacchia et al., 2016), contact points (Lerner et al., 2015; Manal and Buchanan, 2013), contact stiffness (Chen et al., 2014; Zhang et al., 2015), and joint passive stiffness (Dumas et al., 2012; Lundberg et al., 2013).

3.5.2. Muscle model

Several studies also altered the muscle model by introducing some variations in: muscle geometry (*i.e.* insertion sites, muscle path) (Navacchia et al., 2016; Zhang et al., 2015), Hill-type muscle parameters (*e.g.* optimal fibre length, tendon slack length, pennation angle, EMG-driven model parameters) (Manal and Buchanan, 2013; Navacchia et al., 2016; Serrancoli et al., 2016), or level of muscular redundancy (*i.e.* number of lines of action) (Heller et al., 2005; Moissenet et al., 2016).

3.5.3. Optimisation problem

Five studies altered the objective function of the optimisation problem by modifying the objective function (Chen et al., 2014; Modenese et al., 2011; Zhang et al., 2015), or the optimisation weights (Knarr and Higginson, 2015; Steele et al., 2012). Four studies altered the design variables by introducing in the optimisation problem: joint contact forces (Demers et al., 2014; Lin et al., 2010; Modenese et al., 2013; Moissenet et al., 2014), ligament and bone

forces (Moissenet et al., 2014), or reserve actuators (Modenese et al., 2013) instead of only muscle activations or musculo-tendon forces. Three studies altered the optimisation constraints by modifying the boundaries of design variables (Hast and Piazza, 2013; Knarr and Higginson, 2015) or by introducing additional constraints (*e.g.* residual loads acting on the tibia) (Lin et al., 2010). Other cases concerned the alteration of PID gains in a forward dynamics-based optimisation (Guess et al., 2014) and the use of EMG synergies (Walter et al., 2014).

3.6. Level of evidence

All the alterations proposed in the selected studies had an impact on joint contact force accuracy. However, the present review did not allow to highlight any trend defining which alteration has the largest potential to improve accuracy. Moreover, the evidence of this impact highly varied between studies, and could be divided in four categories. The first category corresponded to studies providing quantitative evidence with statistics (*e.g.* t-test, ANOVA, 95% confidence interval, Monte Carlo simulation) (Lerner et al., 2015; Lund et al., 2015; Serrancoli et al., 2016). In these studies, results related to accuracy were reported using the root mean square errors (RMSE) between the measured and estimated joint contact forces, expressed in Newton or in body weight (BW) (to ease comparison, all data have been expressed in BW in Table 1). The second category corresponded to studies providing quantitative evidence with only descriptive statistics (Demers et al., 2014; Heller et al., 2005, 2001; Knarr and Higginson, 2015; Lin et al., 2010; Lundberg et al., 2013, 2012; Manal and Buchanan, 2013; Modenese et al., 2013, 2011, Moissenet et al., 2016, 2014; Walter et al., 2014). In most of these studies, results related to accuracy were reported using the root mean square error (RMSE) between the measured and estimated joint contact forces. The third

category corresponded to studies providing only a partial quantitative evidence. Two cases could be listed in this category. Firstly, only RMSE between baseline and altered joint contact forces was given (Chen et al., 2014; Zhang et al., 2015). Secondly, only quantitative evidence on accuracy at baseline or after alterations was given (Navacchia et al., 2016; Steele et al., 2012; Thelen et al., 2014). In both cases, the results highlighted the alteration impact on accuracy, but do not quantify it. The last category corresponded to studies providing no quantitative evidence (Dumas et al., 2012; Guess et al., 2014; Hast and Piazza, 2013). In this case, results related to accuracy were presented in figures, leading to an approximate interpretation of accuracy.

3.7. Impact of the alterations on joint contact force accuracy

On the whole, by merging all reported quantitative data before alteration (16 out 24 studies), the range of RMSE at baseline (between estimated and measured joint contact forces) extracted from the selected studies providing this value was 0.17 - 1.39 BW for the tibiofemoral medial contact force, 0.18 - 0.81 BW for the tibiofemoral lateral contact force, 0.30 - 0.88 BW for the tibiofemoral total contact force, and 0.45 - 0.59 BW for the hip contact force (Figure 2 and Figure 3: some ranges may differ with these figures where only studies providing baseline and alterations RMSE were reported). After alterations (13 out of 24 studies), the range of RMSE was 0.08 - 0.50 BW for the tibiofemoral medial contact force, 0.09 - 0.63 BW for the tibiofemoral lateral contact force, 0.15 - 0.77 BW for the tibiofemoral total contact force, and 0.18 - 4.52 BW for the hip contact force (Figure 2 and Figure 3). At the hip joint, the use of high power in the objective function had a negative impact on the upper bound of accuracy (Figure 3). At the knee joint, instead, both the minimum and maximum RMSE were decreased for each evaluated contact force component. Moreover,

unsurprisingly, the minimum RMSE was obtained when tracking the measured tibiofemoral joint contact forces (Serrancoli et al., 2016; Walter et al., 2014) by reaching a value of 0.08 BW, 0.09 BW, and 0.15 BW, respectively for the tibiofemoral medial, lateral and total contact forces (Figure 2). It must also be noticed that some alterations mainly have an impact on the medial-lateral tibiofemoral contact forces ratio, with only slight variations on the total contact force accuracy (Moissenet et al., 2016; Thelen et al., 2014).

