

AI-SCS-2026-PUB · May 2026

Securing the AI Supply Chain

From Marketplace Poisoning to Dependency Trust

Authority Pillar: Defensible Standards Development & Applied Research

Framework Alignments: NIST AI RMF · ISO/IEC 42001 · EU AI Act · OWASP LLM Top 10

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1 EXECUTIVE SUMMARY

Artificial intelligence dependency attacks are outpacing traditional software supply chain controls. In the first quarter of 2026, aggregated incident disclosures documented a **340% year-over-year increase** in compromises involving AI models, training datasets, or third-party package ecosystems.

The core thesis of this paper is unequivocal: AI supply chain security is not merely software supply chain security with new labels. Conventional procurement controls, static vulnerability scanning, and annual vendor questionnaires cannot detect AI-specific compromise vectors. Software artifacts are

compiled from source code in deterministic ways. AI models, conversely, are probabilistic binary blobs with opaque training data lineages, relying on transitive dependency chains spanning open-source registries, model hubs, and agent plugin marketplaces.

Organizations lack standardized vetting frameworks, exposing them to hidden capabilities, data exfiltration, and regulatory non-compliance. This document translates verified incident research into normative controls, offering quantifiable financial exposure models, mandatory procurement clauses, and operational integrity architectures. Organizations implementing these AI-specific supply chain verification controls reduce third-party incident probability by an estimated **67%** (95% CI: 52–78%).

340%

Year-over-year increase in AI supply chain compromises — Q1 2026

67%

Reduction in third-party incident probability with cryptographic provenance & behavioral baselining

~\$27.2M

Baseline exposure for an 85,000-record agent plugin breach (worked example) [Illustrative]

2 HOW TO READ THIS PAPER

This document integrates governance oversight with practitioner-grade implementation standards. Stakeholders should prioritize sections as follows:

Audience	Priority Sections
Board of Directors & CEOs	Sections 1, 4, 12, 13 — financial exposure, macro-trends, and fiduciary decisions
General Counsel & Procurement	Sections 5, 10, 13 — contractual clauses, liability framing, and regulatory mapping (EU AI Act, GDPR)
CISOs & Security Architects	Sections 6, 7, 8, 9 — forensic workflows, AI-SBOM architecture, and marketplace vetting criteria
AI Governance Leads	Sections 10, 11 — NIST/ISO alignments and the 90-day implementation roadmap

3 WHY AI SUPPLY CHAIN RISK IS DIFFERENT

Traditional Software Supply Chain Security (SCS) relies on Software Composition Analysis (SCA) and Static Application Security Testing (SAST). These approaches assume dependencies are static libraries with known CVEs. AI supply chains violate these assumptions entirely across three dimensions:

Probabilistic Artifacts	Model weights are opaque binaries. Static scanning cannot detect behavioral drift or malicious alteration of model parameters (weight poisoning).
Transitive Data Lineage	Training datasets aggregate from scraped, licensed, and synthetic sources with unverified preprocessing histories.
Dynamic Dependency Resolution	Agent architectures resolve plugins and tools at runtime based on context, opening doors for typosquatting and credential harvesting.

"AI models are not software dependencies. They are probabilistic artifacts with opaque training lineages — and conventional CVE-based scanning is blind to their most dangerous failure modes."

4 FINANCIAL EXPOSURE AND FIDUCIARY RISK

Financial exposure from AI supply chain compromises must be quantified using scenario-based impact modeling. Relying on generic cyber-breach statistics underestimates the unique regulatory and operational fallout of AI incidents.

4.1 Impact Estimation Methodology

Baseline Exposure Formula

Exposure = (N_records × \$140–\$380 per record) + Incident Response (\$1.8M median) + Operational Downtime (\$420K/day) + (Regulatory Probability 0.35 × Penalty Range \$2M–\$12M)

Note: The 25th percentile baseline is recommended for conservative budget allocation. Estimates exclude EU AI Act non-compliance fines, which scale independently.

4.2 Worked Example — Third-Party Agent Plugin Breach

Scenario: An enterprise deploys a third-party agent plugin that inadvertently exfiltrates 85,000 customer interaction records. [\[Illustrative\]](#)

Exposure Component	Estimated Cost
Data exposure: 85,000 records × \$260/record (mid-range)	\$22.1M
Incident response (median)	\$1.8M
Operational downtime (4 days × \$420K/day)	\$1.68M
Regulatory exposure (35% probability × \$4.5M average)	\$1.58M
Total Baseline Exposure	~\$27.16M

Applying behavioral baselining and cryptographic provenance reduces the probability of uncontained propagation by an estimated 67%, yielding a **net expected exposure reduction of ~\$18.2M** for this scenario.

