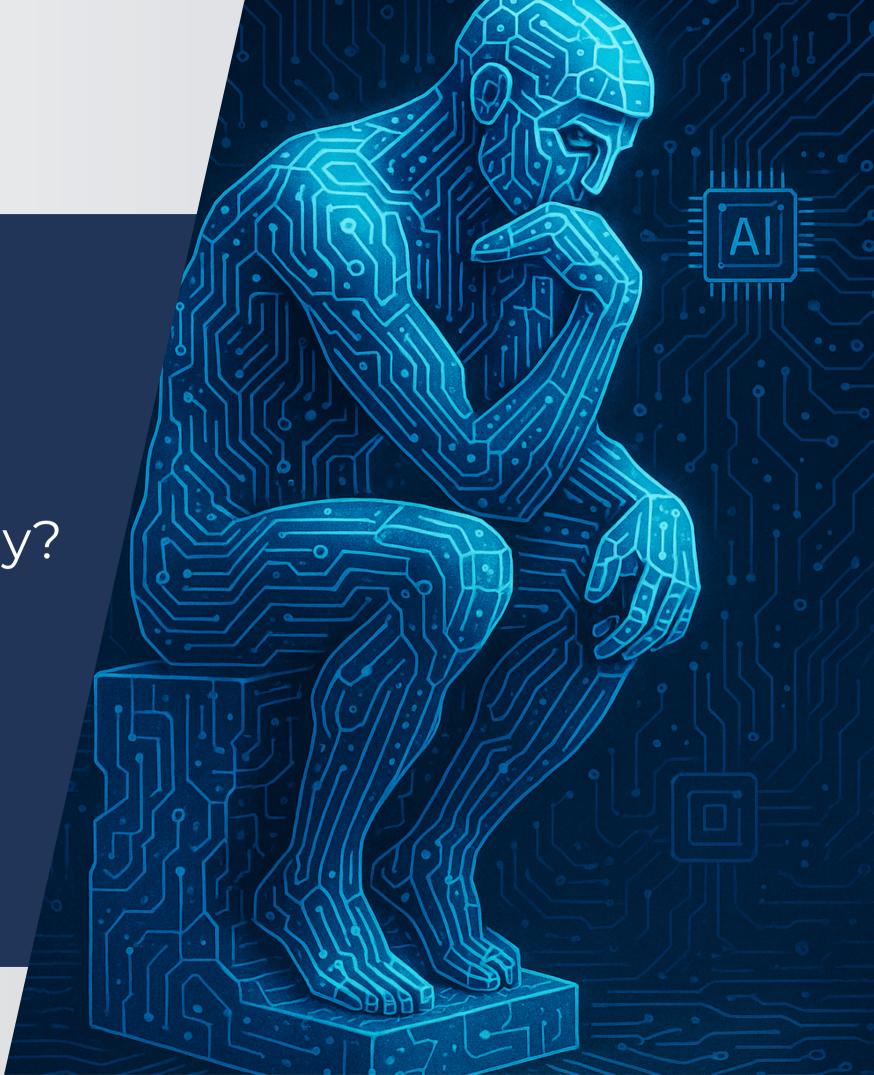


TABLE OF CONTENTS

- Where Did I Come From?
- What Captivates Me?
- What Are My Life Philosophy?
- My Research Footprints
- My Research Interests
- My Role in TUAI project



WHERE DID I COME FROM?

