

I. INTRODUCTION: MY RESEARCH INTERESTS AND PERSPECTIVE

Current AI models have achieved remarkable success across tasks, yet their internal mechanisms remain largely a black box. We know what models can do, and to some extent how they do it in specific details, but we lack an intuitively human-understandable view of their cognitive and reasoning processes. I believe that a key step toward more general and reliable AI lies in understanding and designing the models internal information-processing mechanisms. My research centers on two core questions: first, can we construct within the model a **human-intuitive, stepwise, and organic reasoning trajectory**? Second, how can we enable the model to **understand and align information from different sources** (e.g., images and text) in a human-like manner?

My goal is to explore how to realize the fusion of these two abilities in a high-dimensional latent space, enabling models not only to see and read the world, but also to reason about it in a transparent and trustworthy way.

II. ROBUST FEATURE REPRESENTATIONS (COREFACE)

My research journey began with a deep fascination for self-supervised contrastive learning. It does not rely on expensive human annotations; instead, it learns from the structure inherent in dataan elegance and potential that captivated me. However, the realities of a lab with very limited compute and diverse student directions pushed me to forgo brute-force compute and pursue a more fundamental question: how can **mechanism design** improve model performance?

I found face recognition to be an ideal sandbox. I observed its deep kinship with contrastive learning: both carefully shape the geometry of feature distributions in high-dimensional space. Building on this, I proposed the CoReFace frameworkan instance of explicit design of internal mechanisms. In this fine-grained recognition task, traditional image augmentations can damage identity information. A key insight I had was that Dropout is essentially a **feature-level stochastic perturbation** that can provide the necessary views for contrastive learning without corrupting image semantics. By introducing a contrastive-learning-guided regularizer, I actively **sculpted** the geometric structure of the feature space. Ultimately, this approach **increased the similarity margin between positive and negative pairs by 15%**. Throughout, I led the full lifecyclefrom problem identification and design to independent validation and writingshifting from student to a truly **independent researcher** [1].

III. A UNIFIED FRAMEWORK FOR HETEROGENEOUS DATA (QGFACE)

After CoReFace, I yearned for research with more immediate real-world relevance. This was reinforced when I scrolled through my phones photo album: the built-in face clustering struggled

with family photos. I realized that low-quality data arising from composition, lighting, and other real-world factors differ markedly from our common datasets.

QGFaces **single-encoder architecture** is my solutionan internal **information dispatch system**. It simulates a triage process: high-quality data follow a classification path, low-quality data follow a contrastive path, and gradient truncation prevents contamination across paths. When contrastive learning underperformed due to insufficient positive pairs, I designed a **dynamically updated encoder pool** that significantly boosted performance. Racing against a deadline, I structured my days into intense cycles; GPU time became my rest time. Far from being a burden, this was a final **stress test** of my passion for research. It proved that what drives me is not external expectation, but the intrinsic intellectual joy of solving hard problems. Ultimately, QGFace reached SOTA on low-quality datasets with **only a 0.3% trade-off** on high-quality data, validating that my research philosophy can yield elegant, robust, and practical systems [2].

IV. EXPLORATION AND FOCUS: CLARIFYING MY DIRECTION THROUGH PRACTICE

Despite some academic progress, repeated setbacks and uncertainty in the submission process led to a period of deep confusion and self-doubt. To find clarity, I decided to step into industry and **stress-test** my intrinsic motivation.

I first joined KeyoneAI, a startup led by former IBM China Chief AI Architect Jie Fang. There, I translated frontier generative AI into product, felt the pulse of rapid iteration, advocated technology, and engaged with users.

Later, I served as the sole technical lead on the organizing committee of the Worldwide Educators Conference (WWEC) [3]. Beyond ensuring system reliability, I accelerated the teams digital transformation, developed internal tools to generate 1000+ complex posters, and connected 3,000 attendees, 1,000 exhibitors, 10+ vendors, and many ad-hoc sessionsworking 80 hours per week for nearly three months. While challenging and rewarding, these experiences still could not replace the pure intellectual excitement and flow I feel in researchwitnessing breakthroughs in a field and contributing to them. This deliberate detour granted me unprecedented clarity: my deepest drive is to question fundamentals, refute and rebuild existing solutions, and innovate effectively across disciplines.

More importantly, this exploration helped me reframe what used to be excessive self-scrutiny into a unique research lens. I understand deeply that **a truly intelligent system is not defined by flawless unidirectional reasoning but by its capacity to handle internal conflict, self-examination, and iterative revision**the essence of human reflection and productive struggle.

V. FUTURE RESEARCH BLUEPRINT: ALIGN SEMANTICS, REASON IN LATENT SPACE

I now plan my Ph.D. research with renewed clarity and conviction. I aim to combine my experience in **mechanism design** with reflections on **human cognition**, focusing on two capabilities of large models:

- 1. Building a Unified Semantic Space:** My first goal is to study how to effectively map diverse information into a unified, interpretable latent space. This goes beyond cross-modal mappingit underpins logical reasoning. I believe that efficient cross-modal semantic alignment is a first step toward foundational models that can comprehensively understand the world. My experience with

mixed-quality data in QGFace provides practical grounding for aligning and fusing heterogeneous information.

2. Realizing Structured Reasoning in Latent Space: After achieving semantic alignment, my next goal is to design *structured*, chain-of-thought-like reasoning trajectories within this unified latent space. I want models not only to produce answers, but also to present a decomposable and traceable reasoning process. This improves interpretability and reliability, and may open new avenues for multi-step, open-ended problem solving.

In summary, my research proceeds on two fronts: through **multimodal alignment**, enabling models to see a richer world; and through **latent reasoning**, enabling them to think more clearly and logically.

VI. CONCLUSION

With hands-on experience designing internal mechanisms and a clear plan to integrate *semantic alignment* with *latent reasoning*, I am prepared for the challenges of a Ph.D. I look forward to contributing to the next generation of more capable and trustworthy AI systems in a creative and supportive environment.

REFERENCES

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