Llama 3 Report - Notes

Introduction

- Llama 3 is a foundation model developed by Meta.
- Supports multilinguality, coding, reasoning, and tool usage.
- Largest model: 405B parameters with a 128K token context window.
- Performs comparably to **GPT-4** across various benchmarks.
- Publicly released, including Llama Guard 3 for input and output safety.
- Future plans: Integrate image, video, and speech capabilities.

Post-Training Approaches (Section 4)

Supervised Fine-Tuning (SFT)

- Purpose: Align model outputs with human expectations.
- Data Sources:
 - Human-annotated examples
 - Synthetic data generated by Llama 3 itself
 - Rejection sampling from multiple candidate outputs
- Process:
 - Model is trained on labeled data using a standard cross-entropy loss.
 - Example sources: Code writing, reasoning tasks, multilingual datasets.
- Training Details:
 - Conducted in multiple rounds, improving model behavior iteratively.
 - Applied alongside Direct Preference Optimization (DPO).

Direct Preference Optimization (DPO)

- Alternative to Reinforcement Learning from Human Feedback (RLHF).
- Process:
 - Compares preferred vs. non-preferred responses.
 - Trains the model to **increase** the likelihood of preferred responses.
 - Avoids issues like reward hacking in reinforcement learning.
- Advantages Over RLHF:
 - More stable
 - Less compute-intensive
 - Provides better performance on instruction-following benchmarks.

1

Safety and Llama Guard 3 (Section 5.4)

- Llama Guard 3 is a safety mechanism integrated into Llama 3.
- Functions:
 - Input filtering: Blocks harmful, unethical, or unsafe inputs.
 - Output moderation: Ensures responses align with ethical standards.
- Training Approach:
 - Uses a **fine-tuned classifier** to detect unsafe content.
 - Evaluated via **red teaming** (stress testing for vulnerabilities).
- Comparison to Reinforcement Learning:
 - Unlike RLHF, Llama Guard directly moderates inputs/outputs.
 - More efficient and easier to maintain than traditional RLHF policies.

Comparison with Other Models

Model	Safety Method	Post-Training Approach	Handling Hallucinations
Llama 3	Llama Guard	SFT + DPO	Uses filtering at input/output level
GPT-4	Reinforcement Learning (RLHF)	RLHF + fine-tuning	Uses reward modeling to penalize hallucinations
DeepSeek-R1	Custom reward model	RLHF + supervised learning	Focuses on reward hacking prevention

Key Takeaways

- Llama 3 introduces Llama Guard 3 for improved AI safety.
- Supervised fine-tuning and DPO are used instead of traditional RLHF.
- **Post-training is crucial** for improving factuality, reducing biases, and ensuring ethical AI usage.
- Ethical Considerations:
 - Llama Guard 3 provides proactive filtering rather than reactive moderation.
 - Avoids reinforcement learning pitfalls such as instability and unintended bias.

These notes provide a structured summary of the Llama 3 report for the upcoming class discussion.