University of Cincinnati
National Institute for Occupational Safety
and Health (NIOSH)-supported
Education and Research Center (ERC)

12th Annual 2011 Pilot Research Project (PRP) Symposium

Thursday, Oct. 13, 1 pm-5:15 pm Friday, Oct. 14, 8 am-12:15 pm Kehoe Auditorium, Kettering Laboratory Supported by NIOSH Grant #T42-OHoo8432

Effect of Aging on Human Postural Control: A Predictive Modeling Approach

Carson Willey

Amit Shukla

Renu Sah

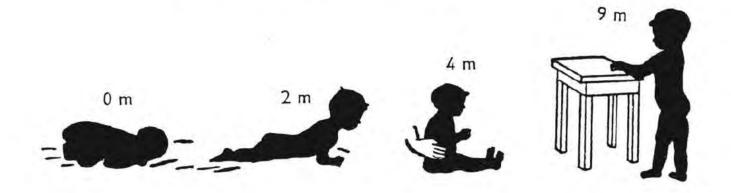


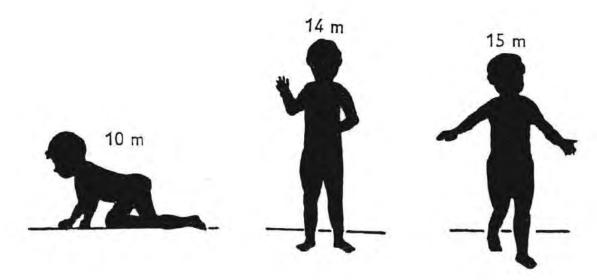
Outline

- Balance and Aging
- Modeling Balance: Dynamics and Control
- L-2 Norm as a balance metric
- Parametric Studies
- Results and Discussion
- Future Work



Balance and Posture

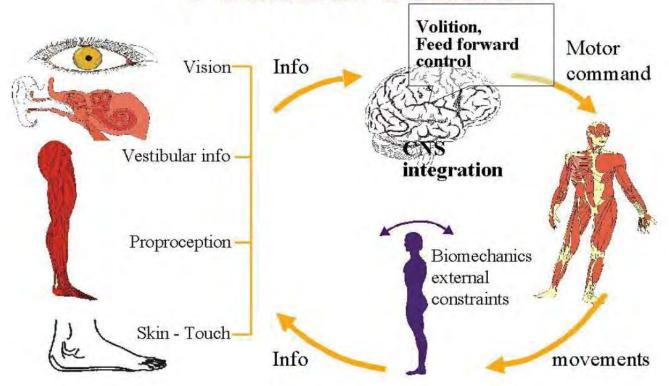






Human Postural Control: Why?

Postural control

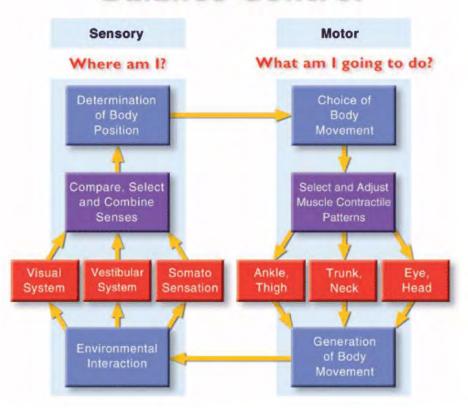


Courtesy: Swedish Medical Research Council



Human Postural Control

Balance Control



Jacobson GP, Newman CW, Kartush JM (1993). *Handbook of Balance Function Testing*. Mosby Year Book, St Louis.



Human Postural Control

Imbalance could be due to:

- cardiovascular dysfunction,
- visual disorientation,
- inadequate limb proprioception,
- vestibular signals or dysfunction.

Complex Nature of Balance Problems



Human Postural Control: Medical Causes

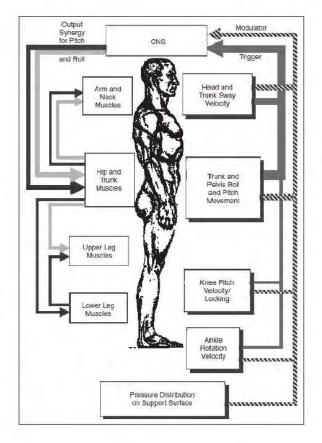
- Aging and Balance Dysfunction
 - Parkinson's Disease
 - Accumulated Injuries
 - Ear infections
 - Diabetes
 - Others

