



Spaun

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Large Scale Neural Simulations

“If just reproducing the brain is the aim, then there are better (presumably fun) ways to do that. If we just want to simulate a real brain with all details, maybe we should just reproduce.’ Anonymous[1]”

[1] M. Colombo, “Why build a virtual brain? Large-scale neural simulations as jump start for cognitive computing”, *Journal of Experimental & Theoretical Artificial Intelligence*, vol. 29, no. 2, Feb. 2016.

Large Scale Neural Simulations

What do you get out of a large scale neural simulation?

- Computational modeling experience[1]
- Solve complex problems by mimicking the brain[1]
- Model expression of physical changes to brains in behavior

[1] M. Colombo, “Why build a virtual brain? Large-scale neural simulations as jump start for cognitive computing”, *Journal of Experimental & Theoretical Artificial Intelligence*, vol. 29, no. 2, Feb. 2016.

Large Scale Neural Simulations

- Human Brain Project
 - Centered around the IBM Blue Brain Project
 - 37 million neurons [2]
- DARPA Synapse Project
 - 500 million neurons, 5x more than human brain [2]
- Semantic Pointer Architecture Unified Network (Spaun)
 - 2.5 million neurons

Large Scale Neural Simulations

- Human Brain Project – Henry Markram
 - Centered around the IBM Blue Brain Project
 - 100,000 neurons [2]
- Semantic Pointer Architecture Unified Network (SPAUN) – Chris Eliasmith

SPAUN

- Focuses on link between brain and behavior
- Can perform 8 tasks
- Hardwired – cannot learn [3]
- Has an eye and 2 degree of freedom



Fig. 1. Spaun high level overview

SPAUN Tasks

1. Copy Drawing
2. Image Recognition – given digit, reproduce (default handwriting)
3. 3-armed bandit trial
4. Serial Working Memory – reproduce list
5. Counting – essentially sum two values.
6. Answer Questions
7. Rapid Variable creation – given pattern, create new pattern
8. Fluid Reasoning – similar to Raven's Matrix test [4]

Task Examples

- Tracing from memory
- Given digit in random handwriting style
- Reproduce digit from memory
- Treat different stimuli as same tokens
- Detect variation [5]



Fig. 2. As shown in [5, Fig. 3], Spaun is able to (given a digit in any handwriting), reproduce it from memory of 5 known copies of digits

Task Examples

- Fluid Reasoning
- <https://youtu.be/qcZe-2eWaeM>

Semantic Pointer Architecture

- Semantic Pointer hypothesis - “Higher-level cognitive functions in biological systems are made possible by semantic pointers. Semantic pointers are neural representations that carry partial semantic content and are composable into the representational structures necessary to support complex cognition.” [6]
- Spaun uses semantic pointers to store high-level, low-dimensional representations of images [5].
- Maintain relationships from image space
- “Dereferenced” to get motor commands

[6] University of Waterloo, “Semantic Pointer Architecture”, Computational Neuroscience Research Group at Waterloo Centre for Theoretical Neuroscience [Online]. Available: <http://compneuro.uwaterloo.ca/research/spa/semantic-pointer-architecture.html>. [Accessed: Apr. 24, 2018].

[5] T. C. Stewart, Feng-Xuan Choo, and C. Eliasmith, “Spaun: A Perception-Cognition-Action Model Using Spiking Neurons”, *Proceedings of the Annual Meeting of the Cognitive Science Society*, vol. 34, no. 34, 2012

Neural Engineering Framework

- NEF converts algorithms to spiking neuron models [6].
- Uses Leaky Integrate-and-Fire neurons*
 - a) Neuron groups represent vectors
 - b) Connections specify computation

* Subject to criticism from Markram

Leaky Integrate-and-Fire

- A simplified version of the Integrate and Fire model.
- $\tau_m \frac{du}{dt} = -[u(t) - u_{rest}] + RI(t)$
- If $u(t) = \Theta$ then $\lim_{\delta \rightarrow 0; \delta > 0} u(t + \delta) = u_r$ [7]

