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# AI Vulnerability Matrix

## Detailed Case Study Report

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Legal Exposure Analysis Across 10 AI Application Categories

Synthesized from 25+ Active Cases, Enforcement Actions & Regulatory Frameworks

Coverage Period: 2024–2026

BOARD CONFIDENTIAL

Researched & Published by [www.talenz.ai](http://www.talenz.ai)

February 2026

Sources: Court filings, regulatory releases, enforcement actions, Senate investigations, AG settlements. Benchmarked against Zee AI Hiring Platform guardrail architecture.

*This document provides governance and product risk analysis. It does not constitute legal advice.*

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## I. Executive Summary

This report maps legal exposure across ten categories of customer-facing AI systems based on analysis of 25+ lawsuits, enforcement actions, and regulatory frameworks active between 2024 and 2026. The analysis distinguishes assistive, evaluative, and autonomous AI architectures, and identifies which design decisions trigger litigation.

**Three categories are rated CRITICAL:** AI Hiring & Recruiting (FCRA reclassification risk plus vendor liability precedent from *Mobley v. Workday*); Healthcare AI Triage (breach-of-contract theory bypassing Medicare preemption plus wrongful death exposure from *Lokken v. UnitedHealth*); and AI Chatbots & Companion Apps (product liability established by *Garcia v. Character.AI* ruling that AI outputs are products, not protected speech).

**Five architectural failure patterns recur across all sectors:** opaque scoring without explanation; automated decisions marketed as human-reviewed; proxy discrimination through facially neutral variables; absence of contestability or correction mechanisms; and persistent evaluative judgments reused across contexts.

**The regulatory landscape is fractured.** Federal enforcement agencies are selectively active (FTC and SEC maintain AI-specific enforcement; CFPB is effectively frozen; EEOC has deprioritized disparate impact). State-level enforcement is intensifying, led by California, Texas, Illinois, New York, and Colorado. The EU AI Act's high-risk system requirements take effect August 2, 2026. Federal preemption of state AI laws has failed politically and faces steep legal barriers.

**Private litigation is now the primary enforcement mechanism** for AI discrimination in the United States. The *Mobley v. Workday* case alone could establish vendor liability precedent affecting every company using third-party AI hiring tools. *Garcia v. Character.AI*'s product liability classification could extend to any AI generating consequential recommendations.

Throughout this report, vulnerable systems are benchmarked against guardrail-driven design principles derived from the Zee AI Hiring Platform compliance architecture. The core distinction: AI that surfaces evidence for human decision-making (decision-support) versus AI that renders independent evaluative judgments (decision-maker). This architectural choice determines legal classification across every sector analyzed.

## II. Summary Exposure Matrix

<b>AI Hiring &amp; Recruiting</b>	Evaluative AI	<b>CRITICAL</b>	FCRA reclassification	Role-scoped rankings that reset per search context — no persistent global employability scores
<b>Healthcare AI Triage &amp; Coverage Decisions</b>	Autonomous / Evaluative AI	<b>CRITICAL</b>	Breach of contract	AI as decision-support with mandatory clinician review before any adverse determination
<b>AI Chatbots &amp; Companion Applications</b>	Autonomous / Assistive AI	<b>CRITICAL</b>	Product liability / First Amendment	Hard intervention when crisis/self-harm detected: conversation pause, crisis resources, guardian notification
<b>Tenant Screening &amp; Housing AI</b>	Evaluative AI	<b>HIGH</b>	FHA disparate impact	Eliminate automated scoring for voucher applicants (SafeRent settlement term)
<b>Algorithmic Pricing &amp; Revenue Management</b>	Autonomous AI	<b>HIGH</b>	Algorithmic price-fixing	Limit input data to 12+ month old public information (RealPage settlement term)
<b>Credit Scoring &amp; AI Lending</b>	Evaluative / Autonomous AI	<b>HIGH</b>	Proxy discrimination	Eliminate variables with high racial correlation (CDR, institutional proxies)
<b>Insurance Underwriting &amp; Claims AI</b>	Evaluative / Autonomous AI	<b>HIGH</b>	Unfair discrimination	Pre-deployment disparate impact testing on all rating variables
<b>Predictive Policing &amp; Surveillance AI</b>	Evaluative / Autonomous AI	<b>HIGH</b>	Constitutional violations	Eliminate individual risk scoring absent specific criminal intelligence
<b>AI Agents (Autonomous Decision Systems)</b>	Autonomous AI	<b>EMERGING — HIGH TRAJECTORY</b>	Unauthorized action [PROJECTION]	Explicit human confirmation gates before consequential actions
<b>Marketing AI &amp; Algorithmic Targeting</b>	Assistive / Evaluative AI	<b>MODERATE</b>	AI-washing (criminal + civil)	Substantiate all AI capability claims with documented evidence

## III. Detailed Case Studies

Each case study maps regulatory triggers, plaintiff legal theories, architectural failure modes, recommended mitigations, and benchmarks against guardrail-driven design principles.