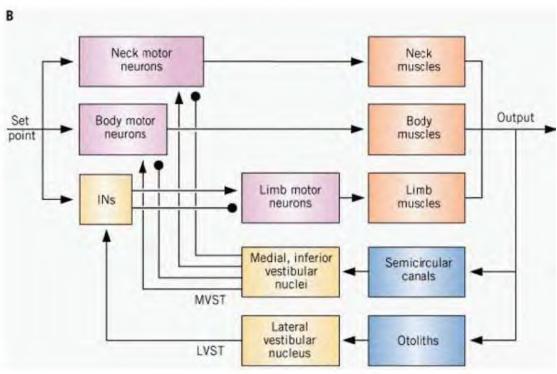
Whether balance disorders result from combinations of subtle problems or obvious disease, clinical studies indicate that elderly fallers are different from their healthy age-matched counterparts and require medical treatment to maintain their functional independence and quality of life.





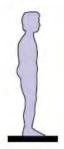
Multiple sensor-multiple inputs



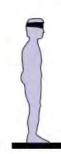




Dynamics posturography with sensory condition number (k)



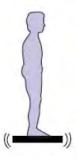
 $\kappa = 1$: Normal



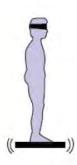
 $\kappa = 2$: Eyes closed



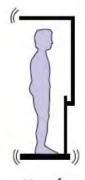
 κ = 3: Vision sway-ref



κ = 4: Platform sway-ref



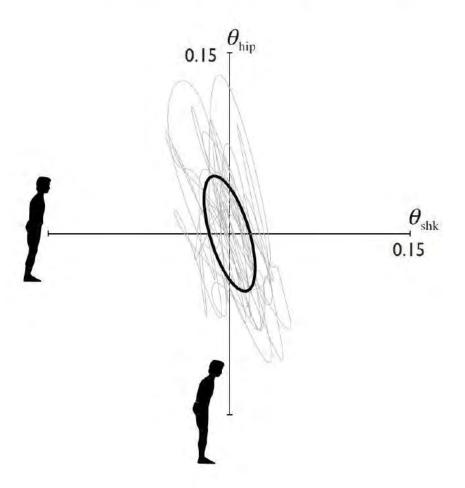
κ = 5: Eyes closed Platform sway-ref



 κ = 6: Vision sway-ref Platform sway-ref



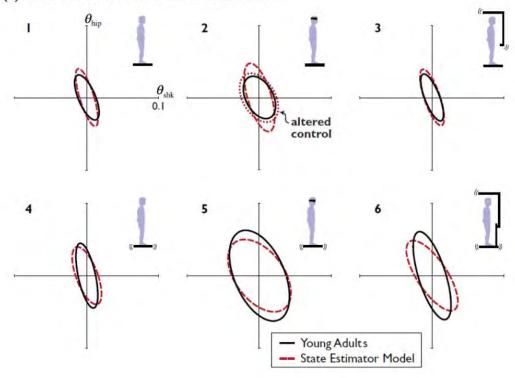
Covariance data: shank angle vs. hip angle





Covariance data:shank angle vs. hip angle

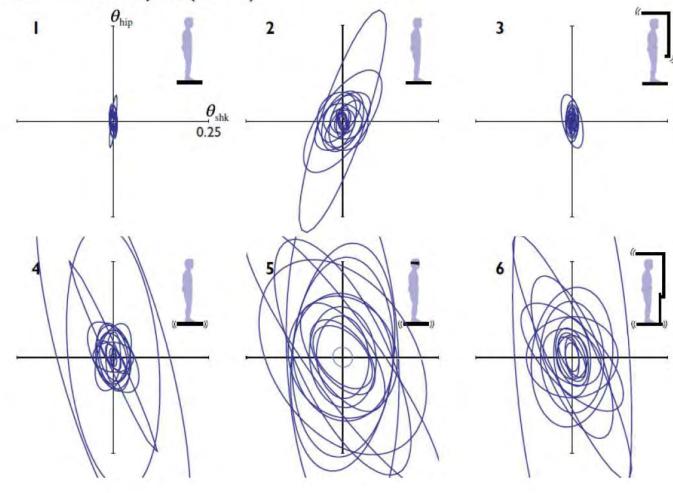
(a) State Estimator Model of Normal Posture



