

Industry White Book

AI-Driven Investigation Framework Based on Behaviour Chain Semantics

Let Ai Investigate, Let Human Decide

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Patent Disclosure and Confidentiality Notice

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The invention was formally filed as a patent application with the China National Intellectual Property Administration (CNIPA) and the United States Patent and Trademark Office (USPTO) on June 19, 2025, with a simultaneous international filing under the Patent Cooperation Treaty (PCT). The filing date constitutes the priority date, protected **under PCT international novelty provisions**.

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I.Executive Summary

The proposed model, named the **Semantic Chain Security Model (SCSM)**, integrates:

- 1.Natural language understanding (LLM/NLP)
- 2.Structured behavior chain modeling
- 3.AI-based semantic reasoning
- 4.Micro-model anomaly scoring
- 5.Human-machine collaboration
- 6.Knowledge feedback and evolution mechanisms

This model transforms isolated logs into structured behavior chains, allowing AI to interpret and investigate attack paths in ways previously only possible for experienced human analysts.

At the core of the model is the belief that AI cannot investigate what it cannot understand—and without structured behavior sequences, AI lacks the semantic substrate required for reasoning.

The Key Innovations of this model Include:

1.Natural Language Interfaces

Analysts use English, Chinese, or German to initiate investigations—no more SPL or KQL.

2.Democratized Threat Investigation

Anyone who can describe a suspicious behavior can launch an AI-powered inquiry.

3.Behavior Chain Construction

Logs are chronologically and semantically transformed into coherent behavioral narratives for each actor (IP, host, user).

4.Semantic Reasoning Engine

AI doesn't just match patterns — it infers intent, detects anomalies in context, and identifies pivotal transitions.

5.Pivot Point Detection (BPP)

Novel concept of pivot strength identifies critical semantic transitions (e.g., from reconnaissance to exploitation), even when actions appear “normal.”

6.Expert Feedback + Knowledge Write-Back

Human-validated behavior chains are written back into a dynamic knowledge base, enabling memory, comparison, and cumulative learning.

7.Micro-Level Anomaly Models

Pluggable, role-specific models analyze single logs for character usage, API paths, time-of-day anomalies, and more.

8.Three-Field Mapping for Accurate Attribution

Introduces external_ip → srcip → dstip mapping to resolve actor ambiguity in edge device logs (e.g., WAF).

9.Investigative Layer Standardization Proposal

Recommends the formal inclusion of an “ Investigative Layer ” into security standards (SIEM/SOC/XDR), bridging the gap between alert generation and incident response.

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II.Problem Statement and Problem Definition

A.Structural Gaps in the Current Industry Framework

1.Existing Security Operations Investigation Architecture:

Mainstream security platforms (such as SIEM, XDR, EDR) generally follow the operational workflow:

Log Collection → Anomaly **Detection** (via Rules/Models) → **Alert** Generation → **Manual Response**

2.Core Issues:

- An alert is **not an investigation** result—it is merely a detection outcome.
- A response is not an analytical judgment—it is often a default action.
- The system lacks the ability to connect behavior fragments into a coherent "attack narrative"—analysts are forced to manually piece everything together.

B.The Missing Investigative Layer: A Semantic Blind Spot in the Industry

Table1: Global Standardization Landscape

| Security Domain | Standardization Status | Representative Standards |
|-----------------|---|--|
| Prevention | Mature / Established | NIST CSF, ISO/IEC 27001 |
| Detection | Mature / Established | MITRE ATT&CK, NIST SP 800-94 |
| Investigation | Absent — No globally unified architectural standard | (Gap — no established global framework) |
| Response | Partially Standardized | NIST SP 800-61 (Computer Security Incident Handling Guide) |

1.No Behavior Chain Structure → Logs Are Fragmented

Each log entry is treated as an isolated "point"—there is no structural connection to form a "line" or "chain".

- There is no structured perspective to answer:
- Who performed the behavior? When did it happen? What actions were taken?

2.No Structural Language → No Standardized Modeling

Without an Entity Behavior Database (EBD), it is impossible to model behavior around a specific actor.

Behavior coordinates—such as which step, what stage, or is this a critical point—are entirely absent.

3.No Attack Path Mapping → No Stage-Based Intent Reasoning

Frameworks like MITRE ATT&CK define tactical stages of attacks, but the industry lacks mechanisms to map raw behavior sequences to these stages.

Can we determine if an attack has escalated? Or if it's preparing for outbound communication?

Without structure, AI cannot make such judgments—it can only passively assign scores.

C.Semantic Disconnection: Alert ≠ Explanation, Detection ≠

Reasoning

Current Reality:

Detection systems are good at identifying anomalous behaviors,

but true investigation requires understanding the behavioral path, intent at each stage, and the actor behind the actions.

The absence of an investigative layer leads to semantic misinterpretation:

Table2: Traditional Detection Conclusion VS SCSM Semantic Interpretation

| Example Behavior Chain | Traditional Detection Conclusion | SCSM Semantic Interpretation |
|--|--|---|
| Failed login ×50 → Successful login | Successful login is treated as “normal” | Successful login is a pivot point, indicating a shift in attack stage |
| 3 failed logins → Brute-force attack detected | Normal fluctuation in login outcome is flagged as “suspicious” | Normal behavior by the same actor is misclassified as an attack |

SCSM is not about detecting anomalies — it's about reconstructing the attack story.