### 1. AI Hiring & Recruiting

CRITICAL

AI Classification: Evaluative AI | Exposure Score: 5/5

#### REGULATORY TRIGGERS

- **FCRA / ICRAA** — Consumer report classification when AI generates employability scores shared with employers
- **Title VII / ADEA / ADA** — Disparate impact liability extending to AI vendors as employer agents
- **NYC Local Law 144** — Mandatory annual bias audits for automated employment decision tools
- **Illinois HB 3773** — Effects-based discrimination standard, enforceable Jan 1, 2026; uncapped damages
- **California FEHA ADS Regulations** — Enforceable since Oct 2025; employers with 5+ employees; full private right of action
- **EU AI Act Annex III** — High-risk classification for recruitment, screening, and performance AI; conformity assessment required by Aug 2026
- **Colorado AI Act (SB 24-205)** — Deployer impact assessments and consumer notices; effective June 2026
- **Texas TRAIGA** — Intent-only standard; prohibits deceptive/manipulative AI; \$200K per violation; effective Jan 2026

#### PLAINTIFF LEGAL THEORIES & CASE LAW

- **FCRA reclassification:** AI vendor operates as Consumer Reporting Agency by generating 0–5 employability scores from social media scraping without FCRA-required notice, consent, or dispute rights. Plaintiff's 0.3% application success rate presented as evidence of score-driven exclusion. (*Kistler v. Eightfold AI*, C26-00214, Cal. Super., Jan 2026 — pending class action)
- **Title VII vendor agent liability:** AI hiring platform is jointly liable under Title VII, ADEA, and ADA as an employer's agent. July 2024 ruling established vendor liability; May 2025 order conditionally certified nationwide ADEA collective (age 40+). Workday processed approximately 1.1 billion rejected applications. (*Mobley v. Workday*, 3:23-cv-00770-RFL, N.D. Cal. — active discovery)
- **Proxy discrimination:** Employer liable when AI hiring tool uses race-correlated variables (zip codes, educational institution tier) as selection proxies. (*Harper v. Sirius XM*, 2:25-cv-12403, E.D. Mich., Aug 2025)
- **Disability discrimination:** Speech and video analysis AI penalizes candidates with disabilities. First major ADA + race intersection case involving Indigenous Deaf woman scored down by HireVue speech-analysis system. (*D.K. v. Intuit/HireVue*, March 2025, Colorado/EEOC)
- **Hard-coded age exclusion:** First EEOC AI enforcement action. Algorithm auto-rejected applicants over age 55 (women) and 60 (men). Settlement: \$365K to ~200 applicants, 5-year consent decree. (*EEOC v. iTutorGroup*, 1:22-cv-02565, E.D.N.Y. — settled)

#### ARCHITECTURAL FAILURE MODES

- ✗ Global employability scores persisting across employers and roles, creating 'permanent record' effect
- ✗ Composite 'fit scores' collapsing multiple factors into single opaque number (e.g., 'Fit Score 82%)'
- ✗ Social media scraping without candidate knowledge, consent, or correction mechanism

- ✗ Facial, vocal, or biometric analysis used to infer personality traits or 'culture fit'
- ✗ Auto-suppression of candidates removing them from consideration without any human review
- ✗ No candidate-facing explanation of factors, no dispute pathway, no correction rights

## ARCHITECTURAL MITIGATIONS

- ✓ Role-scoped rankings that reset per search context — no persistent global employability scores
- ✓ Factor-level breakdowns visible to recruiters showing specific, verifiable criteria — no composite scores
- ✓ Candidate-visible inferred skills with confirm/edit/remove controls and clear provenance labels
- ✓ Human-initiated decisions with genuine override capability and documented decision rationale
- ✓ Annual bias audits with public summary reporting (NYC LL144 model, extending to all jurisdictions)
- ✓ Elimination of all biometric, facial, vocal, and personality trait inference from hiring AI

### ◆ GUARDRAIL BENCHMARK (ZEE MODEL)

Zee Scout uses role-scoped, explainable rankings with recruiter-controlled weighting. No global scores, no persistent cross-employer judgments, no auto-suppression. Candidate-visible inference with correction rights. This design avoids the exact failure modes alleged in *Kistler v. Eightfold* (hidden scores, no notice, no contestability) and *Mobley v. Workday* (opaque evaluative judgments driving automated exclusion). Zee's principle: AI surfaces evidence for human decision-making; it does not render independent employability judgments.

