



# Modeling Quadruped Gait Bifurcations

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## INTRODUCTION

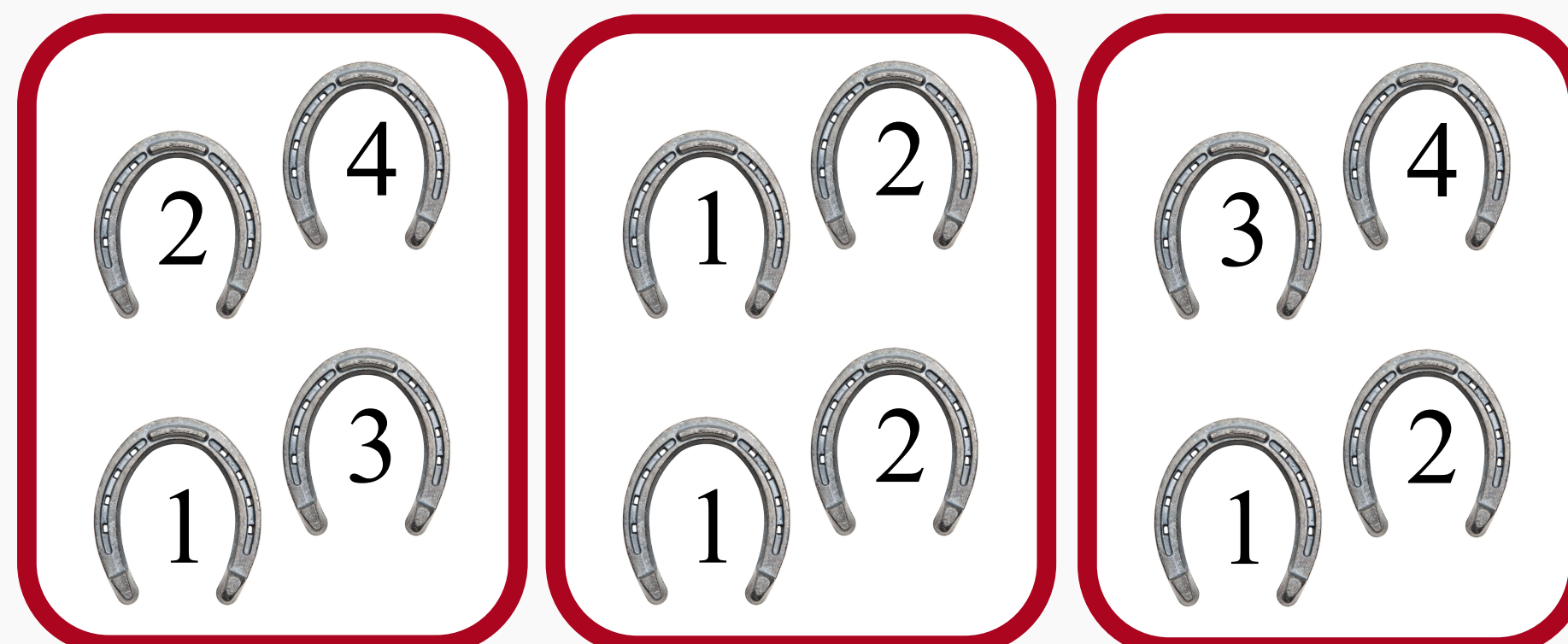
### Why model quadruped gaits?

- **Biology:** Neural gait mechanisms unknown, so model refines hypotheses
- **Diagnostics:** Movement issues found when data shifts away from model
- **Biomimicry:** Robotic movement is natural, flexible, and adaptable

### How do we model quadruped gaits?

- **Central Pattern Generators:** Neural circuits that produce gaits
  - Modeled via cells (a system of ODEs) which gait transition occurs via a **Bifurcation:** Dynamical system behavior changes by varying parameter
  - **Supercritical Hopf Bifurcation:** As a parameter varies, solutions attract from a fixed point (standing) to a stable oscillation (some gait)

**Figure 1:** Spatiotemporal patterns of relevant gaits. Numbers denote order of movement, starting with the left hind. (L.) Walk. (Mid.) Pace. (R.) Transverse Gallop.



