

## Personal Statement

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As a Computer Science student, I was initially interested in designing algorithms and solvers to solve real-world problems. However, it wasn't until I discovered the world of Data Science (DS) that I truly found my calling. At the first class of Machine Learning (ML), my professor [Shikui Tu](#) exhibited how training data boosted *transformers*' potential to generate textual answers based on the images accurately. I was immediately drawn by the power of data to elicit transformers' underlying "wisdom" to analyze the problems. In the following few weeks, I read over 20 papers on the frontier of DS and audited Andrew Ng's Stanford CS229 ML course. I became convinced that DS, especially with ML, was where I belonged to and excited to witness the future developments in DS. I therefore switched my focus to DS and made it my future to become a distinguished DS researcher to explore the potential for real-world applications.

My journey in DS started when I joined the [SJTU-ReThinkLab](#) led by Prof. Junchi Yan. I began my research on graph matching, which aimed at finding point pairs on the given images. With substantial literature review, I found a gap in works on front-end backbones (e.g. CNNs), as most of the existing works focused on back-end modules while neglected the process of initial image. Inspired by the success of attention mechanism, I decided to design a GM-specific backbone based on ViT. However, the obstacle was that original ViT extracts global information, rather than the local key-point information. To address this limitation, I devoted several days studying the paper and code of ViT, discussing with the professor to understand ViT's capabilities. After reflection, I proposed modifying ViT by introducing patches centered around the key-points and adopting the cross-attention mechanism to leverage the global information as auxiliary for the local information. I also harnessed the attention mechanism to enhance the back-end module's ability to capture the graph's underlying structure. Consequently, my proposed method achieved the state-of-the-art (SOTA) performance on extensively used GM benchmarks and the proposed backbone improved most existing frameworks by over 5%. Our paper was accepted by ICASSP and I was the first author.

The experience for my first research imparted to me the knowledge of how DS research was conducted and further sparked my interest in ML. Therefore, I participated in another research in our lab aiming at designing the guided diffusion model for CO, such as TSP and MIS. One of the most challenging problems is the design of guidance. Diffusion model mimics the Brownian motion to denoise a high-quality output from a noised input. Injecting guidance to the generation required reformulating the Bayesian probability distribution, demanding a strong grasp of mathematical skills. I spent substantial efforts learning the structure of diffusion models and conducting mathematical derivation meticulously. After contemplation, I proposed approximating the cost of solutions as guidance using Taylor's First Order Expansion and Energy Function to represent to posterior probability. Nevertheless, I found guided diffusion model showed marginal improvement on TSP and MIS (two NP-hard CO problems) compared with existing works. I discussed the phenomenon with my group members and proposed to use Bernoulli distribution instead of Gaussian distribution to model the generation process so as to better represent the discrete distribution of 0 and 1, which was more associated with our tasks. This solution addressed the issue of continuous probabilistic modeling and improved the performance by 10%. Eventually, our proposed guided-diffusion model achieved the SOTA performance on TSP and MIS benchmarks compared with learning-based models and reduced the solving time by one order of magnitude compared with exact solvers (e.g. from 1h to 4 min). Our paper was accepted by NeurIPS and I was the second author.

The research in CO honed my programming and mathematical skills and boosted my curiosity to explore DS in more domain-specific problems. Therefore, I joined the ML-PL group led by Prof. [Xujie Si](#) affiliated to the University of Toronto and Mila - Quebec AI Institute. I researched on logical puzzles that can be formulated as SAT expressions, such as Sudoku. My work was built on SATNet, a differentiable MaxSAT solver.

Nevertheless, while analyzing the paper of SATNet, I found the learned parameters cannot even represent extremely simple rules such as XOR, contradicting paper's claim. After discussing with group members, we identified a theoretical error in the interpretation. Armed with mathematical knowledge, I intuitively proposed rewriting the SAT conjunctive normal form directly as a parametric matrix. The main challenge in this project was the unfamiliar programming language CUDA, as all my previous projects were written in PyTorch. To implement my idea, I delved into the NVIDIA CUDA tutorial. Although CUDA is a much more obscure programming language compared with PyTorch, I grasped the basic principle and syntax. Within three days, and implemented my idea based on SATNet with group members. The proposed idea substantially reduced the number of trainable parameters while remedying theoretical mistakes in the original design. Notably, once the rules are learned, the model can definitely output correct answers, and this representation made the learned rules entirely interpretable, marking a significant breakthrough in the logic reasoning field. Our paper was accepted by NeurIPS and I was the second author.

The positive impact of my proposed method on the specific field and the exhilaration once my paper was accepted including at NeurIPS strengthened my determination to be a future prominent researcher in the DS field. Now I dream of employing DS knowledge to design ML frameworks that surpass human experts, well-designed algorithms and solvers in terms of both accuracy and efficiency for the logical reasoning field. Therefore, I aspire to acquire advanced ML knowledge and pursue further research on ML such as *Approximating and Reasoning*. The DS, MSE program at UPenn, with its top-notch faculty and impeccably crafted curriculum, is the perfect incubator for this dream. I can not only gain valuable insights into data science via *Data-driven Modeling and Probabilistic Scientific Computing* but also advance my knowledge of ML with *Advanced Topics in Machine Learning*. I am particularly excited to work with the pioneering researcher Prof. Susan B. Davidson on *Approximating and Reasoning about Provenance*, drawing on my previous research experience in ML, particularly in logical reasoning, to help push the boundaries of this interesting field.

Beyond academic research, I also joined various groups and activities which focused on practice, collaborating with diverse teams to hone my leadership skills and create inclusive environments. In the second year's SJTU summer camp, I led a social practice group, with over 100 members, to investigate tourist cities and identified opportunities for technological applications to enhance tourism. We visited local governments to learn the situation, built an online cloud platform to aggregate the information and encouraged team members to brainstorm. We found the deficits lay in the booking system as most tourists can only learn the scenic spots via limited photos. Therefore, we trained a recommendation ML system with sufficient historic data and included videos, VR and tourists' comments to help tourists determine their preference. Our platform was swiftly adopted by local tourism enterprises and received a satisfying feedback. As a result of our hard work and achievement, our team was awarded Third Prize of "Knowledge and Practice Cup" in Shanghai College Students' Social Practice Competition (Top 0.1% in Shanghai) and I was awarded Advanced Individual in the Social Practice Group and the National Scholarship. In my graduate journey, I hope to join the related non-profit student organizations in UPenn to combine my passion for technology with commitment to practical application, volunteer work and societal development.

With a robust quantitative skill-set, extensive experience, and a deep passion for expanding my knowledge of DS, I firmly believe that I am a well-suited candidate for the DS, MSE program. I am confident that my Master's journey at UPenn will equip me with the expertise demanded to make insightful contributions to the DS field, pursue further research as a PhD and finally contribute back to the community at UPenn.