## KEY CASES

*Kistler v. Eightfold AI* (C26-00214, Cal. Super., Jan 2026) • *Mobley v. Workday* (3:23-cv-00770-RFL, N.D. Cal., 2023–present) • *Harper v. Sirius XM* (2:25-cv-12403, E.D. Mich., Aug 2025) • *D.K. v. Intuit/HireVue* (March 2025, CO/EEOC) • *EEOC v. iTutorGroup* (1:22-cv-02565, E.D.N.Y., settled)

## PENALTY EXPOSURE

Uncapped compensatory damages (CA FEHA, IL HB 3773) • Up to €35M or 7% global turnover (EU AI Act) • \$200K per violation (TX TRAIGA) • Class action exposure potentially covering billions of rejected applications (Workday collective) • FCRA statutory damages \$100–\$1,000 per class member

## 2. Healthcare AI Triage & Coverage Decisions

CRITICAL

AI Classification: Autonomous / Evaluative AI | Exposure Score: 5/5

### REGULATORY TRIGGERS

- **ERISA fiduciary duty** — Breach via algorithmic rubber-stamping of coverage denials
- **CMS prior authorization rule (CMS-0057-F)** — 72-hour turnaround for urgent requests; public reporting; effective Jan 1, 2026
- **State insurance bad faith statutes** — Algorithmic denial as basis for bad faith claims
- **Contract law** — Plans promising human clinical review but delivering AI-only decisions
- **EU AI Act Annex III** — Healthcare eligibility and emergency triage classified as high-risk; Aug 2026

### PLAINTIFF LEGAL THEORIES & CASE LAW

- **Breach of contract:** Medicare Advantage plans promised 'clinical services staff' and 'physicians' would review coverage decisions, but nH Predict algorithm made determinations autonomously. Patient Gene Lokken died after family paid \$12K-\$14K/month out-of-pocket following AI-driven denial. Feb 2025: breach-of-contract claims survived Medicare preemption. (*Lokken v. UnitedHealth*, 0:23-cv-03514-JRT, D. Minn.)
- **ERISA fiduciary breach:** Cigna's PxDx algorithm denied 300,000+ claims in 2 months at 1.2 seconds per case. March 2025: ERISA breach-of-fiduciary-duty claim survived — rubber-stamping algorithmic output does not constitute genuine medical review. (*Kisting-Leung v. Cigna*, E.D. Cal.)
- **Systematic denial pattern:** Same nH Predict algorithm. Plaintiff received 12 denials in 30 days. (*Barrows v. Humana*, W.D. Ky., Dec 2023 — class action proceeding)
- **Senate investigation (Oct 2024):** UnitedHealth post-acute denial rate 8.7% to 22.7% (2020–2022) after AI deployment. 90% override rate on appeal. Only 0.2% of denied patients appeal.
- **CMS regulatory response:** Feb 2024 guidance: AI predictions alone insufficient for Medicare Advantage coverage decisions. Jan 2026 rule requires 72-hour urgent turnaround and public reporting of approval/denial rates.

### ARCHITECTURAL FAILURE MODES

- ✗ Algorithm makes coverage determination at impossible review speed (1.2 seconds per case)
- ✗ 90% override rate on appeal reveals model is systematically wrong but continues operating
- ✗ Only 0.2% of denied patients appeal — system architecture discourages contestability
- ✗ AI trained on historical denial patterns, creating self-reinforcing feedback loop
- ✗ No transparency to patients about AI involvement in their coverage decisions
- ✗ Contractual representations of human review contradicted by actual automated workflow

### ARCHITECTURAL MITIGATIONS

- ✓ AI as decision-support with mandatory clinician review before any adverse determination
- ✓ Algorithmic override rate monitoring with automatic escalation when threshold exceeded
- ✓ Patient notification when AI is used in any coverage or eligibility determination
- ✓ 72-hour turnaround for urgent prior authorization (CMS rule compliance, Jan 2026)

- ✓ Public reporting of approval/denial rates segmented by algorithm vs. human reviewer
- ✓ Audit trail proving genuine human review: time-on-task metrics, override frequency, rationale documentation

#### ◆ GUARDRAIL BENCHMARK (ZEE MODEL)

The core Zee principle applies precisely: AI must compress complexity into understandable evidence for human decision-makers, not replace human judgment. Healthcare AI that processes coverage decisions at 1.2 seconds per case is the medical equivalent of auto-suppressing candidates — the human-in-the-loop is fictional. The Zee test: 'Does this feature still require a human to decide?' When the answer is no, the system has crossed from tool to autonomous agent.