## METHODS

- **Idea:** Symmetries of gait  $\leftrightarrow$  symmetries of model
  - Ex: Pace has ipsilateral legs “identical”, so model still makes pace when swapping ipsilateral cells

**Figure 2:** The bottom layer corresponds to legs. 1 is left hind, 2 is right hind, 3 is left front, and 4 is right front. 4 - 8 can be viewed as higher-order neurons. The CPG model is all 8 cells with the gait being  $x_1$  to  $x_4$ .

- Simplify cell as a neuron and model firing by the FitzHugh Nagumo Eqns
  - $f_1$ : membrane potential,  $f_2$ : recovery variable; set  $(a, b, c) = (0.02, 0.2, 0.44)$  reproduce the results of Golubitsky et al. (1993)

$$f_1(x, y) := c(x + y - \frac{x^3}{3}) \quad f_2(x, y) := -\frac{1}{c}(x - a + by)$$

- Then the  $i^{\text{th}}$  cell has the 2D system for  $i = 1, \dots, 8$  and indices modulo 8

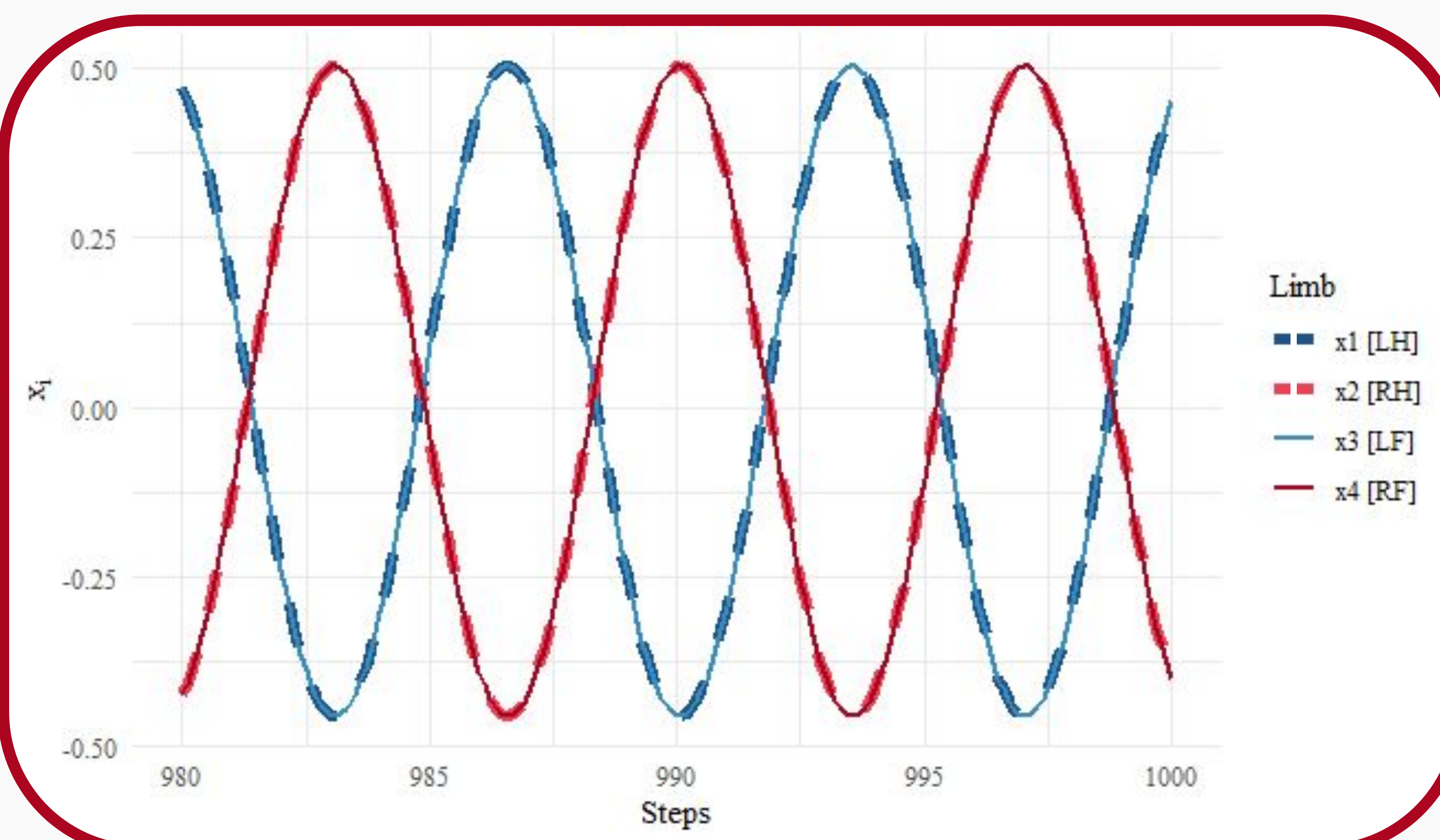
$$\dot{x}_i := f_1(x_i, y_i) + \alpha(x_{i-2} - x_i) + \gamma(x_{i+\epsilon} - x_i) \quad \epsilon = \begin{cases} +1 & i \text{ odd} \\ -1 & i \text{ even} \end{cases}$$

