

## E. Paxon Frady, Ph.D.

---

Intel, Neuromorphic Computing Lab  
e.paxon.frady@intel.com – (770) 380-1770

## Education

---

### UC San Diego

*Ph.D., Neuroscience specialization in Computational Neuroscience*

**2008–2014**

*La Jolla, CA*

**Thesis:** Scalable semi-supervised cell identification reveals canonical swim and preparatory networks.

**Advisors:** William Kristan, Gert Cauwenberghs, Massimo Scanziani, Terry Sejnowski, Tim Gentner

### California Institute of Technology

*B.S., Computation and Neural Systems, Business Economics and Management*

**2004–2008**

*Pasadena, CA*

## Research Experience

---

### Intel, Neuromorphic Computing Lab

*Neuromorphic Algorithms Researcher*

Developed algorithms for spiking neuromorphic computing hardware.

**2019–Pres**

*Santa Clara, CA*

### Redwood Center for Theoretical Neuroscience, UC Berkeley

*Postdoctoral Scientist*

Studied connectionist neural networks, linking algorithms to observations in neuroscience.

**2016–2021**

*Berkeley, CA*

### Inscopix

*Postdoctoral Scientist*

Analyzed large-scale calcium imaging data.

**2015**

*Palo Alto, CA*

### Numenta/UC Berkeley

*Postdoctoral Scientist*

Developed computational models of the cortex.

**2015**

*San Francisco, CA*

### Microsoft Research

*Intern with Eric Horvitz*

Developed computational algorithms to analyze large-scale VSD imaging data.

**2010**

*Redmond, WA*

### Eye-Predict

*Software Engineer*

Applied computational model of visual attention to advertising, product placement, and website layout optimization.

**2007–2009**

*Los Angeles, CA*

## Teaching Experience

---

### Neu 299: Introduction to vector symbolic architectures

*Lecturer*

Organized and lectured new course on modern VSA methods.

**2021**

*Berkeley, CA*

### Neu 299: The art of modeling

*Project Mentor*

Worked with student in designing modeling projects to better understand experimental neuroscience data.

**2021**

*Berkeley, CA*

### VS 265: Neural Computation

*Guest Lecturer*

Taught introduction to connectionist theory and computation with high-dimensional vectors.

**2016, 2018**

*Berkeley, CA*

### Analytical Methods in Computational Neuroscience

*Student Faculty*

Involved in creating a student-run course on computational methods in neuroscience.

- Dimensionality reduction, principal components analysis, independent components analysis.

**2012–2014**

*La Jolla, CA*

### Neural Systems & Behavior

*Teaching Faculty*

Taught fundamentals VSD Imaging in the leech and computational imaging analysis techniques.

**2013–2014**

*Woods Hole, MA*

### UCSD Neurosciences Bootcamp

*Teaching Assistant*

Taught incoming Neuroscience graduate students fundamentals of electrophysiology and neuroscience in intensive two-week course.

- Intracellular recordings: action-potentials, synaptic potentials, and gap-junctions.
- Basics of dynamical systems: systems of differential equations, phase-planes, bifurcations, stability.

**2009–2013**

*La Jolla, CA*

### Bootcamp Computational Special Project

*Head TA*

**2009–2012**

*La Jolla, CA*

Taught basic introduction to computational modeling in neuroscience covering a wide array of projects.

- o Matlab/python simulations: Hodgkin-Huxley, Izhikevich, Morris-Lecar models of action-potential, Calcium channel kinetics in vesicles, sub-cellular dendritic integration (NEURON), gain control in spiking neurons (Brian simulator), multi-stable pattern generating networks.

**Physics 173/BGGN 266: Modern Physics/Biophysics Laboratory**

Teaching Assistant

Lead students through electrophysiology project on leeches.

**2010**

La Jolla, CA

## Grants & Fellowships

---

**2019 DARPA:** Computing in superposition with high-dimensional vectors.

**2018 NIH BRAIN Initiative:** Building analysis tools and a theory framework for inferring principles of neural computation from multi-scale organization in brain recordings.

**2018 Intel Neuromorphic Research Community:** A structured approach to design algorithms for Loihi.

**2018 Berkeley Deep Drive:** Connectionist representations of compositional structure for sensor integration and situational awareness in autonomous vehicles.

**2011–2012 INC Cognitive Neuroscience Fellowship:** Development and validation of functional connectivity algorithms to understand recurrent neuronal circuitry and its relationship to behavior and cognition.

