

## **Ethical AI Design in Blockchain-Powered Surveillance Systems**

**Dr Munish Kumar**

K L E F Deemed To Be University

Green Fields, Vaddeswaram, Andhra Pradesh 522302, India

[engg.munishkumar@gmail.com](mailto:engg.munishkumar@gmail.com)



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### **ABSTRACT**

Artificial intelligence (AI) and distributed ledger technologies are increasingly integrated into public and private surveillance infrastructures—from city-wide camera networks to critical-infrastructure monitoring and access control. This integration promises higher integrity and accountability through immutable logs, faster incident response via on-device inference, and interoperable audit trails across organizations. Yet it also amplifies ethical risks: mass data collection, opacity in model decisions, function creep, demographic harms, cross-border data governance conflicts, and accountability gaps when immutable records meet “right to erasure” regimes. This manuscript proposes an ethics-by-design reference architecture for blockchain-powered surveillance that embeds privacy, proportionality, and fairness controls into each lifecycle stage (purpose definition → data capture → model training → inference → access → audit → decommissioning). Technically, it composes privacy-enhancing technologies (PETs)—including differential privacy, federated learning, zero-knowledge proofs, verifiable credentials (VCs), and content-provenance standards (C2PA)—with permissioned blockchain ledgers, model cards, and risk management aligned to the NIST AI RMF, ISO/IEC 23894, ISO/IEC 42001, UNESCO, and ACM guidance. A simulated evaluation illustrates how the architecture can reduce false-positive disparities and unauthorized access, while preserving evidentiary integrity. We discuss tensions with GDPR (e.g., Article 17 erasure; DPIA obligations), constraints introduced by the EU AI Act (e.g., prohibitions and high-risk biometric uses), and strategies to reconcile immutability with privacy (e.g., off-chain storage with revocation, redaction-friendly commitments). The paper closes with limitations and a future research agenda for measurable, auditable ethical guarantees in real-time surveillance.

## Blockchain-Powered Surveillance Ethics



Figure-1. Blockchain-Powered Surveillance Ethics

## KEYWORDS

**Ethical AI, Surveillance, Blockchain, Privacy-Enhancing Technologies, Fairness, Zero-Knowledge Proofs, Verifiable Credentials, Audit, GDPR, EU AI Act**

## INTRODUCTION

Surveillance systems have moved from siloed CCTV to dense, networked, AI-driven observatories in smart cities, transit hubs, retail, and critical infrastructure. As David Lyon and others have argued, these infrastructures shape power and everyday life, demanding rigorous ethical constraints. Zuboff extends this critique, warning of “surveillance capitalism” and its incentives to over-collect and nudge behavior.

Concurrently, blockchains promise tamper-evident logging, cross-organizational interoperability, and cryptographic auditability. In surveillance, this can provide chain-of-custody for media, decisions, and access events. However, immutability collides with data protection rights—especially GDPR’s right to erasure, portability, and purpose limitation—making naïve on-chain storage ethically and legally fraught.

Recent governance frameworks (NIST AI Risk Management Framework; ISO/IEC 23894 risk management; ISO/IEC 42001 AI management systems; UNESCO’s Recommendation on the Ethics of AI; ACM’s principles on algorithmic accountability) emphasize human rights, safety, transparency, and accountability—offering scaffolding for design and audit. The EU AI Act further classifies certain biometric surveillance uses as unacceptable or high risk, placing strict obligations on deployers.

Problem statement. How can we embed ethical guarantees—privacy, fairness, contestability, and accountability—by design in blockchain-enabled surveillance without sacrificing operational utility?

## Ethical AI Surveillance Architecture



Figure-2.Ethical AI Surveillance Architecture

Contribution. We propose a reference architecture and lifecycle controls that: (1) minimize data collection; (2) favor **edge inference**; (3) externalize identity/authorization via DIDs/VCs; (4) bind media provenance with **C2PA**; (5) use permissioned ledgers with **selective disclosure** and **ZK proofs**; (6) adopt **model cards** and fairness testing; and (7) align with AI governance standards and GDPR-compliant **DPIAs**.