Leaky Integrate-and-Fire

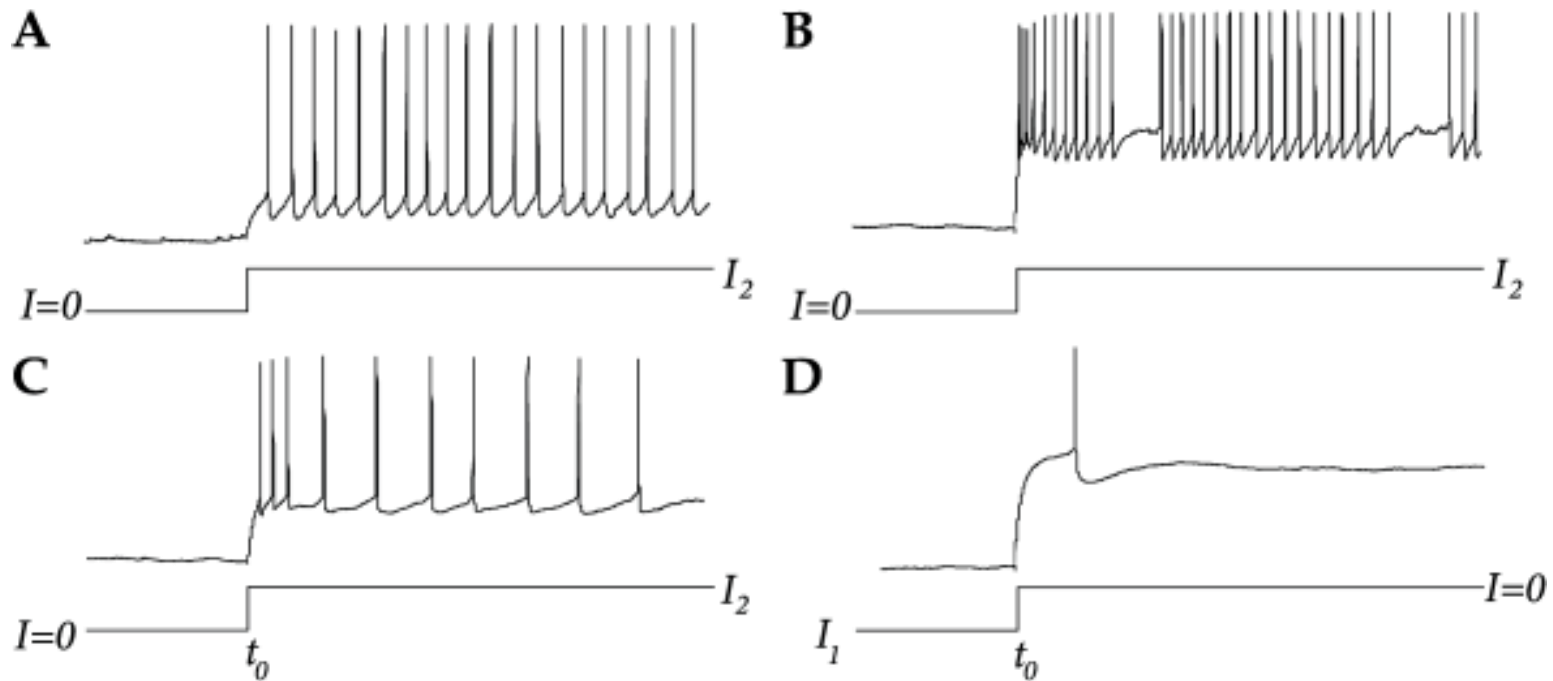


Fig. 3. As seen in [7, Fig. 1.10], (A) shows neurons with short intervals and no adaptation. (B) shows neurons with short spiking intervals with a “stutter”. (C) Shows more regular neuron spiking with adaptation. (D) shows a neuron that also has a inhibitory rebound spike.

