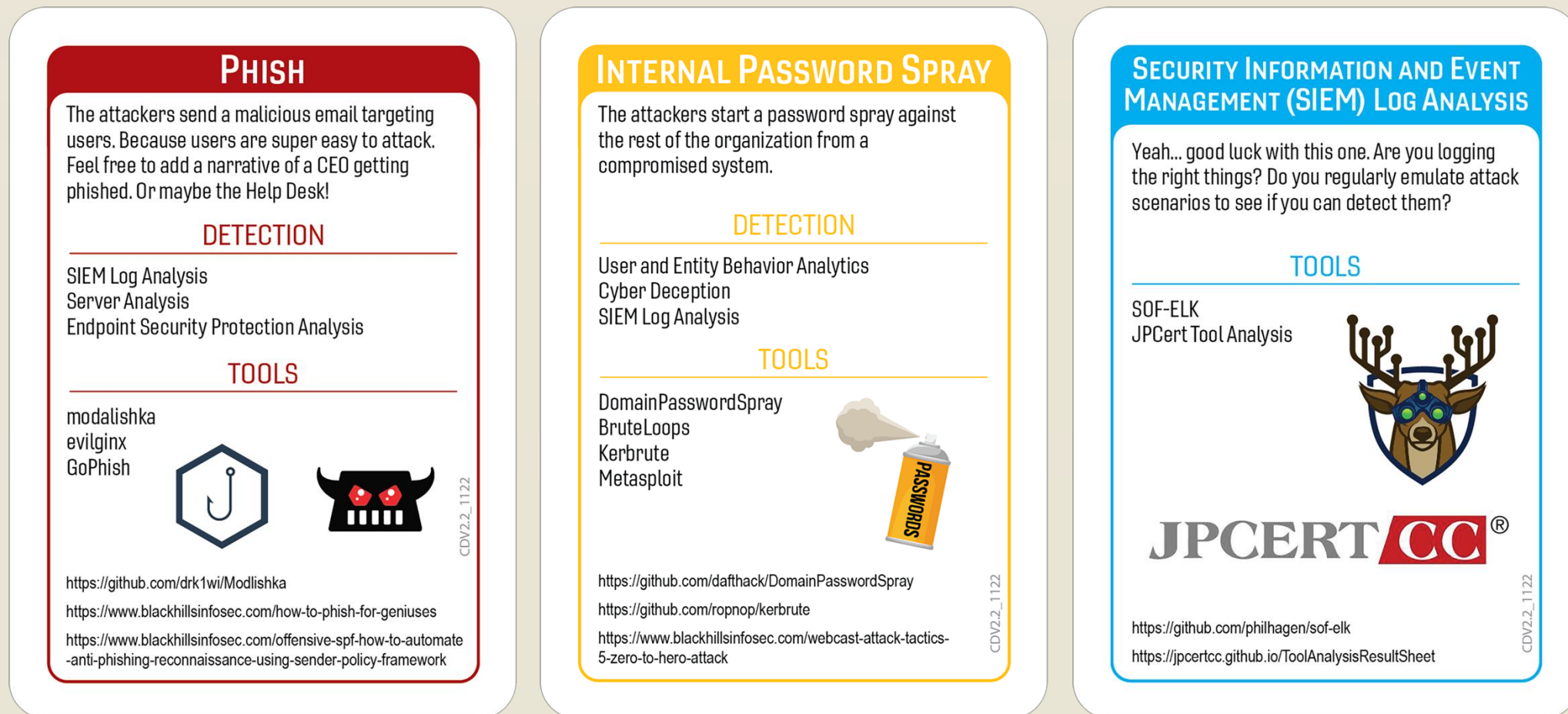


Introduction

- Modern **cyber threats** are complex and multi-stage, demanding rapid detection and coordinated response.
- Effective **Incident Response (IR)** relies on timely decisions, teamwork, and adaptive reasoning.
- Large Language Models (LLMs)** can simulate and support IR teams but often lack access to external knowledge.
- AutoBnB-RAG** extends the AutoBnB framework with **Retrieval-Augmented Generation (RAG)**, enabling agents to fetch and apply external cybersecurity information during collaborative investigations.