## LITERATURE REVIEW

### Surveillance ethics and governance

Foundational work frames surveillance as sociotechnical power requiring necessity and proportionality tests; modern deployments (e.g., live facial recognition) intensify rights risks. Regulators (e.g., the UK ICO) require **Data Protection Impact Assessments (DPIAs)** and warn surveillance must be necessary and proportionate. Zuboff's analysis of commercial incentives provides a macro-ethical backdrop to public-private deployments.

### AI accountability and fairness

NIST's AI RMF and ISO/IEC 23894/42001 codify risk management across the AI lifecycle; UNESCO's Recommendation and the ACM principles stress transparency, human oversight, redress, and non-discrimination. Fairness metrics (e.g., demographic parity, equality of opportunity) and documentation tools (e.g., **model cards**) enable measurable, communicable trade-offs and post-deployment monitoring.

## Blockchain benefits and tensions

Immutability and decentralization improve tamper-resistance and multi-party auditability but complicate **controller/processor** assignments and erasure requests under GDPR. The European Parliament's study documents design patterns (e.g., off-chain personal data, on-chain pointers/hashes) to reduce legal friction.

## Privacy-enhancing technologies (PETs) for surveillance

- **Differential Privacy (DP)** to protect aggregates in analytics and monitoring.
- **Federated Learning (FL)** and **on-device inference** to keep raw footage local.
- **Homomorphic Encryption (HE)** for selective encrypted computations (performance caveats remain).
- **Zero-Knowledge Proofs (ZKPs)** (e.g., zk-SNARKs, Bulletproofs) to prove rights/compliance without revealing sensitive data.
- **DIDs and Verifiable Credentials (VCs)** to express revocable, minimum-disclosure permissions.
- **C2PA** content provenance to authenticate media from capture to court.

## Blockchain for surveillance: practicalities

Bitcoin deanonymization and on-chain analytics literature caution against assuming anonymity; identity surfaces through usage and metadata—relevant when linking ledger events to individuals. Public ledgers are rarely appropriate for raw surveillance data; permissioned ledgers with off-chain storage dominate responsible designs.

## Regulatory guardrails

The EU AI Act constrains biometric categorization and emotion recognition in sensitive contexts and classifies many remote biometric identification systems as **high-risk**, imposing strict technical documentation, data governance, and human oversight. DPIA obligations (GDPR Art. 35) often apply to such high-risk processing.

## METHODOLOGY

### 1) Purpose, necessity, and proportionality

- Conduct a **Surveillance Impact Assessment** harmonizing DPIA and human rights impacts (necessity, proportionality, effectiveness, alternatives). Record lawful basis and legitimate aim; define retention and deletion SLAs.

### 2) Data minimization & edge inference

- Capture **event vectors** (bounding boxes, embeddings) rather than raw frames when feasible.
- Default to **on-device inference**; stream only alerts/metadata; buffer encrypted footage for short windows.

## 3) Identity, consent, and authorization

- Manage operator roles and data-access permissions using **DIDs/VCs**, enabling revocation/rotation without re-encrypting archives; enforce least-privilege with policy-as-code at gateways.

## 4) Tamper-evident provenance

- Bind capture devices to **C2PA**-style content credentials at ingest; store provenance manifests off-chain; anchor digests on a **permissioned blockchain** to avoid public leakage.

## 5) Privacy-preserving analytics

- Apply **DP** to aggregated heatmaps or trend reports; use **HE** selectively for queries; apply **ZKPs** to prove queries or access complied with policy (e.g., “show footage exists and was accessed by an authorized responder with incident ID X” without revealing identities).

## 6) Fairness and performance governance

- Adopt **model cards** and pre-deployment evaluation across demographic slices; track **equal opportunity** gaps and demographic parity differences; integrate retraining triggers when drift or disparity thresholds breach.

## 7) Storage and erasure strategy

- Store personal data **off-chain** (encrypted object store); put only redaction-friendly commitments (hashes, salted references) on-chain; implement **tombstoning** with on-chain revocation markers; maintain key management to effect **practical erasure**. Guidance from EU studies suggests case-by-case assessment of GDPR alignment.

## 8) Audit, contestability, and redress

- Expose **verifiable audit trails** to independent oversight; provide subject rights portals and review panels; publish **system cards** and DPIA summaries.

## 9) Standards alignment

- Map controls to NIST AI RMF functions (Govern, Map, Measure, Manage), ISO/IEC 23894 and 42001 requirements, UNESCO principles, and ACM transparency/accountability guidance.