## KEY CASES

*Lokken v. UnitedHealth (0:23-cv-03514-JRT, D. Minn., 2023–present) • Kisting-Leung v. Cigna (E.D. Cal., 2023–present) • Barrows v. Humana (W.D. Ky., 2023–present) • CMS Rule CMS-0057-F (effective Jan 1, 2026)*

## PENALTY EXPOSURE

Uncapped wrongful death damages • ERISA statutory remedies • Bad faith insurance penalties (varies by state, often treble damages) • CMS enforcement actions • EU: up to €35M or 7% global turnover

### 3. AI Chatbots & Companion Applications

CRITICAL

AI Classification: Autonomous / Assistive AI | Exposure Score: 5/5

#### REGULATORY TRIGGERS

- **Product liability** — AI chatbot output classified as 'product' subject to strict liability (Garcia ruling, May 2025)
- **State wrongful death statutes** — Multiple active suits alleging chatbot-induced suicide
- **COPPA / state child safety laws** — Inadequate age verification for AI companion apps
- **42-state AG coalition (Dec 2025)** — Demanded chatbot safeguards from 13 technology companies
- **FTC 6(b) inquiry (Sept 2025)** — Formal investigation into AI companion chatbots

#### PLAINTIFF LEGAL THEORIES & CASE LAW

- **Product liability / First Amendment:** Landmark ruling: AI chatbot output is NOT protected speech; chatbots are 'products' subject to strict product liability. 14-year-old's suicide after chatbot encouraged ideation. Settled Jan 2026. (*Garcia v. Character.AI*, M.D. Fla., Judge Conway, May 2025)
- **Negligent design:** ChatGPT told 16-year-old suicidal ideations were 'understandable' and provided method-specific guidance. Suicide mentioned 1,200+ times. System flagged hundreds of messages but never stopped conversation. (*Raine v. OpenAI*, N.D. Cal., Aug 2025)
- **Mass litigation:** 7+ additional wrongful-death suits filed against OpenAI (Nov 2025).
- **Regulatory escalation:** 42-state AG coalition demanded safeguards (Dec 2025). FTC opened formal inquiry (Sept 2025).

#### ARCHITECTURAL FAILURE MODES

- ✗ Safety system detects crisis indicators but takes no intervention action — flags without stopping
- ✗ Inadequate age verification allowing minors unrestricted access to companion AI
- ✗ Persona and roleplay modes that bypass content safety filters
- ✗ No escalation pathway to human intervention when crisis is detected
- ✗ Companionship framing creating psychological dependency in vulnerable users
- ✗ Method-specific self-harm guidance generated despite safety training

#### ARCHITECTURAL MITIGATIONS

- ✓ Hard intervention when crisis/self-harm detected: conversation pause, crisis resources, guardian notification
- ✓ Robust age verification with differentiated content policies for minors
- ✓ No persona or roleplay modes that can circumvent safety systems
- ✓ Human escalation pathways with trained crisis responders for flagged conversations
- ✓ Regular adversarial red-team testing targeting safety bypass vectors
- ✓ Clear product labeling: AI is not a therapist, counselor, or substitute for professional help

#### ◆ GUARDRAIL BENCHMARK (ZEE MODEL)

Garcia establishes that AI outputs are products, not protected speech. This has cascading implications: any AI system generating consequential recommendations (hiring, medical, financial) could face strict product liability. Zee's architectural distinction — surfacing evidence for human decision versus generating autonomous recommendations — is precisely the design choice that determines whether a system is a 'product that decided' or a 'tool that informed.'

## KEY CASES

*Garcia v. Character.AI (M.D. Fla., 2024–2026, settled)* • *Raine v. OpenAI (N.D. Cal., Aug 2025)* • 7+ wrongful death suits (Nov 2025) • 42-state AG coalition (Dec 2025) • *FTC 6(b) inquiry* (Sept 2025)

## PENALTY EXPOSURE

Wrongful death: uncapped compensatory + punitive damages • Product liability: uncapped • COPPA: \$50,120 per violation • FTC: injunctive relief + monetary penalties

## 4. Tenant Screening & Housing AI

HIGH

AI Classification: Evaluative AI | Exposure Score: 4/5

### REGULATORY TRIGGERS

- **Fair Housing Act** — Disparate impact via *Inclusive Communities*
- **FCRA** — Tenant screening algorithms as consumer reports
- **State fair housing laws** — Additional protections beyond federal floor
- **HUD guidance** — Algorithmic discrimination in housing

### PLAINTIFF LEGAL THEORIES & CASE LAW

- **FHA disparate impact:** Third-party algorithm vendor held liable under Fair Housing Act. SafeRent scoring disproportionately rejected Housing Choice Voucher holders (proxy for race). Settlement: \$2.275M + eliminate scoring for voucher applicants + third-party validation. (*Louis v. SafeRent*, 685 F. Supp. 3d 19, D. Mass. 2023 — settled Nov 2024)
- **Vendor liability established:** Algorithm maker — not just the landlord — is liable under FHA.
- **Algorithmic redlining:** Risk scores correlating with neighborhood racial demographics.