Confucian Heritage





Spirit of Democracy

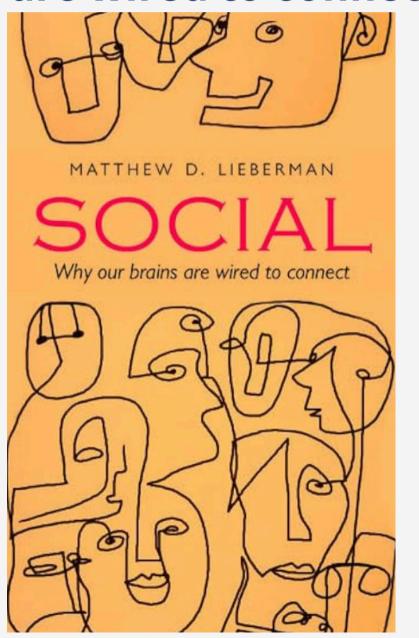


Rich Culinary Traditions

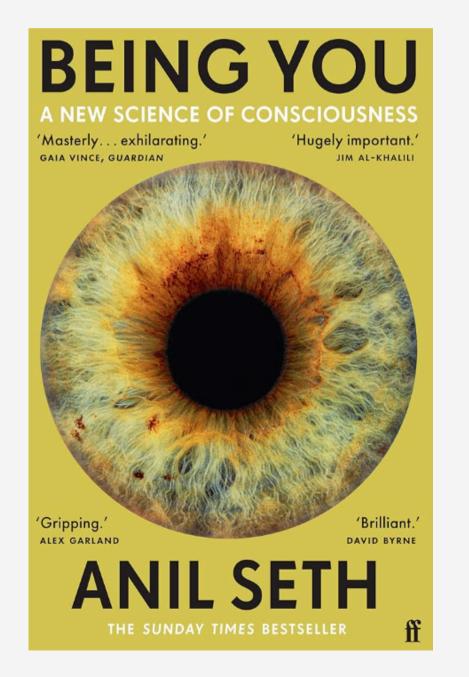


WHAT CAPTIVATES ME? READING

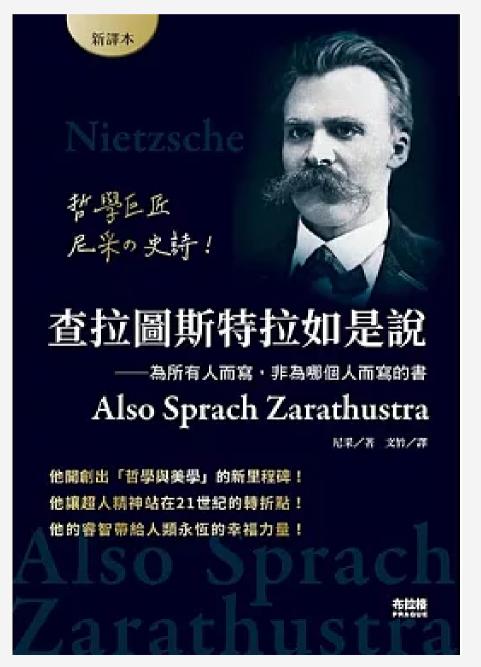
Social- why our brain are wired to connect



Being You



Thus Spake Zarathustr



MY RESEARCH FOOTPRINT

Key milestones

FedCali: Mitigating Overgeneralization for Anomaly Detection in Distributed Sensor Environments

Publisher: IEEE

Cite This



Pi-Wei Chen; Jerry Chun-Wei; Rafał Cupek; Chao-Chun Chen All Authors

Home > Intelligent Information and Database Systems > Conference paper

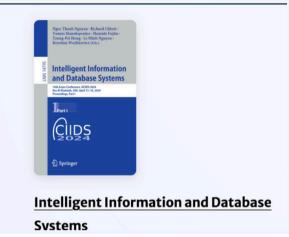
RECALL: Towards Generalized Representations in Unsupervised Federated Learning Under Non-IID Conditions

Conference paper | First Online: 16 July 2024 pp 253–263 | Cite this conference paper

Chen 🖸

Show details

ICCV 2025 Conference Submission



Pi-Wei Chen, Jerry Chun-Wei Lin, Feng-Hao Yeh, Rafał Cupek & Chao-Chun Chen

Feature Purified Transformer With Cross-level Feature Guiding Decoder For Multi-class OOD and Anomaly Detection

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Chao-Chun Chen

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International Conference on Computer Vision 2025