4. Discussion

4.1. Accuracy assessment

The way the joint contact force accuracy was assessed varies between studies, leading to different levels of interpretation. Firstly, the alterations could be simply applied (Demers et al., 2014; Dumas et al., 2012; Guess et al., 2014; Hast and Piazza, 2013; Heller et al., 2005, 2001; Knarr and Higginson, 2015; Lerner et al., 2015; Lin et al., 2010; Lund et al., 2015; Lundberg et al., 2013; Manal and Buchanan, 2013; Modenese et al., 2013, 2011, Moissenet et al., 2016, 2014; Serrancoli et al., 2016; Steele et al., 2012; Thelen et al., 2014; Walter et al., 2014) or explored through a sensitivity analysis (Chen et al., 2014; Lundberg et al., 2012; Navacchia et al., 2016; Zhang et al., 2015). Secondly, it has been showed that the provided level of evidence varied between studies. This was mainly due to differences in the generic musculoskeletal model and in the type of alteration. Indeed, results clearly show that the baseline situation of most of the studies is different, with various modelling approaches (e.g. computation of joint contact forces based on one-point rigid contact model versus deformable contact with viscous damping model) and different levels of details (e.g. number lines of action varying between 11 and 163). Thirdly, the units of the error metric used to report joint

contact forces accuracy also differed between studies (e.g. Newton, BW, percents). Even if RMSE expressed in BW has been used in several works, a consensual methodology would have eased the comparison.

4.2. Strategies of alteration

With some exceptions, most of the strategies of alteration could be gathered in two categories in the selected studies.

On one hand, alterations could have a substantial impact on the model lever arms. This could be done by tuning the position and orientation of joint centres and segmental coordinate systems (*e.g.* indirectly through markers' placement (Lund et al., 2015; Navacchia et al., 2016)). Another option was to modify the position of the contact points by tuning the knee alignment (Lerner et al., 2015; Navacchia et al., 2016; Thelen et al., 2014), or the medial-lateral position of contact points (Lerner et al., 2015; Manal and Buchanan, 2013). In the context of a multi-contact joint (*e.g.* tibiofemoral joint), the lever arms of joint contact forces was then modified, and thus the repartition of forces between contact points. A third possibility was to modify the muscular geometry by tuning the skeletal geometry (*e.g.* femoral anteversion (Heller et al., 2001)), or directly the muscular geometry (*e.g.* insertion sites, muscle path (Navacchia et al., 2016; Zhang et al., 2015)). The muscular lever arms were then directly affected, leading to new estimations of musculo-tendon forces and thus joint contact forces.

On the other hand, alterations could have a substantial impact on the distribution of musculotendon forces (*i.e.* on the solution of the muscular redundancy problem). This could first be done by modifying the number of DoFs (Dumas et al., 2012; Lundberg et al., 2013). Indeed, increasing the number of DoFs may results in the recruitment of muscle aiming to stabilise the

altered joint (Jinha et al., 2006). The sum of musculo-tendon forces related to this joint may thus be increased, leading to a higher joint contact force (Dumas et al., 2012). The muscular redundancy also played a key role in the distribution of musculo-tendon forces (Heller et al., 2005; Moissenet et al., 2016). By increasing the number of muscular lines of action, the load may be shared by several musculo-tendon units with different lever arms, and thus impact joint contact forces. Finally, the way the optimisation problem was defined had a substantial impact on the force sharing problem. By altering the objective function (Chen et al., 2014; Knarr and Higginson, 2015; Modenese et al., 2011; Steele et al., 2012; Zhang et al., 2015), the design variables (Demers et al., 2014; Lin et al., 2010; Modenese et al., 2013; Moissenet et al., 2014), or the optimisation constraints (Hast and Piazza, 2013; Knarr and Higginson, 2015; Lin et al., 2010), the way the muscles were recruited varied and generated different musculotendon forces amplitudes and patterns. This could be explained by the fact that these alterations modified the solution space of the optimisation problem, allowing the identification of new optimal solutions.

