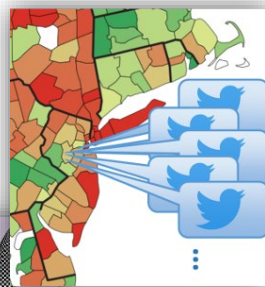
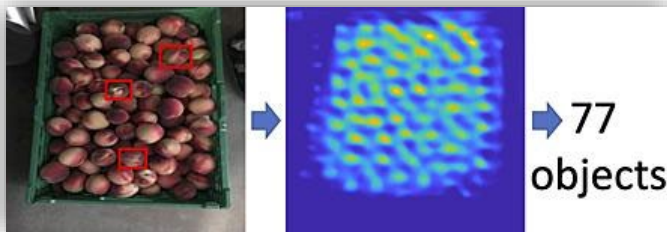
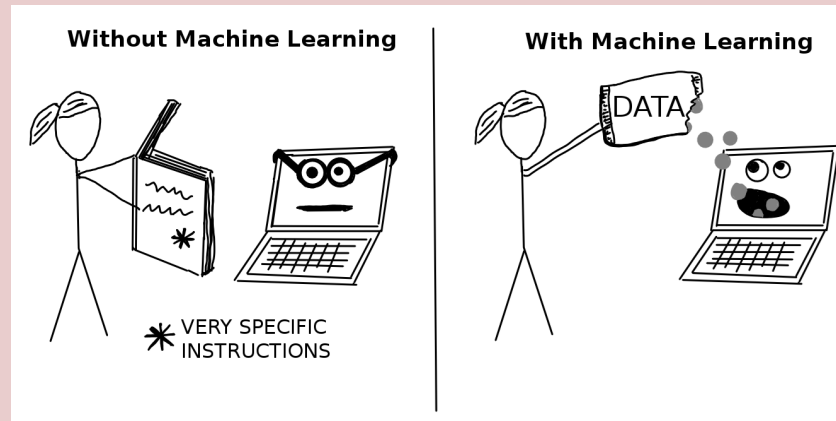


Theoretical Machine Learning @ Stony Brook

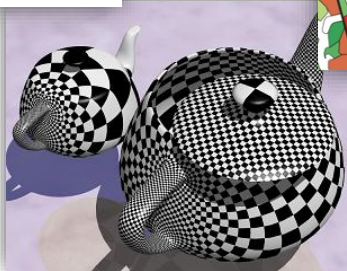
Yifan Sun
SBU-BNL AI Workshop
June 2024



- 0.0852
- 0.8794
- 0.1415
- 0.1996
- 0.4561
- 0.3556
- 0.7532
- 0.2703
- 0.6872
- 0.2623
- 0.3795
- 0.6451
- 0.2032
- 0.4075
- 0.5010
- 0.4783
- 0.9845
- 0.6314

certified fitness zumba gym
vicesession training class trainer
personal basicpotty sassylornipress workout
hahahahaha
haha
haha
btw xp awesome hahahahaha

valley forest total lakepark trailhike grand cave creeksfalls hiking river springs
hahahahaha
hahahahaha
hahahahaha
hahahahaha



Natural language



H. Andrew Schwartz

Assistant Professor

Large and scalable language analyses for psychological and health discovery; computational social science; natural language processing; lexical semantics;... More



Niranjan Balasubramanian

Assistant Professor

Natural Language Processing (NLP) and information retrieval.



Ritwik Banerjee

Research Assistant Professor

Natural Language Processing (NLP), Computational Linguistics, Information Extraction



Steven Skiena

Distinguished Teaching Professor, SUNY Empire Innovation Professor and Director, AI Institute

Algorithms, Computational Biology, Large-scale Text Analytics and Sentiment Analysis, Social Trends Analysis, Combinatorial



I.V. Ramakrishnan

Professor and Associate Dean for Strategic Initiatives

Artificial Intelligence, Computational Logic, Machine Learning/Computational Logic Combination, Information Retrieval, Computer Accessibility



Paul Fodor

Associate Professor of Practice

Artificial Intelligence, Natural Language Processing, Logic Programming, Complex Event Processing, Knowledge Representation for the Semantic Web, Active... More



Praveen Tripathi

Research Assistant Professor

Machine learning, Data mining, Spatio-temporal data analysis, Time series data analysis.

Mining and extraction



Dimitris Samaras

SUNY Empire Innovation Professor

Computer vision; machine learning; computational behavioral sciences; computer graphics; medical imaging; computational photography.



Michael Ryoo

SUNY Empire Innovation Associate Professor

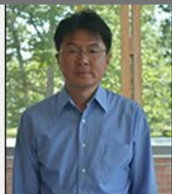
Michael's prime research interest is in the field of deep learning and its applications to computer vision and robotics.