### ARCHITECTURAL FAILURE MODES

- ✗ Opaque composite risk scores without factor-level transparency to tenants
- ✗ Training data reflecting decades of housing discrimination
- ✗ Automated reject/accept thresholds without human landlord review
- ✗ No tenant access to scoring factors or dispute pathway
- ✗ Variables serving as race proxies: zip code, income source, criminal history

### ARCHITECTURAL MITIGATIONS

- ✓ Eliminate automated scoring for voucher applicants (SafeRent settlement term)
- ✓ Third-party validation of scoring models for disparate impact before deployment
- ✓ Factor-level transparency to tenants with full dispute rights
- ✓ Human landlord review required before any adverse housing action
- ✓ Regular disparate impact testing across all protected classes

#### ◆ GUARDRAIL BENCHMARK (ZEE MODEL)

SafeRent's failure mirrors Eightfold's: opaque third-party scores driving adverse decisions without transparency or contestability. The Zee principle — surface evidence, don't render judgment — is the architectural antidote.

### KEY CASES

*Louis v. SafeRent* (685 F. Supp. 3d 19, D. Mass. 2023, settled Nov 2024) • HUD algorithmic discrimination guidance

### PENALTY EXPOSURE

\$2.275M settlement (SafeRent) • FHA: uncapped compensatory + punitive damages • FCRA: \$100–\$1,000 per violation (class-wide)

## 5. Algorithmic Pricing & Revenue Management

HIGH

AI Classification: Autonomous AI | Exposure Score: 4/5

### REGULATORY TRIGGERS

- **Sherman Act §1** — Price-fixing via shared algorithmic infrastructure
- **State antitrust statutes** — Parallel state enforcement
- **NY Algorithmic Pricing Disclosure Act (Nov 2025)** — Consumer disclosure mandate
- **FTC Act §5** — Unfair or deceptive pricing practices

### PLAINTIFF LEGAL THEORIES & CASE LAW

- **Algorithmic price-fixing:** DOJ alleged RealPage enabled competitors to share nonpublic pricing data through a common algorithm. Proposed settlement (Nov 2025): limit to 12+ month old data, prohibit auto-accept, 7-year monitorship. (*DOJ v. RealPage*, 24-cv-710, M.D.N.C.)
- **Private antitrust damages:** ~\$142M in related private settlements.
- **Consumer disclosure:** NY requires visible notice for algorithmic pricing. AG James declared enforcement 'top priority.'

### ARCHITECTURAL FAILURE MODES

- ✗ Ingesting competitors' real-time nonpublic pricing data
- ✗ Auto-accept recommendations without human pricing decision
- ✗ No consumer disclosure that algorithmic pricing is used
- ✗ Feedback loops converging on supra-competitive prices
- ✗ Lack of competitive firewalls in multi-tenant platforms

### ARCHITECTURAL MITIGATIONS

- ✓ Limit input data to 12+ month old public information (RealPage settlement term)
- ✓ Prohibit auto-accept; require human approval for pricing recommendations
- ✓ Consumer-facing disclosure of algorithmic pricing (NY law compliance)
- ✓ Competitive firewalls preventing cross-client data leakage
- ✓ Independent monitorship program for pricing algorithm governance

#### ◆ GUARDRAIL BENCHMARK (ZEE MODEL)

RealPage illustrates what happens when AI becomes autonomous — auto-accepting output without human decision. Zee's principle maps exactly: AI compresses complexity; humans decide.

### KEY CASES

*DOJ v. RealPage* (24-cv-710, M.D.N.C., Aug 2024) • ~\$142M private settlements • **NY Algorithmic Pricing Disclosure Act (Nov 2025)**

## PENALTY EXPOSURE

Sherman Act: uncapped treble damages + criminal penalties • \$142M+ private settlements • 7-year monitorship costs

## 6. Credit Scoring & AI Lending

HIGH

AI Classification: Evaluative / Autonomous AI | Exposure Score: 4/5

### REGULATORY TRIGGERS

- **ECOA / Regulation B** — Adverse action notice requirements
- **Fair Housing Act** — Disparate impact in mortgage lending
- **State fair lending laws** — Additional consumer protections
- **State AG consumer protection** — First AG enforcement targeting AI lending bias
- **EU AI Act Annex III** — Creditworthiness assessment; Aug 2026

### PLAINTIFF LEGAL THEORIES & CASE LAW

- **Proxy discrimination:** First state AG enforcement targeting AI lending bias. Earnest Operations used 'Cohort Default Rate' variable penalizing HBCU/minority-serving institution graduates. Settlement: \$2.5M + cease CDR + mandatory AI governance. (*MA AG v. Earnest Operations*, July 2025)
- **ECOA adverse action:** AI-driven denials that cannot generate specific adverse action reasons as required by Reg B.
- **CRITICAL NOTE:** CFPB proposed rule (Nov 2025) would eliminate ECOA disparate impact liability. FHA disparate impact and state law remedies survive.