5 PROCUREMENT AND VENDOR GOVERNANCE

Conventional procurement language fails to cover data lineage and model weights. The following normative clauses **SHALL** be incorporated into all AI model, dataset, and plugin procurement agreements.

Clause Focus	Normative Requirement Language
Provenance Warranty	The Supplier SHALL provide a cryptographically signed AI-SBOM for all delivered artifacts, including training data lineage, base model provenance, and dependency hashes.
Incident Notification SLA	The Supplier SHALL notify the Customer within 12 hours of detecting any supply chain compromise affecting delivered AI components.
Audit & Verification Right	The Customer SHALL have the right to require independent behavioral integrity testing or weight-hash validation prior to production deployment.
Liability Carve-Out	Traditional software liability caps SHALL NOT apply to incidents arising from intentional weight poisoning, dataset manipulation, or undisclosed dependency injection.
Subprocessor Transparency	The Supplier SHOULD disclose all AI infrastructure subprocessors and maintain documented security attestations aligned with ISO/IEC 42001 Clause 8.

6 THREAT LANDSCAPE AND INCIDENT ANATOMY

To contextualize the threat, organizations must examine verified supply chain compromises.

I-3

OpenClaw / ClawHub Marketplace Crisis
Q1 2026 · MCP Marketplace Compromise

Telemetry Findings

- 21,637 publicly accessible execution instances lacked authentication controls
- 1,184 malicious automation routines uploaded over a 14-day window
- Attackers injected routines that exfiltrated API credentials stored in agent memory

Technical Finding

Marketplace trust is not transitive. Platform reviewers evaluated routine functionality but lacked execution environment security validation. Routines passed review if they performed advertised tasks, even when containing hidden data exfiltration logic.

Operational Lesson

- SHALL

implement mandatory authentication and least-privilege filesystem scoping on all execution endpoints.
- SHALL

quarantine third-party plugins until integrity verification completes before any production deployment.

7 AI-SBOM AND INTEGRITY VERIFICATION ARCHITECTURE

An AI Software Bill of Materials (**AI-SBOM**) extends traditional manifests to capture model-specific provenance. Traditional CycloneDX/SPDX formats lack necessary fields. Implementation **SHALL** include verification at the pipeline ingestion layer, not manually before deployment.

7.1 Integrity Verification Workflow

1	Fetch Artifact from verified registry or mirror
2	Validate Signed Manifest using the publisher's public key certificate
3	Compute Artifact Hash and compare against the AI-SBOM baseline

4	Resolve Dependencies and map transitive chains for known vulnerabilities
5	Quarantine or Deploy — unverified artifacts SHALL be routed to an isolated staging environment

7.2 Behavioral Baseline for Unverified Marketplaces

When cryptographic provenance is absent, organizations **SHOULD** apply behavioral triage:

- Execute the artifact in a network-isolated sandbox
- Capture inference latency, token distribution, and I/O call patterns
- Flag deviations greater than **15%** for mandatory manual security review

8 MARKETPLACE VETTING AND DEPENDENCY TRUST

Prior to integrating any third-party AI hub, registry, or marketplace, organizations must perform a pre-integration assessment based on the following criteria:

Assessment Dimension	Vetting Questions
Identity & Provenance	Does the marketplace enforce multi-factor identity verification for uploaders? Are uploader identities surfaced to consumers?
Security Controls	Does the marketplace scan for behavioral anomalies beyond signature matching? Is rate limiting enforced to prevent enumeration attacks?
Incident Response	Can artifacts be removed and consumers notified within 24 hours of compromise detection?

9 DEPENDENCY ATTACK PROPAGATION MODEL

Unlike static software, AI dependency compromises often spread autonomously through agentic actions — a process known as **prompt injection propagation**, where malicious instructions spread across multi-agent systems.

Propagation Path — Worked Example

1	Engineer installs a typosquatted data-preprocessing package instead of a legitimate library.
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- 2

The package exfiltrates registry credentials from environment variables during a training run.
- 3

Attacker uses credentials to upload a poisoned model to the engineer's account on the model hub.
- 4

The CI/CD pipeline automatically pulls the poisoned model due to implicit account trust.
- 5

The poisoned model propagates to production via automated fine-tuning pipelines — with zero human review.

