






# Musculoskeletal Model of Human Lower Limbs in Gait Simulation

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**Abstract.** One of the most important issues in the context of contemporary research on a human body biomechanics is musculoskeletal models. It turns out that the need for biomechanically correct movement model is enormous not only in medicine and rehabilitation, but also in animations synthesis with computer game engines, where unsightly effects often occur between animation keys. Proposing a musculoskeletal model that could be efficiently adapted to a game development environment would speed up work and solve the difficulties in controlling movement sequence synthesis. The proposed model, using dynamic inversion simulation, includes Usik's equations which have been added to the 6DoF skeletal module. Usik uses the thermomechanics of a continuous medium, which takes into account the cross-effects of mechanical, electrical, chemical and thermodynamic phenomena in muscle tissue which makes it possible to better (closer to the nature) describe the reaction of muscles during movement. The validation value of the model correctness is the GRF (ground reaction force) where the simulated values are compared to the measured one. This work concerns human gait and is the basis for further development in the context of the identified problem in the field of computer game engines. Due to the more accurate GRF results against existing solutions, the proposed one is optimistic for further work on it.

**Keywords:** Musculoskeletal model · Biomechanic · Gait simulation

## 1 Introduction

Simulation of human gait is a complex phenomenon. It is investigated on many levels e.g.: character animations, medical diagnosis or rehabilitation. The most basic form of human locomotion is walking and this movement is analyzed in this article. Healthy gait occurs when the right and left parts of the human body perform similar movements in relation to the anatomical planes of the body. The gait can be analyzed by deriving flat dynamic models describing the movements occurring in the sagittal and frontal plane of the body [1, 4, 6].

The authors consider that simulation due to its use in blending computer animation. Modern game engines offer a number of functionalities for that operation. They hold the key animation frames of skeletal poses. There can still be

unsightly transitions between those keys. These types of problems are so far visible e.g. in sim racing (animation of the driver’s hands) or in the movement of humanoid bots. It is necessary to resort to musculoskeletal modeling to solve it. The authors decided on the already existing 6DOF model because more degrees of freedom are not needed for the previously described situations. The method of placing the muscles is shown on the left side in the Fig. 1. The calculations for controlling movement are contained in the muscle module that the authors have implemented by using the Usik’s mathematical model. It is a novelty because Usik’s model has not been used before for this type solution.

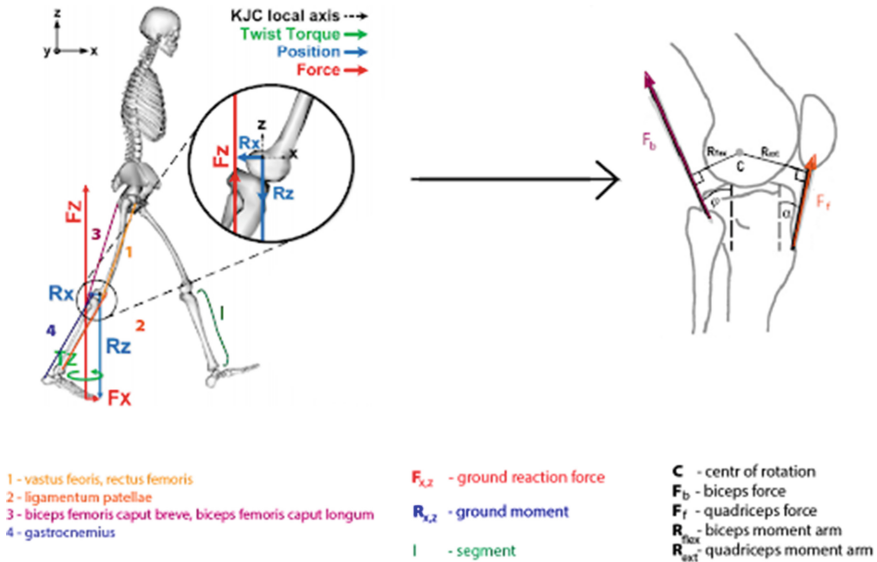


Fig. 1. (From the left) Forces in a lower limb and forces in a joint.