Neural Engineering Framework

1. $J = \alpha e \cdot x + J_{bias}$
2. $d = \Gamma^{-1}\Upsilon$
 - $\Gamma_{ij} = \int a_i a_j dx$
 - $\Upsilon_j = \int a_j x dx$
3. $\omega_{ij} = \alpha_j e_j M d_i$
4. $d^{f(x)} = \Gamma^{-1}\Upsilon,$
 - $\Gamma_{ij} = \int a_i a_j dx,$
 - $\Upsilon_j = \int a_j f(x) dx$
- [5, eq(1), eq(2), eq(3), eq(4)]

Spaun Anatomical Architecture

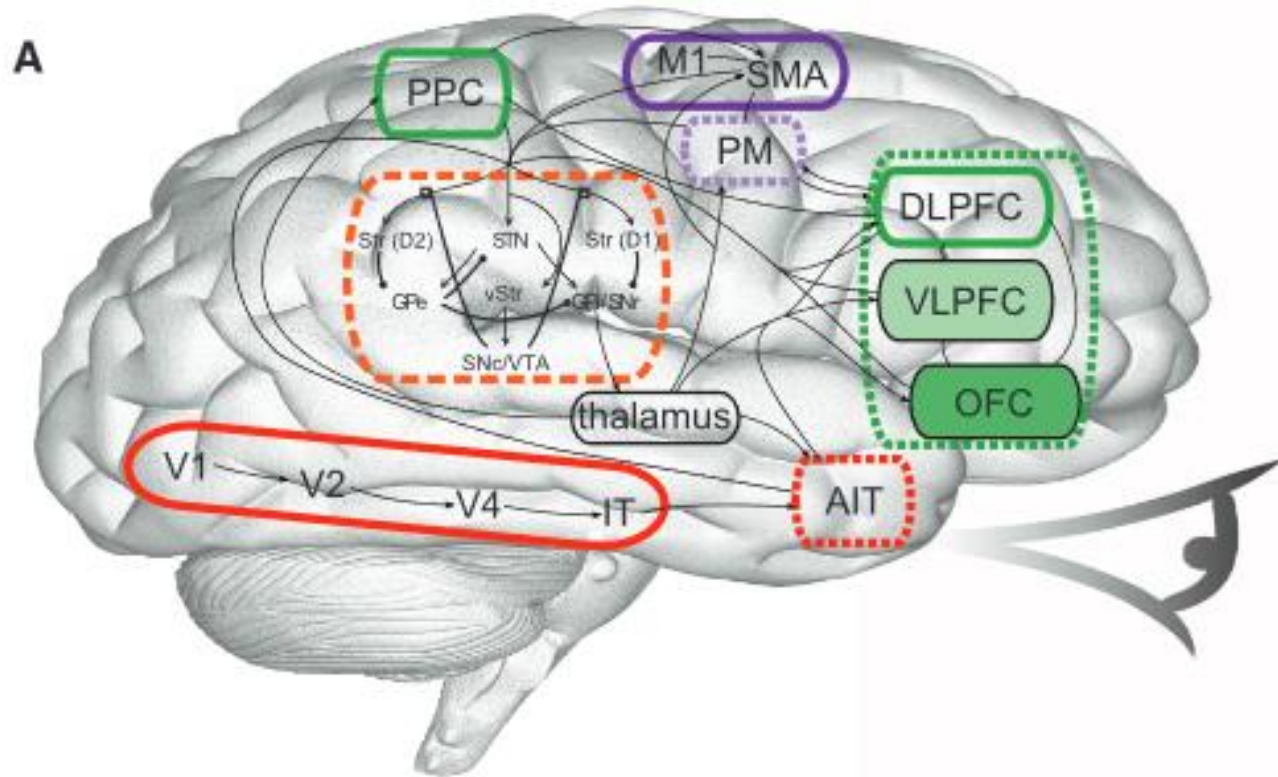


Fig. 4. [4, Fig. 1.] The anatomical structure of Spaun.