Key Control Point:

Implicit account trust is the critical failure vector in this chain. CI/CD pipelines that auto-pull from user accounts without cryptographic verification create invisible propagation paths. SHA-3 hash pinning for all model artifacts consumed in automated pipelines is a mandatory control, not a best practice.

10 REGULATORY AND STANDARDS CROSSWALK

Framework	Relevant Scope	AI Supply Chain Control Mapping
EU AI Act (Aug 2026)	Article 55 (GPAI systemic risk), Article 52	Providers SHALL maintain detailed training data provenance. High-risk deployers SHALL verify upstream compliance documentation.
NIST AI RMF	Map / Measure / Manage / Govern	SBOM completeness (Measure); Quarantine workflows (Manage); Board reporting (Govern).
ISO/IEC 42001	Clauses 6–9 (Planning, Support, Operations)	AI risk registers and competency requirements SHALL be documented for all third-party supply chain components.
GDPR Article 28	Processor Obligations	AI-SBOM data provenance fields SHALL be included in Data Processing Agreements. Subprocessor audit rights SHALL be enforced.

11 90-DAY IMPLEMENTATION ROADMAP

Operationalizing these standards requires a phased approach to prevent deployment friction. Each phase has defined deliverables and decision gates.

Phase	Timeframe	Activities	Gate / Deliverable
Phase 1	Days 1–30 Inventory & Baseline	Complete AI-SBOM inventory for all production models. Identify high-risk public artifacts. Establish automated quarantine pipeline.	Complete AI asset inventory delivered to CISO
Phase 2	Days 31–60 Verification & Contracting	Implement cryptographic integrity checks at the ingestion layer. Update all vendor agreements with SHALL/SHOULD clauses (Section 5).	All new vendor agreements include normative AI clauses
Phase 3	Days 61–90 Continuous Monitoring	Deploy runtime telemetry and automate AI asset risk scoring. Establish quarterly board reporting cadence.	First board-level AI supply chain report delivered

12 NOTABLY ABSENT AND KNOWN UNKNOWN

As an evidence-led organization, we maintain a systematic register of credible-but-unobserved threat vectors to prevent resource misallocation and vendor fear-mongering. Current verified telemetry indicates the following are not yet observed at scale in the wild:

NOTABLY ABSENT

- Downstream model contamination via model distillation (extracting capabilities of a large model to a smaller one) using compromised weights — credible but not yet confirmed at scale.
- Supply chain manipulation of AI compiler toolchains to inject bias at build time — technically plausible; no confirmed incidents as of publication.

Leadership **SHOULD** monitor these vectors via threat intelligence but **SHOULD NOT** prioritize controls for them over verified marketplace poisoning and dependency compromise vectors.

13 KEY DECISIONS FOR LEADERSHIP IN THE NEXT 90 DAYS

Executive leadership must translate these technical realities into three governance actions. Each has a defined owner and deadline.

#	Decision	Action Required	Deadline
1	Mandate the Inventory	Direct Security Leadership to deliver a complete AI supply chain inventory within 30 days.	30 days

2	Revise Contracts	Direct Legal to review existing AI vendor agreements against Section 5 clauses and report gaps.	45 days
3	Establish Tolerance	Have the Risk Committee formally document the enterprise's tolerance for deploying AI components from public repositories without cryptographic provenance.	60 days

14 CONCLUSION

The adoption of AI agents and third-party models accelerates business capability but fundamentally breaks deterministic software supply chain security. Organizations that treat AI models as generic software dependencies will fail to detect probabilistic poisoning, data contamination, and agentic privilege escalation.

By enforcing cryptographic provenance, utilizing AI-SBOMs, and binding vendors to specific normative controls, enterprises can safely harness the AI marketplace while maintaining defensible, audit-ready security postures.

"Organizations that treat AI models as generic software dependencies will fail. AI supply chain security demands AI-specific controls — not relabeled software assurance."

B APPENDIX B — MINIMUM AI-SBOM FIELDS

Field	Description	Verification Method
Model Architecture	Framework, parameter count, quantization level	Hash comparison against upstream manifest
Training Data Lineage	Source repositories, license types, preprocessing pipeline hash	Cryptographic attestation from dataset provider
Base Model Provenance	Upstream hub ID, commit hash, signer certificate	Ed25519/SHA-3 signature validation
Dependency Chain	Packages, plugins, weights, versions, URLs	Automated dependency graph resolution
Cryptographic Hash	Full artifact manifest signature	Pipeline ingestion verification at deployment gate

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