

# Xikun Zhang

443 Via Ortega, Stanford, CA 94305 · [xikunz2@cs.stanford.edu](mailto:xikunz2@cs.stanford.edu)

## RESEARCH FOCUSES

---

I generally focus on **deep learning**, **generative AI**, **large language models**, **computer vision**, and **multimodal models**. Some of my already published work includes

1. Graph machine learning ([JODIE](#), [GreaseLM](#), [DRAGON](#), [AutoPath](#))
2. Understanding large language models ([NumBERT](#))
3. Common-sense reasoning ([NumBERT](#), [GreaseLM](#), [DRAGON](#))
4. Retrieval augmented models for knowledge-intensive tasks ([GreaseLM](#), [DRAGON](#))
5. Foundation models ([Foundation Models](#))
6. AI in biomedicine ([Virtual Cell](#))

## GOOGLE SCHOLAR CITATIONS

---

**6217** (Up to Jan 9, 2025)

## FIRST-AUTHOR PUBLICATIONS

---

1. **Xikun Zhang**, Antoine Bosselut, Michihiro Yasunaga, Hongyu Ren, Percy Liang, Christopher D Manning, and Jure Leskovec, “*GreaseLM: Graph REASONing Enhanced Language Models for Question Answering*,” ICLR 2022 (**Spotlight**, Top 5%) [[paper](#)]
2. **Xikun Zhang\***, Deepak Ramachandran\*, Ian Tenney, Yanai Elazar, and Dan Roth, “*Do Language Embeddings Capture Scales?*” Findings of EMNLP 2020 & EMNLP BlackboxNLP workshop 2020 [[paper](#)]

## OTHER PUBLICATIONS

---

3. Charlotte Bunne, ..., **Xikun Zhang**, ..., and Stephen R. Quake, “*How to Build the Virtual Cell with Artificial Intelligence: Priorities and Opportunities*,” Cell (Dec 2024) [[paper](#)]
4. Jiayuan Mao\*, Xuelin Yang\*, **Xikun Zhang**, Noah Goodman, and Jiajun Wu, “*CLEVRER-Humans: Describing Physical and Causal Events the Human Way*,” NeurIPS Datasets and Benchmarks Track 2022 [[paper](#)]
5. Michihiro Yasunaga, Antoine Bosselut, Hongyu Ren, **Xikun Zhang**, Christopher D Manning, Percy Liang\*, and Jure Leskovec\*, “*Deep Bidirectional Language-Knowledge Graph Pretraining*,” NeurIPS 2022 & AAAI DLG workshop 2023 (**Best Paper Award**) [[paper](#)]
6. Rishi Bommasani, ..., **Xikun Zhang**, ..., and Percy Liang (116 authors), “*On the opportunities and risks of foundation models*,” arXiv 2021 [[paper](#)]
7. Srijan Kumar, **Xikun Zhang**, and Jure Leskovec, “*Predicting Embedding Trajectories for Temporal Interaction Networks*,” KDD 2019 (**Oral**, Top 9%) [[paper](#)]
8. Carl Yang, Mengxiong Liu, Frank He, **Xikun Zhang**, Jian Peng, and Jiawei Han, “*Similarity Modeling on Heterogeneous Networks via Automatic Path Discovery*,” ECML-PKDD 2018 [[paper](#)]
9. Qi Li\*, Meng Jiang\*, **Xikun Zhang**, Meng Qu, Timothy Hanratty, Jing Gao, and Jiawei Han, “*TruePIE: Discovering Reliable Patterns in Pattern-Based Information Extraction*,” KDD 2018 (**Long Presentation**, Top 11%) [[paper](#)]

## EDUCATION

---

### Stanford University

Ph.D. in Computer Science

Sep 2020 - present

M.S. in Computer Science

Sep 2020 - Aug 2023

- Research interests: machine learning, large language models, computer vision, common-sense reasoning, multimodal models, AI agents and tool use, computational biology
- GPA: 4.109/4.3
- Honors: Stanford Data Science Scholar (merit-based, 16 selected out of hundreds of applicants at Stanford)

### University of Illinois at Urbana-Champaign

B.S. in Computer Science

Aug 2016 - May 2019

- GPA: 3.98/4.00, major GPA: **4.00/4.00**
- Honors: Bronze Tablet (highest undergraduate honor, **top 3%** in college, final year), James Scholar (top 5% in department, every semester), Dean’s List (top 5% in college, every semester)
- Classes: **A+**’s in Machine Learning (**top 1** in a class of 300 people), Advanced Information Retrieval, Introduction to Data Mining, and Random Processes

## Nanjing University

B.S. in Electronic Information Science (Transferred out)

Aug 2015 - Jul 2016

- GPA: 3.76/4.00, major GPA: 3.84/4.00
- Rank: 1/203

## RESEARCH EXPERIENCES

---

### AI Resident, Chan Zuckerberg Initiative

Jan 2024 - present

#### AI-powered virtual cell models [Cell]

- Among one of the first two AI resident academic groups at CZI, with the goal of building AI-powered virtual cells to help scientists explore the molecular underpinnings of human health and disease.

