

# DILIP KRISHNAMURTHY

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## EDUCATION

### Carnegie Mellon University

Doctor of Philosophy (Ph.D.) in Mechanical Engineering [GPA: 4.0/4.0]

Pittsburgh, PA

May 2020 (expected)

Advisor: Professor Venkatasubramanian Viswanthan

Interests: Battery Aging Models and Reliability Predictions; Computational Material Design for Next-Generation Batteries; Physics-Aware Machine Learning and Data-Driven Material Discovery; Power Generation Systems Design Electric Aviation; Electrochemical Synthesis of Chemicals and Fuels

### Indian Institute of Technology Madras

Chennai, India

Bachelor of Technology and Master of Technology in Mechanical Engineering [GPA: 9.35/10.0]

June 2015

Research advisor: Professor Sankara J. Subramanian

Interests: Topology Optimization and Mechanical Component Design for Stress Minimization; Finite Element Analysis; Inverse Design of Composite Materials Through Digital Image Correlation.

## SELECT AWARDS & HONORS

<b>Presidential Fellow at Carnegie Mellon University</b> , the highest fellowship at the university level	2019–2020
<b>Bradford &amp; Diane Smith Fellowship</b> , awarded to a department-nominated Ph.D. candidate	2018–2019
<b>Kokes Award</b> by the North American Catalysis Society (NACS)	2017
<b>Neil &amp; Jo Bushnell Fellowship</b> , awarded to one department-nominated Ph.D. candidate	2017–2018
<b>Sundback Graduate Fellowship</b> , awarded to one department-nominated Ph.D. candidate	2016–2017
<b>Institute Merit Prizes</b> , the best academic record across the department at IIT Madras	2014 & 2012
<b>Indian Research Internship Program Scholarship</b> , awarded to 23 students in India	2013
<b>Merit Certificate</b> - Indian National Maths Olympiad, awarded to 50 students in India	2010

## SKILLS

<b>Programming Languages</b>	Proficient: Python, MATLAB, C, C++, Knowledgeable: Bash, SQL, Java, C#
<b>Deep Learning Frameworks</b>	TensorFlow, DeepChem, PyTorch
<b>Materials Simulation</b>	DFT packages (GPAW, VASP), Molecular Dynamics (LAMMPS)
<b>Tools</b>	L <sup>A</sup> T <sub>E</sub> X, Unix, Git

## RESEARCH EXPERIENCE

**Carnegie Mellon University**, Ph.D. candidate (Aug 2015 – present)

Pittsburgh, PA

- **Modeling Li-ion Battery Reliability**: Predictions of battery reliability metrics such as end-of-discharge and end-of-life.
  - Accurately predict the end-of-discharge for aged batteries, and predict the end-of-life of a battery as a function of anticipated usage through an electrochemistry-based aging model.
  - Learnt the evolution over cycles of important aging parameters from experimental discharge curves at different temperatures of operation from our industry collaborators.
- **Engineering the Solid Electrolyte Interface (SEI) in Li-Metal Batteries**: Developed a detailed mechanistic understanding of two high performing electrolytes, Fluoroethylene Carbonate (FEC) and Difluoroethylene Carbonate (DFEC), to demonstrate that minimal structural variations can lead to different decomposition products, and thereby the nature of the SEI.
  - We found different initial bond-breaking mechanisms between FEC and DFEC using density functional theory (DFT) calculations.
  - We suggested that the formation of oligomers instead of long polymers may lead to better SEI compactness.

- **Computational Material Design for Metal-Oxygen Batteries:** Identified promising cathode materials that can enable rechargeable chemistry (mediated through lithium peroxide) within lithium-oxygen batteries.
  - Performed material screening from a range of transition metals to enable selectivity towards lithium peroxide as the discharge product at the cathode.
  - Performed density functional theory (DFT) simulation to assess the energetics associated with various noble metals that can be used for oxygen reduction with lithium ions.
- **Electrolyte Design for Electrochemical Synthesis:** Implemented a data-driven approach based on deep learning to identify novel proton sources that enable electrochemical ammonia synthesis, a cutting-edge alternative to the emissions-intensive Haber-Bosch process.
  - Performed rigorous model selection for robust predictions in the "small-data" regime of machine learning.
  - Developed a framework involving molecular featurization (from SMILES IDs) of proton sources followed by a multi-task neural network to predict solvatochromatic parameters, fed into a trained Leisen-Reimer regression tree to predict ammonia yields.
  - Partnered with my experimental collaborators and identified at least 3 new tested proton sources that surpass the yields of the few known modest proton sources.
- Identified through the inversion of machine learning models (trained on atomic-scale simulation results) the precise nature of active sites where the oxygen reduction reaction occurs on transition metal sulfide materials.
  - Developed a relationship between the atomic-scale structure of the catalyst site and the performance, for which a neural network model and a k-nearest neighbors regression model both perform equally well.
  - Inverted the relationship to identify the optimal nature (3 sulfur nearest neighbors for Ni-S) of active sites.
  - Partnered with experimental collaborators to successfully test the identify optimal catalytic sites. Guiding experimental collaborators to synthesize and test other promising transition metal sulfides.
- Mentored multiple (6 total) students at the undergraduate level and the graduate level.

