Chain Games: Powering Autonomous Threat Hunting

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Threat hunting is a critical part of cyber defense, but the amount of data available to threat hunters is overwhelming.

To develop effective autonomous threat hunting techniques, we are developing Chain Games, a set of games in which threat hunting strategies can be evaluated and refined.

Chain Games: Powering Autonomous Threat Hunting

Motivation and Approach

What is Threat Hunting?

Intrusion Detection/Prevention

• How do we keep the attackers out?

Incident Response

• How do we mitigate what the attackers did?

Threat Hunting

 How do we find/remove the attackers who got in?



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Benefits of Autonomous Threat Hunting

Threat hunting takes time and skill.

Inexpensive, faster hunting could:

- Investigate more data sources
- Coordinate for coverage
- Help triage human threat hunts

The key to faster, less expensive threat hunting is autonomy.



Cyborg Security. *The Threat Hunter's Hypothesis*. https://www.cyborgsecurity.com/library/guides/the-threat-hunters-hypothesis-2/

Approaches to Autonomy

Long-term goal: autonomy

- Predication
- Investigation
- Conclusion

Short-term goal: modeling

- Quantitatively evaluating and developing strategies
- Rapid strategic development
- Capturing the adversarial quality of threat hunting activity

Cyber Deception Games (CDG) and Cyber Camouflage Games (CCG)

2018: Cyber Deception Games [1]

- Situates work in the Cyber Kill Chain
 - Focuses on reconnaissance
- Is a zero-sum game
- Defender is deceiver

2019: Cyber Camouflage [2]

- Is extended to general-sum games
- Defender is still deceiver

Schlenker A, Thakoor O, Xu H, Fang F, Tambe M, Tran-Thanh L, Vayanos P, Vorobeychik Y, "Deceiving cyber adversaries: A game theoretic approach," in Proceedings of the 17th International Conference on Autonomous Agents and MultiAgent Systems, International Foundation for Autonomous Agents and Multiagent Systems, 2018, pp. 892-900.

Thakoor O, Tambe M, Vayanos P, Xu H, Kiekintveld C, Fang F. "Cyber Camouflage Games for Strategic Deception," in Decision and Game Theory for Security, Springer International Publishing, 2019, pp. 525-541.

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Kill/Attack Chains



The Cyber Kill Chain



Ransomware (NotPetya)

APT Campaign (DriftingCloud)

Attack behavior is often conceptualized as chains.

- Decomposes attacks
- Categorizes attack behaviors

ISACA Now Blog. *Ransomware Analysis – Executions Flow and Kill Chain.* https://www.isaca.org/resources/news-and-trends/isaca-now-blog/2017/ransomware-analysisexecutions-flow-and-kill-chain

Lockheed-Martin. *The Cyber Kill Chain*. https://www.lockheedmartin.com/en-us/capabilities/cyber/cyber-kill-chain.html

Volexity. DriftingCloud: Zero-Day Sophos Firewall Exploitation and an Insidious Breach. https://www.volexity.com/blog/2022/06/15/driftingcloud-zero-day-sophos-firewall-exploitation-andan-insidious-breach/

Chain Games: Powering Autonomous Threat Hunting

Simple Chain Games

Chain Games –1

Chain Games are played on state chains.

- States represent positions in the network conveying advantage (or disadvantage) to the attacker.
- The utility and cost of occupying a state can be quantified.
- Progress through the state chain motivates the attacker; stopping progress motivates the defender.



Chain Games –2

Rules

- Two players (Attacker and Defender)
- Fixed number of turns
- General-sum (with zero-sum components)
- Simultaneous action







Chain Game Version 0: Actions and Payoffs

Attacker Actions

- Advance A (Cost: 1)
 - Advances to next state in chain

Defender Actions

- Defend D (Cost: 1)
 - Negates attacker A action

Common Action

- Wait W (Cost: 0)
 - No additional effect

Payoffs

- Attacker gets positional payoff for each advance
- Defender gets negated positional payoff for each advance



Uniform-Value Chain

Chain Game Version 0: Dominant Strategies

Attacker

• Always A

Defender

• Never D (i.e., always W)

Takeaways

- The full value of a strategy is its utility across **all opponent strategies**
- Changes in costs/payoffs lead to different analytic outcomes



	WW	AW	WA	AA
WW	(0, 0)	(-2, 1)	(-2, 1)	(-4, 2)
WD	(-1, 0)	(-3, 1)	(-1, -1)	(0, -3)
DW	(-1, 0)	(-1, -1)	(-3, 1)	(-3,0)
DD	(-2, 0)	(-2, -1)	(-2, -1)	(-2, -2)

Payout Matrix Over Two Turns, Uniform-Value Chain

Introducing Camouflage

Attacker Actions

- Noisy Advance N
- Camouflaged Advance C
- C more costly than N

Defender Actions

- Weak Detect L(ow), Strong Detect H(igh)
- L only detects N
- H more costly than L