4.3. Limitations

The present systematic review was limited to studies exploring only lower limb musculoskeletal models during normal gait, where joint contact estimations were validated against instrumented prosthesis measurements. Several types of musculoskeletal model may thus have been omitted. This was for example the case for finite element models (*e.g.* Adouni et al., 2016; Beillas et al., 2004; Marouane et al., 2017, 2016) for whose no study was found with validation against instrumented prosthesis measurements. Similarly, several alterations may thus have been omitted. This was for example the case for the study of Smith et al. (Smith et al., 2016), studying modified styles of gait (*i.e.* smooth and bouncy), where a variant

of the Arnold et al. model (Arnold et al., 2010) was used in an inverse dynamics-based optimisation to evaluate the impact of alterations on knee alignment and knee ligament stiffness on tibiofemoral joint contact forces accuracy. It was also interesting to note that the studies using subject-specific musculoskeletal models, excluded from this review, specifically personalised the following items: the bone geometry and the origin and insertion of muscle line of actions (Ding et al., 2016; Gerus et al., 2013; Jung et al., 2016; Kia et al., 2014; Marra et al., 2014), the joint kinematics (*i.e.* coupling curves between de DoFs) (Gerus et al., 2013), and the joint geometry (*i.e.* prosthesis design) when deformable joint was introduced (Jung et al., 2016; Kia et al., 2014; Marra et al., 2014).

5. Conclusion

Musculoskeletal modelling is a computational methodology which involves determination of a large number of parameters and several implementation choices. With the objective of a more accurate estimation of the joint contact forces, some of these parameters, typically defining the joint and muscle geometries, have been personalised using medical imaging. According to this systematic review, many of these model parameters and implementation choices have been also altered from a generic musculoskeletal model, *e.g.* number of degrees of freedom, muscle parameters, level of muscular redundancy, objective function, constraints, *etc.* All these alterations had an impact on the accuracy of the hip and tibiofemoral joint contact forces, so demonstrating the potential for improving model prediction without necessarily involving costly and time consuming medical images. However, due to discrepancies in the reported evidence about this impact and despite a high quality of the reviewed studies, it was not possible to highlight any trend defining which alteration had the largest impact. If results of future studies implementing model alterations will be reported with consistent metrics and level of evidence, it might be possible to identify specific

modelling features that, if personalised, will lead to an increased accuracy in the estimation of joint contact forces in the lower limb.

Conflict of interest

The authors declare that the study did not raise any conflict of interest.

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Tables and figures captions

Table 1: Overview of selected studies including generic musculoskeletal model, validation dataset, model alterations, available statistics (other than descriptive), and impact of the alterations on joint contact force accuracy.

Figure 1: Flowchart of the search strategy conducted in this review.

Figure 2: Mean errors on tibiofemoral medial, lateral and total contact forces between baseline and alterations. See Table 1 for the numbering of the alterations.

Figure 3: Mean errors on hip contact forces between baseline and alterations. See Table 1 for the numbering of the alterations.

Table 1

A/P: Anterior-posterior, BW: Body weight, DoF: Degree of freedom, F/E: Flexion-extension, I/E: Internal-external rotation, PF: Patellofemoral, PID: Proportional integral derivative, R2: Determination coefficient, RMSE: Root mean square error, TF: tibiofemoral, VH: Visible human

Article	Joint	Generic musculoskeletal model	Validation dataset	Model alterations	Impact of the alterations on joint contact force accuracy
Chen et al. (2014)	TF	Software: Anybody Model: Variant of the Klein Horsman et al. (2007) model DoFs: 6 pelvis, 3 hip, 1 TF, 1 ankle, 1 subtalar Muscular lines of action: 163 per leg Muscle path: Via points and wrapping surfaces Optimisation: Inverse dynamics-based optimisation Objective function: Min-max muscle recruitment Joint contact model: Force dependent kinematic (Andersen & Rasmusen 2011)	Third edition data of the Grand Challenge Competition to Predict In Vivo Knee Loads (Fregly et al., 2012)	Optimisation problem: - Objective function Joint model: - Contact stiffness - Scaling strategy (alterations performed under a sensitivity analysis)	Conclusion: Alterations slightly impacted TF contact force estimations and thus model accuracy Quantitative evidence: At baseline, RMSE of medial, lateral, and total contact forces was 0.28 BW, 0.23 BW, and 0.45 BW. After alteration, the maximal RMSE with baseline was (1) 0.09 BW, (2) 0.07 BW, and (3) 0.08 BW Statistics: None
DeMers et al. (2014)	TF	Software: Opensim Model: Variant of the Delp et al. (1990) model DoFs: 6 pelvis, 3 hip, 1 TF, 1 ankle Muscular lines of action: 46 per leg Muscle path: Via points Optimisation: Inverse dynamics-based optimisation Objective function: Sum of squared muscle activations and joint loads Joint contact model: One-point rigid contact model (Steele et al., 2012)	First edition data of the Grand Challenge Competition to Predict In Vivo Knee Loads (Fregly et al., 2012)	Optimisation problem: - Design variables: (1) Muscle activations (2) Joint loads	Conclusion: A wide range of TF contact force, and thus model accuracy, can be obtained depending on the objective function Quantitative evidence: First and second force peaks during stance were (1) 0.4 BW and 1.7 BW larger, and (2) <0.1 BW and 1.5 BW lower than validation data Statistics: None
Dumas et al. (2012)	TF	Software: Matlab Model: Variant of the Delp et al. (1990) model DoFs: 6 pelvis, 3 hip, 1 TF, 1 ankle Muscular lines of action: 43 per leg Muscle path: Via points Optimisation: Inverse dynamics-based optimisation Objective function: Sum of squared muscle stresses Joint contact model: Two-point rigid contact (Feikes et al., 2003)	Second edition data of the Grand Challenge Competition to Predict In Vivo Knee Loads (Fregly et al., 2012)	Joint model: - Number of DoFs - Passive stiffness	Conclusion: The use of hinge joints with coupled DoFs at knee and ankle, and the introduction of passive joint moments, improved model accuracy Quantitative evidence: No quantitative measurement of model accuracy Statistics: None
Guess et al. (2014)	TF	Software: ADAMS and Simulink/Matlab Model: Variant of the Delp et al. (1990) model DoFs: 6 pelvis, 3 hip, 6 TF, 6 PF, 3 ankle, 1 toe Muscular lines of action: 44 per leg Muscle path: Via points Optimisation: Forward dynamics-based optimisation Objective function: Track inverse dynamics joint angles with PID Joint contact model: Deformable contact with viscous damping (Machado et al., 2012)	First edition data of the Grand Challenge Competition to Predict In Vivo Knee Loads (Fregly et al., 2012)	Optimisation problem: - PID gains	Conclusion: A decrease of PID gain improved model accuracy Quantitative evidence: No quantitative measurement of model accuracy Statistics: None