### Graduate Research Assistant, Stanford Lundberg Lab, Prof. Emma Lundberg

Dec 2022 - present

#### Comprehensive benchmark of amino-acid-sequence-based protein location predictors

- Constructed a new comprehensive benchmark dataset with validated protein subcellular location labels from the Human Protein Atlas, OpenCell and UniProt.
- Evaluated all existing multi-label protein localization predictors including all state-of-the-art protein language models (ProtT5, ESM2, ESM3, etc) on the new benchmark.

#### Generative models of spatial proteomics images

- Proposed and built a novel deep learning model based on latent diffusion models, protein language models and variational autoencoders to generate spatial proteomics images at the subcellular resolution.
- The method enabled in-silico profiling of subcellular protein localization to understand disease mechanisms and screen drugs in a high-throughput manner.

### Graduate Research Assistant, Stanford Newman Lab, Prof. Aaron Newman

Dec 2022 - present

#### Cellular deconvolution at the single-cell resolution

- Proposed and built a novel computational model that combines deep learning with the DeepSHAP feature attribution method to deconvolve bulk tissue transcriptomics profiles into single-cell RNA sequencing data.
- The method enabled scaling up the power of single-cell biology in the clinic and downstream development of next-generation diagnostics and personalized treatment of diseases like cancer.
- Cleaned, preprocessed, normalized, scaled, analyzed and visualized different types of omics datasets including single-cell RNA-seq (10x Chromium, SMART-seq), bulk RNA-seq, CITE-seq (RNA+protein) and Flow Cytometry data using bioinformatics packages including Seurat and Scanpy.

### Graduate Research Assistant, Stanford Kundaje Lab, Prof. Anshul Kundaje

Aug 2021 – Nov 2022

#### Chromosome scale deep learning models of gene expression

- Proposed graph neural networks to predict cell-type resolved gene expression from DNA sequences spanning megabase-scale regulatory domains.
- Interpreted the trained model using the DeepSHAP feature attribution method to infer cis-regulatory syntax of regulatory elements (e.g. transcription-factor binding motifs) and long-range regulatory interactions (e.g. enhancer-gene linking).

#### DNA-sequence-based deep learning models of human epigenomic data at base resolution

- Developed novel deep learning architectures that map 1-5kb DNA sequence inputs to base-resolution TF ChIP-seq profiles jointly across different kinds of TFs and cell types in one model.

### Graduate Research Assistant, Stanford Bassik Lab, Prof. Michael Bassik

Aug - Nov 2021

#### Synthetic lethality prediction using the gene coessentiality network

- Designed new graph-neural-network-based deep learning architectures to learn gene embeddings using the gene coessentiality network.
- Used the learned model and the gene embeddings to predict which gene pairs are synthetic lethal.

### Graduate Research Assistant, Stanford Network Analysis Project group, Prof. Jure Leskovec

Apr - Oct 2021

#### Blueprint of foundation models

- Wrote a whitepaper where we created the now widely adopted term “foundation model” to refer to models trained on broad data that can be adapted to a wide range of downstream tasks and provided a thorough account of the opportunities and risks of foundation models, ranging from their capabilities and technical principles to their applications and societal impacts.
- Was a core contributor to the Vision section of the whitepaper, where we discussed the potential of foundation models for vision to distill raw, multimodal sensory information into visual knowledge.

Augmenting large language models with external knowledge bases [ICLR 2022 (**Spotlight**)]

- Proposed GreaseLM, a new question-answering model that enhances pretrained language models with graph neural networks reasoning over an external knowledge graph and lets the two components interact with each other over multiple layers of modality interaction operations.
- Showed that GreaseLM boosted performance on answering questions in domains that need external knowledge, like commonsense reasoning, or biomedical question answering.

Scaling language-knowledge foundation models [NeurIPS 2022 & AAAI DLG workshop 2023 (**Best Paper Award**)]

- Proposed Dragon, a self-supervised method to pretrain a deeply joint language-knowledge foundation model (GreaseLM) from text and knowledge graphs at scale.
- Pretrained this model by unifying two self-supervised reasoning tasks, masked language modeling and knowledge graph link prediction.
- Showed that Dragon outperforms existing language models and language model+knowledge graph models on diverse downstream tasks including question answering across general and biomedical domains, with +5% absolute gain on average.

**Graduate Research Assistant, Stanford CogAI group & Stanford CoCoLab**, Prof. Jiajun Wu & Prof. Noah Goodman  
Aug 2020 - Mar 2021

Constructing a video reasoning dataset for causal judgment [NeurIPS Datasets and Benchmarks Track 2022]

- Constructed the CLEVRER-Humans benchmark, a video reasoning dataset for causal judgment of physical events with human labels and demonstrated its challengingness for existing question-answering models.
- Employed two techniques to improve data collection efficiency: first, a novel iterative event cloze task to elicit a new representation of events in videos, which we term Causal Event Graphs; second, a data augmentation technique based on neural language generative models.

