Assignment: Advanced AI Systems for Yelp Reviews

Overview

This assignment explores how AI systems can go beyond simple text classification to demonstrate reasoning, robustness, and actionable insights in the context of **Yelp reviews**.

You will begin with baseline sentiment classification and progressively tackle harder challenges: structured prompting, reasoning vs direct answers, multi-task outputs, and robustness to domain shift. Some tasks may not have perfect solutions — part of the evaluation is how you design, experiment, and justify your approaches.

Dataset

- **Primary**: Yelp Review Full (650k train, 50k test)
- Optional domain-shift test sets: Amazon Reviews, IMDB Reviews (available on HuggingFace)
- Labels: {1, 2, 3, 4, 5} stars

Tasks

1. Zero-Shot & Few-Shot Prompting Under Constraints

- Use an instruction-tuned LLM (e.g., GPT-4, Claude, Llama-3).
- Prompt the model to classify reviews into 1–5 stars.
- Output must be in strict JSON format:

```
{ "stars": 3, "explanation": "The review was neutral, with mild praise but no enthusiasm." }
```

- Do it for both zero-shot and few-shot prompting and compare.
- Evaluate:
 - Accuracy / Macro-F1 on a sample subset
 - Format compliance rate (valid JSON %)

2. Chain-of-Thought vs Direct Answers

- Compare two prompting strategies:
 - Direct: "This review is X stars."
 - Reasoning: "Explain reasoning, then output stars."
- Analyze: Does chain-of-thought improve or hurt classification?
- Report both accuracy and error types (e.g., reasoning mismatch).

3. Multi-Objective Al Assistant

Extend the task to make the system useful for businesses:

- For each review, generate:
 - Star rating (1–5)
 - Extracted key complaint/compliment
 - o A short, polite business response
- Since ground truth doesn't exist, design an evaluation protocol:
 - You may use human annotation, or
 - LLM-as-judge evaluation (e.g., GPT evaluating outputs for faithfulness/actionability).
- Report findings and examples of success/failure.

4. Domain Shift & Robustness

- Train/fine-tune on Yelp → test on Yelp AND Amazon/IMDB reviews.
- Report:
 - Drop in performance (Yelp → other domains)
 - How adversarial inputs affect predictions
- Suggest at least one mitigation

Deliverables

- 1. Code / Notebook(s)
 - o All experiments, prompts, and evaluation scripts
 - Robustness/adversarial dataset creation
- 2. Report (4-5 pages)
 - o Prompt iterations & insights
 - Results tables (accuracy, JSON compliance, robustness, etc.)

- o Comparative analysis: direct vs CoT, classification vs assistant outputs
- Evaluation of domain shift
- Discussion of trade-offs and limitations

Notes

- Some tasks are intentionally **open-ended** and may not have a perfect solution.
- The goal is not just accuracy, but **designing**, **analyzing**, **and justifying approaches**.
- Use **smaller subsets** for fast iteration if needed.