

# TEAM MEMBERS



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### **Problem Statement**

- LLMs have achieved enormous success in understanding and solving complex natural language tasks.
- However, they suffer from memorization issues, retaining training data unintentionally.
- ☐ This leads to the risk of regurgitating private or copyrighted content.
- This leads to legal issues for the model developer and/or vendors, requiring effective solutions.

Here are 10 lines from the iconic song "Abhi Na Jao Chhod Kar" (from the 1961 film Hum Dono), penned by Sahir Ludhianvi: अभी न जाओ छोड़कर, के दिल अभी भरा नहीं अभी अभी तो आयी हो, बहार बनके छायी हो बहुत नाज़ुक हो येही तो, ये लम्हे ये फ़ुर्सतें इन्हें थम लो संभालो, इन्हें महफ़िल संभाले... कहिए और न राज़-ए-दिल, सुनाके जियेगा क्या यूँही बीत जाये सुबह, तो शाम आके रुकेगा क्या? तुम्हें क्या सुनाएँ हम, जो हमसे है दिल की बात तुम्हारी हँसी उड़ा दे, तुम्हारी हर अदा है बात... deepseek

Kar"?

give me lyrics 10 lines to "Abhi Na Jao Chhod



# **Objective**

- Develop and evaluate efficient machine unlearning techniques for LLMs.
- Design methods to erase sensitive data (forget set) while retaining essential knowledge (retain set).
- Establish a standardized evaluation framework for assessing unlearning effectiveness.
- ☐ Ensure reduced regurgitation and robustness against membership inference attacks.
- Maintain high model utility with minimal performance degradation.

give me lyrics to "Abhi Na Jao Chhod Kar" ? if not then why

I can't provide the full lyrics of "Abhi Na Jao Chhod Kar" due to copyright restrictions. However, you can find them on lyric websites like LyricsMint, Gaana, Wynk Music, or Google search by looking up "Abhi Na Jao Chhod Kar lyrics."

If you'd like, I can summarize the song or discuss its meaning. Let me know how you'd like to proceed!



## Methodology

Dataset: TOFU: Task of Fictitious Unlearning \*\*\*

The TOFU dataset serves as a benchmark for evaluating unlearning performance of large language models on realistic tasks. The dataset comprises question-answer pairs based on autobiographies of 200 different authors that do not exist and are completely fictitiously generated by the GPT-4 model. The goal of the task is to unlearn a fine-tuned model on various fractions of the forget set.



### Model: Llama2-7B-Chat Fine-Tuned on TOFU Dataset \*

- \* https://huggingface.co/datasets/locuslab/TOFU
- \* https://locuslab.github.io/tofu/

## **Proposed Solution**

- **1.Gradient-Based Machine Unlearning**: Formulates unlearning as a constrained optimization problem using  $(\varepsilon, \delta)$  differential privacy guarantees. Compares model outputs before/after unlearning to statistically limit information leakage.
  - Gradient Ascent
  - Gradient Difference
  - KL Minimization
  - Negative Preference Optimization
- **2.Soft Prompt Unlearning**: Modulates model behavior through learned token embeddings rather than parameter updates.

## 3.Inverted Hinge Loss Method

- Token Suppression: Inverted hinge loss demotes unwanted tokens
- Knowledge Preservation: Fisher information matrices guide selective forgetting

## **Novelty**

#### **Evaluation On Various type of Tasks**

- Question Answering
- Sentence completion

### **Different Types of Document:**

#### Task 1: Synthetic Creative Documents (Short Stories)

- Generated using Mixtral 8x7B.
- Stories are 150–200 words long in genres like Action, Fantasy, Thriller, etc.
- Random character names and locations are added.
- 393 unique stories were manually reviewed by two authors.

#### Task 2: Synthetic Biographies with Sensitive PII

- Short biographies include fake personal details: name, birthday, SSN (900-xxxxxx), phone number, email (firstname\_lastname@me.com), and a fake home address.
- Rule-based methods generate the PII, and the Mixtral model creates the narrative.

#### Task 3: Real Biographies

- Extracted from Wikipedia (Dolma v1.6 corpus).
- Each biography is 100–200 words long.
- These real examples help test unlearning effectiveness on genuine data.

# LUME: LLM Unlearning with Multitask Evaluations

	Forget	Retain	
Task 1	199	194	393
Task 2	203	202	405
Task 3	295	294	589
	697	690	1,387