Article	Joint	Generic musculoskeletal model	Validation dataset	Model alterations	Impact of the alterations on joint contact force accuracy
Hast and Piazza (2013)	TF	Software: SIMM Model: Variant of the Delp et al. (1990) model DoFs: 6 pelvis, 3 hip, 6 TF, 6 PF, 1 ankle Muscular lines of action: 13 per leg Muscle path: Via points and wrapping surfaces Optimisation: Computed muscle control Objective function: Track inverse dynamics joint angles with PID (forward dynamics level) and sum of squared residuals between muscle activations and normalised EMGs (inverse dynamics level) Joint contact model: Rigid body spring model (Li et al., 1997)	Second edition data of the Grand Challenge Competition to Predict In Vivo Knee Loads (Fregly et al., 2012)	Optimisation problem: - Optimisation constraints: maximum isometric forces	Conclusion: A decrease of maximum isometric forces led to a decrease of TF contact forces and an improvement of model accuracy Quantitative evidence: No quantitative measurement of model accuracy Statistics: None
Heller et al. (2001)	Hip	Software: Custom-made software Model: Own model based on VH project Ackerman (1998) DoFs: 6 pelvis, 3 hip, 6 TF, 3 ankle Muscular lines of action: 95 per leg Muscle path: Via points Optimisation: Inverse dynamics-based optimisation Objective function: Sum of musculo-tendon forces Joint contact model: One-point rigid contact model	HIP98 (Data of subjects HSR, KWR, PFL, IBL) (Bergman et al., 2001)	Joint model: - Femoral anteversion	Conclusion: An increase of femoral anteversion led to an increase of hip contact forces and a decrease in model accuracy, while a decrease of femoral anteversion led to little or no change Quantitative evidence: Hip superior-inferior contact force increased by up to 24% compared to validation data Statistics: None
Heller et al. (2005)	Hip	Software: Custom-made software Model: Own model based on VH project Ackerman (1998) DoFs: 6 pelvis, 3 hip, 6 TF, 3 ankle Muscular lines of action: 95 per leg Muscle path: Via points Optimisation: Inverse dynamics-based optimisation Objective function: Sum of musculo-tendon forces Joint contact model: One-point rigid contact model	HIP98 (Typical patient, averaging procedure based on Fourier analysis) (Bergman et al., 2001)	Muscle model: - Muscular redundancy	Conclusion: A decrease of the number of muscular lines of action led to a decrease in model accuracy Quantitative evidence: Hip superior-inferior contact force differed by 1% with baseline model and 7% with altered model compared to validation data Statistics: None
Knarr and Higginson (2015)	TF	Software: Opensim Model: Variant of the Delp et al. (1990) model DoFs: 6 pelvis, 3 hip, 1 TF, 1 ankle, 1 toe Muscular lines of action: 46 per leg Muscle path: Via points Optimisation: Inverse dynamics-based optimisation Objective function: Sum of squared muscle activations Joint contact model: One-point rigid contact model (Steele et al., 2012)	First, second and third editions data of the Grand Challenge Competition to Predict In Vivo Knee Loads (Fregly et al., 2012)	Optimisation problem: - Objective function: (1) Sum of squared muscle activations (baseline) (2) Weighted sum with uniform weights (3) And with subject-specific weights - Optimisation constraints: (4) Baseline with subject-specific musculo-tendon forces boundaries	Conclusion: Parameters' personalisation improved model accuracy Quantitative evidence: RMSE range of the total contact force was (1) 0.37-0.67 BW, (2) 0.43-0.48 BW, (3) 0.33-0.41 BW, (4) 0.37-0.40 BW Statistics: None