Kuo A D 1995 An optimal control model for analyzing human postural balance *IEEE Trans. Biomed. Eng.* **42** 87–101



Older Adult Subjects (N = 15)





The One Degree of Freedom Model

Parameter	Value	Units
m	80	kg
g	9.81	m/s^2
h	1	m
K	.25 mgh	N/rad
В	4	N/rad*s
P	.8 mgh	N/rad
D	14	N/rad*s
a	4	rad/s/rad
r	.004	

$$I\ddot{\theta} = mgh\sin(\theta) - T$$

$$T = K\theta + B\dot{\theta} + P\theta_{\Delta} + D\dot{\theta}_{\Delta}$$

$$\theta_{\Delta} = \theta(t - \tau)$$

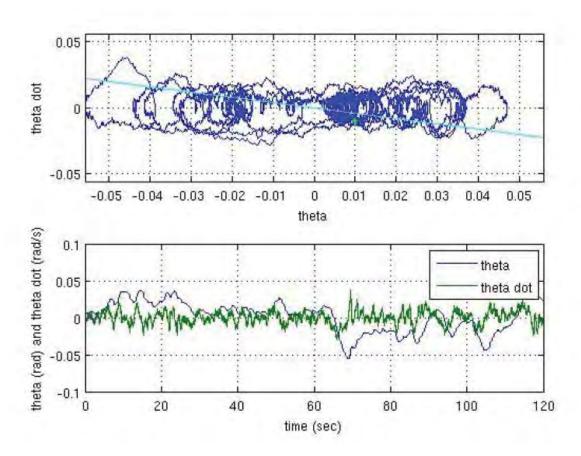
$$\dot{\theta}_{\Delta} = \dot{\theta}(t - \tau)$$

$$\sqrt{\dot{\theta}_{\Delta}^{2} + \theta_{\Delta}^{2}} < r \text{ and } \theta_{\Delta}(\dot{\theta}_{\Delta} - a\theta_{\Delta}) > 0$$

[1] Y. Asai, Y. Tasaka, K. Nomura, T. Nomura, and M. Casadio, "A model of postural control in quiet standing: Robust compensation of delay-induced instability using intermittent activation of feedback control," PLoS ONE, vol. 4, no. 7, 2009.



One DOF Model Output





The Two Degree of Freedom Model

Pendulum Dynamics
$$\frac{\ddot{\theta_1}(2\alpha - (\delta^2)/(2\beta)\cos(\theta_1 - \theta_2)^2) = -(\delta^2)/(2\beta)\dot{\theta}_1^{\,2}\sin(\theta_1 - \theta_2)\cos(\theta_1 - \theta_2)}{-(\delta\zeta)/(2\beta)\sin(\theta_2)\cos(\theta_1 - \theta_2) - \delta\dot{\theta}_2^{\,2}\sin(\theta_1 - \theta_2) + \gamma\sin(\theta_1) - T_1}$$

$$= \frac{\ddot{\theta}_2(2\beta - (\delta^2)/(2\alpha)\cos(\theta_1 - \theta_2)^2) = (\delta^2)/(2\alpha)\dot{\theta}_2^{\,2}\sin(\theta_1 - \theta_2)\cos(\theta_1 - \theta_2)}{-(\delta\gamma)/(2\alpha)\sin(\theta_1)\cos(\theta_1 - \theta_2) + \delta\dot{\theta}_1^{\,2}\sin(\theta_1 - \theta_2) + \zeta\sin(\theta_2) - T_2}$$

$$= \frac{(\delta\gamma)/(2\alpha)\sin(\theta_1)\cos(\theta_1 - \theta_2) + \delta\dot{\theta}_1^{\,2}\sin(\theta_1 - \theta_2) + \zeta\sin(\theta_2) - T_2}{\alpha = (m_1l_1^2)/8 + (m_2l_1^2)/2 + I_1/2}$$

$$= \frac{\beta = (m_2l_2^2)/8 + I_2/2}{\beta = (m_2l_2^2)/8 + I_2/2}$$

$$= \frac{(gm_1l_1)/2 + gm_2l_1}{\beta = (m_2l_2^2)/2}$$

$$= \frac{(gm_2l_2)/2}{\delta = (m_2l_2l_1)/2}$$

$$= \frac{I_1 = (m_1l_1^2)/12}{I_2 = (m_2l_2^2)/12}$$

$$= \frac{I_2 = (m_2l_2^2)/12}{I_2 = (m_2l_2^2)/12}$$

$$= \frac{I_1 = K_1\theta_1 + K_2(\theta_1 - \theta_2) + B_1\dot{\theta}_1 + B_2(\dot{\theta}_1 - \dot{\theta}_2) + P_1\theta_{1\Delta} + P_2(\theta_{1\Delta} - \theta_{2\Delta}) + D_1\dot{\theta}_{1\Delta}}{\beta = (m_2l_2^2)/12}$$

