Multi-Agent Collaboration in Incident Response with Large Language Models

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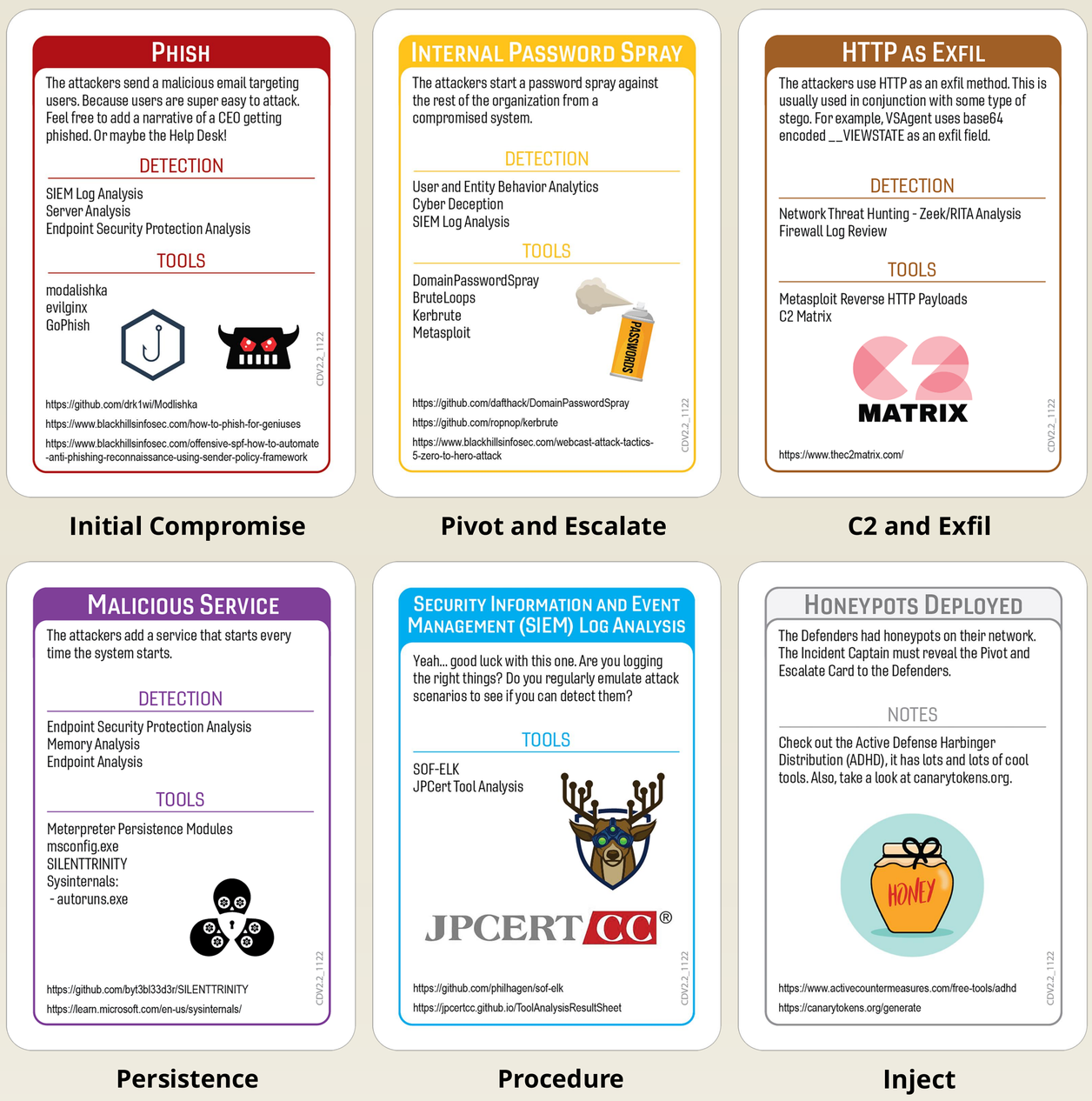
Overview

- **Incident response (IR)** is essential for cybersecurity, requiring **quick decision-making and coordination**.
- **Large Language Models (LLMs)** can serve as **intelligent agents** to enhance collaboration and efficiency.
- **LLM-based multi-agent collaboration** in **cybersecurity** is explored using **Backdoors & Breaches**, a tabletop game designed for IR training.
- Different **team structures (centralized, decentralized, hybrid)** are analyzed to evaluate their impact on IR effectiveness.