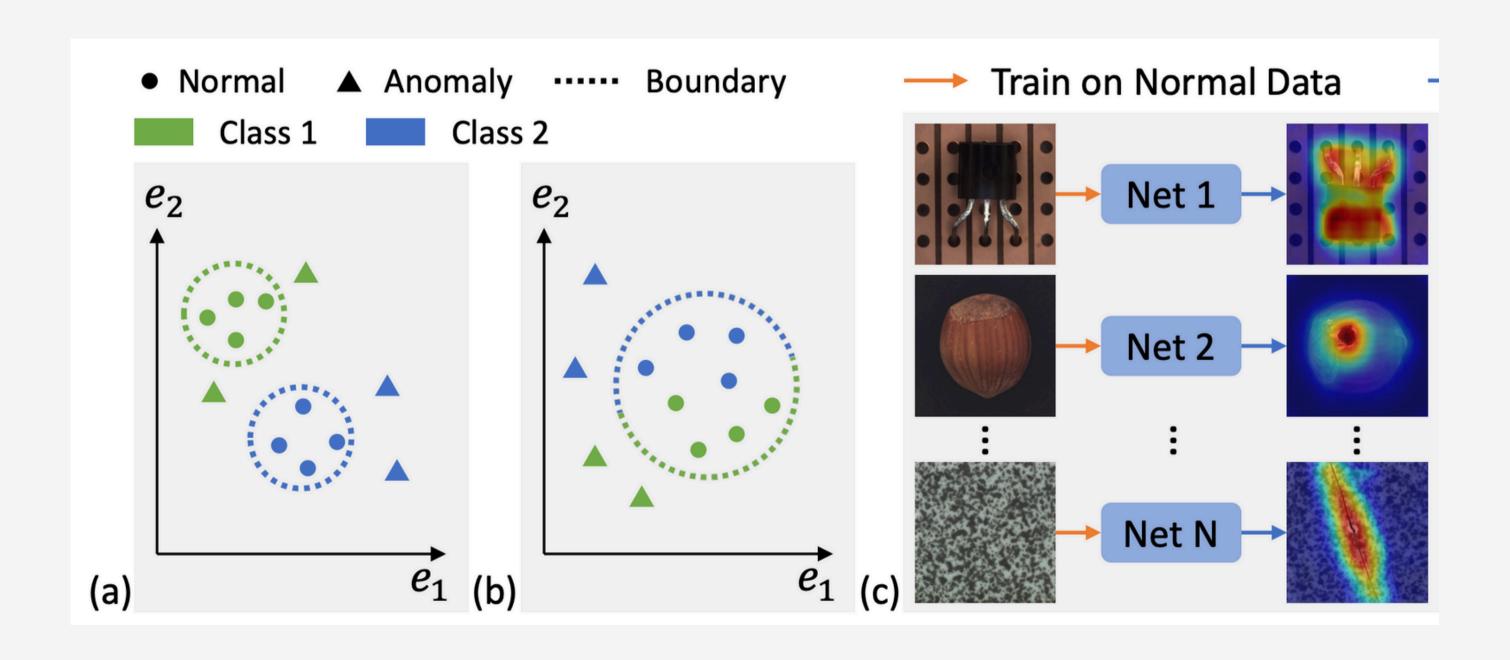
Beyond Human-prompting: Adaptive Prompt Tuning with Semantic Alignment for Anomaly Detection

PI-WEI Chen , Chao-Chun Chen , Jerry Chun-Wei Lin , Feng Hao Yeh , JIA JI , Zih-Ching Chen , Wei-Han

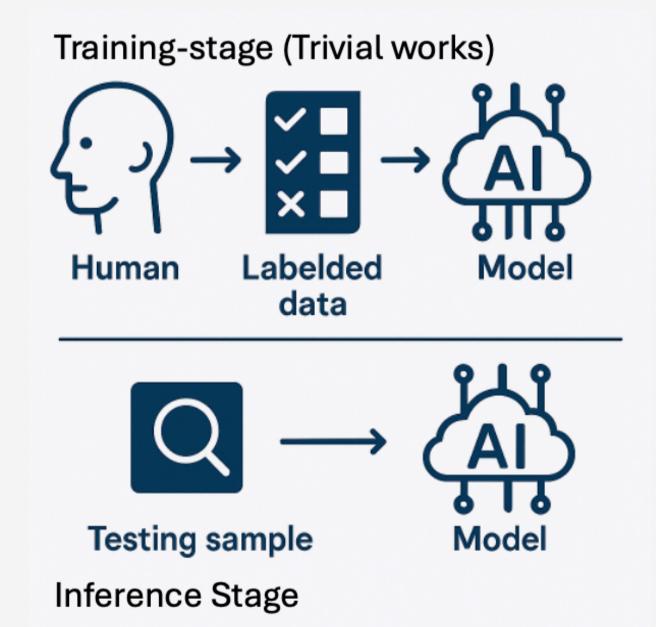
Average Preliminary Recommendation: 3.00 (Min: 2,

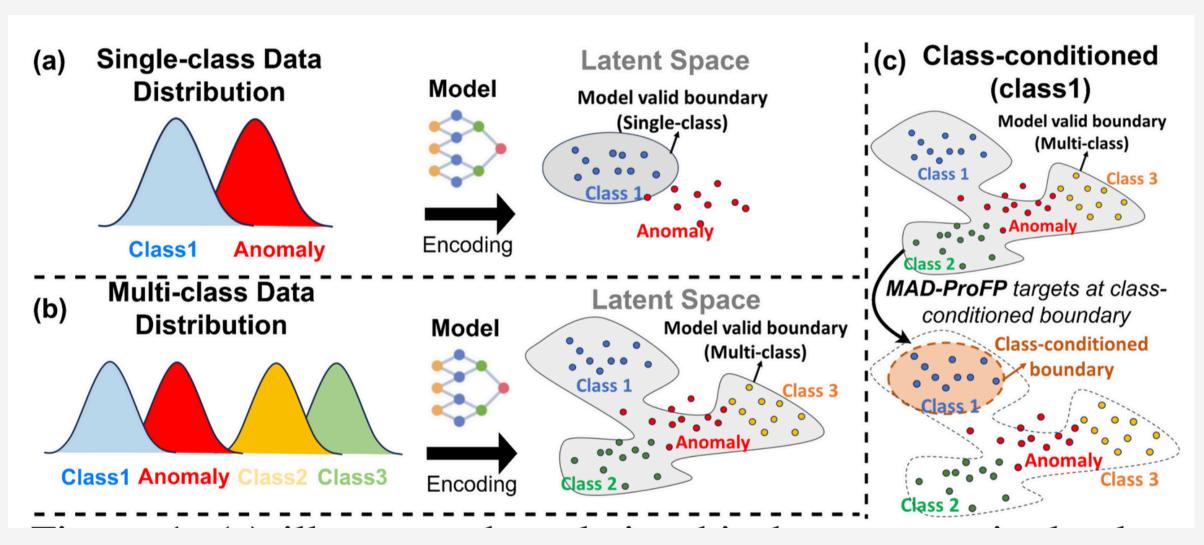
Average Confidence: N/A (Min: N/A, Max: N/A)