Table3: Seven Consequences of a Missing Investigative Layer

| Missing Structural Element | Immediate Symptom | Systemic Impact |
|-----------------------------|--------------------------------|---|
| No Behavior Chain | Alerts are fragmented | No context; AI cannot reconstruct a coherent "story" |
| No Role-Based Path Mapping | Actor confusion across logs | Cannot determine if actions belong to the same entity |
| No Stage Coordinates | Logs lack positional reference | AI cannot judge whether behavior belongs to an attack phase |
| No Pivot Point Recognition | Phase transitions are vague | Attack escalations blend with normal activity |
| Missing Structural Element | Immediate Symptom | Systemic Impact |
| No Semantic Query Interface | No semantic search | Analysts must rely on handwritten |

| | capability | KQL/SPL queries |
|------------------------------------|-----------------------|--|
| No Expert Feedback Mechanism | Human insight is lost | Judgments cannot be structured and written back for reuse |
| No Knowledge Base Growth Mechanism | AI has no "memory" | Cannot reuse or evolve historical investigative structures |

Let's take another look at three major unsolved challenges in today's cyber security landscape & Industry Exploration Trends :

Table4: Three major unsolved challenges

| No. | Threat Type / Problem | Current Limitation | Root Cause Analysis | Reasoning |
|-----|---|--|--|--|
| 1 | Zero days False Positives / Missed Alerts / Alert Overload | Single-rule decisions lack context, easily bypassed or become ineffective | No behavior chain structure; rules operate without semantic context | SCSM does not rely on known signatures or predefined indicators. Instead, it identifies abnormal behavior chain structures and intent transition signals by reasoning over behavior chain paths and attack phases—even when the individual actions may appear normal, as is often the case in 0-day attacks. |
| 2 | APT (Advanced Persistent Threats) | Log fragmentation, no context, no chronological action reconstruction | No unified behavior chain per entity; semantic gaps remain | SCSM constructs role-centric temporal chains through the Entity Behavior Database (EBD) , ensuring that the attack path remains continuous even across devices and systems. |
| 3 | DDoS (Distributed Denial of Service) | Knowledge can't accumulate; no reusable templates for identifying patterns across actors | No knowledge feedback or generalization mechanism; behaviors remain isolated | SCSM does not determine DDoS attacks based on access frequency alone; instead, it evaluates behavior chain similarity and overlapping multi-source activity paths to identify coordinated role-based attack behaviors. |

While some vendors claim to offer "investigation" capabilities, Actually, their systems typically only aggregate or correlate data after alerts have been triggered. In the SCSM (Semantic Chain Security Model) framework, **true investigation requires:**

- 1.Reconstruction of behavioral sequences — not isolated or single-point queries
- 2.Role continuity across actions — not raw IP matching
- 3.Path-based semantics and event coordinates — not flat or surface-level correlations
- 4.Reasoning over tactical phase transitions — not rigid rule-based triggers
- 5.Feedback-integrated memory and structured knowledge base

Therefore, these so-called "investigation" features in commercial platforms do not constitute

semantic chain investigation. They remain alert-centric, descriptive, and superficial, lacking the structural layer that should exist between detection and response.

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Table5: Global Vendor Gap vs SCSM

| Vendor | Investigation Capability Claim | Actual Layer Coverage | Semantic Behavior Chain Modeling | Pivot Strength / Phase Reasoning | Expert Feedback + Knowledge Memory | Comments |
|-----------------------|--|-------------------------------|------------------------------------|---------------------------------------|---|--|
| Google Chronicle | Natural language search + threat hunting | Alert Correlation Layer | No | No | No | Primarily enhances Alert Searchability |
| Microsoft Sentinel | KQL + Workbooks + Hunting Queries | Alert Correlation Layer | No | No | No | Focused on Alert Enrichment & Hunting |
| Splunk Security Suite | SPL-based correlation & dashboards | Alert/Detection Layer | No | No | No | Powerful detection but lacks Behavior Chain semantics |
| CrowdStrike Falcon | EDR + Threat Graph | EDR-level Process Tracking | Partial (Process Chains) | No | No | Strong EDR focus, lacks cross-system Behavior Chain and semantic investigation |
| Palo Alto Cortex XDR | Analytics + Playbooks | XDR Alert Handling | Partial (limited path correlation) | No | No | Emphasizes playbook-driven response, no semantic layer |
| SCSM | Semantic Chain Modeling + AI Reasoning | Dedicated Investigation Layer | Full Behavior Chain Modeling | Yes, Pivot Strength + Phase Reasoning | Yes, Expert Feedback + Knowledge Memory | Fills global architecture gap between Detection and Response |

In Summary:

Whether it's the gradual failure of traditional rule-based systems against novel attack scenarios, or the current industry's fragmented attempts to integrate AI, both trends point to a fundamental root cause: the absence of an investigative layer and structured behavior chains.