$$= \frac{I_1 = K_1\theta_1 + K_2(\theta_1 - \theta_2) + B_1\dot{\theta}_1 + B_2(\dot{\theta}_1 - \dot{\theta}_2) + P_1\theta_{1\Delta} + P_2(\theta_{1\Delta} - \theta_{2\Delta}) + D_1\dot{\theta}_{1\Delta}}{\beta = (m_2l_2^2)/12}$$

$$= \frac{I_1 = K_1\theta_1 + K_2(\theta_1 - \theta_2) + B_1\dot{\theta}_1 + B_2(\dot{\theta}_2 - \dot{\theta}_1) + P_2(\theta_{2\Delta} - \theta_{1\Delta}) + D_2(\dot{\theta}_{2\Delta} - \dot{\theta}_{1\Delta}) + \sigma\eta}{\beta = (m_2l_2^2)/2}$$

$$= \frac{I_1 = K_1\theta_1 + K_2(\theta_1 - \theta_2) + B_1\dot{\theta}_1 + B_2(\dot{\theta}_1 - \dot{\theta}_2) + P_1\theta_{1\Delta} + P_2(\theta_{1\Delta} - \theta_{2\Delta}) + D_1\dot{\theta}_{1\Delta}}{\beta = (m_2l_2^2)/12}$$

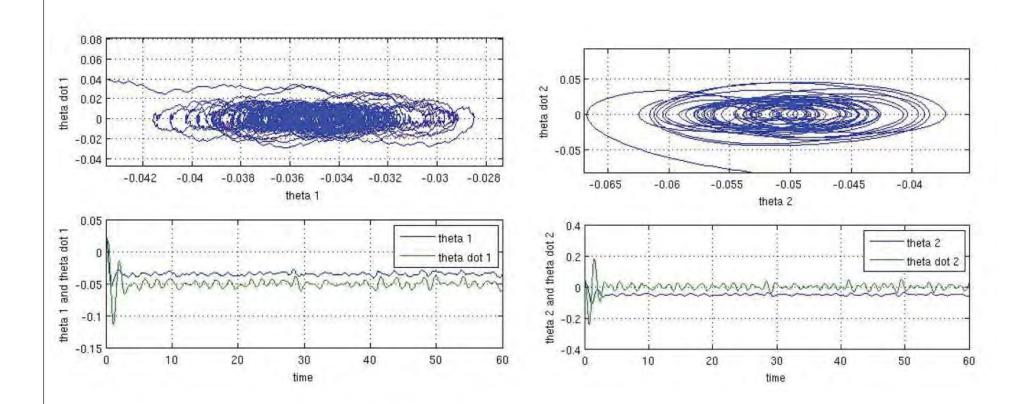
$$= \frac{I_1 = K_1\theta_1 + K_2(\theta_1 - \theta_2) + B_1\dot{\theta}_1 + B_2(\dot{\theta}_1 - \dot{\theta}_2) + P_1\theta_{1\Delta} + P_2(\theta_{1\Delta} - \theta_{2\Delta}) + D_1\dot{\theta}_{1\Delta}}{\beta = (m_2l_2^2)/12}$$

$$= \frac{I_1 = K_1\theta_1 + K_2(\theta_1 - \theta_2) + B_1\dot{\theta}_1 + B_2(\dot{\theta}_1 - \dot{\theta}_2) + P_1\theta_{1\Delta} + P_2(\theta_{1\Delta} - \theta_{2\Delta}) + D_1\dot{\theta}_{1\Delta}}{\beta = (m_2l_2^2)/12}$$

$$= \frac{I_1 = K_1\theta_1 + K_2(\theta_1 - \theta_2) + B_1\dot{\theta}_1 + B_2(\dot{\theta}_1 - \dot{\theta}_2) + P_1\theta_{1\Delta} + P_2(\theta_{1\Delta} - \theta_{1\Delta}) + D_2(\dot{\theta}_2 - \dot{\theta}_{1\Delta}) + D_2(\dot{\theta}_2 - \dot{$$