Article	Joint	Generic musculoskeletal model	Validation dataset	Model alterations	Impact of the alterations on joint contact force accuracy
Lerner et al. (2015)	TF	Software: Opensim Model: Variant of the Delp et al. (1990) model DoFs: 6 pelvis, 3 hip, 1 TF, 1 ankle, 1 subtalar Muscular lines of action: 46 per leg Muscle path: Via points Optimisation: Inverse dynamics-based optimisation Objective function: Sum of squared weighted muscle activations, with subject-specific weights minimising peaks error in TF contact forces (calibrated with validation data) Joint contact model: Two-point rigid contact (Steele et al., 2012)	Second edition data of the Grand Challenge Competition to Predict In Vivo Knee Loads (Fregly et al., 2012)	Joint model: - Knee alignment - Contact points (1) Baseline (2) Knee alignment and contact points (3) Knee alignment only (4) Contact points only	Conclusion: Improvement of knee alignment and contact points location increased model accuracy Quantitative evidence: RMSE of the total contact force was (1) 0.51 BW, (2) 0.33 BW, (3) 0.37 BW, (4) 0.45 BW (peak errors also reported for each alteration) Statistics: 95% confidence intervals reported to determine if statistically significant differences exists at first and second TF contact force peaks
Lin et al. (2010)	TF	Software: Matlab Model: Own model of the knee and surrounding structures derived from medical imaging DoFs: 6 TF, 6 PF Muscular lines of action: 11 Muscle path: Via points Optimisation: Inverse dynamics-based optimisation Objective function: Sum of squared muscle activations Joint contact model: Surrogate contact modelling (Lin et al., 2009)	Gait data collected on an adult male with instrumented knee implant Zhao et al. (2007)	Optimisation problem: - Design variables: (1) Muscle activations (2) Compressive contact forces - Optimisation constraints: (a) F/E moment residual (b) F/E moment and A/P force residuals (c) F/E moment and I/E moment residuals (d) Both (baseline: 1a)	Conclusion: The introduction of residual loads on joint forces and/or moments improved model accuracy. In most cases, the objective functions provided similar results for lateral contact, while (2) provide a better model accuracy when A/P force residual was used as constraint Quantitative evidence: RMSE were (1a) 0.17 and 0.51 BW, (1b) 0.32 and 0.37 BW, (1c) 0.23 and 0.35 BW, (1d) 0.26 and 0.35 BW, (2a) 0.20 and 0.45 BW, (2b) 0.12 and 0.40 BW, (2c) 0.24 and 0.39 BW, and (2d) 0.24 and 0.41 BW, respectively for medial and lateral contact forces Statistics: None
Lund et al. (2015)	TF	Software: Anybody Model: Variant of the Klein Horsman et al. (2007) model DoFs: 6 pelvis, 3 hip, 1 TF, 1 PF, 2 ankle Muscular lines of action: 159 per leg Muscle path: Via points and wrapping surfaces Optimisation: Inverse dynamics-based optimisation Objective function: Sum of cubed muscle activations Joint contact model: One-point rigid contact	Third edition data of the Grand Challenge Competition to Predict In Vivo Knee Loads (Fregly et al., 2012)	Joint model: - Markers' placement - Scaling strategy: (1) Linear scaling (baseline) (2) Anatomical scaling (3) Kinematic scaling	Conclusion: Kinematic scaling (3) provided the most accurate TF total contact force and was the method the less affected by marker placements errors Quantitative evidence: RMSE of the total contact force was (1) 0.36 BW, (2) 0.64 BW, and (3) 0.28 BW Statistics: Monte Carlo simulations to study the effect of variations in manual markers placements
Lundberg et al. (2012)	TF	Software: Matlab Model: Knee model, variant of the Delp et al. (1990) model DoFs: 6 TF Muscular lines of action: 15 (reduced to 3) Muscle path: Via points Optimisation: Not used. A model reduction is applied to cancel muscular redundancy, and equations governing dynamic equilibrium and relationship between medial and lateral contact forces are solved Objective function: Not used Joint contact model: Linear function of the varus-valgus moment	Gait data collected on four adults with instrumented knee implant Mundermann et al. (2008)	Optimisation problem: - Optimisation constraints: muscle activation boundaries (alterations performed under a sensitivity analysis)	Conclusion: The alterations introduced variability in TF contact forces estimations impacting model accuracy Quantitative evidence: RMSE averaged across trials of the total contact force ranged between 0 and 0.09 BW at the first force peak, -0.03 BW and -0.30 BW at the local minimum at mid-stance, and 0 BW and -0.10 BW at the second force peak, for patients without gait deviation (Subjects 1, 2 and 4). Values of an adjusted R ² are also given Statistics: None