III.Solution Theory&Solution Framework

Table6: Industry Paradigm Evolution for cyber security

| Evolution Stage | Paradigm Innovation | Representative Technologies/Products | Industry Architectural Layer | SCSM Contribution | Impact |
|----------------------------|---|--------------------------------------|----------------------------------|---|--|
| First Stage (1990s) | Perimeter Security (Firewall) | Cisco PIX, Checkpoint FW | Perimeter Layer | — | Defined boundary security |
| SecondStage (2000s) | Real-time Intrusion Detection (IDS/IPS) | Snort, Suricata, Bro/Zeek | Detection Layer | — | Introduced real-time detection |
| Third Stage (2005–2015) | Centralized Log Analysis (SIEM) | Splunk, ArcSight | Detection → Alert Layer | — | Enabled cross-device correlation |
| Fourth Stage (2015–2020) | Tactical Phase Modeling (MITRE ATT&CK) | MITRE ATT&CK Framework | Detection → Attack Understanding | — | — |
| Fifth Stage (2017–Present) | Automated Response (SOAR) | Cortex XSOAR, DFLabs | Response Layer | — | Enabled response automation |
| Sixth Stage (2025.6) | Investigation Layer | SCSM | New Investigation Layer | First structured investigation layer definition | Bridges Detection → response gap |
| Seventh Stage (2025.6) | Knowledge Loop / AI-Evolving Layer | SCSM | New Knowledge Memory Layer | First expert feedback + AI learning loop | Establishes AI-driven investigative paradigm |

Over the past three decades, the cybersecurity field has undergone multiple paradigm shifts — each introducing a new architectural layer: perimeter defense, real-time detection, centralized analysis, tactical phase modeling, and automated response.

Yet **two critical architectural gaps remain**: the absence of a native Investigation Layer and a structured, AI-driven Knowledge Memory Layer. The Semantic Chain Security Model (SCSM) addresses **both gaps simultaneously** — defining the world's **first structured Investigation Layer**, and **introducing an AI-evolving Knowledge Loop** that enables dynamic learning and memory within security operations. This marks not one, but two paradigm breakthroughs in cybersecurity architecture — moving beyond traditional detection and response toward **AI-driven investigation** and **knowledge-based adaptive security**.

A.Solution Theory

Original Semantic Constructs Introduced for AI-Driven Security Investigation Architecture.As the original author, I hereby introduce the following core semantic constructs into the AI-driven security incident investigation system. These foundational concepts are designed to support behavior chain modeling, path inference, human-AI consensus, and knowledge base evolution through write-back mechanisms.

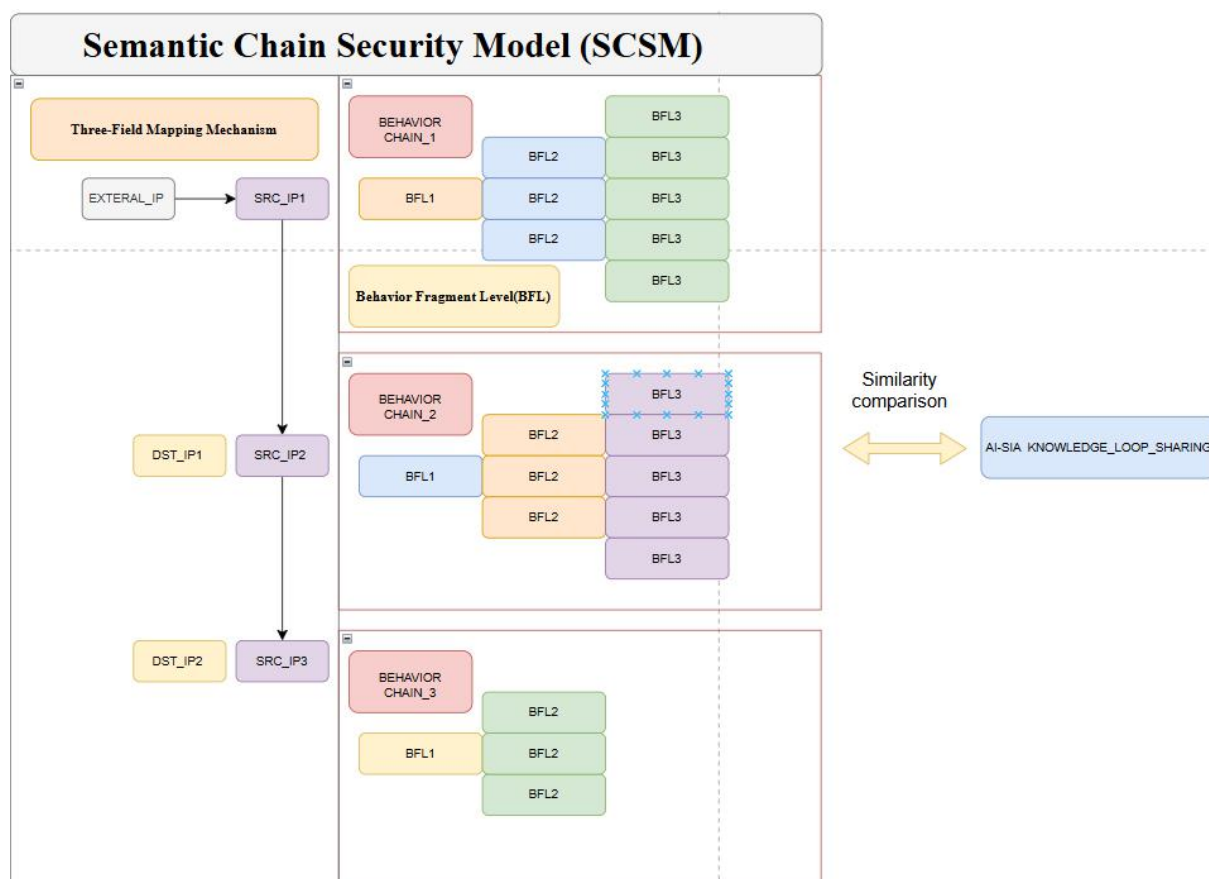


Figure1: Semantic Chain Security Model

1.Semantic Chain Security Model (SCSM)

The Semantic Chain Security Model (SCSM) is an AI-driven security operations framework centered on behavioral semantics. It establishes a full-cycle investigation process that spans from log ingestion, behavior reconstruction, and semantic inference to expert validation and knowledge write-back.