Two DOF Model Output

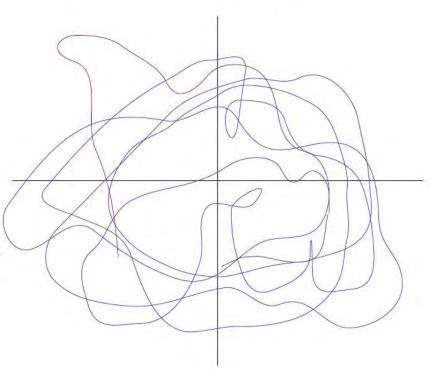




Average L2 Norm Analysis

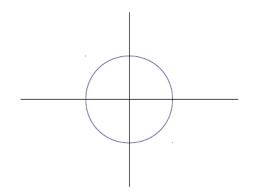
- . Ast → ∞
- Case1: Avg L₂ Norm < Ref L₂ Norm
- Case2: Avg L₂ Norm ≈ Ref L₂ Norm
- Case3: Avg L₂ Norm > Ref L₂ Norm
- Ref L2 is the average norm taken with respect to the delay history (red).
- Avg L2 Norm is the same measure taken with respect to the projected curve (blue).

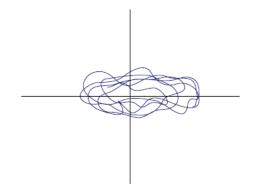
$$L_{2 \text{ avg}} = \frac{1}{N} \sum_{i=1}^{N} L_{i} \text{ where } L_{i} = \sqrt{x_{i}^{2} + y_{i}^{2}}$$



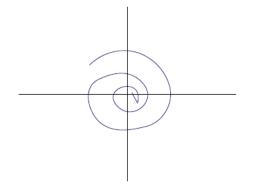


Average L2 Norm Analysis

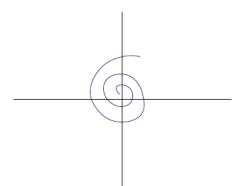




Case 2 : constant energy



Case 2 : realistic phase plot



Case 1 : Decreasing Energy

Case 3: Increasing Energy

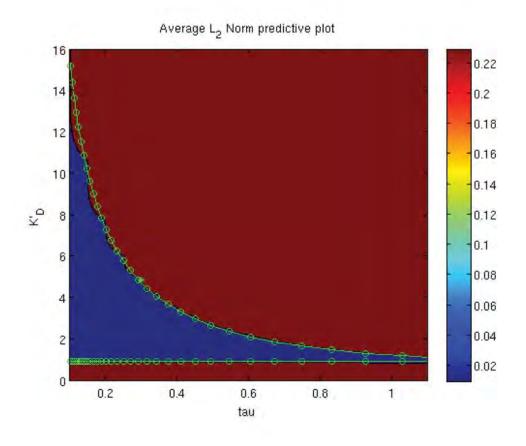


In this research

- Hypothesis 1: Aging can be modeled as change in parameters.
- Hypothesis 2: L2-Norm can be used as a balance metric.