**AI Resident, Google Research**, Michael Collins, David Reitter

Oct 2019 - Aug 2020

Multimodal foundation models

- Pretrained Transformer-based multimodal models via self-supervised learning on large-scale image caption datasets (scale of millions) on hundreds of TPUs, which powers many downstream vision-and-language tasks, including visual question answering.
- Applied this model to significantly improve matching between dish photos and menu items on Google Maps restaurants through jointly modeling dish photos, user reviews, and menus.
- Because we trained the model with a massive dataset, on a massive number of TPUs on a super complex Google internal infrastructure, I wrote the code in a highly optimized way so that the computational resources were efficiently utilized and jobs finished in a reasonable amount of time. Models were implemented using Python and large-scale parallel data processing was done using C++ and the Flume framework (a more advanced version of MapReduce).

**Research Intern, Google Research**, Dan Roth, Ian Tenney, Deepak Ramachandran

Jun - Sep 2019

Understanding what large language models learn [Findings of EMNLP 2020 & EMNLP BlackboxNLP workshop 2020]

- Trained linear prediction models on top of frozen large language models like BERT and ELMo to understand to what extent they encode information about the scalar magnitudes of real-world objects.
- Identified contextual information in pre-training and numeracy as two key factors affecting the performance of this task.
- Pretrained a new BERT model with a way of canonicalizing numbers I came up with (called NumBERT) on the English Wikipedia (6.9 million articles, 4.7 billion words) and the Books corpus (11k books, 985 million words) across hundreds of TPUs and it achieves better performance than the original BERT.

**Research Assistant, Stanford Network Analysis Project group**, Prof. Jure Leskovec      May – Aug 2018  
Deep-learning-based recommender systems [KDD 2019 (Oral)]

- Proposed a recurrent-neural-network-based recommender system called JODIE
- Proposed the t-Batch algorithm which can create temporally coherent mini-batches of interactions, making JODIE highly scalable and 9 times faster than existing co-evolution models.
- Conducted experiments to show that JODIE achieved state-of-the-art results on two prediction tasks, future interaction prediction and state change prediction, on four different datasets containing up to 1.29 million interactions.

**Research Assistant, UIUC Data Mining Group**, Prof. Jiawei Han      Sep 2017 - Apr 2018  
Deep learning models on heterogenous graphs [ECML-PKDD 2018]

- Proposed an algorithm called Autopath that can automatically discover useful paths to find similar pairs of nodes using both structure and content information from a heterogeneous network by combining reinforcement learning and deep embedding into a novel semi-supervised joint learning framework.

Truthful text mining [KDD 2018 (Long Presentation)]

- Proposed a novel method, called TruePIE, that finds reliable text patterns to not only extract information related to a certain task based on the statistics of the patterns' individual contents (e.g., length, frequency), but also correct information from a text corpus.
- Adopted a self-training framework that repeats the training-predicting-extracting process to gradually discover more and more reliable patterns.

**Research Assistant, UIUC KnowEnG BD2K center**, Prof. Saurabh Sinha      May - Aug 2017  
Genomic knowledge network construction

- Automatically constructing the KnowEnG Knowledge Network, a massive heterogeneous network composed primarily of genes and their annotations as well as their mutual relationships, by identifying online relevant knowledge bases, and automatically cleaning and consolidating those disparate resources.
- Integrating and fusing heterogeneous, high-dimensional, and noisy biological data.

## **PROFESSIONAL ACTIVITIES**

---

- Top 5% teaching assistant across the Stanford computer science department during fall 2023 (Head TA in [CS224W](#) )
- Organizing a workshop "[Tools for assembling the cell: Towards the era of cell structural bioinformatics](#)" at the [Pacific Symposium on Biocomputing \(PSB\)](#) 2024
- Invited talk at Amazon AI Lab on the GreaseLM paper (Apr 2022)
- Conference reviewer: AAAI, AACL-IJCNLP
- Journal reviewer: BMC Bioinformatics, IEEE Transactions on Computational Social Systems
- Workshop reviewer: ACL CSRR workshop, CIKM FedGraph workshop, AAAI-GCLR workshop, AAAI-UDM workshop

## **SKILLS**

---

- Programming languages: Proficient in Python, C / C++, Java, OCaml, Javascript, R
- Machine learning frameworks: PyTorch, PyTorch Lightning, TensorFlow, Caffe, Keras, scikit-learn, XGBoost
- Bioinformatics packages: Seurat, Scanpy
- Web development technologies: HTML, CSS, Javascript, JQuery, Bootstrap, Python Flask