Feature Purified Transformer With Cross-level Feature Guiding Decoder For Multi-class OOD and Anomaly Detection

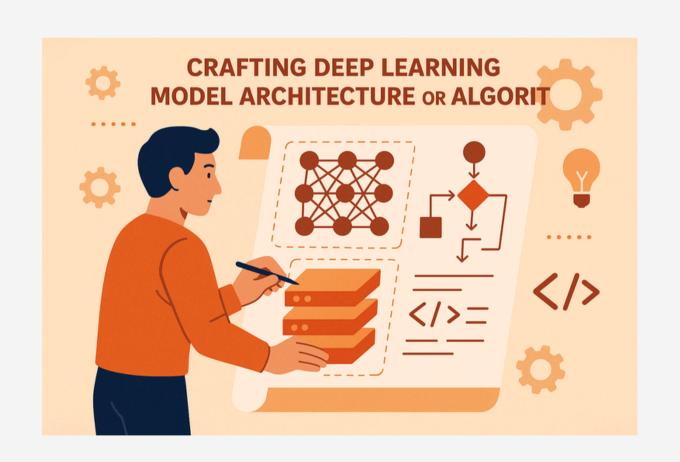


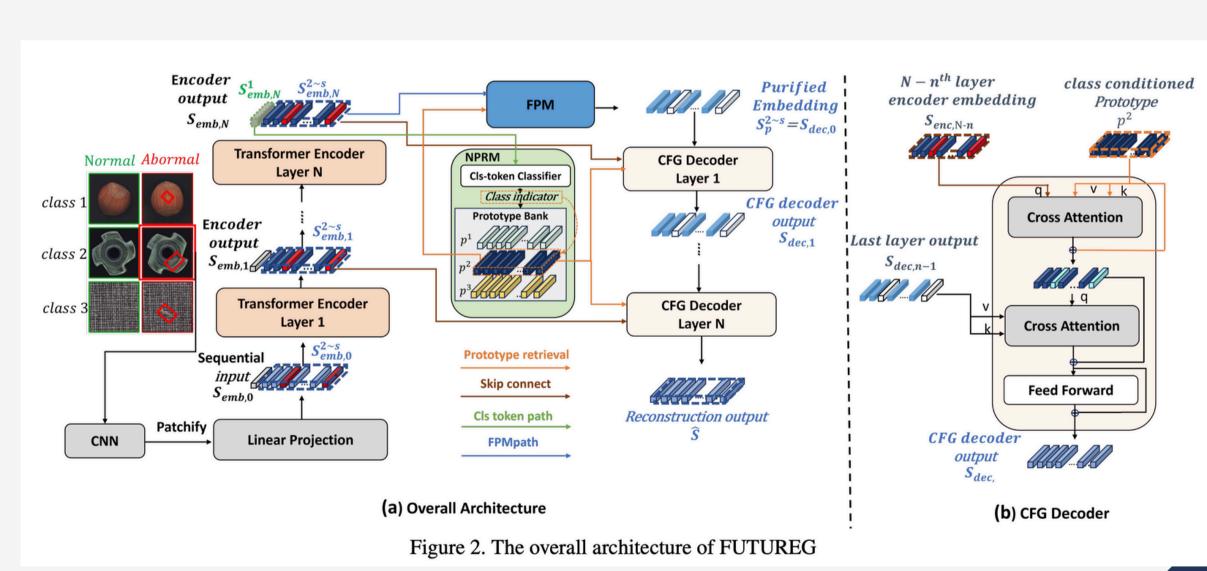
Feature Purified Transformer With Cross-level Feature Guiding Decoder For Multi-class OOD and Anomaly Detection