**2009–2011 UCSD Interfaces Fellowship:** Inferring homologous cells across animals to analyze neural circuits.

**2007 Frank W. Wood SURF Fellow:** Understanding the visual saliency of faces and text.

**2006 Summer Undergraduate Research Fellowship:** The relation between phase-noise and overt attention.

## Publications

---

### Peer-Reviewed Articles

---

1. Orchard, G., Frady, E.P., Ben-Dayan Rubin, D., Sanborn, S., Srestha, S., Sommer, F., Davies, M. (2021). Efficient neuromorphic signal processing with Loihi 2. *IEEE 2021 International Workshop on Signal Processing Systems*.
2. Frady, E.P., Kleyko, D., Sommer, F. (2021). Variable binding for sparse distributed representations: theory and applications. *IEEE Transactions on Neural Networks and Learning Systems*.
3. Kleyko, D., Frady, E.P., Kheffache, M., Osipov, E. (2020b). Integer Echo State Networks: Efficient Reservoir Computing for Digital Hardware. *IEEE Transactions on Neural Networks and Learning Systems*.
4. Frady, E.P., Kent, S., Olshausen, B.A., Sommer, F.T. (2020). Resonator Networks, 1: An Efficient Solution for Factoring High-Dimensional, Distributed Representations of Data Structures. *Neural Computation* 32(12): 2311-2331.
5. Kent, S., Frady, E.P., Sommer, F.T., Olshausen, B.A. (2020). Resonator Networks, 2: Factorization Performance and Capacity Compared to Optimization-Based Methods. *Neural Computation* 32(12): 2332-2388.
6. Kleyko, D., Kheffache, M., Frady, E.P., Wiklund, U., Osipov, E. (2020a). Density Encoding Enables Resource-Efficient Randomly Connected Neural Networks. *IEEE Transactions on Neural Networks and Learning Systems*.
7. Frady, E. P., Orchard, G., Florey, D., Imam, N., Liu, R., Mishra, J., Tse, J., Wild, A., Sommer, F. T., Davies, M. (2020). Neuromorphic nearest neighbor search using Intel's Pohoiki Springs. In *Proceedings of the Neuro-Inspired Computational Elements Workshop*, NICE '20, New York, NY, USA. Association for Computing Machinery.
8. Frady, E.P. & Sommer, F.T. (2019). Robust computation with rhythmic spike patterns. *PNAS* 116(36): 18050-18059.
9. Frady, E.P., Kleyko, D., Sommer, F.T. (2018). A theory of sequence indexing and working memory in recurrent neural networks. *Neural Computation* 30(6): 1449-1513.
10. Rahimi, A., Datta, S., Kleyko, D., Frady, E.P., Olshausen, B.A., Kanerva, P., Rabaey, J.M. (2017) High-dimensional computing as a nanoscale paradigm. *IEEE Trans. on Circuits and Systems* 64(9): 2508-2521.
11. Lippi, G., Fernandes, C.C., Ewell, L.A., Romoli, B., Curia, G., Frady, E.P., Jensen, A.B., Chaabane, M.M., Belal, C., Nathanson, J.L., Zoli, M., Leutgeb, J.K., Biagini, G., Yeo, G.Y., Berg, D.K. (2016) MicroRNA-101 Regulates Multiple Development Programs to Constrain Excitation in Adult Neural Networks. *Neuron* 92(6): 1337-1351.
12. Frady, E.P., Kapoor, A., Horvitz, E., Kristan, W.B. (2016). Scalable semi-supervised functional neurocartography reveals canonical neurons in behavioral networks. *Neural Computation* 28(8): 1453-1497.
13. Berdycheva, T. K., Frady, E. P., Nassi, J. J., Aluisio, L., Cherkas, Y., Otte, S., Wyatt, R., Dugovic, C., Ghosh, K.K., Schnitzer, M.J., Lovenberg, T., Bonaventure, P. (2016). Direct Imaging of Hippocampal Epileptiform Calcium Motifs Following Kainic Acid Administration in Freely Behaving Mice. *Frontiers in Neuroscience*, 10, 53.
14. Woodford, C.R., Frady, E.P., Smith, R., Morey, B., Canzi, G., Araneda, R., Kristan, W.B., Kubiak, C.P., Miller, E.M., Tsien, R.Y. (2015). Improved PeT molecules for optically sensing voltage in neurons. *J. Am. Chem. Soc.* 137: 1817.
15. Miller, E.W., Lin, J.Y., Frady, E.P., Steinbach, P.A., Kristan, W.B., Tsien, R.Y. (2012). Optically monitoring voltage in neurons by photo-induced electron transfer through molecular wires. *PNAS* 109(6): 2114-2119.
16. Cerf, M.\*, Frady, E.P.\*, Koch, C. (2009). Faces and text attract gaze independent of task: Experimental Data and Computer Model. *Journal of Vision* 9(12): 1-15.
17. Einhauser, W., Rutishauser, U., Frady, E.P., Nadler S., Konig, P., Koch, C. (2006). The relation of phase noise and luminance contrast to overt attention in complex visual stimuli. *Journal of Vision* 6: 1148-58.