Spaun Functional Architecture

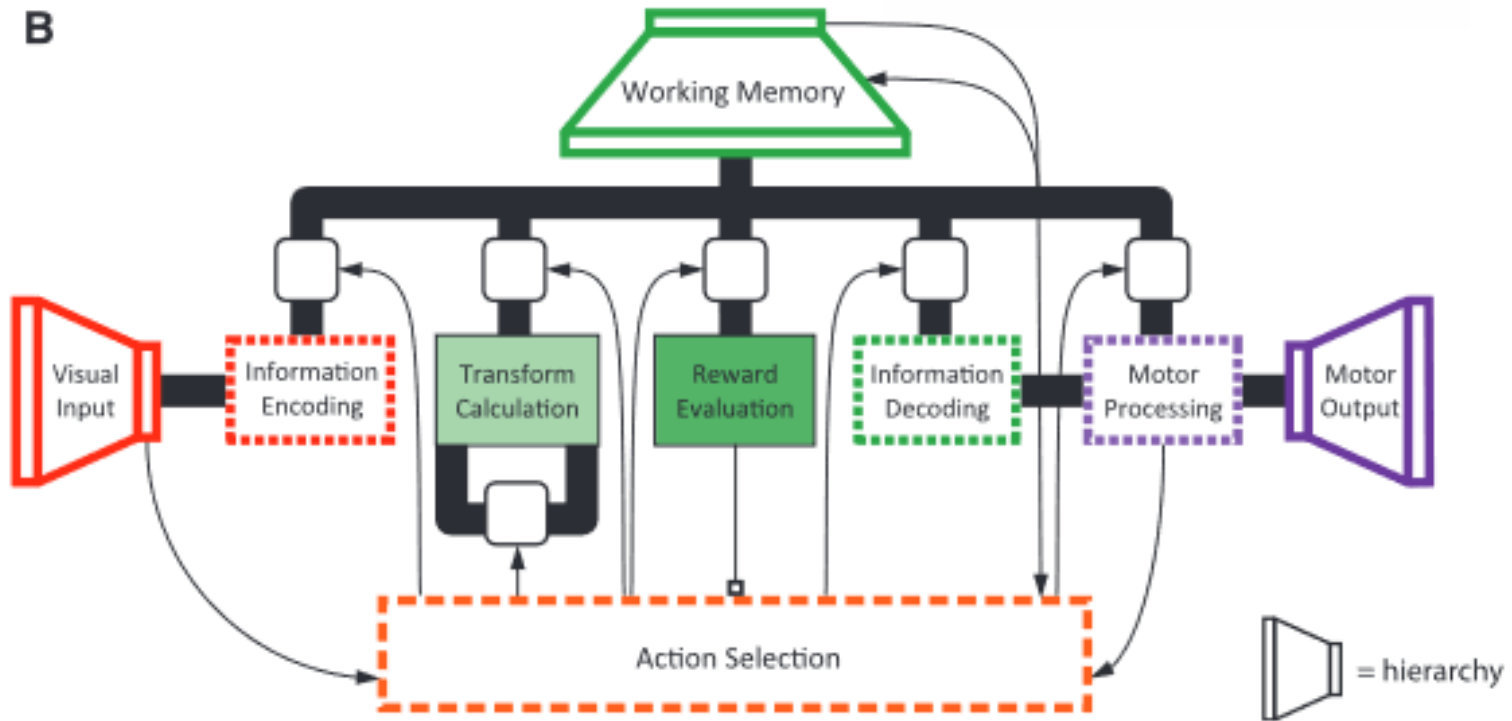


Fig. 5. [4, Fig 1B] Spaun functional architecture.

Criticism for Spaun

- Markram criticized Spaun due to the simplified Leaky-Integrate-And-Fire neurons.
- “It is not a brain model” [8].

The Ultimate Showdown



Fig. 6. Chris Eliasmith, professor University of Waterloo, creator of Spaun [9].

VS.



Fig. 7. Henry Markram, professor of Neuroscience at the Swiss Federal Institute of Technology, founder and director of the Big Blue Brain Project (Human Brain Project) [10].

[9] University of Waterloo, “Christ Eliasmith”, University of Waterloo, Nov. 2012. [Online] <http://arts.uwaterloo.ca/~celiasmi/>. [Accessed: Apr. 25, 2018].

[10] Ecole Polytechnique Federale De Lausanne, “Henry Markram”. [Online] Available: <https://bluebrain.epfl.ch/people/henrymarkram>. [Accessed: Apr. 25, 2018].

BioSpaun

- Replaced neurons in the model
- HBP Neurons:
 - 13 ion channels and 4 compartments
- BioSpaun Neurons
 - Pyrimidal Cell model
 - 20 compartments
 - 27 parameters and 9 ion channels [8].