$$\dot{y}_i := f_2(x_i, y_i) + \beta(x_{i-2} - x_i) + \delta(x_{i+\epsilon} - x_i)$$

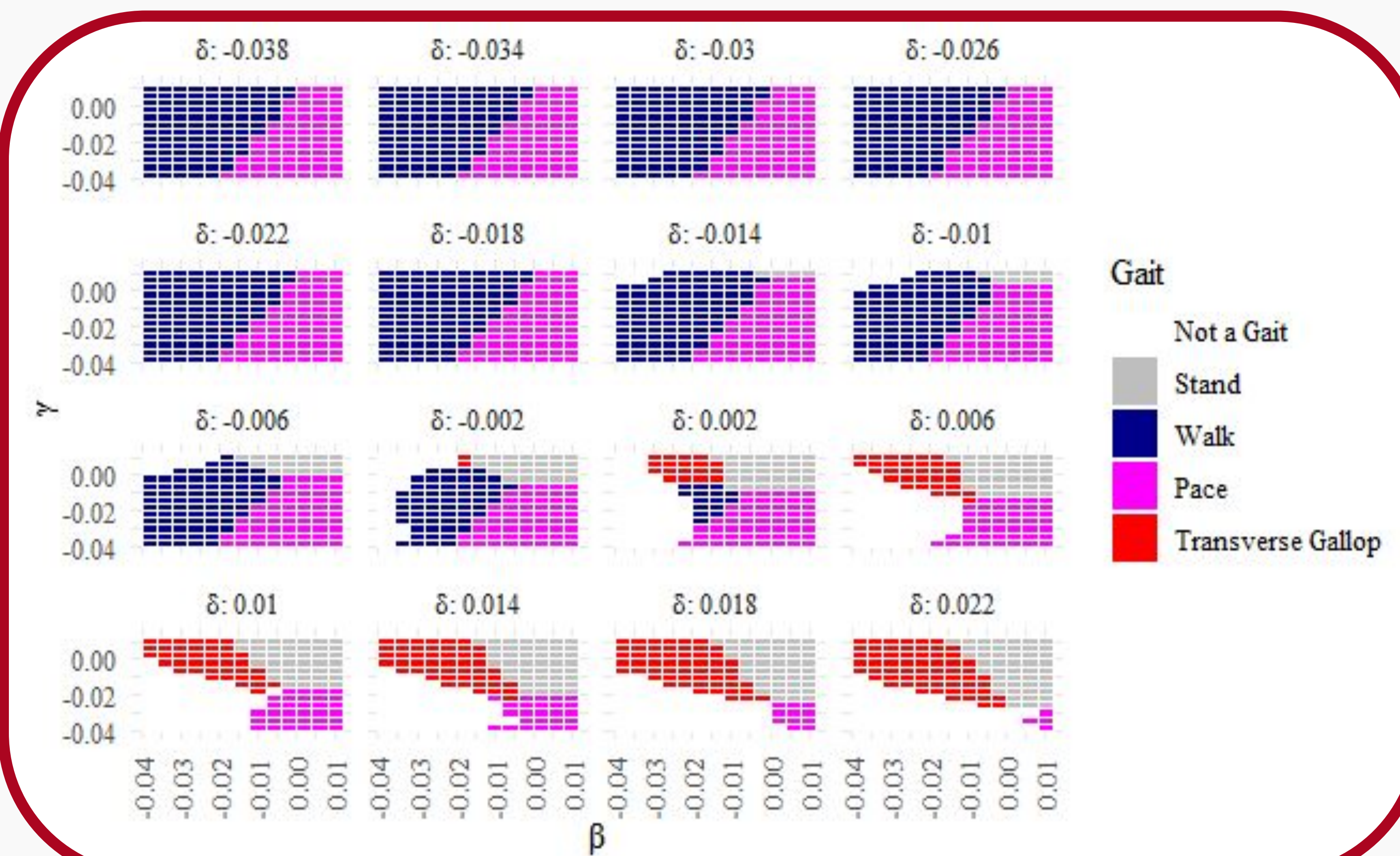
Ipsilateral and Contralateral Coupling Strengths