SCSM breaks away from the traditional SOC paradigm built on rule-based detection and manual log querying. By introducing a structural semantic layer, it empowers AI to understand behavior—enabling proactive risk assessment, attack path recognition, and continuous self-evolution.

2.Character/Entity Behavior Database (EBD)

The Entity Behavior Database (EBD) is a structured, role-centric data repository that

chronologically records and organizes all observable actions performed by a specific entity—such as a user, IP address, host, or device.

It provides the foundational structure for behavior reconstruction and actor-centric semantic reasoning in security investigations.

3. Knowledge Write-Back Mechanism

The Knowledge Write-Back Mechanism refers to the process by which human security experts review, validate, and optionally revise the AI-generated behavior chains and risk evaluations.

This mechanism is the cornerstone of human-AI collaboration. It ensures that expert judgments are not lost but instead structurally recorded and fed back into the system's evolving knowledge base, thereby improving semantic inference over time.

4. Three-Field Mapping Mechanism

The Three-Field Mapping Mechanism addresses semantic ambiguities in traditional binary log field mappings (e.g., srcip, dstip), particularly in edge device logs (e.g., WAF, proxy).

It introduces a third semantic field, external_ip, to explicitly represent the origin of the access request. This results in a clarified behavioral structure:

External_ip → srcip → dstip

enabling accurate actor attribution and path-level modeling in a three-stage behavior chain format.

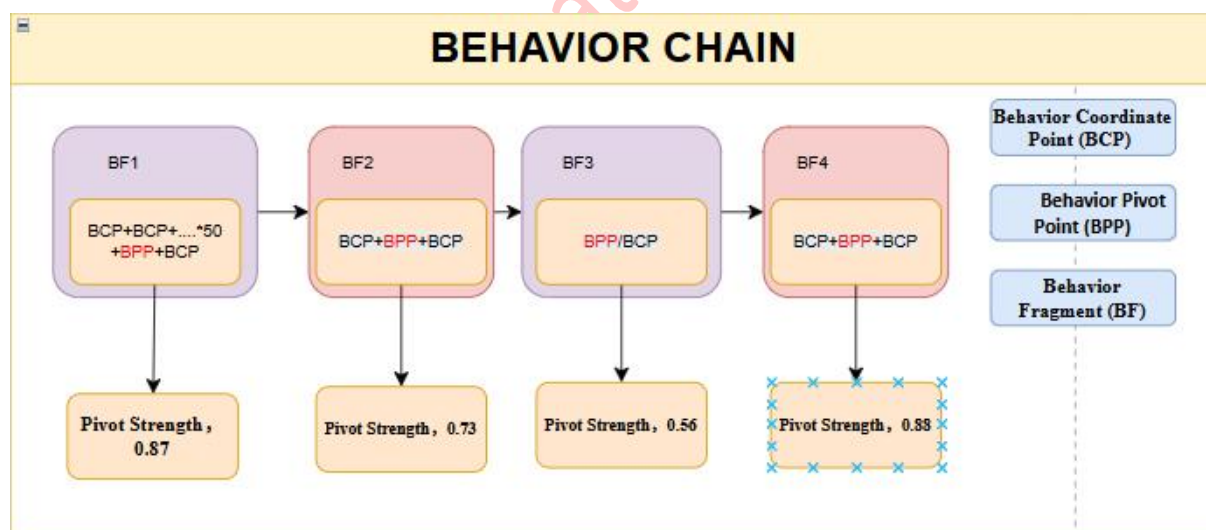


Figure2: Behavior Chain

5. Behavior Chain / User Behavior Chain (BC / UBC)

A Behavior Chain is a time-ordered, semantically consistent sequence of behavioral events linked to a single entity (e.g., user, IP, host, or device) within a defined time window. It integrates both normal and abnormal behaviors and provides the structural foundation for AI to perform reasoning, identify anomalous stages, and infer malicious intent. It is the precondition for building an attack chain.

6. Behavior Chain Path (BP)

A Behavior Chain Path (BP) represents the semantic linkage of behavior **chains across different hosts** or entities, describing how an intrusion or action propagates between them. When tracking activities across systems — for example, **from Host A to Host B** — the sequential transition $A \rightarrow B$ forms a Behavior Chain Path. This cross-host reconstruction enables end-to-end tracing of attacker behavior. Each BP is composed of multiple Behavior Chains (BCs) connected through source – destination relationships identified by the Three-Field Mapping Mechanism (e.g., SRC_IP, DST_IP).

Behavior Fragment (BF) Structural Components:

Behavior States: Different outcomes or phases of the same behavior (e.g., success, failure).

Behavior Coordinate Point (BCP): An individual state event in the process (e.g., one failed login).

Behavior Pivot Point (BPP): A special type of BCP representing a decisive transition to another phase (e.g., multiple failures followed by a final success).

Conditions: Contextual rules for evaluating the fragment (e.g., time window, frequency thresholds, file type).