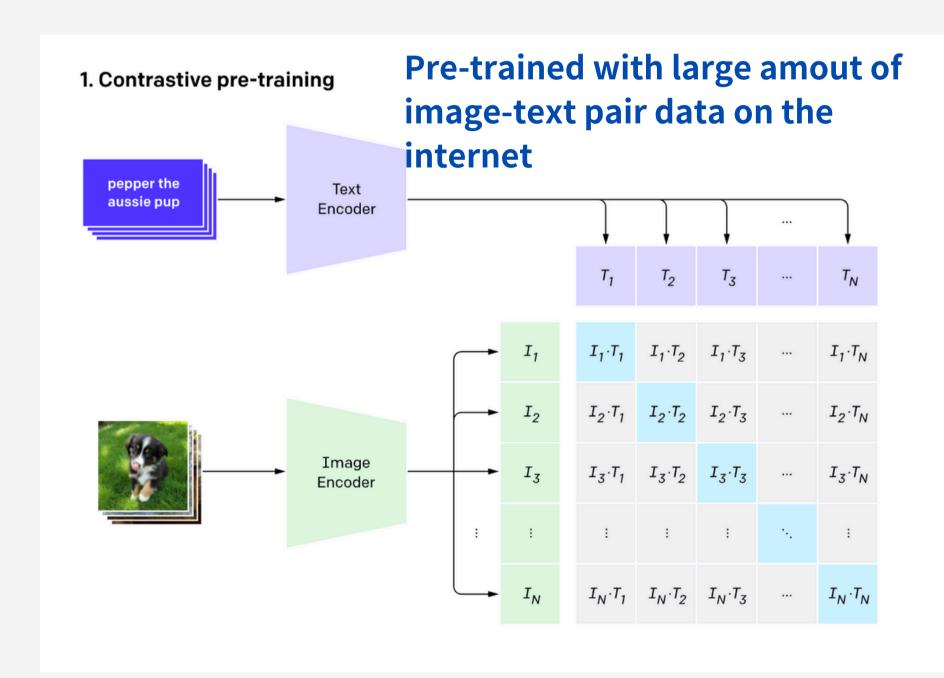


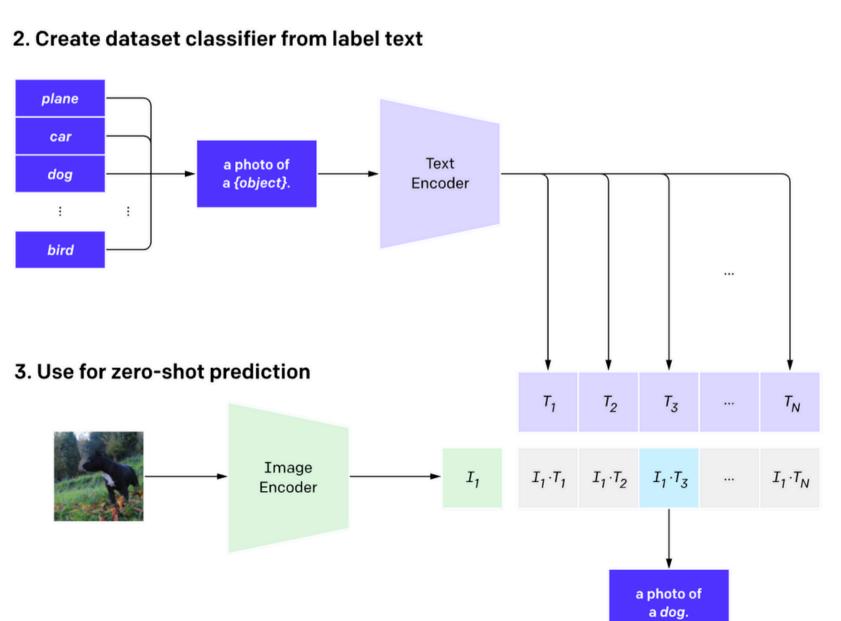
Feature Purified Transformer With Cross-level Feature Guiding Decoder For Multi-class OOD and Anomaly Detection