Dynamic inversion has been chosen for that solution. The input data were downloaded from the available database [2]. The output data was compared with the measurement one (Rigid Body Dynamic Library [2]) by comparing the ground reaction forces (vertical and horizontal one) (Fig. 1). This method of validation was chosen to compare the authors’ results to other solutions from recent years such as Belaise et al. [5], Sartori M. et al. [8], Yamasaki et al. [7], Wojnicz W. [6], Moissenet’s et al. [2]. The authors concluded that adding muscles using the Usik’s model can improve simulation results, especially for the initial and final phase of gait. Thanks to that it gives an optimistic, potential solution to the problem of biomechanically correct blending for human movement animation.

## 2 Methodology

As part of this work the following were carried out:

- implementation of the 6DoF model in C++ with Vulkan API,
- creation a parser for the .c3d format, in which the input data was stored,
- creation of two gastrocnemius muscles based on the author’s model,
- set the relationship between the 6DoF model and the muscle model,
- saving parameters on charts during a gait simulation.

### 2.1 Skeletal Model

The selected biomechanical model is defined by two 6DoF models: open for one-legged support and closed for two-legged support. The complete systems of the equation are contained in publication [6]. The first of them was presented in this work in Eq. 1 and 2 to indicate the occurrence of the influence of mass (as a parameter) on a given segment.

$$\begin{aligned}
 & A_{11} \cdot \ddot{\alpha}_1 + A_{12}(\alpha_1, \alpha_2) \cdot \ddot{\alpha}_2 + A_{13}(\alpha_1, \alpha_3) \cdot \ddot{\alpha}_3 \\
 & \quad + A_{14}(\alpha_1, \alpha_4) \cdot \ddot{\alpha}_4 + A_{15}(\alpha_1, \alpha_5) \cdot \ddot{\alpha}_5 \\
 & \quad + A_{16}(\alpha_1, \alpha_6) \cdot \ddot{\alpha}_6 + A_{17}(\alpha_1, \alpha_7) \cdot \ddot{\alpha}_7 \\
 & = M_1 - L_1 \cdot \sin(\alpha_1) \cdot F_y + L_1 \cdot \cos(\alpha_1) \cdot F_z
 \end{aligned} \tag{1}$$

where:

$$\begin{aligned}
 M_1 = & M_{ex1} - M_{12} - \cos(\alpha_1) \cdot (S_1 \cdot m_1 \cdot g + L_1 \cdot g \cdot \sum_{i=2}^7 m_i) \\
 & + L_1 \cdot \sin(\alpha_j - \alpha_1) \cdot (\dot{\alpha}_j)^2 \cdot (S_j \cdot m_j + L_j \cdot \sum_{k=j+1}^7 m_k) \\
 & \quad + L_1 \cdot \sin(\alpha_6 - \alpha_1) \cdot (\dot{\alpha}_6)^2 \cdot S_6 \cdot m_6 \\
 & \quad + L_1 \cdot \sin(\alpha_7 - \alpha_1) \cdot (\dot{\alpha}_7)^2 \cdot S_7 \cdot m_7 \\
 & \quad \text{for } j = 2, 3, 4, 5 \\
 & \quad \dots
 \end{aligned} \tag{2}$$

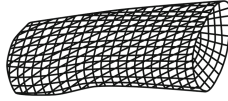
$A_i$  -  $i$ -th segment,  $L_i$  - length of the segment,  $S_i$  - segment radius,  $M_i$  - moment derived from the ground reaction components,  $\alpha$  - rotational displacement of the segment,  $\dot{\alpha}_i$  - segment angular velocity,  $\ddot{\alpha}_j$  - segment angular acceleration,  $m$  - mass of the segment.

### 2.2 Muscle Modeling

The process of the muscle modeling had a three steps: generate a muscle fiber representation using a B-spline solid, create a model with virtual muscle fibers, apply Usik’s model as a behavior model [3].