Minh Hoai Nguyen

Associate Professor

Computer Vision; Machine Learning; Human Activity Recognition; Semantic Video



Xianfeng Gu

SUNY Empire Innovation Professor

Computational Conformal Geometry, Computer Graphics, Visualization, Computer Vision, Geometric Modeling, Networking, Medical Imaging, Digital Geometry... More



Yifan Sun

Assistant Professor

Convex and nonconvex optimization for machine learning and scientific computing

Computer vision



Zhaozheng Yin

SUNY Empire Innovation Associate Professor

Biomedical Image Analysis, Computer Vision, Machine Learning, Cyber-Physical Systems, Human-Robot Collaboration



Haibin Ling

SUNY Empire Innovation Professor

Computer Vision, Medical Image Analysis, Augmented Reality and Human-Computer Interaction.



Ting Wang

SUNY Empire Innovation Associate Professor

Computer Security and Machine Learning



Stanley Bak

Assistant Professor

Verification of Neural Networks, Cyber-Physical Systems, Formal Analysis of Hybrid Systems

Theoretical foundations and verification

@ SBU

Theoretical foundations and verification



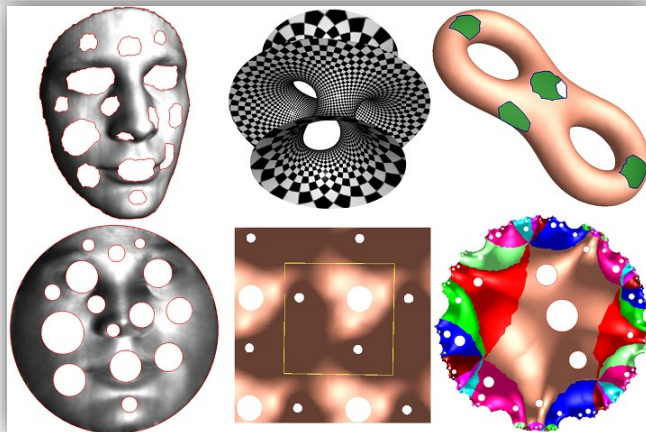
Xianfeng David Gu

SUNY Empire Innovation Professor
[Department of Computer Science](#)
[Department of Applied Mathematics](#)
Stony Brook University

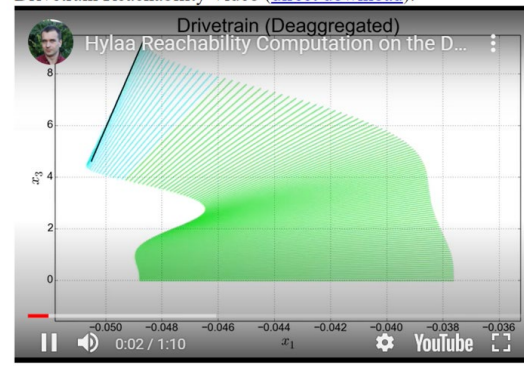
Room 147 New Computer Science Building
State University of New York at Stony Brook
Stony Brook, New York 11794-2424

Phone: (631) 632-1828 (Office)
Fax: (631) 632-8334
gu at cs.stonybrook.edu

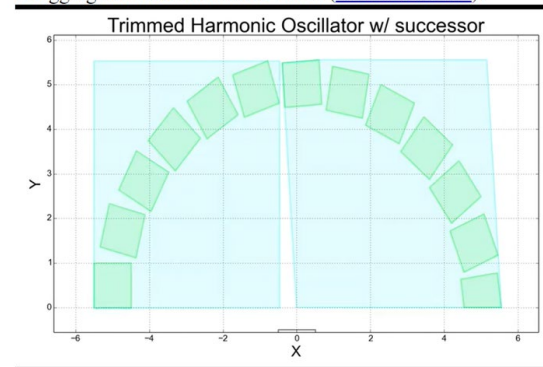
Director of 3D Scanning Laboratory
<http://www.cs.stonybrook.edu/~gu>



Drivetrain Reachability Video ([direct download](#)):



Deaggregation Demonstration Video ([direct download](#)):



Decision Intelligence & Emerging Networked Systems Laboratory

--- PI: Jian Li (jian.li.3@stonybrook.edu)



❑ **Decision Intelligence:** intersection of sequential decision making (DM) and AI/ML

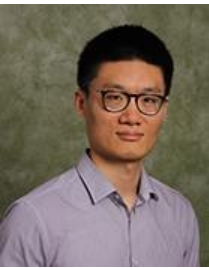
❑ Theory foundation and algorithmic solutions to DM in large-scale AI/ML and data science problems, including algorithm design and analysis, optimization, and implementation, using state-of-the-art mathematical techniques and system technology:

- ❑ Reinforcement learning (RL) and representation learning
 - ❑ RL for generative models, e.g., LLMs w/RL, diffusion models
- ❑ Robust/Adversarial and risk-aware online decision making (online learning, e.g., multi-armed bandits)
- ❑ Learning-augmented network optimization
- ❑ Distributed/federated/trustworthy learning and optimization