Table1: Structural Variables Usable as Trigger Conditions

| Category | Description | Example |
|-------------------|--|---|
| Time-based | Specific time windows or event intervals | 1:00–5:00 AM; high-frequency behavior within 10 minutes |
| Count / Threshold | Total number of BCPs; existence of a BPP | More than 100 failed logins AND at least one successful login |
| Category | Description | Example |
| Keyword / Regex | Presence of sensitive commands, paths, or indicators in logs | curl, /tmp/myscript.sh, wget, port |
| Log Type Ratio | Proportion of specific log types among all BCPs | 60% from endpoint logs, 40% from network logs |
| Logical Formula | Boolean expression to define abnormal behavior patterns | (BPP = True) AND (BCP > 10) |
| AI Micro-Model | AI-based scoring and/or tag evaluation | Score > 0.75 AND tag = "Suspicious behavior" |

Behavior Fragment (BF) Key Features:

Captures the **entire lifecycle of a behavior**, not just isolated events.

Can be benign or malicious, depending on the combination of states.

Serves as the fundamental analytic unit for mapping behaviors to the attack lifecycle.

Examples of Behavior Fragments

Example 1: Login Behavior (BF) — Abnormal Brute-Force Behavior

States included:

Multiple failed login attempts (BCPs)

One successful login (BPP)

Condition: ≥ 50 failed attempts within 24 hours

Interpretation: A complete login behavior fragment \rightarrow successful brute-force attempt.

Example 2: Login Behavior (BF) — Normal Login Behavior

States included:

A few failed login attempts (BCPs)

One successful login (BPP)

Condition: < 10 failures within 1 hour

Interpretation: A complete login behavior fragment \rightarrow normal login activity.

Behavior Fragment Level (BFL)

Behavior Fragment Level (BFL) is a semantic classification system that assigns each **Malicious Behavior Fragment (MBF)** to a distinct phase in the typical attack lifecycle.

Each level reflects a different stage of adversarial activity, semantic severity, and corresponding MITRE ATT&CK tactics.

It serves as a bridge between micro-level behavior detection and macro-level kill-chain reasoning, enabling:

1. Progressive threat scoring
2. AI reasoning sequence validation
3. Visualization of attack stage distribution

Level 1 (**Early Stage / Initial Access & Discovery**) Representative BF: **Login (Brute-Force)**

MITRE ATT&CK Mapping: Initial Access, Credential Access

Level 2 (**Mid Stage / Execution, Persistence, Lateral Movement**) Representative BF: **File Upload**

MITRE ATT&CK Mapping: Execution, Persistence, Lateral Movement

Level 3 (**Late Stage / Command & Control, Exfiltration**) Representative BF: **Command Execution**

MITRE ATT&CK Mapping: Command & Control, Exfiltration, Impact

Table2: Semantic Phase Mapping Between Behavior Fragment Levels (BFL) and MITRE ATT&CK Framework

| BFL Semantic Phase | Definition | Corresponding MITRE ATT&CK Tactic(s) |
|--------------------|---|---|
| Level 1 | Also known as Recon Phase. This level describes the process from external probing to successful access. For example, brute-force attempts followed by successful login. | <i>Initial Access, Credential Access, Discovery, Reconnaissance</i> |
| Level 2 | Represents the stage where the attacker leverages obtained access to execute tools, scripts, or malicious payloads within the environment. | <i>Execution, Privilege Escalation, Defense Evasion, Persistence (early)</i> |
| Level 3 | After tool execution, multiple paths may emerge including lateral movement, credential theft, communication with C2 servers, or data exfiltration. | <i>Lateral Movement, Collection, Command & Control, Exfiltration, Impact, Persistence (sustained)</i> |

The following three structural elements form the foundational prerequisites for enabling AI-driven knowledge base feedback:

1.A behavior chain must exist

Providing the structural backbone that organizes actions chronologically and semantically around a specific actor (user, IP, host, etc.).

2.Each behavior must have a semantic coordinate

Marking its position and meaning within the overall context (e.g., login attempt, file upload, lateral movement), enabling role-aware interpretation.

3.AI must identify pivot points

Detecting critical transitions that signify a shift in the attack phase, such as a successful login following multiple failures.

Only when these conditions are met can a human analyst validate the event severity based on the reconstructed path and perform accurate knowledge write-back. This is not just a technical process—it is the semantic essence of the system. These structural anchors are what allow the AI to truly “understand the attack story” and evolve from detection to reasoning.

(Note:Pivot Point Identification: From Binary Judgment to Weighted Confidence)

4.Pivot Strength

Pivot Strength refers to the semantic confidence score assigned to a specific behavior node when it is identified as a pivot point within an attack behavior chain. It reflects the logical support for judging whether the action represents a tactical phase transition—such as a shift from reconnaissance to execution—and serves as a key signal indicator in AI inference pathways.

In semantic investigation, whether a behavior constitutes a pivot point should not be treated as

a binary decision (yes/no), but rather as a probabilistic weight known as Pivot Strength.

Three Key Decay Factors Impact Pivot Strength:

1. **Time interval since the most recent failure** — the longer the interval, the greater the decay (no $\times 1$ multiplier applied).

2. **Number of previous failed actions (BCPs) compared with the predefined malicious behavior fragment threshold** — for example, the preset brute-force limit is 200 attempts, but the current attack contains 500 attempts, indicating a larger deviation.

3. **Average time interval between consecutive failures**, reflecting the attack's persistence and intensity.