Simulation Framework

- The simulation is built on **Backdoors & Breaches (B&B)**, a tabletop game for training **cyber incident response** teams.
- In B&B, defenders uncover a hidden **four-stage attack path** by selecting investigative procedure cards and rolling a **20-sided die**; rolls of **11 or higher** reveal an attack card.
- The four stages are **Initial Compromise**, **Pivot and Escalate**, **Command and Control (C2)** and **Exfiltration**, and **Persistence**.
- Each team has **10 turns** to reveal all stages and win the game; otherwise, the incident remains unresolved.
- AutoBnB-RAG** digitizes this process with **LLM agents** replacing human players.
- The system automates game logic, card handling, and dice rolls to ensure consistent, repeatable evaluations.



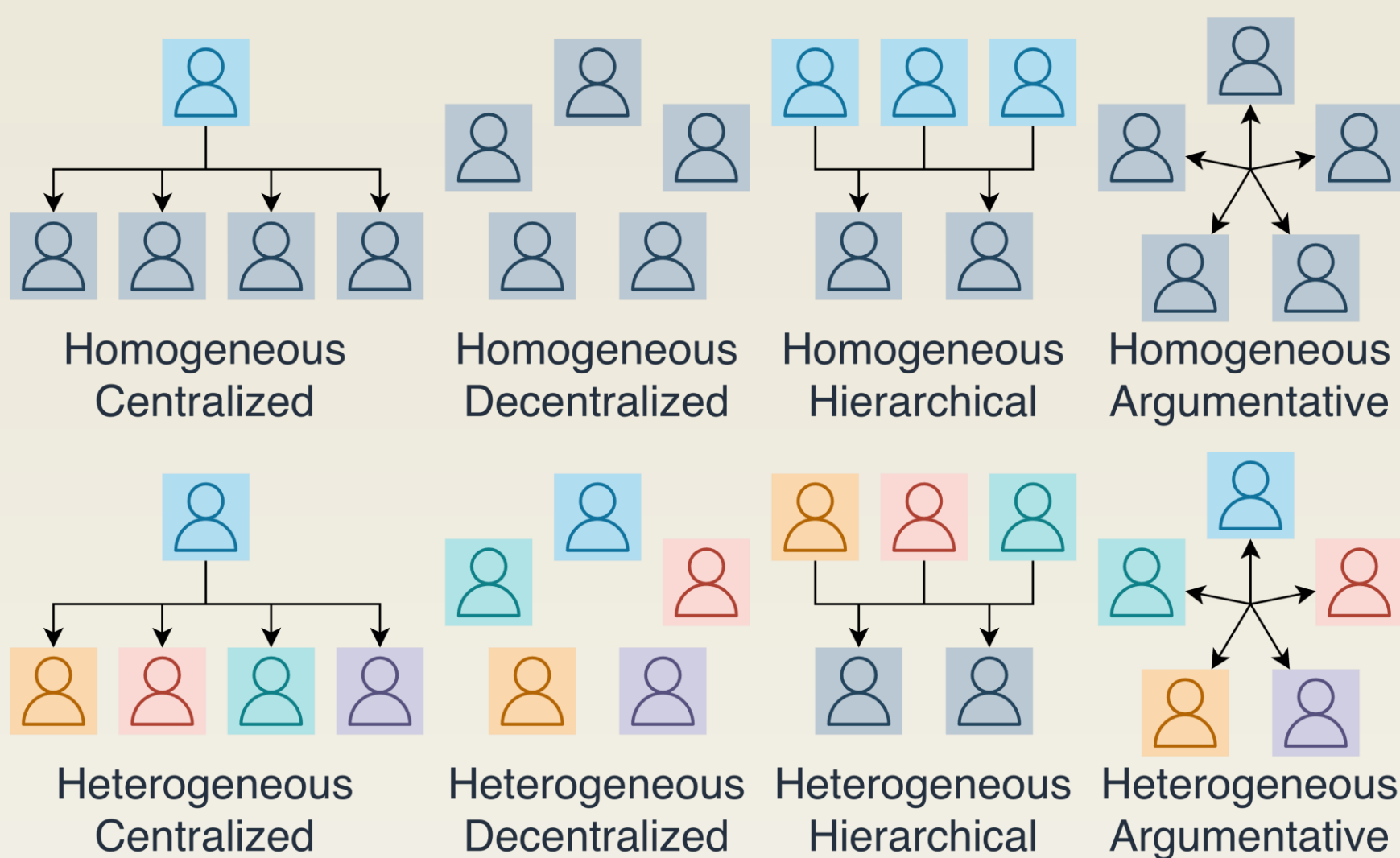
Initial Compromise

Pivot and Escalate

Procedure

Team Structures

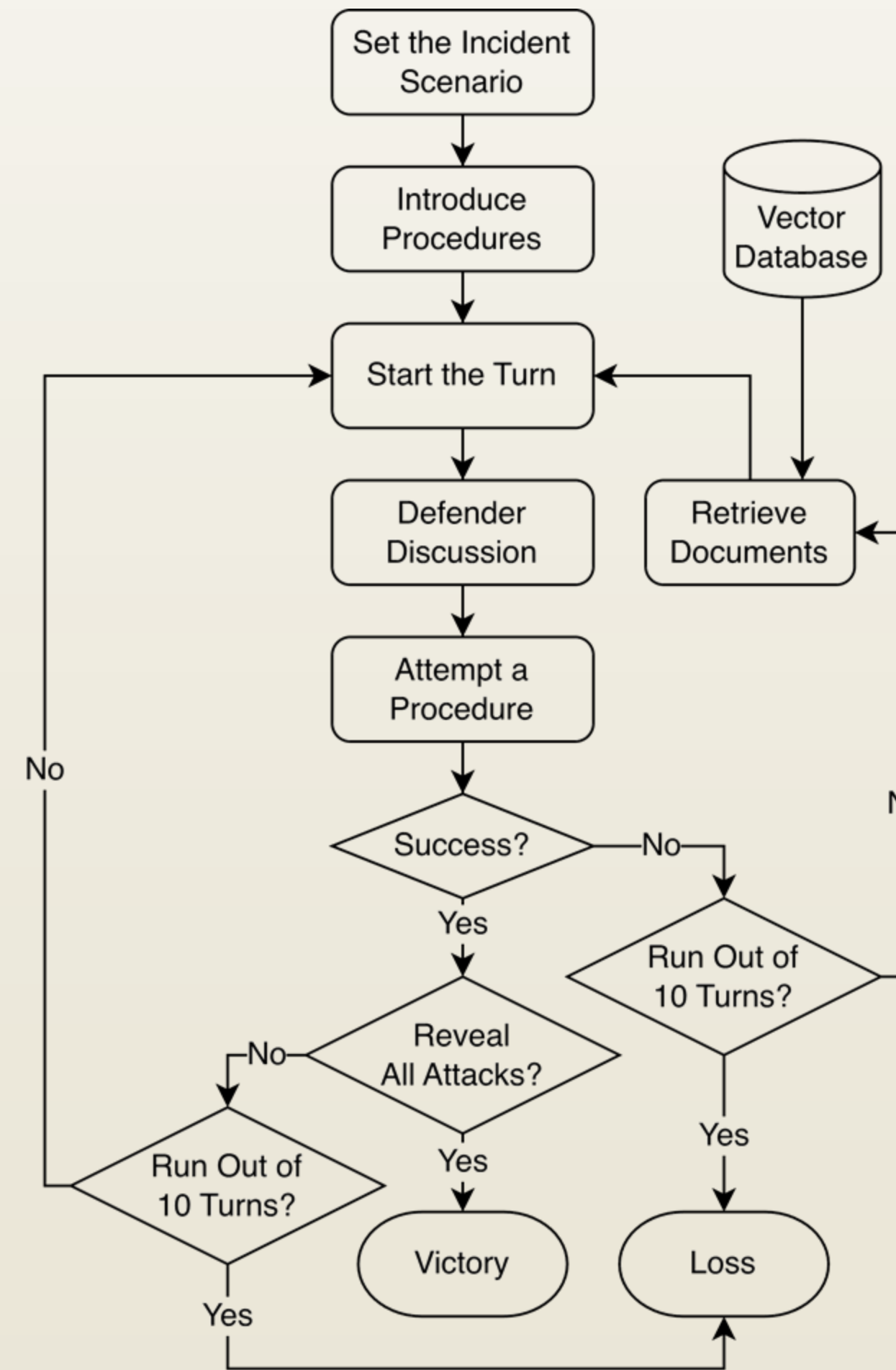
- Eight organizational models** were tested to examine how coordination style affects IR.
- Structures included **centralized**, **decentralized**, **hierarchical**, and **argumentative** variants, each in **homogeneous** or **heterogeneous** form.
- Homogeneous teams** shared the same role expertise, while **heterogeneous teams** combined defenders with different specialties.