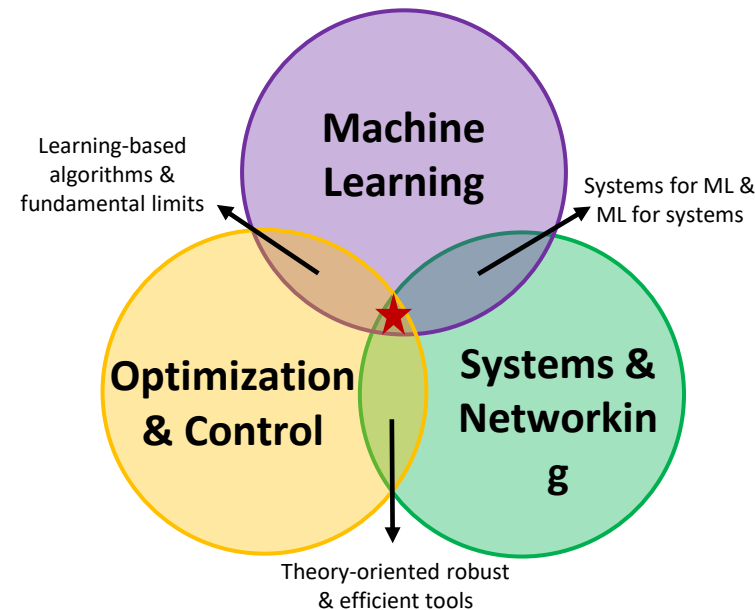
❑ **Emerging Networked Systems**

❑ Resource allocation in general centered on fundamental issues on optimality, scalability, and trustworthiness (e.g., communication-computation efficiency, resilience)

- ❑ AI for wireless networking: O-RAN, mmWave networks, etc
- ❑ Autonomous systems: robotics, swarm/drones control
- ❑ Edge/Cloud systems: serverless computing, content delivery
- ❑ Cyber-physical systems



★ **Decision Intelligence for Robust and Efficient NextG Data Systems**



- **Multiple funded positions for Ph.D. and/or MS students. Contract Prof. Li for details**
- **Collaboration/internship with industry research labs: AT&T, IBM, MERL, etc**



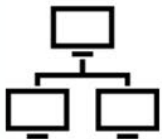
Thrust 1: Leveraging the advances in ML for cyber-security applications