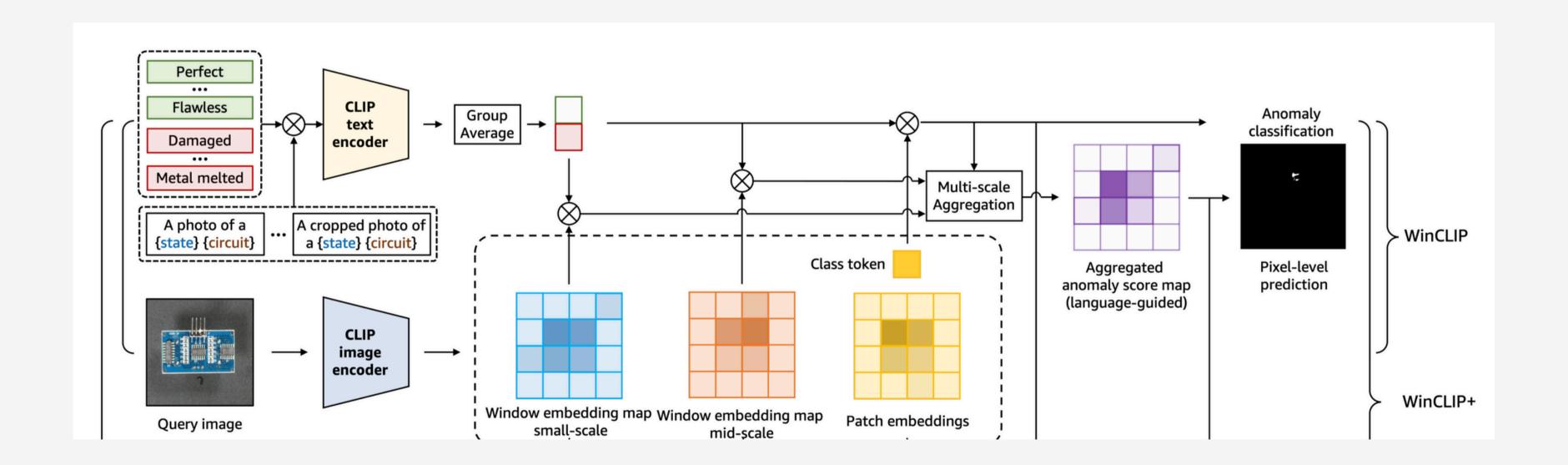


The 1'st Revolution from Pre-trained VIsual-Language Model

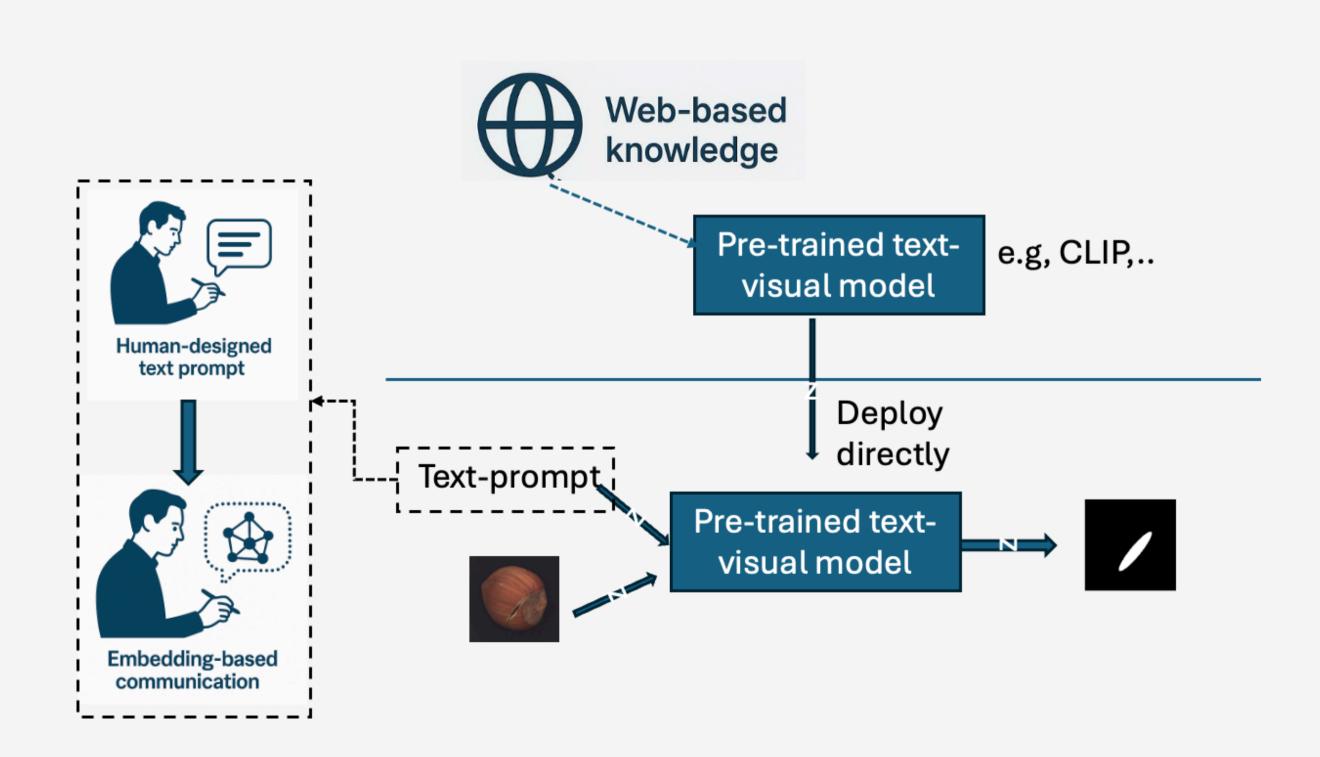




Zero/few-shot Anomaly Detection via prompting



Beyond Human-prompting: Adaptive Prompt Tuning with Semantic Alignment for Anomaly Detection