B.Solution Framework

Table8: Five-Layer Foundation

| Level | Name | Function Description |
|-------|---|---|
| 1 | Field Normalization Layer | Standardizes heterogeneous log formats from multiple sources; constructs key fields (e.g., srcip, dstip, eventid, source_from, external_ip) |
| 2 | Micro-Scoring Layer | Applies multiple AI models to each log entry to generate threat score fields (model1 to model4) for downstream semantic computation |
| 3 | Behavior Chain Modeling Layer | Reconstructs behavior chains for the same entity based on temporal, semantic, and role consistency, forming inference material |
| 4 | AI Semantic Reasoning Layer | Performs path inference, attack chain recognition, and pivot point detection on behavior chains; supports natural language interface via LLM |
| 5 | Expert Feedback & Knowledge Loop | Integrates human feedback into a knowledge base with behavior chains, pivot points, labels, and attack names, enhancing future reasoning capabilities |

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IV. Use Cases

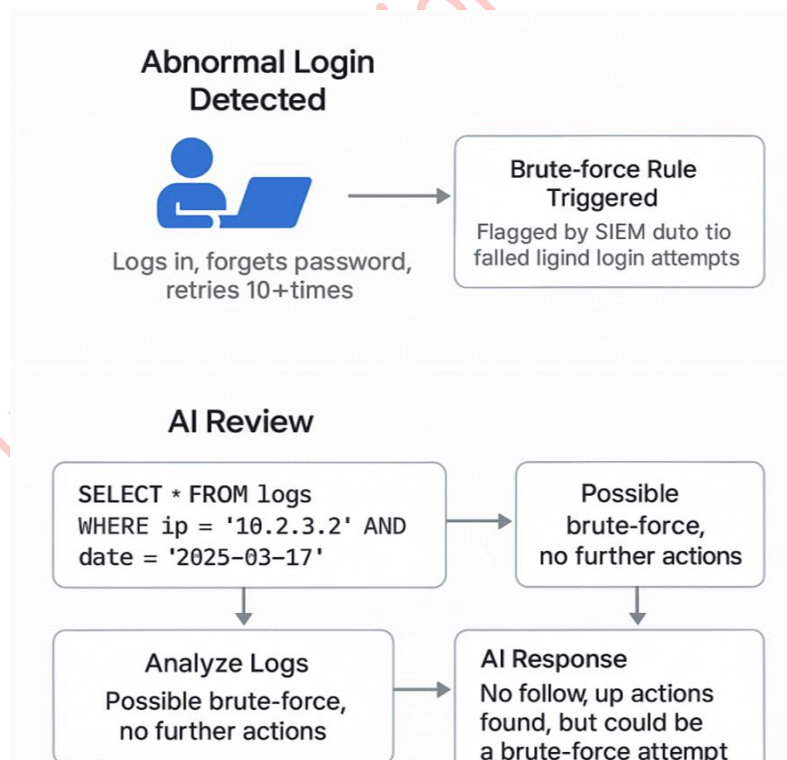
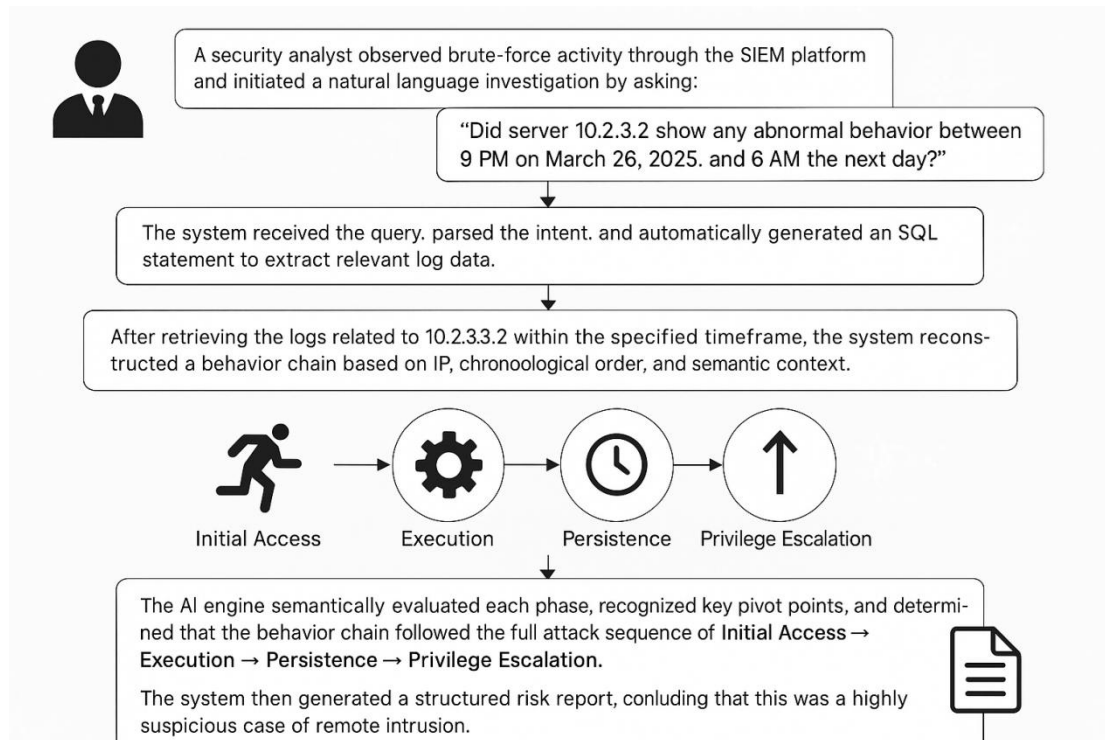


Figure8:AI-Driven Behavior Chain Analysis Flowchart(1.1)