Software



- Reverse engineering
- Code clone detection
- Fuzzing

Network

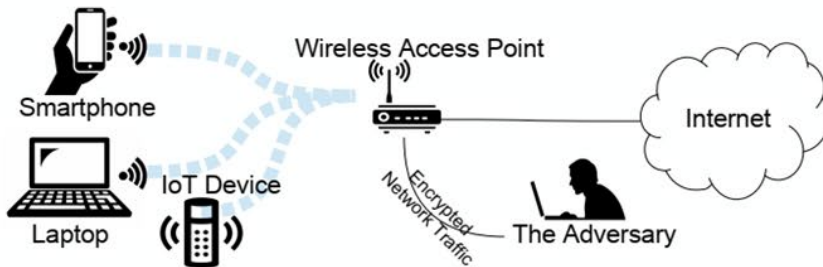


- Attack investigation
- Exploit detection

Web App



- App fingerprinting
- UI analysis



App fingerprinting over encrypted wireless traffic



ML → Security





Thrust 2: Investigating security & privacy issues arising in ML techniques

Security



- Adversarial robustness
- Poisoning robustness
- Backdoor robustness

Privacy

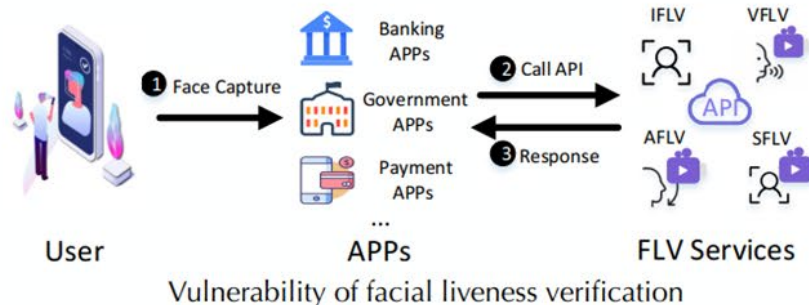


- Privacy inference
- Federated learning
- Differential privacy

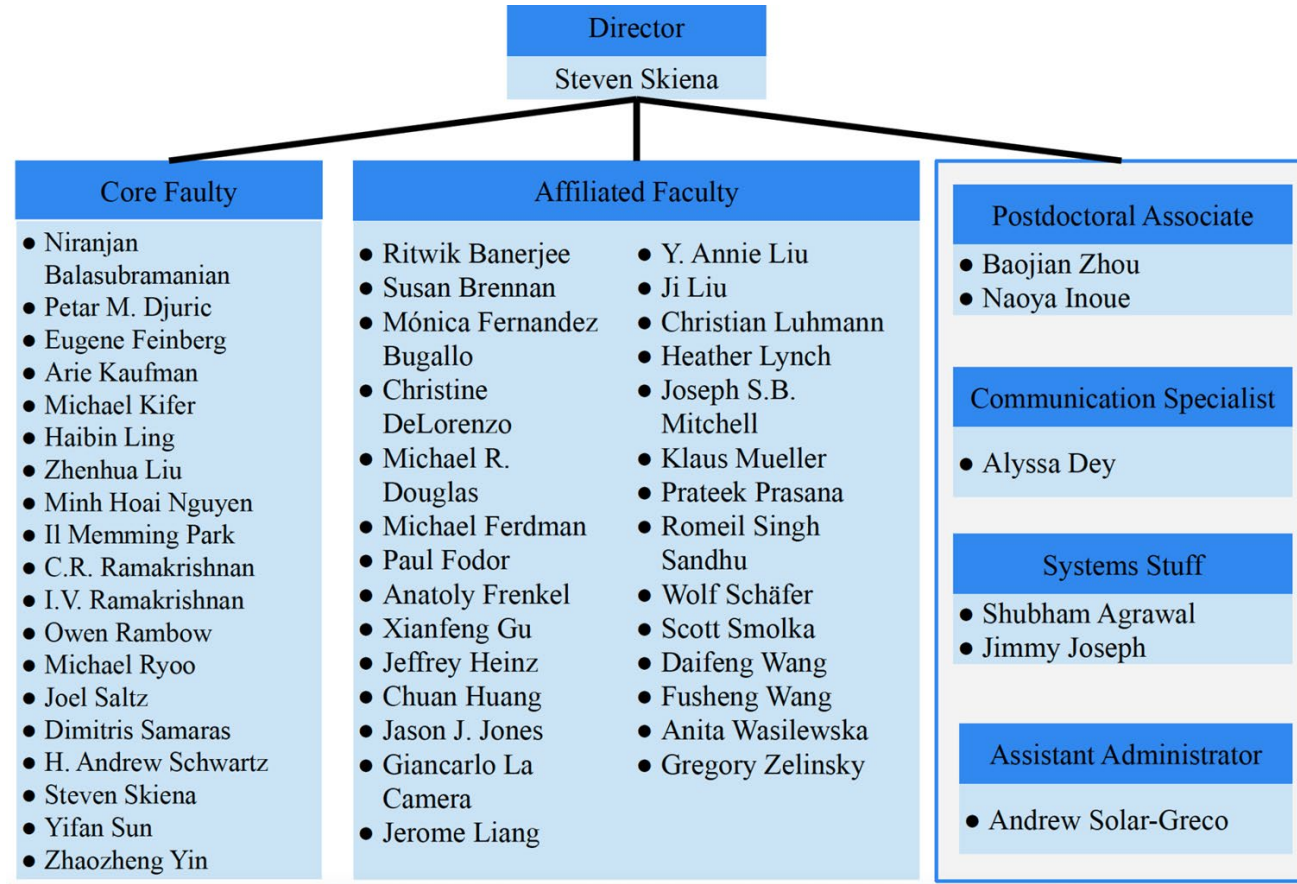
Transparency



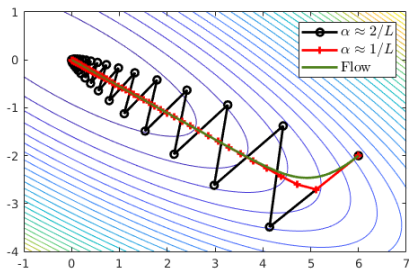
- Interpretation reliability
- Human in the loop