Beyond Human-prompting: Adaptive Prompt Tuning with Semantic Alignment for Anomaly Detection

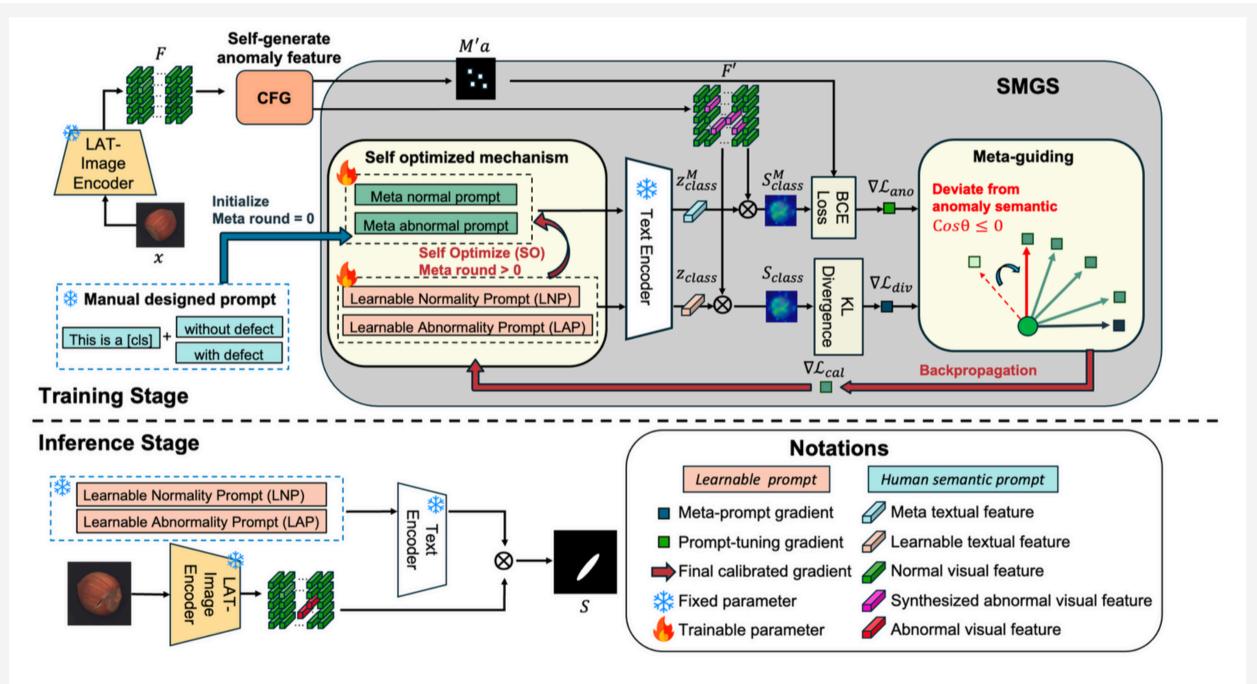


Figure 2. The overall architecture of the proposed Adaptive Prompt Tuning (APT) framework, featuring three main components: (1) Self-Optimizing Meta-Prompt Guiding Scheme (SMGS), which aligns learnable prompts with anomaly semantics through gradient calibration; (2) Contextual Feature Generation (CFG), which generates context-aware anomalies to enhance prompt adaptability; and (3) Locality-Aware Transformer (LAT), which focuses on locality-sensitive feature extraction. These components collaboratively optimize prompt alignment and improve anomaly detection robustness across diverse contexts.