## Pre-prints, Views & Proceedings.....

1. Frady, E.P., Kleyko, D., Kymn, C.J., Olshausen, B.A., Sommer, F.T. (2021). Computing on functions using randomized vector representations. arXiv:2109.03429 [cs.LG]
2. Kleyko, D., Davies, M., Frady, E.P., Kanerva, P., Kent, S.J., Olshausen, B.A., Osipov, E., Rabaey, J.M., Rachkovskij, D., Rahimi, A., Sommer, F.T. (2021). Vector Symbolic Architectures as a Computing Framework for Nanoscale Hardware. arXiv:2106.05268 [cs.AR]
3. Kleyko, D., Rosato, A., Frady, E.P., Panella, M., Sommer, F.T. (2020). Perceptron Theory for Predicting the Accuracy of Neural Networks. arXiv:2012.07881 [cs.LG]
4. Kleyko, D., Frady, E.P., Sommer, F.T. (2020). Cellular automata can reduce memory requirements of collective-state computing. arXiv:2010.03585 [cs.NE]
5. Frady, E.P., Kleyko, D., Sommer, F.T. (2020). Variable binding for sparse distributed representations: theory and applications. arXiv:2009.06734 [cs.NE]
6. Kleyko, D., Kheffache, M., Frady, E.P., Wiklund, U., Osipov, E. (2019). Density Encoding Enables Resource-Efficient Randomly Connected Neural Networks. arXiv:1909.09153 [cs.LG]
7. Kent, S., Frady, E.P., Sommer, F.T., Olshausen, B.A. (2019). Resonator circuits for factoring high-dimensional vectors. arXiv:1906.11684 [cs.NE]
8. Frady, E.P., Kanerva, P., Sommer, F.T. (2019). Robust computation with rhythmic spike patterns. arXiv:1901.07718 [cs.NE]
9. Kleyko, D., Frady, E.P., Osipov, E. (2018). Integer Echo-State Networks: Hyperdimensional Reservoir Computing. arXiv:1706.00280 [cs.NE]
10. Frady, E.P., Kleyko, D., Sommer, F.T. (2017). Theory of the superposition principle for randomized connectionist representations in neural networks. arXiv:1707.01429 [cs.NE]
11. Frady, E.P., Kristan, W.B. (2015). The Imaging Computational Microscope: A Matlab tool for automatic analysis and visualization of large-scale imaging data. arXiv:1502.07009 [q-bio.NC]
12. Kapoor, A., Frady, E.P., Jegelka, S., Kristan, W.B., Horvitz, E. (2015). Inferring and Learning from Neural Correspondences. arXiv:1501.05973 [q-bio.NC]
13. Frady, E.P., Kristan, W.B. (2013). Computation with Population Codes. In: Jaeger D., Jung R. (Ed.) Encyclopedia of Computational Neuroscience: Springer Reference (www.springerreference.com). Springer-Verlag Berlin Heidelberg.
14. Frady, E.P., Palmer, C.R., Kristan, W.B. (2012). Sexual Attraction: Sex-Specific Wiring of Neural Circuitry. Current Biology 22(22): R953-R956.
15. Cerf, M., Frady, E.P., Koch, C. (2008). Using semantic content as cues for better scanpath prediction. ETRA 143-146.
16. Frady, E.P., Kleyko, D., Trann, Q., Kanerva, P. (in prep.). On inverting compound representations of English verbs.
17. Frady, E.P., Yudice, J., Konanur, V., Todd, K.L., French, K.A., Kristan, W.B. (in prep.). Shunting and inhibitory currents co-regulate the input-output function of an identified leech neuron.
18. Bybee, C., Frady, E.P., Sommer, F.T. (in prep.). Deep learning in spiking phasor neural networks.