### 2.3 3D Fiber Representation

The muscle fiber was chosen as the basic element and was implemented virtually as the volumetric B-spline solid (Fig. 2). A detailed description of the fiber formation method using by the authors was described in publication [3].



**Fig. 2.** A tessellated B-spline solid for a fiber muscle representation [3].

### 2.4 Muscle Model

For this solution, the gastrocnemius muscle was selected. It was created using the finite volume method with B-spline curves and Usik's equations. This volume was a fiber representation with two attachment points. Control point was represented by the Eq. 3:

$$V(u, v, w) = \sum_{i=0}^l \sum_{j=0}^m \sum_{k=0}^n B_i^u(u) B_j^v(v) B_k^w(w) C_{ijk} \quad (3)$$

where each  $C_{ijk} \in R^3$ .

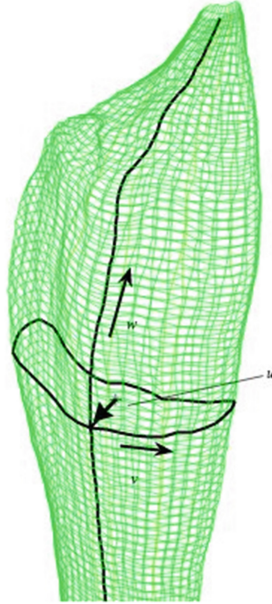
$C = C_{ijk}$  - the set of points form a control point lattice which will influence the shape of the B-spline solid,  $V$  - a parametric solid given by the tritenor product of these B-spline basis functions (in this case, the polynomials  $B_i^u(u) B_j^v(v) B_k^w(w)$ ) with the control points in  $C$  (Fig. 3).

### 2.5 Combination of Muscle Module with Skeletal Biomechanical Module

A module with two gastrocnemius muscles was added to the biomechanical skeletal model. During gait simulation, the forces were applying on each joint. They were as the base parameter of calf control points located at the muscle joints.

The use of the molecular weight of the  $k$ -th component is important in the proposed solution. This parameter is included in one of the equations from the Usik's model, specifically in the mass balance equation of the  $k$ -th component (Eqs. 4).

$$\begin{aligned} \rho \delta y_k^1 \div \delta t &= Q_k^{out} - Q_k^{in} \\ \rho \delta y_k^2 \div \delta t &= Q_k^{in} + \sum_{k=1}^n M_k V_{kj} I_j \end{aligned} \quad (4)$$



**Fig. 3.** A fragment of a gastrocnemius muscle with the  $u, v, w$  coordinates for any control point. Those values are the Cartesian coordinates that compose the boundary and volume of the solid

where  $k$ -  $k$ -th element,  $\rho$  - mass density,  $y$ - ,  $t$ - time ,  $Q^{out}$ - source density in exchange with the environment ,  $Q^{in}$ - source density in inter-phase exchange,  $M$ - molecular weight,  $v$ - speed ,  $I$ - chemical reaction rate.

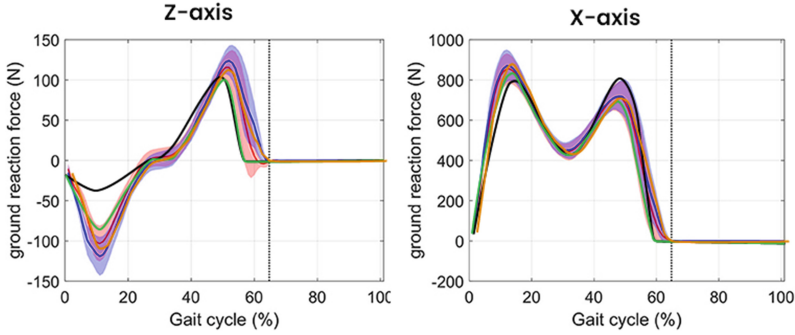
$K$ -th components are control points in the finite volumes of muscle fibers (Fig. 2). The sum of the molecules masses  $M_k$  from Eqs. 4 was treated as the mass  $m_i$  of the each knee joint segment in Eqs. 2 and 3 [9].