Backdoors & Breaches

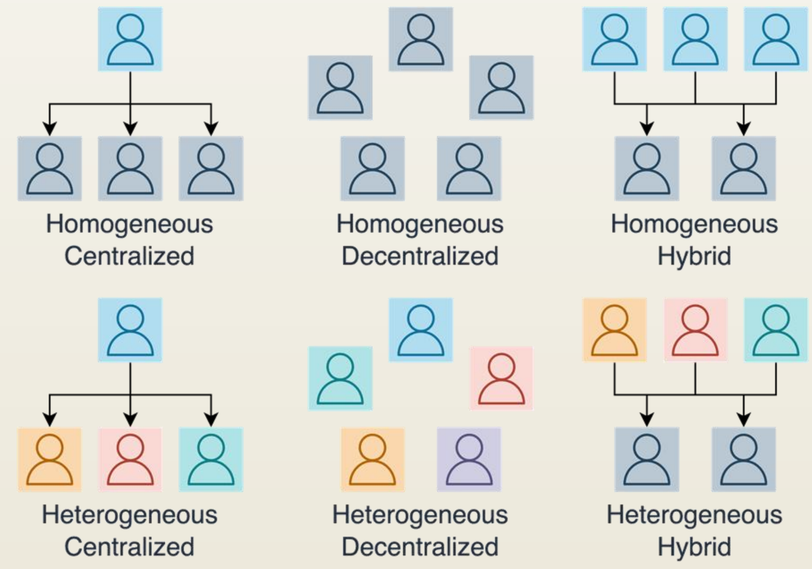
- A **tabletop game** that simulates **real-world cybersecurity incidents**.
- Players take on roles as **incident captain** and **defenders**, working together to **uncover and mitigate attack vectors**.
- The game includes different card types:
  - **Attack Cards:** Represent stages of a cyberattack (e.g., initial compromise, pivot and escalate, command and control (C2) and exfiltration, and persistence).
  - **Procedure Cards:** Defensive strategies used to detect and counter threats.
  - **Inject Cards:** Unexpected events that introduce new challenges.
- The goal is to **reveal all hidden attack cards** within limited turns through strategic decision-making.



Experimental Setup

- **LLM-based agents:** Implemented using **AutoGen**, with structured roles and interactions.
- **Team structures:**
  - **Centralized:** Leadership-driven decision-making.
  - **Decentralized:** Equal decision-making across all agents.
  - **Hybrid:** Mix of leadership and collaboration.
- **Evaluation metrics:**
  - **Success rate** in uncovering all attack stages.
  - **Failure patterns** across different team structures.

Team Structures



Experimental Results

- **Centralized teams** perform well due to clear leadership but may struggle with adaptability.
- **Decentralized teams** leverage diverse expertise but can have coordination challenges.
- **Hybrid teams** balance structured leadership with flexibility, leading to strong performance.
- **LLM-based agents** facilitate IR processes by assisting in decision-making and coordination.

Team	Success	Failure	Pentest	Invalid
Homo-Cen	14	1	2	3
Heter-Cen	13	3	3	1
Homo-Dec	13	1	1	5
Hetero-Dec	12	3	3	2
Homo-Hyb	14	3	2	1
Hetero-Hyb	13	1	2	4

Case Studies

- **Homogeneous Centralized:** Over-reliance on high-modifier procedures led to poor adaptability.
- **Heterogeneous Centralized:** Struggled with prioritization and aligning expert inputs.
- **Homogeneous Decentralized:** Slow decision-making and overuse of standard procedures.
- **Heterogeneous Decentralized:** Lack of coordination led to missed attack indicators.
- **Homogeneous Hybrid:** Misprioritized investigations delayed threat detection.
- **Heterogeneous Hybrid:** Expertise misalignment caused early attack stages to be overlooked.

Conclusion & Future Work

- **LLMs** demonstrate strong potential in **multi-agent collaboration** in **incident response** and **cybersecurity**.
- **Future directions** includes:
  - Improving adaptability of LLMs for unpredictable cyber threats.
  - Extending simulations to real-world cybersecurity environments.
  - Exploring human-LLM hybrid teams for incident response.

References

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Wu, Qingyun, et al. "AutoGen: Enabling Next-Gen LLM Applications via Multi-Agent Conversation." *ICLR 2024 Workshop on Large Language Model Agents*.
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