- We also performed sensitivity and bifurcation analyses

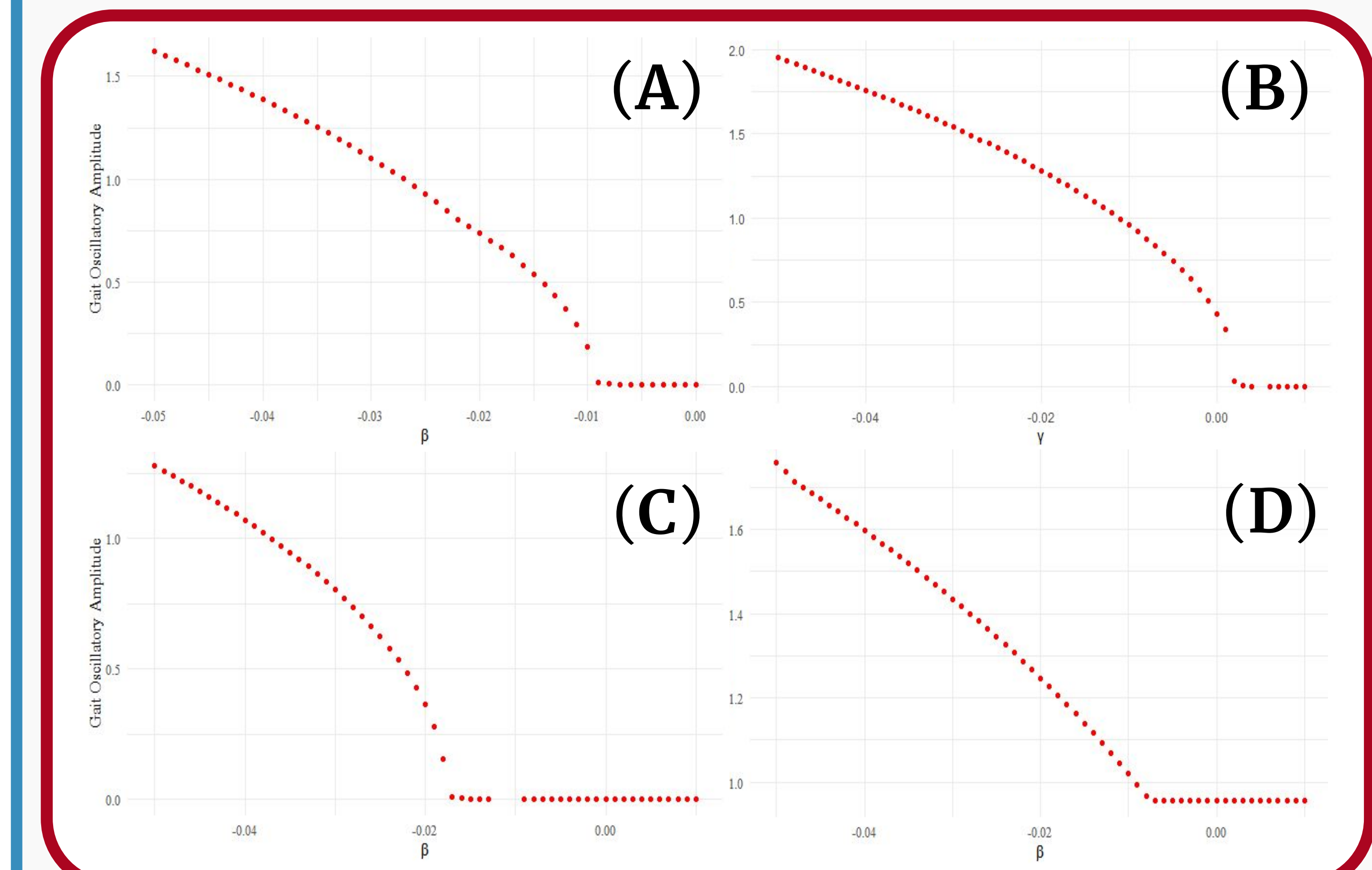
## RESULTS



**Figure 3:** Example gait. Gait is defined spatiotemporally: right order and timing. Pace has ipsilateral legs moving at the same time (order) in a 2 beat cycle (timing). This model had strong ipsilateral and weak contralateral couplings  $(\alpha, \beta, \gamma, \delta) = (0.1, 0.1, -0.1, -0.1)$ , producing pace as expected.



**Figure 4:** Strengthening / weakening coupling strengths matches expected gait. We fix  $\alpha = 0.1$  and vary  $\beta, \gamma$ , and  $\delta$  like a cube, which then we cut  $\beta\gamma$ -planes. With  $\alpha > 0$ , pace requires the other ipsilateral coupler  $\beta > 0$ . As walk has the left legs move before the right and transverse gallop has the back legs move before the front, then walk prioritizes ipsilateral coupling and transverse gallop contralateral. As  $\delta$  increases, walk is swapped for transverse gallop. All couplers positive makes stand.



**Figure 5:** Transition of (A) walk, (B) pace, (C) transverse gallop (left side of plots) to stand (right) are smooth. When fixing parameters and varying only one, each resembles a Supercritical Hopf Bifurcation. The y-axis is the max-min of  $x_1$  to  $x_4$  where the gait converges (namely: gait oscillatory amplitude error  $< 0.002$ ). (D) Transition of walk (left) to pace (right) is similar.

## DISCUSSION

- Model is **good** as it is