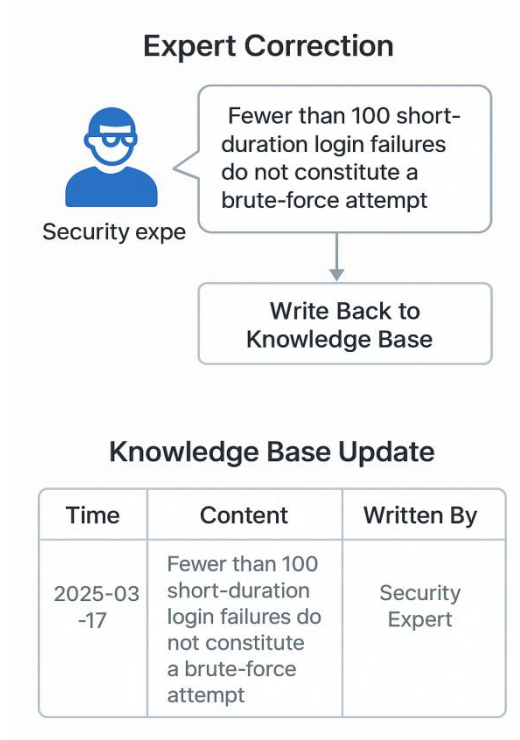


Figure9:Expert Feedback and Knowledge Base Update Diagram(1.2)

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VI. Industry Impact

A. Impact

To address the lack of structured investigative semantics between “alert” and “response” in current SOC/SIEM architectures, it is recommended that the “Investigative Layer” be formally introduced as a distinct architectural component in industry standards. It mainly includes:

1. Inputs: Sequences of behavioral logs and alert events
2. Outputs: Structured behavior chains, attack path graphs, and AI-generated reasoning suggestions
3. Core Functions: Behavior modeling, pivot point identification, and expert knowledge feedback integration
4. Necessity: Enables the response layer to execute precise, explainable, and automated decisions

B. Systemic Value

1. Language replaces experience: Transitions from manual, experience-driven investigation to AI-driven semantic reasoning and interpretation.

2. Structure captures expertise: Uses behavior chains to structurally store human expertise and AI feedback

3. AI understands behavior: Enables AI to understand not just “logs” but actual “behaviors”

4. Adaptive to attack-defense evolution: Supports continuous evolution, knowledge accumulation, and reasoning over behavioral variants

5. Transforms traditional SIEM investigation: Fully replaces the manual “log piecing” paradigm—AI investigates, humans validate

Table 10: Structured Solutions to Core Industry Challenges

| Industry Challenge | Root Cause | Structured Solution |
|-----------------------|--|---|
| 0day Attacks | 0days fundamentally bypass fixed detection rules; in traditional defenses, rule bypass = full defense failure | Structural-semantic language does not rely on static rules. It identifies anomalies through behavior chain paths and AI semantic reasoning, making it resilient to rule bypasses |
| Pivot Point Detection | Traditional systems misclassify post-exploit “normal” behaviors as benign, failing to detect lateral movement stages | Based on the ATT&CK tactical framework, any “normal behavior” following a labeled attack stage is flagged as a pivot point. The system includes decay mechanisms and path-based scoring |
| Industry | Root Cause | Structured Solution |

| Challenge | | |
|--------------|---|---|
| APT Attacks | Logs are fragmented and lack contextual continuity, making it impossible to reconstruct the full attack chain | Builds a Role-Based Behavior Database (EBD), storing logs by entity and sorting them chronologically. AI semantic analysis is applied to reconstruct the complete attack path |
| DDoS Attacks | Distributed, multi-source, concurrent access depletes resources. Traditional frequency-based thresholds are easily bypassed | Trains micro-models targeting specific DDoS features to detect patterns in path, frequency, and distribution. AI synthesizes these into behavior chains for organizational defense. Detection strategy includes: <ol style="list-style-type: none"> 1. Micro-models for fine-grained analysis, macro-logic for attack inference 2. Behavior chain knowledge base with expert feedback integration |

Table11:SCSM vs Traditional SOC: Structural Innovation Comparison

| Innovation Point | Traditional SOC | SCSM Advantages |
|-----------------------------|---|---|
| Log Structure | Non-standard fields, inconsistent formats | Field semantic normalization with triple-field mapping (e.g., external_ip) |
| Detection Method | Rule-based matching, signature-dependent | AI-based micro-scoring, multi-dimensional models, self-evolving detection |
| Behavior Understanding | Evaluates each log independently | Constructs semantic chains via temporal and role-based behavior linking |
| Attack Recognition | Relies on static attack rules | Dynamically identifies attack chains and pivot points; supports variant inference |
| Expert Knowledge Retention | High loss of individual insights | Writes back into structured knowledge chains, enhancing system memory |
| LLM Integration | Absent | Supports full flow: natural language → SQL/semantic reasoning → structured output |
| Structural Layer Completion | Detection → Alert → Response (Investigation is manual) | Completes the gap: Detection → Alert → Investigation → Response |

Table12:SCSM vs. Three Historical Protocol Standards

| Protocol | Initial Phase | Structural Release | Adopted by Standards Bodies | Global Ecosystem Expansion |
|-------------|----------------------------------|---|---|---|
| HTTP | Defined at CERN as URL structure | Adopted by W3C → Unified browser language | Referenced by all web protocols | Became the universal language of web communication |
| TLS | Netscape's proprietary SSL | Extended by IETF as TLS standard | Adopted by all secure browser platforms | Became the de facto encrypted communication standard for the Internet |
| MITRE ATT&C | MITRE research | Released as an open tactical | Referenced by NIST, widely adopted by | Became the factual standard for threat detection and |

| K | project | structure model | security companies | behavior modeling |
|------|---|---|--------------------|-------------------|
| SCSM | Independently invented + patent-defined | Released via industry white paper + five supporting documents | | |

