

## Kevin Kaichuang Yang

Senior Researcher, BioML  
Microsoft Research New England

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<http://yangkky.github.io>

### EDUCATION

2018	PhD in Chemical Engineering	California Institute of Technology
2011	B.S. in Chemical Engineering	Ohio State University

### INVITED TALKS

- Biomedical Informatics Seminar, Columbia University. October 2022.
- Bioinforange. October 2022.
- Merck. August 2022.
- Journal Club, OpenBioML. May 2022.
- Genomics Seminar, UC Riverside. May 2022.
- AI for Drug Discovery Seminar, University of Washington. May 2022.
- Bioinformatics Student Symposium, Boston University. May 2022.
- Molecular Maker Lab Institute Conference, UIUC. March 2022.
- AI4Science Seminar, Chalmers University. March 2022.
- Machine learning seminar, IBM. December 2021.
- Protein Engineering Congress Global. October 2021.
- GSK Data Forum. Feb 2021.
- Boğaziçi University Biotech Conference. Jan 2021.
- Ohio State University Society for Biological Engineering. Nov 2020.
- Janelia Research Center. May 2019.
- Gray-Hill Lecture, Occidental College, June 2018.

### PUBLICATIONS

#### Peer-Reviewed Papers

[\*co-first authors]

17. Exploring evolution-based &-free protein language models as protein function predictors. Mingyang Hu, Fajie Yuan, **Kevin K. Yang**, Fusong Ju, Jin Su, Hui Wang, Fei Yang, Qiuyang Ding. NeurIPS 2022, to appear.
16. Randomized gates eliminate bias in sort-seq assays. Brian L. Trippe, Buwei Huang, Erika A. DeBenedictis, Brian Coventry, Nicholas Bhattacharya, **Kevin K. Yang**, David Baker, Lorin Crawford. Protein Science, 2022.
15. Evolutionary velocity with protein language models. Brian L. Hie, **Kevin K. Yang**, and Peter S. Kim. Cell Systems, 2022. 10.1016/j.cels.2022.01.003
14. Machine learning modeling of family wide enzyme-substrate specificity screens. Samuel Goldman, Ria Das, **Kevin K Yang**, Connor W Coley. PLoS computational biology, 2022. 10.1371/journal.pcbi.1009853

13. A topological data analytic approach for discovering biophysical signatures in protein dynamics. Wai Shing Tang, Gabriel Monteiro da Silva, Henry Kirveslahti, Erin Skeens, Bibo Feng, Timothy Sudijono, **Kevin K. Yang**, Sayan Mukherjee, Brenda Rubenstein, Lorin Crawford. PLoS computational biology, 2022. 10.1371/journal.pcbi.1010045
12. Adaptive machine learning for protein engineering. Brian L. Hie and **Kevin K. Yang**. Current Opinion in Structural Biology, 2022. 10.1016/j.sbi.2021.11.002
11. FLIP: Benchmark tasks in fitness landscape inference for proteins. Christian Dallago, Jody Mou, Kadina E. Johnston, Bruce J. Wittmann, Nicholas Bhattacharya, Samuel Goldman, Ali Madani, **Kevin K. Yang**. NeurIPS 2021 Datasets and Benchmarks Track. 10.1101/2021.11.09.467890
10. Protein sequence design with deep generative models. Zachary Wu, Kadina E. Johnston, Frances H. Arnold, and **Kevin K. Yang**. Current Opinion in Chemical Biology, 2021. 10.1016/j.cbpa.2021.04.004
9. Learned embeddings from deep learning to visualize and predict protein sets. Christian Dallago, Konstantin Schütze, Michael Heinzinger, Tobias Olenyi, Maria Littmann, Amy X. Lu, **Kevin K. Yang**, Seonwoo Min, Sungroh Yoon, James T. Morton, Burkhard Rost. Current Protocols, May 2021. 10.1002/cpz1.113
8. Signal Peptides Generated by Attention-Based Neural Networks. Zachary Wu, **Kevin K. Yang**, Michael J. Liszka, Alycia Lee, Alina Batzilla, David Wernick, David P. Weiner, and Frances H. Arnold. ACS Synthetic Biology, 10 July 2020. 10.1021/acssynbio.0c00219
7. Machine learning-guided channelrhodopsin engineering enables minimally-invasive optogenetics. Bedbrook CN, **Yang KK**, Robinson JE, Gradinaru V, Arnold FH. Nature Methods, October 14, 2019. 10.1038/s41592-019-0583-8.
6. Machine-learning-guided directed evolution for protein engineering. **Yang KK**, Wu Z, Arnold FH. Nature Methods, July 15, 2019. 10.1038/s41592-019-0496-6.
5. Batched stochastic Bayesian optimization via combinatorial constraints design. **Yang KK**, Chen Y, Lee A, Yue Y. AISTATS 2019. arxiv.
4. The Generation of Thermostable Fungal Laccase Chimeras by SCHEMA-RASPP Structure-Guided Recombination in Vivo. Mateljak I, Rice A, **Yang KK**, Tron T, Alcalde M. ACS Synthetic Biology, March 21, 2019. 10.1021/acssynbio.8b00509
3. Learned protein embeddings for machine learning. **Yang KK**, Wu Z, Bedbrook CN, Arnold FH. Bioinformatics. 23 March 2018. 10.1093/bioinformatics/bty178.
2. Machine learning to predict eukaryotic expression and plasma membrane localization of engineered integral membrane proteins. Bedbrook CN\*, **Yang KK\***, Rice AJ, Gradinaru V, Arnold FH. PLOS Computational Biology 13(10): e1005786 (2017). 10.1371/journal.pcbi.1005786.
1. Structure-Guided SCHEMA Recombination Generates Diverse Chimeric Channelrhodopsins. C. N. Bedbrook, A. J. Rice, **K. K. Yang**, X. Ding, S. Chen, E. M. LeProust, V. Gradinaru, F. H. Arnold. Proceedings of the National Academy of Sciences 114, E2624-E2633 (2017). 10.1073/pnas.170026911.