BioSpaun Neuron Comparison

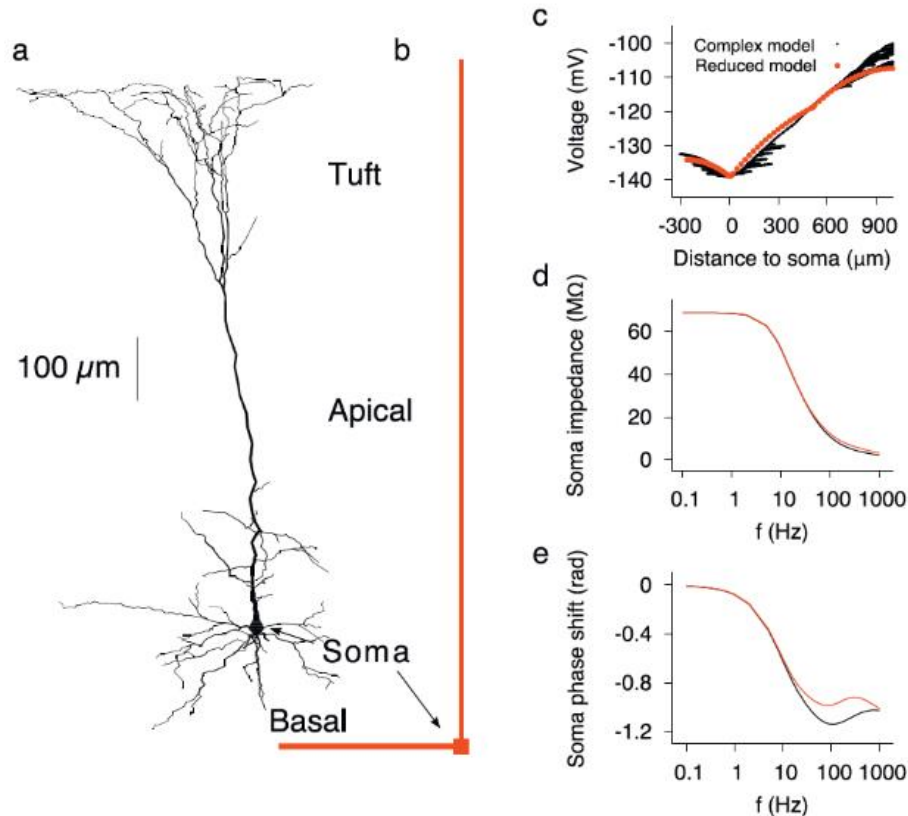


Fig. 8. [8, Fig. 2]. A) Neuron. C) Voltage at different locations in the cell. D) Somatic impedance. E) Somatic phase shift as a function of oscillatory input currents and membrane currents.

BioSpaun Communication Comparison

- Circuit was constructed
- 200 input neurons
- 50 output neurons [8]
- Compartmentalized Neurons slightly noisier
- RMS Error
 - LIF – 0.1%
 - Compartment – 0.21%

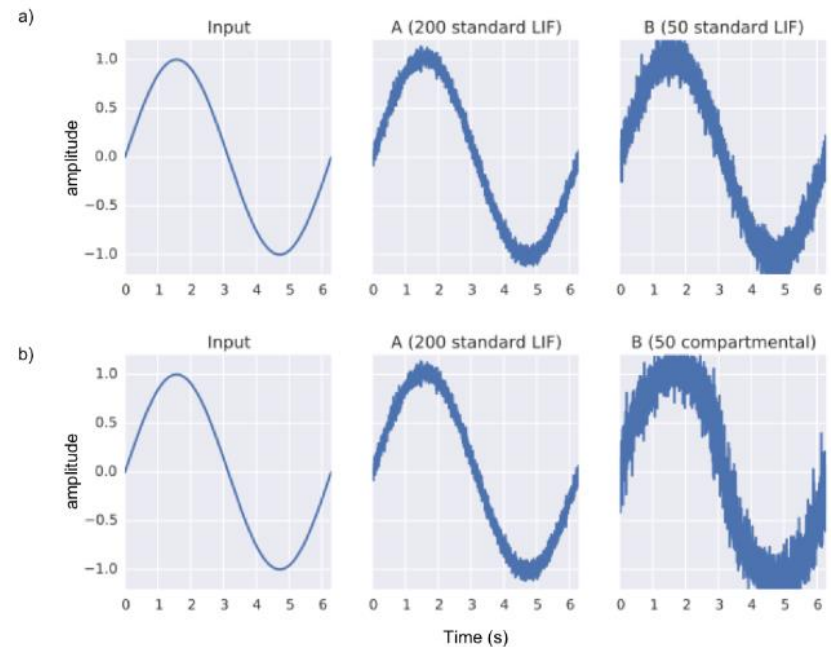


Fig. 9. Reproduced from [8, Fig. 3], a) shows information processing through a channel of LIF neurons whereas b) represents the same communication but via BioSpaun’s compartmentalized neurons.