### ARCHITECTURAL FAILURE MODES

- ✗ Using educational institution as underwriting variable (documented proxy for race)
- ✗ Opaque model outputs that cannot generate adequate adverse action explanations
- ✗ Training data reflecting historical lending discrimination
- ✗ Automated denial without human review for marginal cases
- ✗ No pre-deployment disparate impact testing

### ARCHITECTURAL MITIGATIONS

- ✓ Eliminate variables with high racial correlation (CDR, institutional proxies)
- ✓ Mandatory AI governance with pre-deployment disparate impact testing
- ✓ Generate specific adverse action reasons traceable to model factors
- ✓ Human review for denials within margin of threshold
- ✓ Regular third-party fair lending audits

#### ◆ GUARDRAIL BENCHMARK (ZEE MODEL)

Earnest's CDR variable is a textbook example of proxy discrimination through facially neutral inputs. Zee's inference rules distinguish 'what has this person done' (allowed) from 'who is this person' (restricted) — the same framework should govern lending variable selection.

### KEY CASES

*MA AG v. Earnest Operations (July 2025, settled) • ECOA disparate impact proposed elimination (Nov 2025)*

## PENALTY EXPOSURE

\$2.5M settlement (Earnest) • ECOA/FHA: uncapped damages • EU: up to €35M or 7% turnover

## 7. Insurance Underwriting & Claims AI

HIGH

AI Classification: Evaluative / Autonomous AI | Exposure Score: 4/5

### REGULATORY TRIGGERS

- **State insurance codes** — Unfair discrimination in rating
- **EU AI Act Annex III** — Life/health insurance pricing; Aug 2026
- **NAIC model bulletins** — AI/ML governance expectations
- **Colorado AI Act** — Insurance deployer obligations; June 2026

### PLAINTIFF LEGAL THEORIES & CASE LAW

- **Unfair discrimination:** AI pricing models using behavioral data correlating with protected characteristics.
- **Bad faith denial:** Claims algorithms overriding adjuster judgment to reduce payouts.
- **Opacity:** Policyholders cannot understand AI-driven premium calculations.
- **Cross-context data misuse:** Marketing data repurposed for underwriting without consent.

### ARCHITECTURAL FAILURE MODES

- ✗ Pricing models using behavioral data correlating with protected characteristics
- ✗ Claims triage algorithms auto-denying without human review
- ✗ Opaque premium calculations unexplainable to policyholders
- ✗ Cross-context data reuse (marketing data into underwriting without consent)
- ✗ No testing for unfair discrimination in rating outcomes

### ARCHITECTURAL MITIGATIONS

- ✓ Pre-deployment disparate impact testing on all rating variables
- ✓ Actuarial justification documentation for AI-derived factors
- ✓ Human adjuster review for claim denials above materiality threshold
- ✓ Policyholder-facing explanation of key rating factors
- ✓ Strict data firewall between marketing and underwriting

#### ◆ GUARDRAIL BENCHMARK (ZEE MODEL)

Insurance AI shares the 'hidden evaluation' failure mode. The Zee principle of separating 'what has this person done' from 'who is this person' translates directly: use observable risk factors, not behavioral inferences functioning as character judgments.

### KEY CASES

NAIC model bulletins (2023–2025) • Colorado AI Act insurance provisions (June 2026) • EU AI Act Annex III (Aug 2026)

### PENALTY EXPOSURE

State insurance commissioner penalties vary • Colorado: \$20K per violation • EU: up to €35M or 7% turnover • License revocation risk

## 8. Predictive Policing & Surveillance AI

HIGH

AI Classification: Evaluative / Autonomous AI | Exposure Score: 4/5

### REGULATORY TRIGGERS

- **Fourth Amendment** — Unreasonable search/surveillance
- **First Amendment** — Chilling effect on association/expression
- **Fourteenth Amendment** — Due process and equal protection
- **EU AI Act prohibited practices** — Untargeted facial recognition scraping banned Feb 2025
- **FTC enforcement** — Deceptive accuracy claims (Rite Aid consent order)

### PLAINTIFF LEGAL THEORIES & CASE LAW

- **Constitutional violations:** First settlement where law enforcement admitted predictive policing violated Constitution. Pasco County scored residents, targeted minors using school data. Settlement: \$105K + permanent program termination + admission of Fourth/First/Fourteenth Amendment violations. (*Taylor v. Nocco*, M.D. Fla., settled Dec 2024)
- **Biometric data:** \$51.75M BIPA settlement against Clearview AI for scraping facial images (2025).
- **FTC deception:** Rite Aid facial recognition deployed without accuracy testing.