## Preprints

4. Protein structure generation via folding diffusion. Kevin E. Wu, **Kevin K. Yang**, Rianne van den Berg, James Y. Zou, Alex X. Lu, Ava P. Amini

3. Deep self-supervised learning for biosynthetic gene cluster detection and product classification. C Rios-Martinez, N Bhattacharya, AP Amini, L Crawford, **KK Yang**.
2. Masked inverse folding with sequence transfer for protein representation learning. **Kevin K. Yang**, Niccolò Zanichelli, Hugh Yeh.
1. Convolutions are competitive with transformers for protein sequence pretraining. **Kevin K. Yang**, Alex X. Lu, Nicolo Fusi.

## **HONORS, AWARDS, AND FELLOWSHIPS**

- |      |   |
|------|---|
| 2017 | Caltech Chemistry and Chemical Engineering Teaching Assistantship Award |
| 2015 | Caltech Biotechnology Leadership Program                                |
| 2011 | National Science Foundation Graduate Research Fellowship                |

## **TEACHING AND MENTORING**

### **Teaching Assistantships**

- |      |   |
|------|---|
| 2018 | Caltech ChE/BE 163 Introduction to biomolecular engineering |
| 2016 | Caltech ChE/BE 163 Introduction to biomolecular engineering |
| 2015 | Caltech ChE 101 Chemical reaction engineering               |

## **EXPERIENCE**

### **Senior Researcher, BioML**

*Microsoft Research New England, Cambridge, MA (April 2020 – Present)*

- Develop methods for protein sequence pretraining and generation
- Develop benchmarks for protein function prediction
- Lead project teams that include interns, data scientists, and researchers

### **Machine Learning Scientist**

*Generate Biomedicines, Cambridge, MA (January 2019 – April 2020)*

- Develop methods for quantifying the uncertainty of protein function predictions
- Develop methods for model-based optimization of protein function
- Build and train a language model for proteins that can be fine-tuned for specific prediction tasks.

### **Graduate Research Assistant, Professor Frances Arnold's Group**

*California Institute of Technology, Pasadena, CA (August 2014 – December 2018)*

- Used Gaussian process models ([github.com/yangkky/gpmodel](https://github.com/yangkky/gpmodel)) to design channelrhodopsins with improved properties
- Designed embedded representations of protein sequences based on doc2vec to streamline machine learning pipelines ([github.com/fha\\_lab/embeddings\\_reproduction](https://github.com/fha_lab/embeddings_reproduction))
- Built neural machine translation models on PyTorch to predict signal peptides from their corresponding mature protein sequences

### **Computational Intern**

*Ambry Genetics, Aliso Viejo, CA (June 2017 – September 2017)*

- Developed and implemented neural network models in Keras and PyTorch to predict outcomes of genetic variation by transferring information across paralogous proteins
- Incorporated model into a pipeline that finds paralogs for variants of interest and then uses paralogs and model to predict variant outcomes