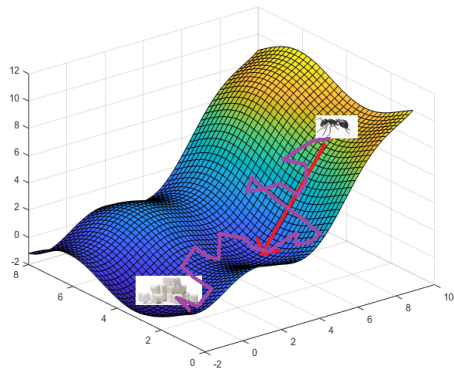
AI Institute



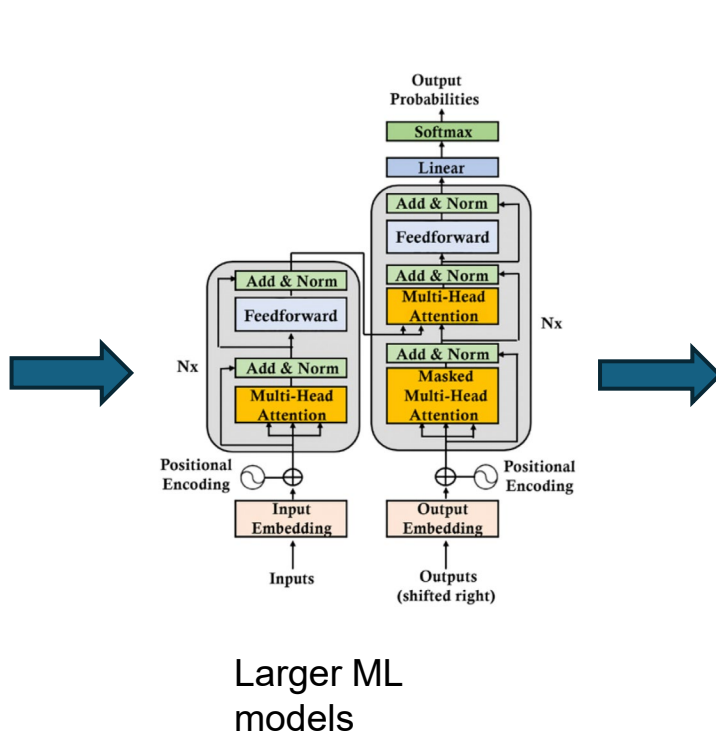
Optimization theory + machine learning



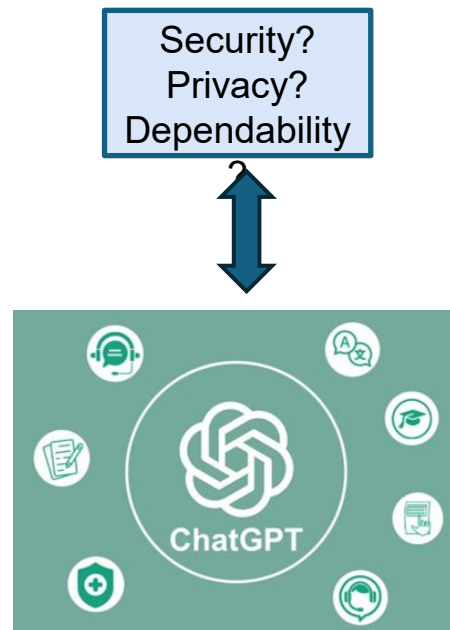
Optimization method behavior



Loss function behavior

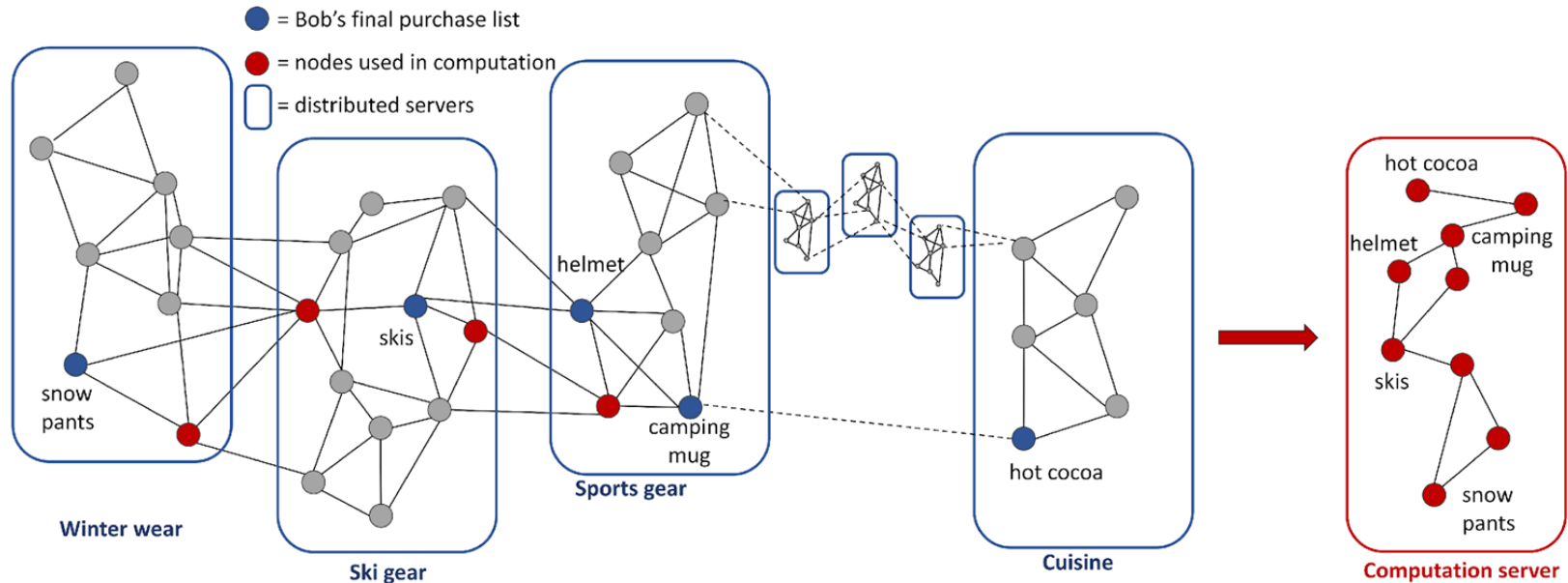


Larger ML models

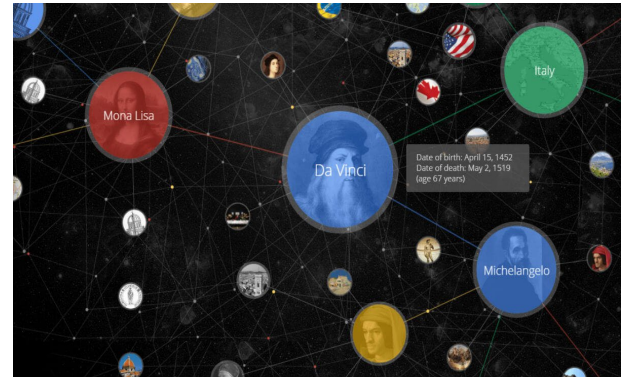


Modern day applications

Machine learning over graphs



Machine learning over graphs