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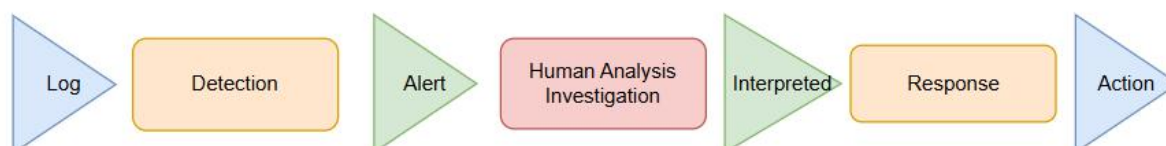
VII.Problem Solved

1.1 Security Operations Architecture Issue

Typical Process:

Log Collection → Detection (via Rules/Models) → (Human analysis) → Response

Core Problem:



The core task defined by the investigation layer is to Explain and reconstruct **"how events occur in the corporate environment"**. Without the investigation layer, we can never explain **"what is the basis for response"**. Investigation layer is neither a single event, nor an alert(log), nor a response(action), nor an intelligence technique(ioc,tactic). It output **Structured Evidence of investigative logic**. (Ref. Page 25. Miner Script Json)

1.2 Excessive False Positives in SOC's

Root Causes:

1. Rule-driven and event-centric: Each log entry is treated as an isolated event and analyzed statically.
2. No mechanism exists for identifying Behavior Pattern Points (BPP).

SCSM Approach:

SCSM reduces false positives not because it "uses AI" or "adds more rules," but because it **fundamentally changes the unit of meaning** through which the entire security system interprets alerts.

Semantics-driven and role-centric: Each log is interpreted **as an action** performed by a subject (Behavior Chain Point – BCP, Behavior Pattern Point – BPP), enabling dynamic contextual analysis.

BPP recognition enables semantic transition points — linking isolated events into meaningful behavioral chains for reasoning and investigation.

Example:

Traditional model: 500 failed login attempts → 500 alerts

SCSM model: 500 failed logins + 1 success → 1 incident, with Behavior Fragment Level (BFL) =

1& Attck T1110

In **SCSM**, investigation is **not rule-based** but **behavior-segment-based**. For example, the "login" behavior exists in the local behavior repository in both **normal** and **abnormal** states (e.g., brute-force attack, see Px). In the **role-behavior database**, these 500 failed logins and one successful login are

recognized as actions belonging to the same role.

When this entire **behavior segment** matches a **behavior fragment** in the local repository, it is classified as a single alert.

In summary:

Traditional model: 500 failed login attempts → 500 alerts

SCSM model: 500 failed logins + 1 success → **1 alert**, with **Behavior Fragment Level (BFL) = 1**

1.3 Advanced Persistent Threat (APT) attacks

Traditional SOC or SIEM systems can hardly *understand* long-chain APT attacks, because an APT is a **cross-stage, cross-device, and cross-time behavioral semantic chain**.

Each phase of an APT campaign appears in a traditional SIEM as a set of unrelated log entries.

The purpose of **SCSM** is to **reconstruct these fragmented logs into a coherent semantic behavior chain**.

Example:

Day 1: External IP performs port scanning

Day 5: Abnormal VPN login

Day 8: Internal host executes PowerShell

Day 10: Archive uploaded to an external address

A traditional SOC would treat these as **four isolated events**.

SCSM interprets them as **four behavioral segments of a single incident**, recognizing them as one **behavior chain**:

Reconnaissance (port-closed → BCP, port-open → BPP) →

Initial Access (login failed → BCP, login successful → BPP) →

Execution (PowerShell failed → BCP*(0-xx), PowerShell executed → BPP) →

Exfiltration (connection failed → BCP*(0-xx), connection successful → BPP)

The system links all actions of the **same role** (user / IP / host) according to **temporal** and **semantic continuity**, constructing a complete **Behavior Chain**.

SCSM is designed to solve this problem:

SCSM defines the **smallest logical unit of a security event** not as a *log entry*, but as a **Behavior Chain**.

1.4 0DAY attacks

Zero-Day Attacks and Unknown Vulnerabilities

Root-Cause Analysis:

Zero-day attacks exploit vulnerabilities that are not yet known to vendors or security communities.

Because there is no existing signature, IOC, or patch, traditional detection and response pipelines cannot recognize them in time.

Problem:

Rule-based systems rely on prior knowledge; when facing a 0-day exploit, they generate no alert or produce ambiguous anomalies.

SCSM Detection Philosophy: Semantic-Driven, Not Signature-Driven

SCSM detects threats **based on semantics rather than signatures**.

Its detection logic does not ask whether *a single event matches a vulnerability rule*; instead, it determines whether a **sequence of events forms a causal and self-contained behavior chain**.

Even when an attacker gains system access through a **zero-day exploit**, as long as subsequent actions occur, the **Entity Behavior Database (EBD)** will reveal a clear and traceable **attack path represented as a behavior chain**.

1.5 In Traditional SOC, BPP Identification Relies Entirely on Human Expertise

In conventional SOC operations, the identification of Behavior Pattern Points (BPPs) is not formally defined within the system. It exists only in the analyst's mind, relying on individual experience and intuition rather than a structured or machine-recognizable coordinate.