## Conference Presentations.....

1. Frady, E.P. (2021). Hyperdimensional computing with complex representations. VSA Workshop ([online](#)).
2. Frady, E.P. (2020). VSAs and Resonator Networks: Towards Cognitive Computing on Neuromorphic Hardware. VSA Workshop ([online](#)).
3. Frady, E.P., Sommer, F.T. (2019). Robustly encoding multiple variables on smooth manifolds in a spiking model of hippocampus. Society for Neuroscience (Chicago, USA).
4. Frady, E.P., Davies, M., Sommer, F.T. (2019). Robust computation with rhythmic spike patterns on neuromorphic hardware. Joint Symposium for Neural Computation (Los Angeles, USA).
5. Frady, E.P., Kanerva, P., Sommer, F.T. (2019). A framework for linking computation and rhythm-based timing patterns in neural firing. Computational and Systems Neuroscience (Lisbon, Portugal).
6. Frady, E.P., Kent, S., Kanerva, P., Olshausen, B.A., Sommer, F.T. (2018). Cognitive neural systems for disentangling compositions. Cognitive Computing (Hannover, Germany).
7. Frady, E.P., Kanerva, P., Sommer, F.T. (2018). A framework for linking computation and rhythm-based timing patterns in neural firing. Cognitive Computing (Hannover, Germany).
8. Frady, E.P., Kanerva, P., Sommer, F.T. (2018). Spike-timing computation with complex vectors: emergence of theta oscillations, place-fields and phase-precession in a model of hippocampus. Society for Neuroscience (San Diego, USA).
9. Frady, E.P., Kanerva, P., Sommer, F.T. (2018). A memory network model using spike phase-precession. Asilomar Conference on Signals, Systems, and Computers (Asilomar, USA). [[oral](#)]
10. Frady, E.P., Kanerva, P., Sommer, F.T. (2018). A framework for linking computations and rhythm-based timing patterns in neural firing, such as phase-precession in hippocampal place cells. Cognitive Computational Neuroscience (Philadelphia, USA). [[oral](#)]
11. Frady, E.P., Kent, S., Tran, Q., Kanerva, P., Olshausen, B.A., Sommer, F.T. (2018). Resonator circuits: a model for inferring compositional structure in distributed representations. Computational and Systems Neuroscience (Denver, USA).
12. Kleyko, D., Frady, E.P., Osipov, E. (2018). Echo state networks based on hyperdimensional computing. Neural Inspired Computational Elements (Hillsboro, USA).

13. **Frady, E.P., Kleyko, D., Kanerva, P., Sommer, F.T.** (2016). The channel capacity and scaling of distributed neural activity. Society for Neuroscience (San Diego, USA).
14. **Berdyeva, T.K., Frady, E.P., Aluisio, L., Otte, S., Wyatt, R.M., Dugovic, C., Shelton, J., Ghosh, K., Schnitzer, M.J., Lovenberg, T., Bonaventure, P.** (2015). Direct imaging of calcium pathology preceding kainic acid induced seizure activity in freely behaving mice. Society for Neuroscience (Chicago, USA).
15. **Gulati, S., Frady, E.P., Cao, V., Joshi, P., Otte, S.L.** (2015). Multilayer cortical imaging in freely behaving animals. Society for Neuroscience (Chicago, USA).
16. **Sturgill, J.F., Frady, E.P., Isaacson, J.** (2015). Addition by division: a recurrent circuit explains cortical odor response regulation by SOM cells. Computational and Systems Neuroscience (Salt Lake City, USA). [oral]
17. **Frady, E.P., Kristan, W.B.** (2014). Scalable semi-supervised framework for activity mapping the leech nervous system. Computation and Systems Neuroscience (Salt Lake City, USA).
18. **Frady, E.P., Yudice, J., Konanur, V., Todd, K.L., French, K.A., Kristan, W.B.** (2013). Shunting and inhibitory currents co-regulate DE-3's input-output function. Society for Neuroscience (San Diego, USA).
19. **Frady, E.P., Yudice, J., Kristan, W.B.** (2013). Gain-control via shunting inhibition in a spiking model of the leech local bend reflex. Computational and Systems Neuroscience (Salt Lake City, USA).
20. **Frady, E.P., Kristan, W.B.** (2012). Inferring Homologous Cells Across Animals to Analyze Neural Circuits. Organization for Computational Neurosciences (Atlanta, USA). [oral]
21. **Frady, E.P., Kapoor, A., Horvitz, E., Kristan, W.B.** (2011). Utilizing multi-functional neuronal responses during different behaviors to uniquely identify all neurons in the leech ganglion. Society for Neuroscience (Washington D.C., USA).
22. **Frady, E.P., Kapoor, A., Horvitz, E., Kristan, W.B.** (2011). Isomap analysis of neuronal populations during decision-making. Computational and Systems Neuroscience (Salt Lake City, USA).
23. **Frady, E.P., Cauwenberghs, G.** (2009). Using dopamine as a modulator for STDP for reinforcement learning applied to Tic-Tac-Toe in a temporal difference framework. Society for Neuroscience (San Diego, USA).
24. **Frady, E.P., Cauwenberghs, G.** (2009). Multi-compartmental model of synaptic plasticity. Joint Symposium for Neural Computation (Los Angeles, USA).
25. **Cerf, M., Frady, E.P., Koch, C.** (2008). Subjects' inability to avoid looking at faces suggests bottom-up attention allocation mechanism for faces. Society for Neuroscience (Washington D.C., USA).