Article	Joint	Generic musculoskeletal model	Validation dataset	Model alterations	Impact of the alterations on joint contact force accuracy
Lundberg et al. (2013)	TF	Software: Matlab Model: Knee model, variant of the Delp et al. (1990) model DoFs: 6 TF Muscular lines of action: 15 (reduced to 3) Muscle path: Via points Optimisation: Not used. A model reduction is applied to cancel muscular redundancy, and equations governing dynamic equilibrium and relationship between medial and lateral contact forces are solved Objective function: Not used Joint contact model: Linear function of the varus-valgus moment	Third edition data of the Grand Challenge Competition to Predict In Vivo Knee Loads (Fregly et al., 2012)	Joint model: - Number of DoFs - Medial-lateral contact force ratio - Passive stiffness	Conclusion: The alterations improved the model accuracy in estimating medial and lateral contact forces, while keeping a similar total contact force Quantitative evidence: At baseline, RMSE of medial, lateral, and total contact forces was 0.38 BW, 0.32 BW, and 0.37 BW. After alterations, it was 0.31 BW, 0.13 BW, and 0.38 BW Statistics: None
Manal and Buchanan (2013)	TF	Software: SIMM Model: Variant of the Delp et al. (1990) model DoFs: 6 pelvis, 3 hip, 3 TF, 3 ankle Muscular lines of action: 12 Muscle path: Via points Optimisation: EMG-driven Objective function: Sum of squared differences between estimated and inverse dynamic-based joint moments Joint contact model: Two-point rigid contact	Third edition data of the Grand Challenge Competition to Predict In Vivo Knee Loads (Fregly et al., 2012)	Joint model: - Contact points (tibial plateau width and points' location) Muscle model: - Muscle parameters (EMG-driven parameters)	Conclusion: An iterative process of parameters' variations allowed to improve the model accuracy Quantitative evidence: At baseline, RMSE of medial and lateral contact forces was 0.28 BW and 0.18 BW. After alterations, it was 0.16 BW and 0.22 BW. R ² and peak difference are also given Statistics: None
Modenese et al. (2011)	Hip	Software: Opensim Model: Variant of the Klein Horsman et al. (2007) model DoFs: 6 pelvis, 3 hip, 1 TF, 1 ankle Muscular lines of action: 163 per leg Muscle path: Via points and wrapping surfaces Optimisation: Inverse dynamics-based optimisation Objective function: Sum of muscle stresses at power p Joint contact model: One-point rigid contact	HIP98 (Data of subjects HSR, KWR, PFL, IBL) (Bergman et al., 2001)	Optimisation problem: Objective function: (1) Power p = 1 (2) Power p = 2 (3) Power p = 3 (4) Power p = 5 (5) Power p = 10 (6) Power p = 15	Conclusion: A quadratic objective function provided a better model accuracy than other powers in terms of hip contact force and musculo-tendon forces estimation Quantitative evidence: RMSE averaged range across subjects was (1) 0.25-0.47 BW, (2) 0.23-0.52 BW, (3) 0.33-0.66 BW, (4) 0.50-0.91 BW, (5) 1.77-2.85 BW, (6) 2.88-4.52 BW Statistics: None
Modenese et al. (2013)	Hip	Software: Opensim Model: Variant of the Klein Horsman et al. (2007) model DoFs: 6 femur, 1 TF, 1 ankle Muscular lines of action: 163 per leg Muscle path: Via points and wrapping surfaces Optimisation: Inverse dynamics-based optimisation Objective function: Sum of squared muscle stresses Joint contact model: One-point rigid contact	HIP98 (Data of subjects HSR, KWR, PFL, IBL) (Bergman et al., 2001)	Optimisation problem: - Design variables: (1) Without reserve actuators (baseline) (2) With reserve actuators	Conclusion: The introduction of reserve actuators improved model accuracy, showing a need for improving the muscular geometry (i.e. lever arms) Quantitative evidence: RMSE was (1) 0.59 BW, (2) 0.18 BW Statistics: None