  - **Interpretable:** Relative coupling strengths successfully predicted gaits
  - **Stable:** Perturbing initial conditions didn't change 3 tested gaits (60 trials)
    - Walk also stabilizes without chaotic transitions during the bifurcation
- Model can be **improved** as it is
  - **Simplistic:** one cell is likely many, not one, neurons
  - ... and **complicated:** there's 7 parameters with many couplings possible
  - **Abstract:** a real life gait has no obvious pairing to specific parameters
  - **Lacking Feedback:** Would (1) mark “higher” v. “lower” cells, (2) adapt to movement needs, and (3) model top-down and bottom-up processes
    - Proprioceptive for lower and vestibular for higher to balance

## ACKNOWLEDGEMENTS

- Work supported by the National Science Foundation through DMS-1937229
- Also supported by the Data Driven Discovery RTG at the University of Arizona
- Guided by Dr. Kevin Lin and Dr. Shay Gilpin

## REFERENCES

- Collins, J. J. & Stewart, N. (1992). Coupled Nonlinear Oscillators and the Symmetries of Animal Gaits. *Nonlinear Science*, 3(1), 349-392. <https://doi.org/10.1007/BF02429870>
- Golubitsky, M., Stewart, I., Buono, P., & Collins, J. J. (1998). A Modular Network for Legged Locomotion. *Physica D*, 115(1-2), 56-72. [https://doi.org/10.1016/S0167-2789\(97\)00222-4](https://doi.org/10.1016/S0167-2789(97)00222-4)

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