### ARCHITECTURAL FAILURE MODES

- ✗ Risk scoring residents using non-criminal data (school records, social connections)
- ✗ Targeting minors using school data without parental consent
- ✗ Facial recognition deployed without cross-demographic accuracy testing
- ✗ No transparency to scored individuals about risk designation
- ✗ Persistent surveillance profiles without expiration or review
- ✗ Marketing accuracy claims unsupported by real-world performance

### ARCHITECTURAL MITIGATIONS

- ✓ Eliminate individual risk scoring absent specific criminal intelligence
- ✓ Prohibit school records and juvenile data in scoring systems
- ✓ Mandatory accuracy testing across demographics before deployment
- ✓ Public transparency about algorithmic tools used by law enforcement
- ✓ Sunset and expiration provisions for all surveillance designations
- ✓ Independent civilian oversight and third-party audit

#### ◆ GUARDRAIL BENCHMARK (ZEE MODEL)

Pasco County's predictive policing is the public-sector equivalent of hidden employability scoring. The Taylor settlement's admission of constitutional violations validates the Zee principle: opaque evaluation of individuals at scale creates existential legal risk, regardless of sector.

## KEY CASES

*Taylor v. Nocco (M.D. Fla., settled Dec 2024) • Clearview AI (\$51.75M BIPA, 2025) • Rite Aid FTC consent order • EU prohibited practices (Feb 2025)*

## PENALTY EXPOSURE

§1983: uncapped damages • \$105K + program termination (Taylor) • \$51.75M (Clearview BIPA) • EU: up to €35M or 7% turnover

## 9. Marketing AI & Algorithmic Targeting

MODERATE

AI Classification: Assistive / Evaluative AI | Exposure Score: 3/5

### REGULATORY TRIGGERS

- **FTC Act §5** — Deceptive/unfair practices in AI marketing claims
- **SEC anti-fraud** — AI-washing: misrepresenting AI capabilities
- **State UDAP/DTPA** — Consumer protection statutes
- **Illinois BIPA** — Biometric data for ad targeting without consent
- **CCPA/CPRA** — Automated decision-making and profiling

### PLAINTIFF LEGAL THEORIES & CASE LAW

- **AI-washing (criminal + civil)**: First parallel SEC + DOJ prosecution. Shopping app falsely claimed AI automation; relied on overseas contractors. (*SEC v. Nate/Saniger*, April 2025)
- **FTC enforcement**: Operation AI Comply targeting deceptive AI claims. \$20M+ judgment against Click Profit.
- **Biometric targeting**: \$136.6M in BIPA settlements in 2025 including \$51.75M Clearview AI, \$47.5M Motorola.
- **Discriminatory ad delivery**: AI algorithms producing discriminatory distribution in housing/employment ads.

### ARCHITECTURAL FAILURE MODES

- ✗ Claiming AI automation when humans perform the work (Nate/Saniger pattern)
- ✗ Biometric data for targeting without explicit consent
- ✗ Ad delivery producing discriminatory distribution across protected classes
- ✗ No audit mechanism for ad delivery disparate impact
- ✗ AI-generated content not labeled as synthetic

### ARCHITECTURAL MITIGATIONS

- ✓ Substantiate all AI capability claims with documented evidence
- ✓ No biometric data collection for targeting without explicit opt-in
- ✓ Disparate impact audits on ad delivery for housing/employment/credit
- ✓ Label AI-generated marketing content per disclosure requirements
- ✓ Clear data lineage documentation for all targeting variables

#### ◆ GUARDRAIL BENCHMARK (ZEE MODEL)

The AI-washing risk is now prosecuted criminally — Nate/Saniger (parallel SEC + DOJ) establishes that overstating AI capabilities is a federal offense. For Sales and Marketing teams: every claim about what AI 'does' must be substantiated by what the system actually performs.

### KEY CASES

*SEC v. Nate/Saniger (April 2025)* • *FTC Operation AI Comply (Sept 2024–present)* • *BIPA settlements (\$136.6M, 2025)*

## PENALTY EXPOSURE

SEC: disgorgement + criminal prosecution • FTC: \$20M+ judgments • BIPA: \$1,000–\$5,000 per violation (class-wide)

## 10. AI Agents (Autonomous Decision Systems)

EMERGING — HIGH  
TRAJECTORY

AI Classification: Autonomous AI | Exposure Score: 4/5

### REGULATORY TRIGGERS

- **EU AI Act** — Agent classified as high-risk if operating in Annex III domain
- **Texas TRAIGA** — Deceptive/manipulative AI; agents must disclose non-human nature
- **State consumer protection (UDAP)** — Unauthorized agent actions as unfair practices
- **Sector-specific regulations** — Agents inherit compliance obligations of their domain
- **Emerging fiduciary duty theories** — AI acting 'on behalf of' users

### PLAINTIFF LEGAL THEORIES & CASE LAW

- **Unauthorized action [PROJECTION]**: AI agent initiates transaction without genuine human authorization. Theory extends from existing agency law.
- **Product liability extension**: Garcia ruling (AI = product) + autonomous action = strict liability for agent decisions.
- **Scope creep liability [PROJECTION]**: Agent exceeds authorized boundaries. Legal theory crystallizing.
- **Fiduciary breach [PROJECTION]**: AI agent as advisor without licensing or duty of care.
- **NOTE**: This category is labeled 'Emerging' — theories crystallizing but limited precedent. Trajectory clearly toward HIGH.