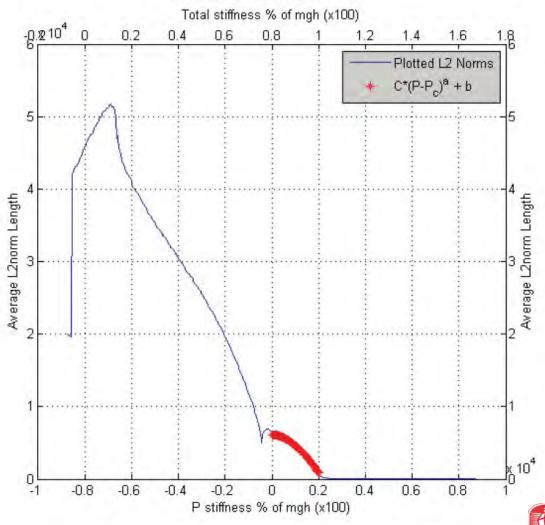


Linear Stability Analysis of 1 DOF

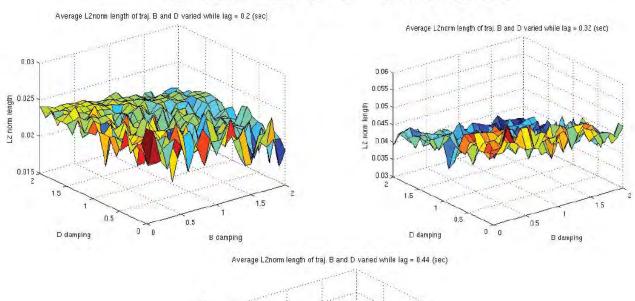


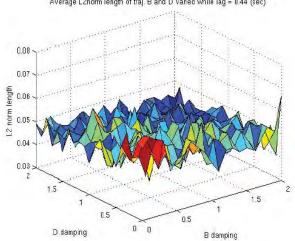


Parametric Study of One DOF Model



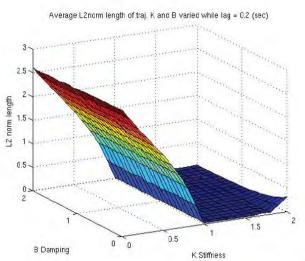
Parameter Combination Study of One DOF Model

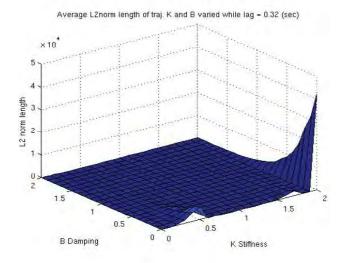


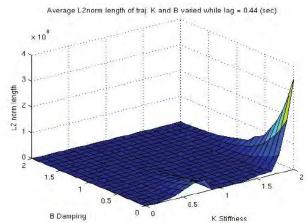




Parameter Combination Study of One DOF Model (cont'd)

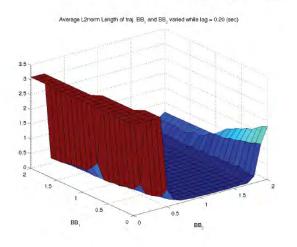


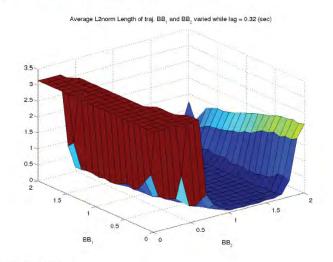


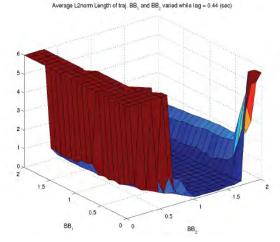




Parameter Combination Study of Two DOF Model

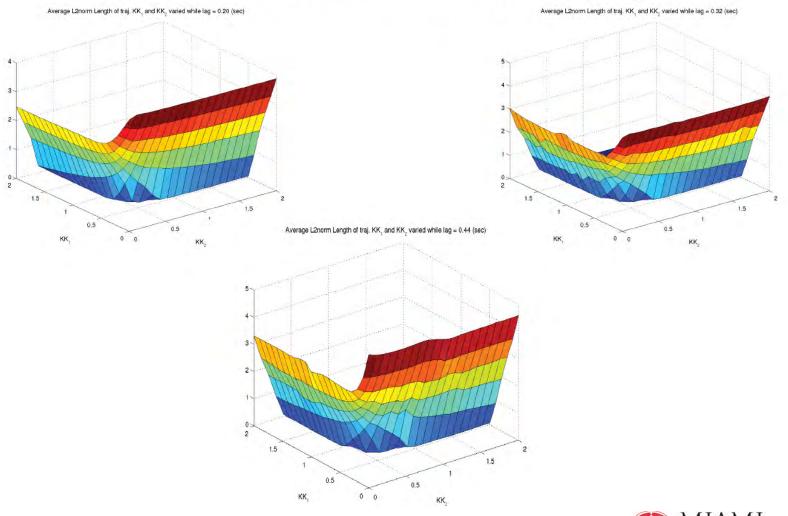








Parameter Combination Study of Two DOF Model (cont'd)



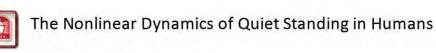


Discussion

- L2 –Norm is useable metric for balance.
- Parametric studies show definite trends and sensitivities.
- Model choice dictates predictive capability as the strategy of balance can change with age.

Many more results in MSThesis by Carson Willey (see Ohio Link)

OhioLINK ETD Center



Willey, Carson Landis



Future Work

- Correlation with clinical/experimental data.
- Inclusion of normal forms into the parametric variations for balance.