Article	Joint	Generic musculoskeletal model	Validation dataset	Model alterations	Impact of the alterations on joint contact force
Moissenet et al. (2014)	TF	Software: Matlab Model: Variant of the Delp et al. (1990) model DoFs: 6 pelvis, 3 hip, 1 TF, 1 PF, 1 ankle Muscular lines of action: 43 per leg Muscle path: Via points Optimisation: Inverse dynamics-based optimisation Objective function: Sum of squared musculo-tendon forces Joint contact model: Two-point rigid contact (Feikes et al., 2003)	First, second, third and fourth edition data of the Grand Challenge Competition to Predict In Vivo Knee Loads (Fregly et al., 2012)	Optimisation problem: - Design variables: (1) Musculo-tendon forces (baseline) (2) Musculo-tendon forces and joint contact, ligament, and bone forces	Conclusion: The introduction of joint contact, ligament, and bone forces in the muscular redundancy problem improved model accuracy Quantitative evidence: RMSE ranges of medial and lateral contact forces across subjects were (1) 0.91-1.39 BW and 0.44-0.81 BW, (2) 0.31-0.50 BW and 0.22-0.43 BW Statistics: None
Moissenet et al. (2016)	TF	Software: Matlab Model: Variant of the Delp et al. (1990) model DoFs: 6 pelvis, 3 hip, 1 TF, 1 PF, 1 ankle Muscular lines of action: 43 per leg Muscle path: Via points Optimisation: Inverse dynamics-based optimisation Objective function: Sum of squared and weighted musculo-tendon forces and joint contact, ligament, and bone forces Joint contact model: Two-point rigid contact (Feikes et al., 2003)	Sixth edition data of the Grand Challenge Competition to Predict In Vivo Knee Loads (Fregly et al., 2012)	Muscle model: - Muscular redundancy: (1) 43 lines of action (baseline) (2) 163 lines of action (variant of the Klein Horsman et al. (2007) model)	Conclusion: An increase of muscular redundancy modified medial-lateral contact forces balance but did not impact the accuracy of the model to estimate total contact force Quantitative evidence: RMSE of medial, lateral, and total contact forces was (1) 0.65 BW, 0.45 BW, and 0.70 BW, (2) 0.26 BW, 0.63 BW, and 0.77 BW Statistics: None
Navacchia et al. (2016)	TF	Software: Opensim Model: Variant of the Delp et al. (1990) model DoFs: 6 pelvis, 3 hip, 2 TF, 1 ankle Muscular lines of action: 92 per leg Muscle path: Via points Optimisation: Inverse dynamics-based optimisation Objective function: Sum of squared muscle activations Joint contact model: One-point rigid contact model (Steele et al., 2012)	Data of three undefined subjects of the Grand Challenge Competition to Predict In Vivo Knee Loads (Fregly et al., 2012)	Joint model: - Markers' placement - Knee alignment - Inertial parameters Muscle model: - Muscle geometry - Muscle parameters (alterations performed under a sensitivity analysis)	Conclusion: Each alteration impacted the model accuracy Quantitative evidence: At baseline, RMSE of the total contact force ranged between 0.30 BW and 0.35 BW. After alteration, no quantitative measurement of model accuracy is reported Statistics: A Monte-Carlo simulation was conducted. The validation data were within the predicted 5-95% confidence bounds for 77%, 83%, and 75% of stance phase, respectively.
Serrancoli et al. (2016)	TF	Software: Opensim Model: Variant of the Arnold et al. (2010) model DoFs: 6 pelvis, 3 hip, 6 TF, 6 PF, 2 ankle Muscular lines of action: 44 per leg Muscle path: Via points and wrapping surfaces Optimisation: Inverse dynamics-based optimisation Objective function: Sum of squared muscle activations plus sum of six squared reserve activations Joint contact model: Validated regression equation converting superior-inferior force and varus-valgus moment to medial and lateral contact forces (Fregly et al., 2012)	Fourth edition data of the Grand Challenge Competition to Predict In Vivo Knee Loads (Fregly et al., 2012)	Muscle model: - Muscle geometry - Muscle parameters Variations are managed though a two-level optimisation: (1) Without tracking validation data (baseline) (2) By tracking validation data	Conclusion: Tracking validation data improved model accuracy Quantitative evidence: RMSE of the medial, lateral and total contact forces averaged across trials was (1) 0.51 BW, 0.55 BW, and 0.88 BW, (2) 0.08 BW, 0.09 BW, and 0.15 BW Statistics: Two-tailed t-test were used to evaluate statistical significant differences in results