Payout Matrix Over Three Turns

Introducing Camouflage – Dominant Strategies

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Attacker

• Always W

Defender

• HLH



	WWW	WWN	WWC	
WWW	(0, 0)	(-1, 0)	(-2, 0)	
WWL	(-3, 2)	(-1, -1)	(-2, -1)	
WWH	(-3, 1)	(-4, 1)	(-2, -2)	
WLW	(-3, 2)	(-4, 2)	(-5, 2)	
WLL	(-6, 4)	(-4, 1)	(-5, 1)	
WLH	(-6, 3)	(-7, 3)	(-5, 0)	
WHW	(-3, 1)	(-4, 1)	(-5, 1)	
WHL	(-6, 3)	(-4, 0)	(-5, 0)	
WHH	(-6, 2)	(-7, 2)	(-5, -1)	
LWW	(-3, 2)	(-4, 2)	(-5, 2)	
LWL	(-6, 4)	(-4, 1)	(-5, 1)	
LWH	(-6, 3)	(-7, 3)	(-5, 0)	
LLW	(-6, 4)	(-7, 4)	(-8, 4)	
LLL	(-9, 6)	(-7, 3)	(-8, 3)	
LLH	(-9, 5)	(-10, 5)	(-8, 2)	
LHW	(-6, 3)	(-7, 3)	(-8, 3)	• • •
LHL	(-9, 5)	(-7, 2)	(-8, 2)	
LHH	(-9, 4)	(-10, 4)	(-8, 1)	
HWW	(-3, 1)	(-4, 1)	(-5, 1)	
HWL	(-6, 3)	(-4, 0)	(-5, 0)	
HWH	(-6, 2)	(-7, 2)	(-5, -1)	
HLW	(-6, 3)	(-7, 3)	(-8, 3)	• • •
HLL	(-9, 5)	(-7, 2)	(-8, 2)	• • •
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Payout Matrix Over Three Turns (detail)

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More Complex Chains

- There is no dominant pure strategy for attacker or defender.
- Non-uniform chains represent more realistic attack conditions.
- Initial infection is valuable.
- Some positions of advantage may have value that justifies taking on intermediate risk.

 $(s_0) \xrightarrow{2} (s_1) \xrightarrow{-1} (s_2) \xrightarrow{4} (s_3)$

2: WWW	2: WWA	2: WAW	2: WAA	2: AWW	2: AWA	2: AAW	2: AAA
0	1 3	1 3	0	1 3	0	0	0

1: WWW	1: WWD	1: WDW	1: WDD	1: DWW	1: DWD	1: DDW	1: DDD
*	1 33	1 33	0	1 33	0	0	7



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Future Work

Enriching the Game Space

Evidence

- The game is augmented with an information vector (IV)
 - Indicators of attacker activity
- Different kinds of attacker actions change different parts of the IV
- Defender actions collect evidence from IV
- New Defender *R*(emediate) actions stop attacker advances or evict the attacker



Simulation

Simulation is a way to model activities that are difficult to analyze exhaustively.

Simulation can model behavior that violates assumptions of rationality.

```
# Information common to all games of this type
_GAME_TYPE = pyspiel.GameType(
    short name="chain game v0".
    long_name="chain game version 0".
    dynamics=pyspiel.GameType.Dynamics.SIMULTANEOUS,
    chance mode=pvspiel.GameTvpe.ChanceMode.DETERMINISTIC.
    information=pyspiel.GameType.Information.IMPERFECT_INFORMATION.
    utility=pyspiel.GameType.Utility.ZERO_SUM,
    # The other option here is REWARDS, which supports model-based
    # Markov decision processes. (See spiel.h)
    reward_model=pyspiel.GameType.RewardModel.TERMINAL,
    # Note again: num players doesn't count Chance
   max num players=len(Players).
   min_num_players=len(Players),
    provides_information_state_string=False,
    provides information state tensor=False.
    provides_observation_string=False.
    provides_observation_tensor=False,
    provides_factored_observation_string=False,
    # We can worry about parameters later
    parameter_specification={},
```

Game Specification with OpenSpiel [4]

[4] Deepmind. OpenSpiel: A Framework for Reinforcement Learning in Games. https://github.com/deepmind/open_spiel

Mapping to the Problem Domain

- Reflect patterns of adversary behavior in chains
 - Distribution of positional payoffs
 - Introduce attack graphs and attacker choice
- Reflect relationships between network activities (Attacker advances) and evidence in IV
- Evaluate real-world threat hunting strategies in simulation





Chain Games: Powering Autonomous Threat Hunting

Extra Slides

Introducing Camouflage –2

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Attacker Actions

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- Camouflaged Advance C
- C more costly than N

Defender Actions

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Payout Matrix Over Three Turns (detail)

Kill/Attack Chains



With 'Hands on Keyboard' access, intruders accomplish their original goals

Attack behavior is often conceptualized as chains.

- Decomposes attacks
- Categorizes attack behaviors

Attack graphs are a composition of attack chains.

The Cyber Kill Chain graphic is reused with permission from Lockheed Martin Corporation. [3]

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[3] Lockheed-Martin. The Cyber Kill Chain. https://www.lockheedmartin.com/enus/capabilities/cyber/cyber-kill-chain.html