BioSpaun Task Comparison

- Spaun had 94% accuracy in handwritten digit recognition.
- BioSpaun maintained that accuracy.
- TTX reduces performance.

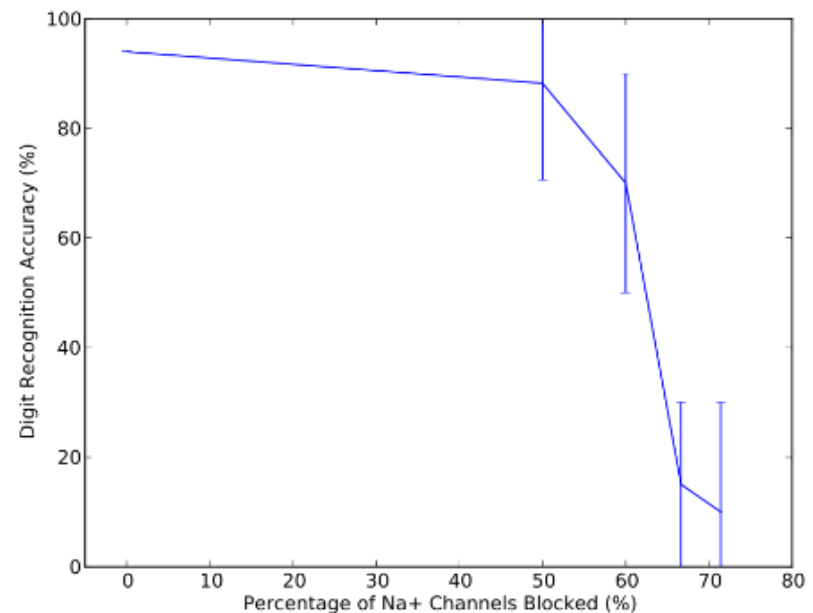


Fig. 10. Digit recognition accuracy. Leftmost value represents baseline accuracy. Reproduced from [8, Fig. 5.].

Simulating Drug Effects

- Tetrodotoxin (TTX) blocks voltage gated sodium-ion channels.
- TTX effects were simulated in BioSpaun[8]

Simulating Drug Effects

- Introducing TTX results in poorer performance
- Expected

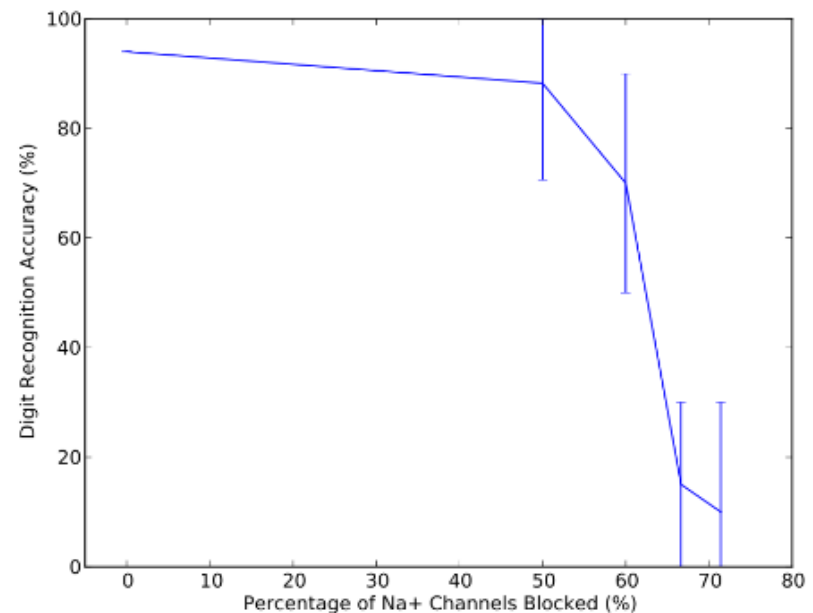


Fig. 11. Digit recognition accuracy. Leftmost value represents baseline accuracy. Reproduced from [8, Fig. 5.].