- Amazon's product graph: 12 million products
- Google knowledge graph: 500 million entities
- Paypal's fraud detection: 500 million users

These graphs don't fit on one server!

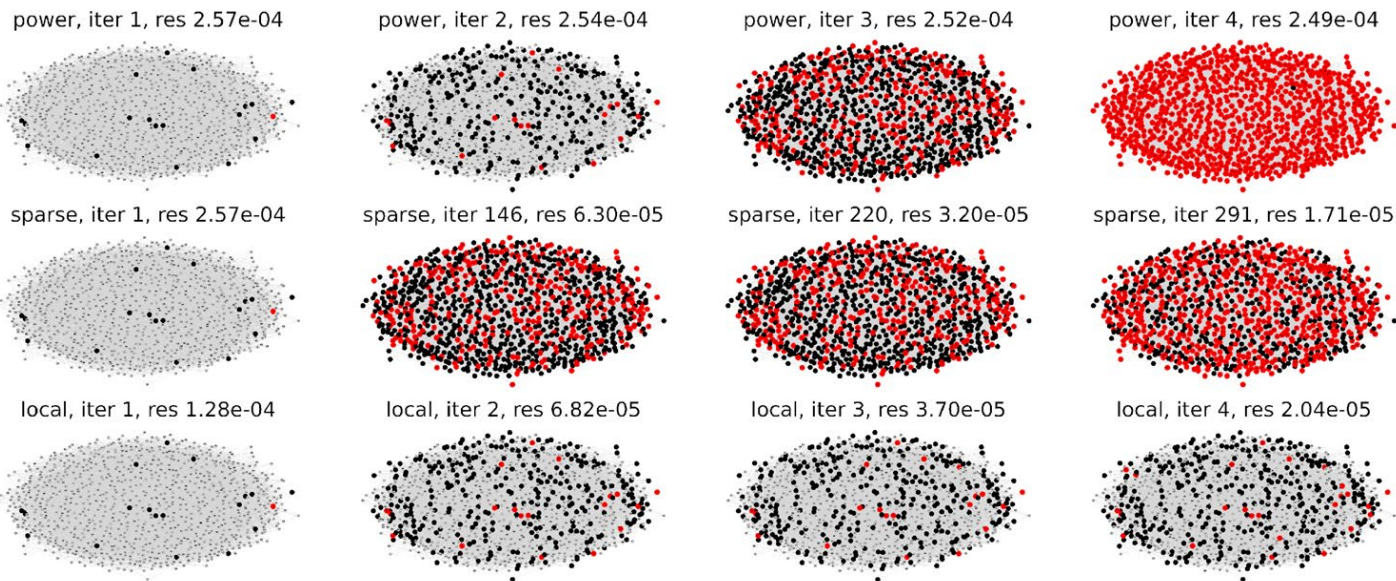
Machine learning over graphs



- Cybersecurity or web services: Internet consists of billions of websites, connects billions of devices, all heterogeneous nodes
- Complex scenarios, many independent small entities

