

Are virtual muscles are reflex controllers capable of describing variations within the human gait cycle? Sandra K. Hnat and Antonie J. van den Bogert Parker Hannifin Human Motion and Control Lab (hmc.csuohio.edu) Department of Mechanical Engineering, Cleveland State University



INTRODUCTION

- Various strategies for controlling modern prosthetic devices have been proposed, though able-bodied locomotion is usually *not* achieved
- Virtual muscles and autonomous reflex control produces realistic human locomotion in simulations and hardware [1]
- Reflex models are not usually tuned or validated using human walking data obtained through experiments



RESULTS

Results from one test subject (male, age = 21, mass = 64 kg) are shown here, where the control parameters were optimized using 5 gait cycles and the 1.2 m/s walking speed.

The model reproduces the joint torque observed in the experiment (Fig. 1):

• Produces variations in peak

Walking Speed = 0.8 m/s					
	L	L	L	L	Experiment

Objectives:

- Use optimization to tune the parameters of the Virtual Muscle Reflex (VMR) system to produce realistic joint torques using human experimental data
- Evaluate the performance of reflex models in describing the variations 2 within and between gait cycles under the effect of mechanical perturbations

METHODS

Ten subjects walked on an instrumented treadmill (0.8, 1.2, and 1.6 m/s) while being longitudinally perturbed by Gaussian white noise [2]. Joint angles and joint torques were obtained through traditional inverse dynamics [3].

Mathematical Muscle Model:

• Three Hill-type muscles: Gastrocnemius, Soleus, and Tibialis Anterior



- moment between gait cycles, correlating to the torque exhibited by the subject
- Matches the amplitude and timing of ankle push-off
- Mimics the shape of the experimental torque during the swing phase



Fig 1: Experimental joint torque (black) and VMR joint torque using optimized reflex control parameters (red). Negative torque corresponds to plantarflexion.

The model produces realistic muscles forces (Fig. 2) that agree with those reported in literature:

• Gastrocnemius and Soleus produce the large peak required for ankle push-off



• Activation dynamics





 $F_{SEE} = \left(aF_{max} \cdot f_{FL}(L_{ce}) \cdot f_{FV}\left(\dot{L_{ce}}\right) + F_{PEE}\right)cos\phi + F_{D}$

- Implicit Formulation solved by fixed-step Rosenbrock solver [4] in MATLAB
- Torque (τ) is obtained by multiplying muscle force by the moment arm

Reflex Control:

Muscle control signals (u) were generated by the reflex controller of Geyer et al. [1].

- Stance Phase
 - $u_{SOL} = S_{0,SOL} + G_{SOL}F_{SOL}(t t_{SOL})$

 $u_{TA} = S_{0,TA} + G_{TA} [(l_{CE,TA}(t_d) - l_{OFF,TA})] G_{SOLTA} F_{SOL}(t - t_{TA})$

- Soleus is activated before the Gastrocnemius
- Tibialis Anterior is minimally activated

Fig 1: Estimated muscle forces of the Gastroc. (blue), Soleus (green), and Tibialis Anterior (red), at 1.2 m/s walking speed

CONCLUSION

Completed Objectives:

- The VMR produces realistic joint torques produced by test subjects
- Initial findings suggest that the controller is also capable of replicating human response to perturbations and may describe important aspects of the human control system

Future Work:

Compare results across multiple subjects

 $u_{GAS} = S_{0,GAS} + G_{GAS}F_{GAS}(t - t_{GAS})$

• Swing Phase

 $u_{GAS} = S_{0,GAS} + G_{GAS}F_{GAS}(t - t_{GAS})$

 $u_{TA} = S_{0,TA} + G_{TA} [(l_{CE,TA} (t - t_{TA}) - l_{OFF,TA})]$

 $u_{GAS} = S_{0,GAS}$

Particle Swarm Optimization [5]:

- Optimize 8 control parameters, series/parallel element slack-lengths, and muscle reflex delays
- Minimize the multi-objective cost function:

$$C = W_1 \left(\sqrt{\frac{1}{N} \sum \left(\tau_{VMR} - \tau_{exp} \right)^2} \right) + W_2 \left(\sqrt{\frac{1}{N} \sum a^2} \right)$$

- Allow longer computation time for optimizations (greater accuracy)
- Expand model to include the knee and hip
- Implement and test in hardware (INDEGO exoskeleton)



1. H. Geyer et al., IEEE Trans. Neural Syst. Rehab. Eng., 2010 2. J. K. Moore et al., *PeerJ*, 2015 3. D.A. Winter, Biomechanics of Human Movement, 1979 4. A. J. van den Bogert et al., *Procedia IUTAM*, 2011 5. D. Simon, Evolutionary Optimization Algorithms, 2013

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