Beyond Human-prompting: Adaptive Prompt Tuning with Semantic Alignment for Anomaly Detection

Pixel-wise Anomaly Detection Comparison Table											
Method		MVTec		VisA		MPDD		SDD		CVC	
		w/o VG	w/ VG	w/o VG	w/ VG	w/o VG	w/ VG	w/o VG	w/ VG	w/o VG	w/ VG
1-shot	WinCLIP [12]	81.4	86.2	85.67	89.96	89.56	92.83	68.02	70.91	63.24	62.32
	AnomalyGPT [11]	-	95.3	-	96.20	_	-	-	-	-	-
	PromptAD [15]	91.32	95.0	92.34	96.12	94.77	95.53	92.01	93.4	62.46	54.43
	Ours	92.32	95.32	93.43	96.09	95.11	95.82	95.73	95.47	72.46	64.4
2-shot	WinCLIP [12]	81.4	89.65	85.67	90.26	89.56	92.78	68.02	70.02	-	-
	AnomalyGPT [11]	_	95.6	_	96.40	_	-	-	-	-	-
	PromptAD [15]	91.34	96.03	92.52	96.43	94.3	96.0	91.0	90.8	-	-
	Ours	92.12	97.0	93.61	96.89	95.28	96.32	96.68	92.43	-	-
4-shot	WinCLIP [12]	81.43	90.63	85.67	91.3	89.56	93.26	68.02	71.10	-	-
	AnomalyGPT [11]	_	96.25	_	96.70	_	-	-	-	-	-
	PromptAD [15]	91.63	96.34	92.33	96.90	93.53	96.52	90.64	94.23	-	-
	Ours	92.16	97.1	94.68	97.1	95.14	96.9	96.33	95.3	-	-

Limitation of Visiual text model

Insensitive to environmental variation



The Mighty LLM?



provide detail deciption of the image

The image shows a silver-blue Toyota Wish, a compact MPV (multi-purpose vehicle), captured from an elevated angle. The vehicle has a sleek design with a sunroof, tinted windows, and alloy wheels.

There are visible patches of paint damage and scratches on various parts of the car, particularly on the front bumper, side fenders, and rear doors. These areas appear to have been manually edited or marked, possibly indicating areas requiring repair.

The car is parked outdoors, reflecting buildings and trees on its glossy surface. The background has been removed, making the car appear isolated against a white or transparent backdrop. The lighting conditions suggest a natural daylight setting.

New Paradigm Shift











What Role Can We Play in This New Landscape?

Potential LLM Research Area

RAG System Design

Retrieval Robustness to Distribution Shift

Knowledge-Graph -Aware RAG

Reinforcement learning

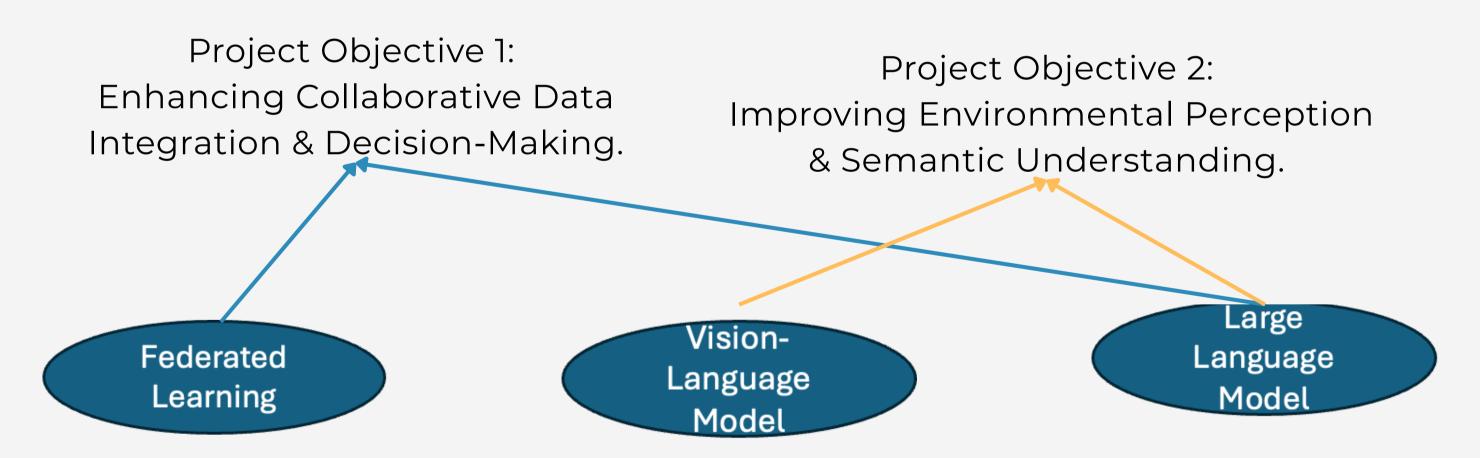
Synthetic Preference Bootstrapping

Multimodal RLHF for Vision Tasks

My Role in TUAI project

TUAI project-

Collaborative Sensor Fusion System: Concept, Modelling and Applications



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Email



Patrick Chen

WhatsApp 連絡人