## INDUSTRY EXPERIENCE

**MRF Tires**, Intern (Jan 2014 – Jan 2015)

Chennai, India

- Developed a robust methodology for material property (hyperelasticity) characterization of carbon-filled rubbers using the eigenfunction virtual fields method.
- Carried out experiments with multiple relaxation steps for obtaining hyperelastic parameters; captured full-field strain data using 3D Digital Image Correlation (3D-DIC). Designed and optimized a novel specimen to obtain heterogeneous strains using a planar test setup.

## PUBLICATIONS

[Google Scholar page](#): [Citation Metrics](#): [h-index: 7](#), [i10-index: 6](#), [total citations: 142](#)

*Material Design for Next-Generation Batteries:*

Y. Zhang, **D. Krishnamurthy**, and V. Viswanathan, "Engineering Solid Electrolyte Interphase Composition by Assessing Decomposition Pathways of Fluorinated Organic Solvents in Lithium Metal Batteries" [J. Electrochem. Soc.](#) **167**, 070554 (2020).

A. Lee<sup>†</sup>, **D. Krishnamurthy**<sup>†</sup>, and V. Viswanathan, "Exploring MXenes as Cathodes for Non Aqueous Lithium Oxygen Batteries: Design Rules for Selectively Nucleating Li<sub>2</sub>O<sub>2</sub>." [ChemSusChem](#) **11**, 1911 (2018).

A. Khetan, **D. Krishnamurthy**, and V. Viswanathan. "Towards Synergistic Electrode-Electrolyte Design Principles for Nonaqueous Li-O<sub>2</sub> batteries." [Top. Curr. Chem](#) **376**, 11 (2018).

**D. Krishnamurthy**, H. A. Hansen, and V. Viswanathan, "Universality in Nonaqueous Alkali Oxygen reduction on Metal Surfaces: Implications for Li-O<sub>2</sub> and Na-O<sub>2</sub> Batteries." [ACS Energy Lett.](#) **94**, 162 (2016).

*Electrocatalysis for Energy Conversion Devices:*

D.S. Roman<sup>†</sup>, **D. Krishnamurthy**<sup>†</sup>, R. Garg, H. Hafiz, N.T. Nuhfer, V. Viswanathan, and T. Cohen-Karni, "Engineering Three-Dimensional (3D) Out-of-Plane Graphene Edge Sites for Highly-Selective Two-Electron Oxygen Reduction Electrocatalysis." [ACS Catal.](#) **10**, 1993 (2020)

Y. Kim, S. Xu, J. Park, A. Lal Dadlani, O. Vinogradova, **D. Krishnamurthy**, M. Orazov, D. Lee, S. Dull, H. Han, Z. Wang, T. Graf, T. D. Schladt, J. E. Mueller, R. Sarangi, R. Davis, V. Viswanathan, D. Higgins, T. F. Jaramillo, F. Prinz, "Atomic Layer Deposition Prepared Platinum-Titanium Alloys for Oxygen Reduction Reaction" (submitted)

**D. Krishnamurthy**, V. Sumaria, and V. Viswanathan, "Quantifying Robustness of DFT Predicted Pathways and Activity Determining Elementary Steps for Electrochemical Reactions." *J. Chem. Phys.* **150**, 041717 (2019)

G. Houchins<sup>†</sup>, **D. Krishnamurthy**<sup>†</sup>, and V. Viswanathan, "The Role of Uncertainty Quantification and Propagation in Accelerating the Discovery of Electrochemical Functional Materials." *MRS Bull.* **44**, 204 (2019)

O. Vinogradova, **D. Krishnamurthy**, V. Pande, and V. Viswanathan, "Quantifying Confidence in Density Functional Theory Predicted Surface Pourbaix Diagrams at Solid-Liquid Interfaces and its Implications for Electrochemical Processes." *Langmuir* **34**, 12259 (2018)

V. Sumaria, **D. Krishnamurthy**, and V. Viswanathan, "Quantifying Confidence in DFT Predicted Surface Pourbaix Diagrams and Associated Reaction Pathways for Chlorine Evolution." *ACS Catal.* **8**, 9024 (2018).

**D. Krishnamurthy**<sup>†</sup>, V. Sumaria<sup>†</sup>, and V. Viswanathan, "Maximal predictability approach for identifying the right descriptors for electrocatalytic reactions." *J. Phys. Chem. Lett.* **9**, 588 (2018).

B. Yan<sup>†</sup>, **D. Krishnamurthy**<sup>†</sup>, C. H. Hendon, S. Deshpande, Y. Surendranath, and V. Viswanathan, "Surface Restructuring of Nickel Sulfide Generates Optimally Coordinated Active Sites for Oxygen Reduction Catalysis." *Joule* **1**, 600 (2017). [Highlight Article Link](#)

#### *Machine Learning Driven Material Design:*

**D. Krishnamurthy**, H. Weiland, A.B. Farimani, E. Anton, J. Green, and V. Viswanathan, "Accelerating Energy Materials Discovery and Optimization through Machine Learning based Approaches." *ACS Energy Lett.* **4**, 187 (2018)

(equally contributing authors<sup>†</sup>)

#### PATENT

Y-M Chiang, V. Viswanathan, L. Li, V. Pande, **D. Krishnamurthy**, Z. Ahmad, and W. H. Woodford. "Lithium Metal Electrodes and Batteries Thereof." *U.S. Patent 20170288281*, *WO Patent 2017176936*, October 5, 2017.  
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