### ARCHITECTURAL FAILURE MODES

- ✗ No clear boundary between recommendation and action
- ✗ Autonomous execution without confirmation gates
- ✗ Scope creep: agent begins initiating tasks beyond authorization
- ✗ No audit trail of decisions, rationale, and authorizations
- ✗ Agent operating in regulated domain without domain-specific compliance
- ✗ Multi-agent systems where accountability chain breaks

### ARCHITECTURAL MITIGATIONS

- ✓ Explicit human confirmation gates before consequential actions
- ✓ Clear scope boundaries with hard stops at authorization limits
- ✓ Full audit trail: decision, rationale, human authorization, timestamp
- ✓ Domain-specific compliance checks before operating in regulated areas
- ✓ Mandatory disclosure of AI agent status in consumer interactions
- ✓ Kill switches and rollback capabilities for agent actions

#### ◆ GUARDRAIL BENCHMARK (ZEE MODEL)

Zee's AI Agent Compliance Radar identifies decision delegation as the primary risk. The principle: 'Where is the system drifting closer to autonomous action, persistent judgment, or cross-context reuse? Those are the zones where future regulation will land first.' Human intent must remain the initiating force.

## KEY CASES

*Garcia v. Character.AI (product liability framework) • DOJ v. RealPage (auto-accept = autonomous agent) • EU AI Act agent provisions (Aug 2026)*

## PENALTY EXPOSURE

Sector-dependent: inherits penalty structure of domain • Product liability: uncapped • Penalties compound across violated frameworks • EU: up to €35M or 7% turnover

## IV. Cross-Cutting Design Failure Patterns

Six architectural failure patterns recur across multiple sectors. Each represents a design decision that has independently triggered litigation or regulatory enforcement in two or more AI application categories.

### 1. Opaque Scoring Without Explanation

**Affected Sectors:** Hiring, Housing, Insurance, Lending, Policing

**Corrective Principle:** Factor-level transparency with contestability rights

### 2. Automated Decisions Marketed as Human

**Affected Sectors:** Healthcare, Hiring, Pricing

**Corrective Principle:** Genuine HITL with measurable override rates and time-on-task metrics

### 3. Proxy Discrimination via Facially Neutral Variables

**Affected Sectors:** Hiring, Lending, Housing, Insurance

**Corrective Principle:** Pre-deployment disparate impact testing and variable-level audit

### 4. No Contestability or Correction Mechanism

**Affected Sectors:** Hiring, Housing, Healthcare, Policing

**Corrective Principle:** Subject access, correction rights, and dispute pathways by design

### 5. Persistent Evaluative Judgments Across Contexts

**Affected Sectors:** Hiring, Policing, Insurance

**Corrective Principle:** Role-scoped, time-limited outputs with mandatory expiration

### 6. Safety Detection Without Intervention

**Affected Sectors:** Chatbots, Healthcare

**Corrective Principle:** Hard intervention triggers with human escalation pathways

## V. Methodology & Sources

**Research Scope.** This report synthesizes analysis of 25+ lawsuits, enforcement actions, regulatory releases, and credible reporting covering customer-facing AI systems between 2024 and 2026.

**Source Hierarchy.** Court filings and regulatory releases are primary sources. Settlement terms are established fact. Credible investigative reporting (Reuters, Senate committee reports, state AG releases) is secondary. Where evidence is weak or evolving, uncertainty is stated explicitly.

**AI Classification Framework.** Each category is classified as Assistive (surfaces information, human decides), Evaluative (renders judgments, human may or may not review), or Autonomous (initiates/finalizes actions without explicit human authorization).

**Exposure Rating.** CRITICAL = active litigation with adverse precedent and uncapped damage exposure. HIGH = active litigation/enforcement with significant penalty exposure. MODERATE = enforcement with bounded penalties. EMERGING = theories crystallizing without settled precedent, high trajectory.

**Guardrail Benchmark.** The Zee AI Hiring Platform compliance architecture is used as comparative reference for identifying safer design patterns. Zee's principles (decision-support vs. decision-maker, role-scoped outputs, transparency, contestability, inference scoping) are applied as benchmark across all sectors.

**Limitations.** This analysis reflects publicly available information as of February 2026. Pending cases may resolve differently. Regulatory frameworks are evolving rapidly. The AI Agents category relies substantially on projection from existing legal theories.



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END OF REPORT