We want it now, we want it fast

Fast graph solvers with low memory footprint



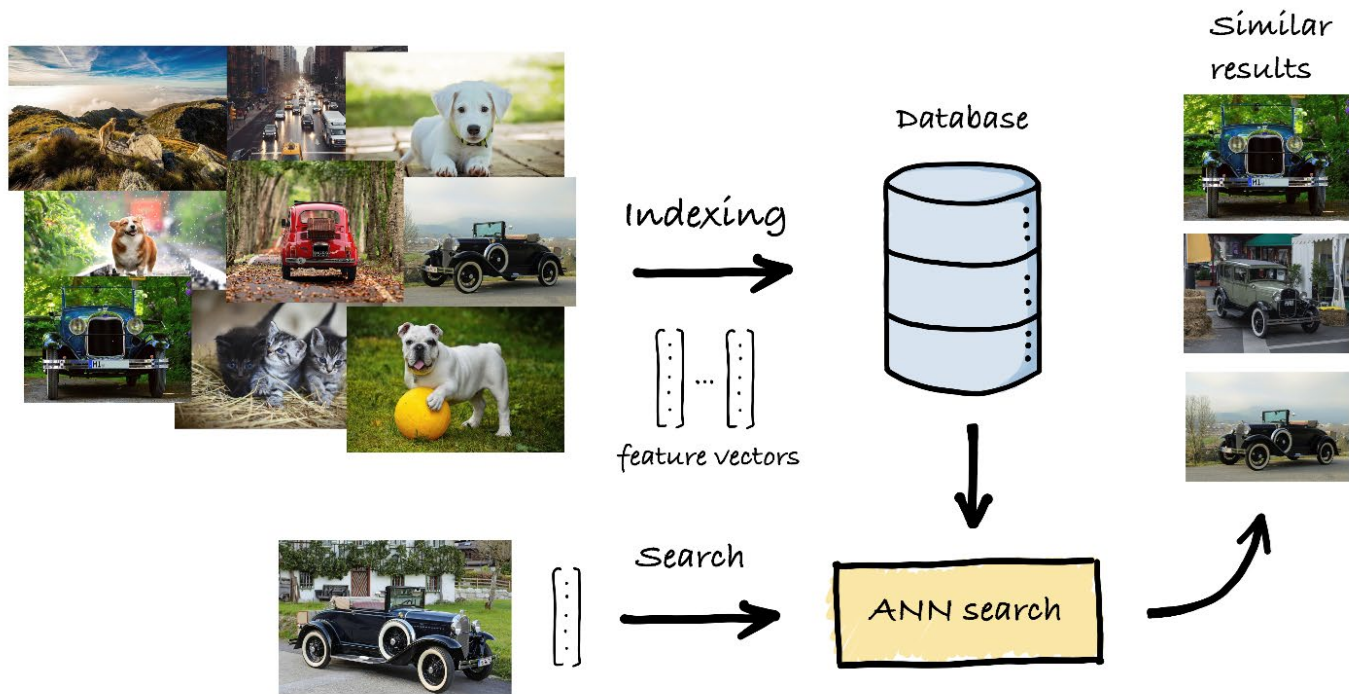
Baojian Zhou and Yifan Sun. Approximate frank-wolfe algorithms over graph-structured support sets. In *International Conference on Machine Learning*, pages 27303–27337. PMLR, 2022.

Baojian Zhou, Yifan Sun, and Reza Babanezhad Harikandeh. Iterative methods via locally evolving set process for large graphs. Under review.

Baojian Zhou, Yifan Sun, and Reza Babanezhad Harikandeh. Fast online node labeling for very large graphs. In *International Conference on Machine Learning*, pages 42658–42697. PMLR, 2023.

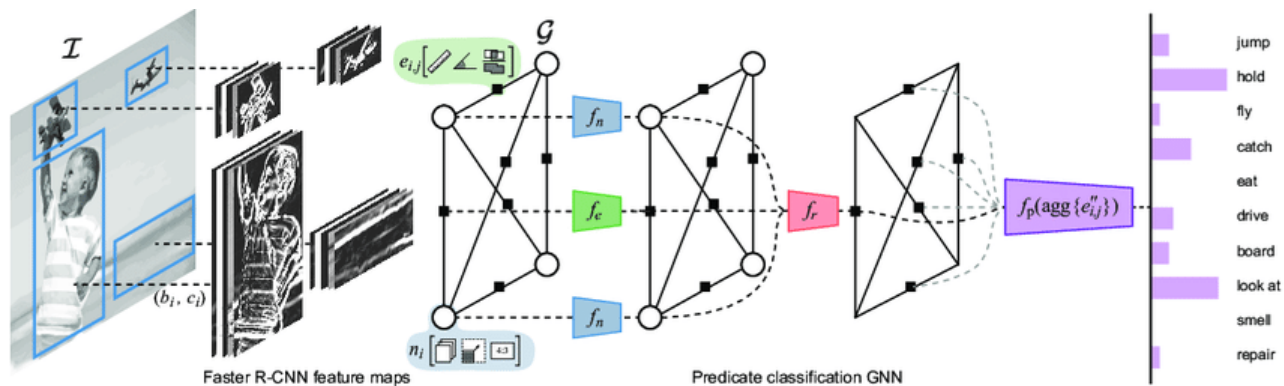


Application: Information retrieval



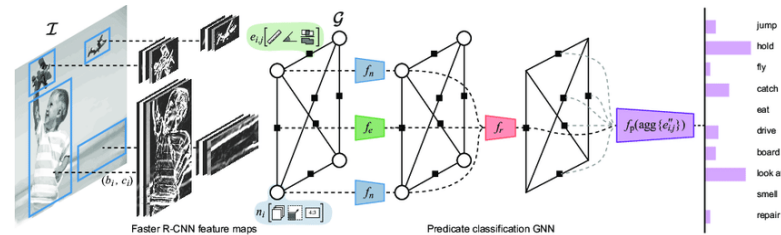
Winning the NeurIPS BillionScale Approximate Nearest Neighbor Search Challenge
Mariano Tepper, Cecilia Aguerreberre, Ted Willke, Sourabh Dongaonkar, Jawad B Khan, Mark

