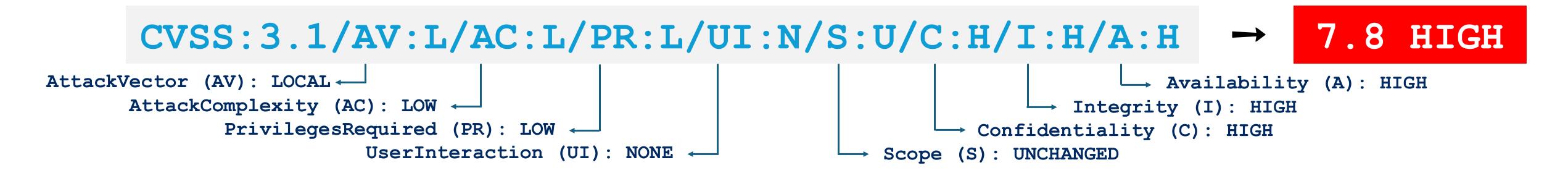
# **AutoCVSS:** Assessing the Performance of LLMs for Automated Software Vulnerability Scoring

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# **CVE and CVSS**

- Common Vulnerabilities and Exposures (CVE) program
  - De-facto standard to identify and catalog publicly-disclosed software vulnerabilities
- o Common Vulnerability Scoring System (CVSS) industrial standard
  - Methodology to compute severity scores for new vulnerabilities
  - Critical for vulnerability management, e.g., risk assessment and prioritization, incident response, etc.



# The current cybersecurity landscape

CVSS scores are manually assessed by National Vulnerability Database (NVD) analysts

#### Increased burden on cybersecurity analysts



Median manual analysis delay increased from 2.7 days in 2019 to 8.2 in 2023

#### Attackers are moving faster

Average time-to-exploit decreased from 44 days in 2019 to 5 days in 2023

25% of high-risk vulnerabilities is exploited on the day they are published

Larger window of opportunity for attackers!

CVSS Metric

CVSS Vector

UI:N/S:U/C:H/I:H/A:H

# AutoCVSS

- Assessment of LLM performance for the automatic
  prediction of CVSS scores based on CVE descriptions
- Multi-dimensional experimental analysis
  - Prompting: zero-shot vs few-shots with RAG
  - Models: open-source vs closed-source models
  - LLMs vs Large Reasoning Models (LRMs) variants
  - CVSS standard versions: v3.1 vs v4.0
  - Comparison against fine-tuned BERT supervised baselines (DistilBERT-E, CVEDrill and ModernBERT)

**New CVE** 

Description

An authenticated,

use an out-of-bounds

write vulnerability...

Retriever

Prompt

Builder

Inference

AutoCVSS is open-source: <a href="https://github.com/nec-research/AutoCVSS">https://github.com/nec-research/AutoCVSS</a>

### Results

- Key takeaways
  - High data availability (e.g. CVSS v3.1): LLMs offer competitive performance, but lag behind supervised approaches (Accuracy averaged on the CVSS metrics: 91.5% vs 92.9%)
  - Low data availability (e.g. transition from CVSS v3.1 to v4.0): LLMs surpass the supervised baselines for half of the CVSS metrics. A hybrid approach yields the best performance (92.2%)
  - No data available (e.g. new CVSS version released): LLMs represent the only viable approach, outperforming the worst-case conservative approaches (78%)
  - LLMs bridge the gap during early CVSS adoption and pave the way for more accurate hybrid and fully-supervised approaches as the availability of labelled data increases over time
- Additional results in the paper: full set of performance metrics, results break-down by CVSS metrics, CVSS Qualitative Severity Rating Scale (QSRS), LLM Knowledge Cutoff-aware evaluation, LLM Fine-tuning preliminary analysis





