

Introduction

- Gait parameters such as the mean and standard deviation of speed correlate with human falls [1].
- However, they can require hundreds of steps to accurately calculate using brute force methods [2].
- Further, brute force methods need to be repeated if testing conditions change.
- We introduce a novel method of directly calculating gait parameters using an absorbing Markov Chain.

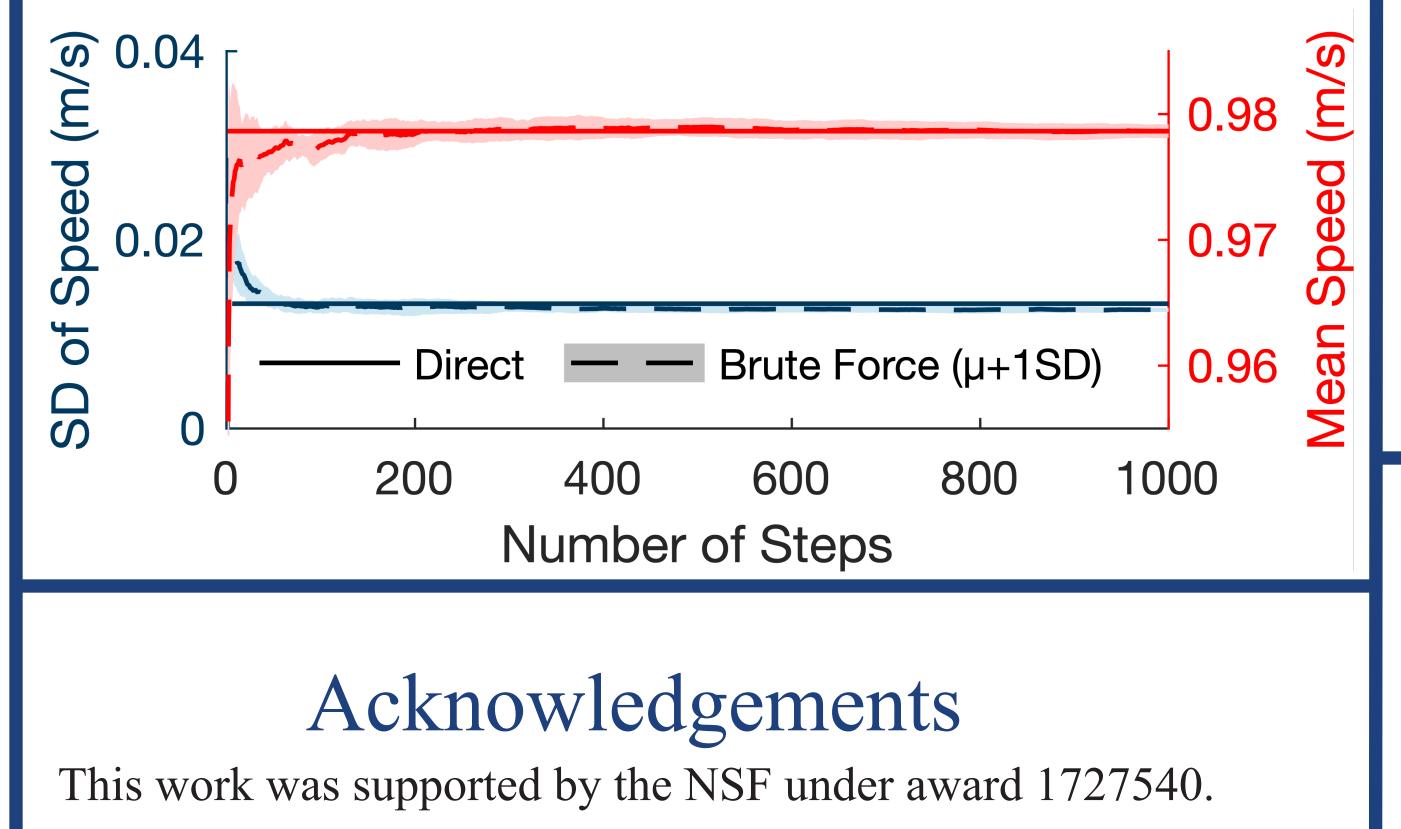
Methods

- Bipedal gait was represented as an absorbing Markov chain.
- First, a deterministic state-transition matrix T^d was created [3]:
 - The states of the Markov chain were the positions and velocities at heelstrike.
 - The transition from one state to another was dependent on the perturbation.
- This T^d can be used to analyze many testing conditions and only needs to be created once.
- By combining T^d and a specific distribution of random



Results

- We compared this novel method to a brute force method calculating walking speed.
 - The brute force method used 10 simulations of 1000 steps each.
- Testing was done using a six link, planar biped for three perturbation conditions.
- Brute force was inaccurate for low numbers of steps
- The difference in means between methods were < 0.5%when the brute force method was used for 1000 steps.
- The difference in SD were 0.80% to 12%.



perturbations the absorbing Markov chain T^s was created.

 $T^{s} = \begin{pmatrix} \mathbf{Q} & \mathbf{R} \\ \mathbf{Q} & \mathbf{1} \end{pmatrix}$

• T^s was used to find the expected number of times the biped will be in each state before falling.

 $N = (I - Q)^{-1}$

where $n_{ij} \in N$ is the expected number of times the biped will be in state p_i if it starts in state p_i .

- Weighted mean μ_i and standard deviation (SD) σ_i for each step that starts at state p_i were calculated using gait data saved when calculating T^d and N.
- The overall parameters were found using

$$\mu = \frac{\sum_{i=1}^{t} n_{1i} \mu_i}{\sum_{i=1}^{t} n_{1i}}, \quad \sigma^2 = \frac{\sum_{i=1}^{t} n_{1i} (\sigma_i^2 + (\mu_i - \mu)^2)}{\sum_{i=1}^{t} n_{1i}}$$



- [1] Hamacher, D, et al. J. Roy. Soc. Interface 8, 1682-1698, 2011. • [2] Owings, TM & Grabiner, MD. J. Biomech. 36, 1215-1218, 2003
- [3] Saglam, CO & Byl, K. Robotic. Sci. Syst., 2014.
- [4] Grinstead, CM & Snell, JL, Amer. Math. Soc., 2012