Graphical neural networks



Baldassarre, Federico & Smith, Kevin & Sullivan, Josephine & Azizpour, Hossein. (2020). Explanation-based Weakly-supervised Learning of Visual Relations with Graph Networks.

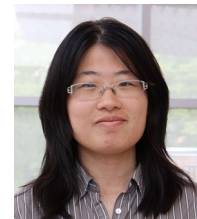
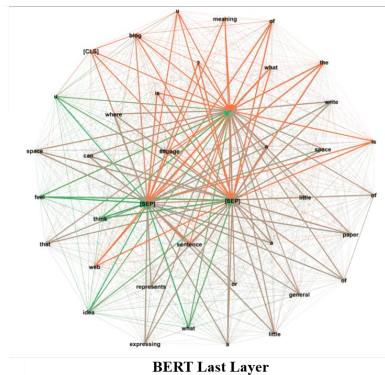
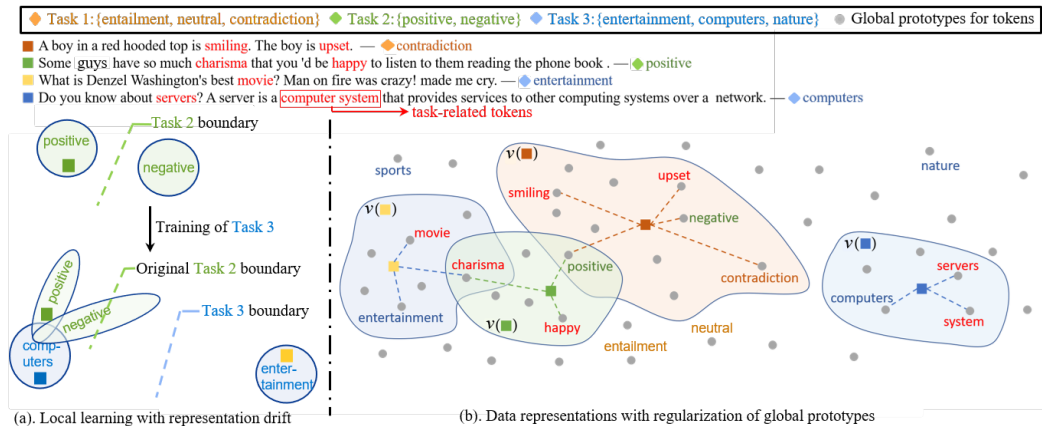
Application: train deeper graphical neural networks



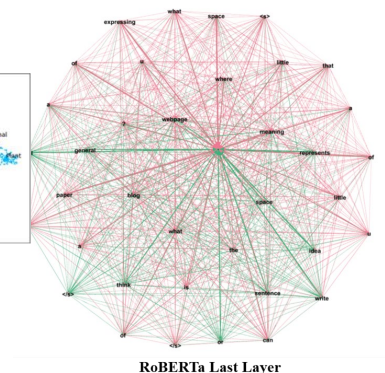
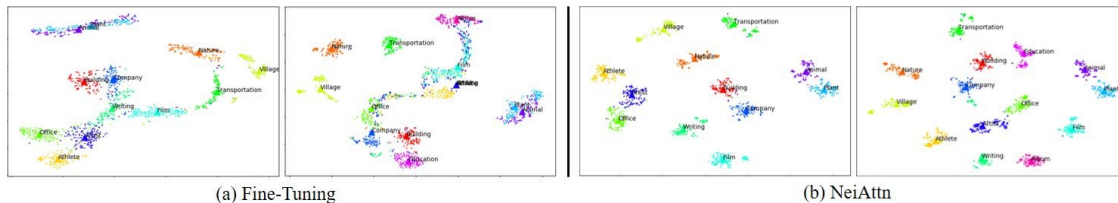
- GNNs cannot be deep
 - 3-4 layers deep
 - In contrast, CNNs are 50-100 layers deep
- Memory complexity
 - “width” = number of nodes
 - Even a sparse graph becomes dense in a few layers
- Oversmoothing
 - Output performance degrades with depth

Better message passing protocols for GNNs?

Modeling transformers with graphs



Xueying Bai



Niranjn Balasubramanian

Neighborhood attention for continual learning

Thank you!