Experimental Setup

- Simulations run with **GPT-4o** via **AutoGen** (temperature 0.7).
- Each simulation includes **1 incident captain** and **5 defenders**.
- Each of the 8 team structures evaluated over **30 runs**.
- Top 3 documents** are retrieved using **LangChain** and **Chroma** (OpenAI embeddings).

Retrieval



- Retrieval** is triggered after failed procedures to simulate real-world reference consultation.
- A **retrieval agent** returns relevant context to support the team's reasoning.
- Two knowledge sources**:
 - RAG-Wiki**: 125 curated webpages (Wikipedia, MITRE ATT&CK, Microsoft Learn, etc.)
 - RAG-News**: 100 synthetic narrative incident reports

Experimental Results

- Retrieval improved success rates** across all 8 team structures in simulated incident response.
- Argumentative team configurations**, newly introduced in this study, also benefited from retrieval integration.
- Retrieval-augmented teams** showed clearer investigative progress and fewer repeated failed actions compared to non-retrieval runs.

Win Rates (%) and Performance Gains

Team	Base	RAG-Wiki	RAG-News
Homogeneous Centralized	20.0	50.0 (+30.0)	60.0 (+40.0)
Heterogeneous Centralized	30.0	43.3 (+13.3)	63.3 (+33.3)
Homogeneous Decentralized	33.3	40.0 (+6.7)	43.3 (+10.0)
Heterogeneous Decentralized	26.7	50.0 (+23.3)	50.0 (+23.3)
Homogeneous Hierarchical	23.3	40.0 (+16.7)	43.3 (+20.0)
Heterogeneous Hierarchical	30.0	36.7 (+6.7)	70.0 (+40.0)
Homogeneous Argumentative	23.3	43.3 (+20.0)	46.7 (+23.4)
Heterogeneous Argumentative	30.0	46.7 (+16.7)	53.3 (+23.3)

Ablation Studies

- Retrieval performance remained stable across different **top-k values** and improved with larger **chunk sizes**.

Numbers of Documents

Setting	Top-1	Top-3	Top-5
RAG-Wiki	46.7	50.0	46.7
RAG-News	60.0	60.0	63.3

Document Chunk Sizes

Setting	1k Chars	5k Chars
RAG-Wiki	33.3	50.0
RAG-News	63.3	60.0

Conclusion

- AutoBnB-RAG** integrates retrieval-augmented generation into multi-agent incident response simulations.
- Retrieval grounding improves **decision quality**, **coordination**, and **overall success rates** across diverse team structures.
- The framework demonstrates how **external knowledge** can enhance **reasoning**, **adaptability**, and **realism** in LLM-driven cybersecurity research.

References

- Wu, Qingyun, et al. "AutoGen: Enabling Next-Gen LLM Applications via Multi-Agent Conversation." *ICLR 2024 Workshop on Large Language Model Agents*.
- Liu, Zefang. "Multi-Agent Collaboration in Incident Response with Large Language Models." *AAAI 2025 Workshop on Multi-Agent AI in the Real World*.
- Liu, Zefang. "AutoBnB: Multi-Agent Incident Response with Large Language Models." *2025 13th International Symposium on Digital Forensics and Security (ISDFS)*. IEEE, 2025.



SCAN ME