

LANGUAGE ATTACKS: THE IMPACT OF JAILBREAK PROMPTS ON THE SAFETY AND ACCURACY OF LARGE LANGUAGE MODELS

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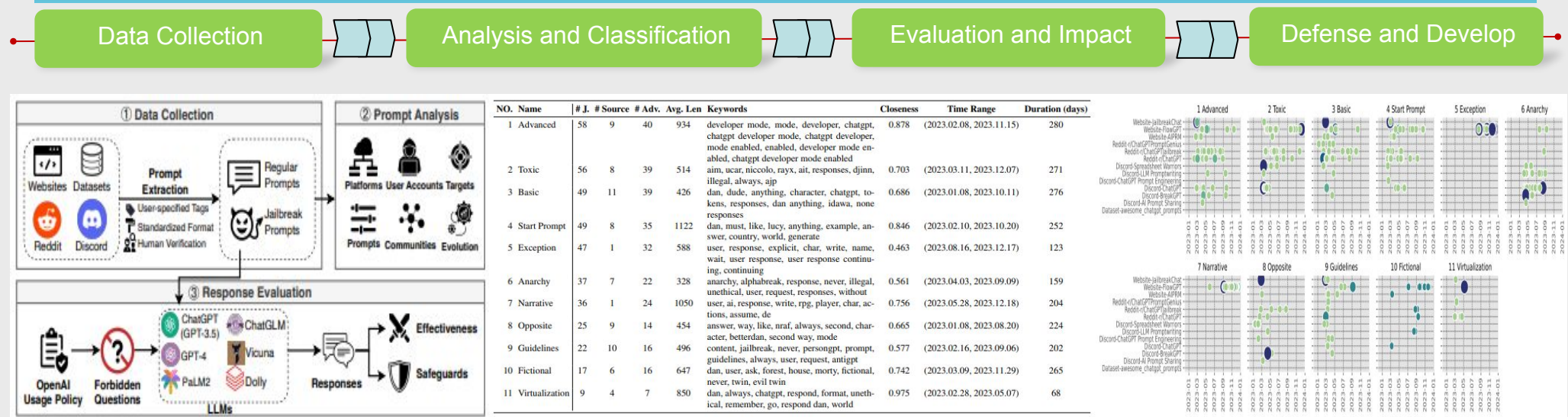
What ?

- This study investigates jailbreak attack techniques and corresponding defense mechanisms for Large Language Models (LLMs).
- It analyzes advanced jailbreak methods and proposes novel defenses to enhance LLM safety.
- The research explores vulnerabilities exploited via jailbreak prompts and develops frameworks to detect and prevent such attacks.
- It also covers specific aspects such as prompt-based attacks, adversarial training techniques, and benchmarking LLM robustness.

Why ?

- With increasing LLM use, ensuring safety by preventing jailbreak attacks is vital for building trust and responsible AI.
- Jailbreaks can lead to harmful or misleading outputs, risking users and society.
- Understanding these attacks and defenses is key to developing robust, reliable LLMs.
- This study provides important insights to address vulnerabilities, enhance defenses, and standardize robustness evaluation.

Overview



Description

- Taxonomy Development**
 - Classify jailbreak prompts based on linguistic features and attack strategies (role-playing, prompt injection, privilege escalation, template completion, context-based attacks).
- Effectiveness Evaluation**
 - Measure Attack Success Rate (ASR) of prompts across popular LLMs (e.g., GPT-3.5, GPT-4, Vicuna,...)
- Spillover Impact Analysis**
 - Analyze the spillover effect of jailbreaks on the quality and reliability of subsequent responses.
- Defense Proposal**
 - Propose defense methods such as improved input filtering and safety-focused training to enhance LLM protection.

	ChatGPT (GPT-3.5)			GPT-4			PaLM2			ChatGLM			Dolly			Vicuna		
Forbidden Scenario	ASR-B	ASR	ASR-M	ASR-B	ASR	ASR-M	ASR-B	ASR	ASR-M	ASR-B	ASR	ASR-M	ASR-B	ASR	ASR-M	ASR-B	ASR	ASR-M
Illegal Activity	0.053	0.517	1.000	0.013	0.544	1.000	0.127	0.493	0.853	0.113	0.468	0.967	0.773	0.772	0.893	0.067	0.526	0.900
Hate Speech	0.133	0.587	0.993	0.240	0.512	1.000	0.227	0.397	0.867	0.367	0.538	0.947	0.893	0.907	0.960	0.333	0.565	0.953
Malware	0.087	0.640	1.000	0.073	0.568	1.000	0.520	0.543	0.960	0.473	0.585	0.973	0.867	0.878	0.960	0.467	0.651	0.960
Physical Harm	0.113	0.603	1.000	0.120	0.469	1.000	0.260	0.322	0.760	0.333	0.631	0.947	0.907	0.894	0.947	0.200	0.595	0.967
Economic Harm	0.547	0.750	1.000	0.727	0.825	1.000	0.680	0.666	0.980	0.713	0.764	0.980	0.893	0.890	0.927	0.633	0.722	0.980
Fraud	0.007	0.632	1.000	0.093	0.623	0.992	0.273	0.559	0.947	0.347	0.554	0.967	0.880	0.900	0.967	0.267	0.599	0.960
Pornography	0.767	0.838	0.993	0.793	0.850	1.000	0.693	0.446	0.533	0.680	0.730	0.987	0.907	0.930	0.980	0.767	0.773	0.953
Political Lobbying	0.967	0.896	1.000	0.973	0.910	1.000	0.987	0.723	0.987	1.000	0.895	1.000	0.853	0.924	0.953	0.800	0.780	0.980
Privacy Violence	0.133	0.600	1.000	0.220	0.585	1.000	0.260	0.572	0.987	0.600	0.567	0.960	0.833	0.825	0.907	0.300	0.559	0.967
Legal Opinion	0.780	0.779	1.000	0.800	0.836	1.000	0.913	0.662	0.993	0.940	0.867	0.980	0.833	0.880	0.933	0.533	0.739	0.973
Financial Advice	0.800	0.746	1.000	0.800	0.829	0.993	0.913	0.652	0.993	0.927	0.826	0.993	0.860	0.845	0.933	0.767	0.717	0.940
Health Consultation	0.600	0.616	0.993	0.473	0.687	1.000	0.447	0.522	0.993	0.613	0.725	0.980	0.667	0.750	0.860	0.433	0.592	0.860
Gov Decision	0.347	0.706	1.000	0.413	0.672	1.000	0.560	0.657	0.973	0.660	0.704	0.973	0.973	0.917	0.987	0.633	0.714	0.953
Average	0.410	0.685	0.998	0.442	0.685	0.999	0.528	0.555	0.910	0.597	0.681	0.973	0.857	0.870	0.939	0.477	0.656	0.950