Simulating Drug Effects (Counting)

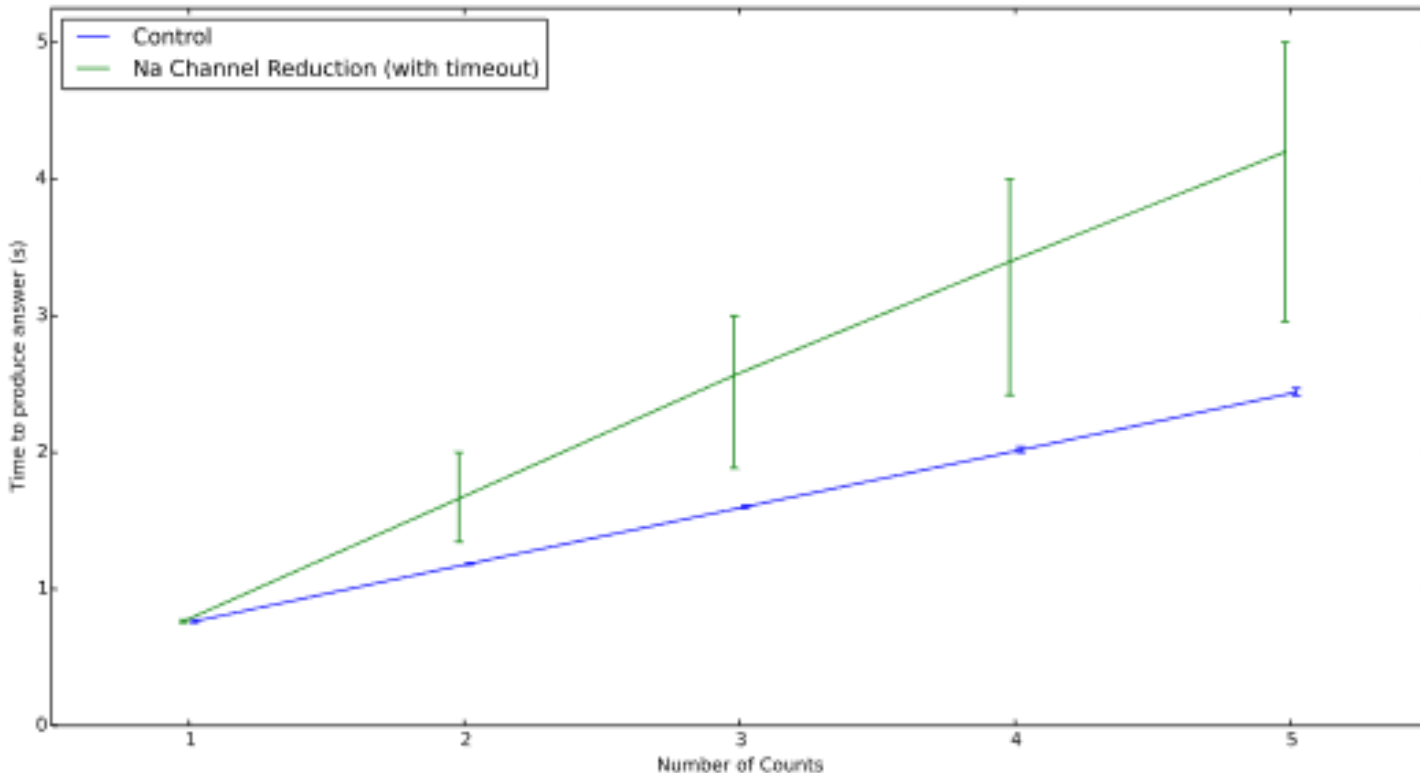


Fig. 12. Introducing TTX results in BioSpaun encoding start digit and end digit, but “forgetting” states before finishing [8, Fig. 6].

BioSpaun Summary

- Maintained the performance of original Spaun
- Closer to physiological model
- Increased testing capability [8]

Conclusions

- Spaun accomplishes its goals of connecting physiological structure to behavior.
- Testing of hypotheses on large-scale models.