Article	Joint	Generic musculoskeletal model	Validation dataset	Model alterations	Impact of the alterations on joint contact force accuracy
Steele et al. (2012)	TF	Software: Opensim Model: Variant of the Delp et al. (1990) model DoFs: 6 pelvis, 3 hip, 1 TF, 1 ankle Muscular lines of action: 92 per leg Muscle path: Via points Optimisation: Inverse dynamics-based optimisation Objective function: Sum of weighted squared muscle activations Joint contact model: One-point rigid contact model	First edition data of the Grand Challenge Competition to Predict In Vivo Knee Loads (Fregly et al., 2012)	Optimisation problem: - Optimisation weights	Conclusion: Using an iterative process, the authors were able to improve model accuracy by finding the best set of optimisation weights Quantitative evidence: At baseline, no quantitative measurement of model accuracy is reported. After alteration, RMSE of total contact force was 0.28 BW Statistics: None
Thelen et al. (2014)	TF	Software: Opensim Model: Variant of the Arnold et al. (2010) model DoFs: 6 pelvis, 3 hip, 6 TF, 6 PF, 2 ankle Muscular lines of action: 44 per leg Muscle path: Via points and wrapping surfaces Optimisation: Computed muscle control Objective function: Track joint angles computed during inverse dynamics with PID (forward dynamics level) and sum of weighted squared muscle activations (inverse dynamics level) Joint contact model: One-point rigid contact model	Fourth edition data of the Grand Challenge Competition to Predict In Vivo Knee Loads (Fregly et al., 2012)	Joint model: - Knee alignment	Conclusion: Adjustment of knee alignment modified medial-lateral contact forces balance but did not impact the accuracy of the model to estimate total contact force Quantitative evidence: At baseline, RMSE of medial, lateral, and total contact forces was 0.26 BW, 0.42 BW, and 0.51 BW. After alteration, no quantitative measurement of model accuracy is reported Statistics: None
Walter et al. (2014)	TF	Software: Opensim Model: Variant of the Arnold et al. (2010) model DoFs: 6 pelvis, 3 hip, 6 TF, 6 PF, 2 ankle Muscular lines of action: 44 per leg Muscle path: Via points and wrapping surfaces Optimisation: Mixed optimisation combining inverse skeletal dynamics and forward muscle activation and contraction dynamics Objective function: Sum of squared differences between estimated and inverse dynamic-based joint moments Joint contact model: Validated regression equation converting superior-inferior force and varus-valgus moment to medial and lateral contact forces (Fregly et al., 2012)	Third edition data of the Grand Challenge Competition to Predict In Vivo Knee Loads (Fregly et al., 2012)	Optimisation problem: - Use of synergies: (1) Independent controls without EMG tracking (baseline) (2) Independent controls with EMG tracking (3) Synergy controls without EMG tracking (4) Synergy controls with EMG tracking	Conclusion: EMG tracking decreased model accuracy when using synergy controls, but increased model accuracy when using independent controls (EMG tracking is then critical). However, synergy controls always produced more accurate contact forces Quantitative evidence: RMSE of medial and lateral contact forces averaged across was (1) 0.29 BW, (2) 0.20 BW, (3) 0.15 BW, and (4) 0.19 BW. Statistics: None
Zhang et al. (2015)	Hip	Software: Anybody Model: Variant of the Klein Horsman et al. (2007) model DoFs: 6 pelvis, 6 hip, 1 TF, 1 ankle, 1 subtalar Muscular lines of action: 163 per leg Muscle path: Via points and wrapping surfaces Optimisation: Inverse dynamics-based optimisation Objective function: Sum of squared musculo-tendon forces Joint contact model: Force dependent kinematic (Andersen & Rasmusen 2011)	HIP98 (Data of subjects HSR, KWR, IBL) (Bergman et al., 2001)	Optimisation problem: (1) Objective function Joint model: (2) Contact stiffness Muscle model: (3) Muscle geometry (alterations performed under a sensitivity analysis)	Conclusion: Only alteration of the objective function had a marked impact on accuracy. Quantitative evidence: At baseline, RMSE averaged across subjects was 0.45 BW. After alteration, the maximal RMSE with baseline was (1) 0.18 BW, (2) 0.03 BW, and (3) 0.05 BW Statistics: None

Figure 1

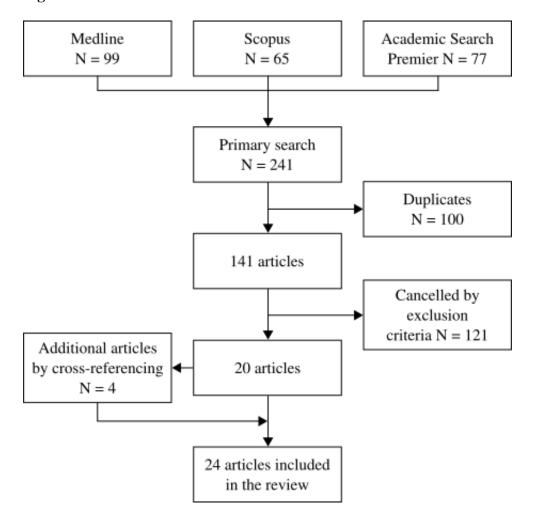


Figure 2

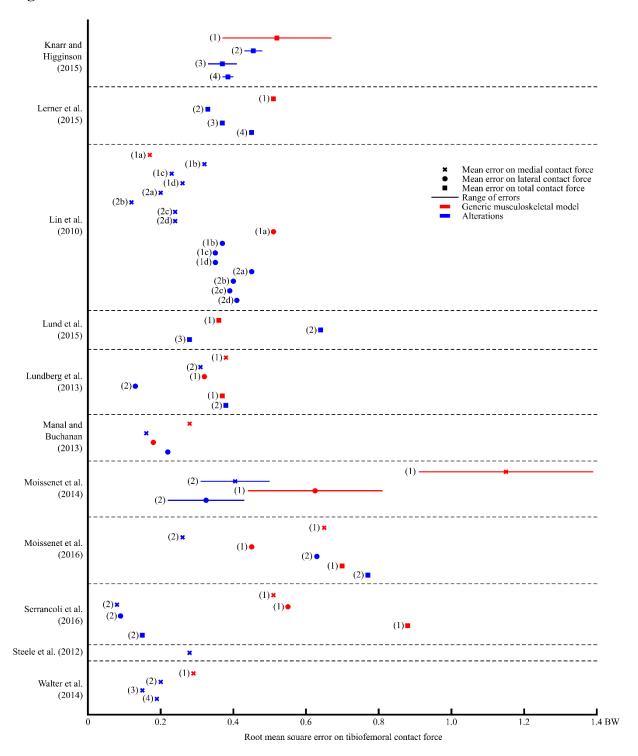


Figure 3