**2.6 Results**

Below is a summary of the author’s charts with the calculated data obtained after adding the author’s muscle. Table 1 and 2 contain values of the area under the curve of a given force during one walking cycle. The cycle has been divided into 100 compartments (Fig. 4).

Interpolation of 4 and 5 degree polynomial was used. The charts provide an illustrative course of validation parameters.

Table 1 and 2 compare the values of the fields below the chart. This method was used due to the fact that the graphs from the simulations intersected with the graph from the measurements only in 3 places, which gave ease in applying the method of counting the areas under the graph as the final validation method.



**Fig. 4.** The upper graph shows the ground force response values for the Z-axis and the lower graph for the X-axis. Black curve - measurement data from [6], blue curve - data from [6], red curve - data from [2], orange curve - authors' data. Red and blue areas are the possible limit of error. (Color figure online)

**Table 1.** Validation parameters

Reference model	Z-axis - area under the graph		
	Simulation	Measuring	Difference
[6]	48.56	42.24	6.32
[2]	46.15	42.24	3.91
Authors	40.12	42.24	2.12

**Table 2.** Validation parameters

Reference model	X-axis - area under the graph		
	Simulation	Measuring	Difference
[6]	56.50	51.43	5.07
[2]	54.70	51.43	3.27
Authors	49.88	51.43	1.55

### 3 Conclusions

It can be seen in Table 1 and 2 that the proprietary solution increased the correctness of the final results for the values of ground reaction forces. To further improve the results (bring the values from the simulation closer to the measured values), it is worth breaking the muscles into smaller sections, which would allow for a more accurate representation of the behavior of soft tissues.

Due to the better results, the next planned step is to create plug-in based on that solution for UE5 game engine as an open-source tool which would solve the problem described in the Introduction. If it turned out to be helpful, it would be worth extending this solution with other parts of the human body.

## References

1. Ezati, M., Ghannadi, B., McPhee, J.: A review of simulation methods for human movement dynamics with emphasis on gait. *Multibody Syst. Dyn.* **47**(3), 265–292 (2019). <https://doi.org/10.1007/s11044-019-09685-1>
2. Moissenet, F., Bélaïse, C., Piche, E., Michaud, B., Begon, M.: Centre National de Rééducation Fonctionnelle et de Réadaptation-Rehazenter. Luxembourg, Luxembourg (2019)
3. Pietka, E., Badura, P., Kawa, J., Wieclawek, W. (eds.): ITIB 2018. AISC, vol. 762. Springer, Cham (2019). <https://doi.org/10.1007/978-3-319-91211-0>
4. Bednarski, R., Bielak, A.: Use of motion capture in assisted of knee ligament injury diagnosis. *J. Appl. Comput. Sci.* **26**(1), 7–32 (2018)
5. Bélaïse, C., Dal Maso, F., Michaud, B., Mombaur, K., Begon, M.: An EMG-marker tracking optimisation method for estimating muscle forces. *Multibody Syst. Dyn.* **42**(2), 119–143 (2017). <https://doi.org/10.1007/s11044-017-9587-2>
6. Wojnicz, W.: Biomechanical models of the human musculoskeletal system, Gdansk University of Technology (2018)
7. Yamasaki, T., Idehara, K., Xin, X.: Estimation of muscle activity using higher-order derivatives, static optimization, and forward-inverse dynamics. *J. Biomech.* **49**, 2015–2022 (2016). <https://doi.org/10.1016/j.jbiomech.2016.04.024>
8. Sartori, M., Farina, D., Lloyd, D.G.: Hybrid neuromusculoskeletal modeling to best track joint moments using a balance between muscle excitations derived from electromyograms and optimization. *J. Biomech.* **47**, 3613–3621 (2014). <https://doi.org/10.1016/j.jbiomech.2014.10.009>
9. Usik, P.I.: Continual mechanochemical model of muscular tissue. *PMM* **37**(3), 448–458 (1973)