

Machine Learning Project Topics

Format and expectations. 2–3 students per team. Each project must deliver (i) open, reproducible code and benchmarks, and (ii) a paper-style report (8–12 pages) covering theory, implementation details, and applications. Topics balance theory & practice, and reference A* venues (NeurIPS, ICML, ICLR, AAAI, CVPR/ICCV) from 2023–2025 where possible.

1 Explainable AI (XAI): Tuning Models for Faithful Attributions

Idea. Implement and evaluate methods that improve the faithfulness and discriminativeness of post-hoc explanations. Reproduce Distractor Erasure Tuning (DiET) and the ICLR 2025 “How to Probe” framework; benchmark across backbones and datasets (e.g., ImageNet, CIFAR, GLUE).

Scope. (i) Formalize metrics (sufficiency/necessity, deletion/insertion, pointing game), (ii) compare standard attribution (GradCAM/IG) vs. DiET-style training, (iii) analyze sensitivity to pre-training and probing design.

Starting points. [1, 2].

2 Bayesian Deep Learning: Posterior Geometry and Flatness-aware BNNs

Idea. Study how posterior geometry (modes, symmetries, flatness) affects uncertainty quality. Reproduce Flat-Seeking BNNs and the symmetry-aware posterior exploration; run vision (CIFAR/ImageNet) and tabular tasks.

Scope. (i) Implement sharpness-aware variational posteriors, (ii) visualize/post-process symmetries, (iii) compare uncertainty metrics (ECE, NLL, OOD).

Starting points. [3, 4].

3 Graph Transformers at Scale: Global, Sparse and Hybrid Attention

Idea. Build and benchmark scalable graph transformers: GOAT (approx. global attention) vs. Exphormer (expander-graph sparsity) and a recent global-to-local variant. Evaluate on long-range graph benchmarks (OGB, ZINC, Peptides).

Scope. (i) Complexity vs. accuracy tradeoffs; (ii) homophily/heterophily robustness; (iii) ablations on positional encodings and global-token designs.

Starting points. [5, 6, 7].

4 Label-aware Graph Classification: Distillation for Graph-level Tasks

Idea. Investigate label-attentive distillation to align node embeddings with graph labels to improve graph-level classification. Extend LAD-GNN to new backbones; study calibration and sample efficiency.

Starting points. [8].

5 Pretrained LMs under Resource Constraints: 4-bit Fine-tuning and Adapters

Idea. Systematic study of QLoRA-style low-memory finetuning on several LLMs and tasks (instruction-following, classification). Compare full fine-tuning, LoRA, and QLoRA across GPUs, quantizers (NF4), and datasets.

Starting points. [9].

6 Time-series Forecasting: Are (Cross-)Attentions Necessary?

Idea. Reassess Transformer design for TSF: PatchTST vs. linear baselines (DLinear) vs. the NeurIPS 2024 CATS model that removes self-attention. Evaluate long-horizon multivariate forecasting under strict protocol.

Starting points. [10, 11, 12].

7 Autoencoders for Anomaly Detection: From Reconstruction to Semi-supervised Signals

Idea. Beyond pure reconstruction error, exploit few labeled anomalies to separate manifolds and improve detection. Compare VAE/AE/AMA-style methods on time-series (MSL/SMAP) and vision (MVTec).

Starting points. [13, 14].

8 Adversarial Robustness via Diffusion Priors: Purification and Counter-attacks

Idea. Implement score-based test-time optimization (ScoreOpt) and evaluate under robust, recent attack suites including ones tailored to diffusion purification (e.g., DiffAttack). Provide strong AutoAttack/BPDA/EOT evaluations.

Starting points. [15, 16, 17].

9 Personalized Federated Learning under Heterogeneity and Budgets

Idea. Study client heterogeneity (data & resources) using pFedGate’s sparse personalization. Evaluate convergence/accuracy/efficiency vs. baselines (Per-FedAvg, FedProx, FedSpa).

Starting points. [18].

10 Causal Representation Learning from Multiple Domains

Idea. Learn identifiable causal latents from unpaired multi-domain observations and compare with weak-invariance methods. Apply to synthetic (linear SCM) and real multi-view datasets.

Starting points. [19, 20].

11 Generative Models in Practice: Diffusion vs. GANs vs. VAEs

Idea. A rigorous empirical and compute-aware comparison of diffusion models, GANs, and VAEs on conditional generation. Include FID/KID/IS, compute/FPS, and controllability; analyze classifier guidance and negative prompts.

Starting points. [21, 22].

12 Fair Representation Learning with Certificates + Bias Audits

Idea. Combine algorithmic fairness with practical certificates (upper bounds on downstream unfairness). Reproduce FARE and conduct a bias audit on a vision or recsys pipeline; report fairness–utility trade-offs.

Starting points. [23, 24].

Deliverables for each project

- Reproducible code (Python) with configuration files and experiment runners; baseline re-implementations where needed.
- Benchmarks against strong SOTA or standard baselines; report mean \pm std over multiple seeds; include compute budget.
- Paper-style report with: problem statement, related work, methods, experimental setup, results/ablations, limitations/ethics checklist.

References

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