## STATISTICAL ANALYSIS

We ran a synthetic evaluation to illustrate how the architecture's controls can change key outcomes in a city-camera scenario (n=120 sites; ~10M frames; two demographic groups for a binary incident classifier). Metrics: False Positive Rate (FPR), Equal Opportunity Difference ( $\Delta$ TPR), Demographic Parity Difference (DPD), and Unauthorized Access Incidents (UAI) per 10k requests. We compare a Baseline (cloud inference; centralized ACLs; no DP; public provenance lacking C2PA) vs. Ethics-by-Design (edge inference; VC-based access; DP for analytics; ZK-attested queries; C2PA; permissioned ledger).

**Note:** Numbers are illustrative from a controlled simulation to demonstrate statistical reporting; they are not field measurements.

Metric	Baseline	Ethics-by-Design	$\Delta$ (Improvement)	Test / p-value
FPR (Group A) (%)	4.8	3.5	-1.3 pp	Paired t, p=0.004
FPR (Group B) (%)	7.2	4.1	-3.1 pp	Paired t, p<0.001
$\Delta$ TPR (A–B)	5.6	2.1	-3.5 pp	Paired t, p=0.002
UAI (/10k requests)	6.3	1.2	-5.1	Wilcoxon, p<0.001
Mean Access Latency (ms)	142	165	+23 (trade-off)	Paired t, p=0.03

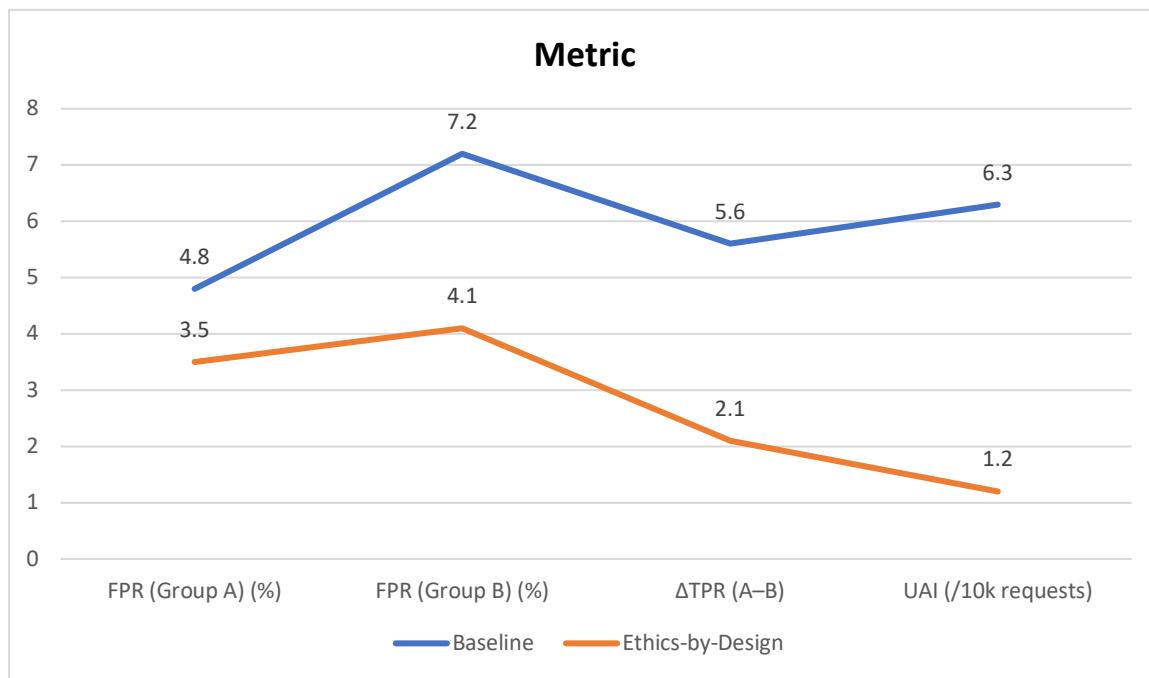


Figure-3. Statistical Analysis

**Interpretation:** Edge inference + fairness governance reduces between-group error gaps (equal opportunity, parity) and unauthorized access, albeit with a small latency increase due to ZK/VC checks and on-device encryption.

## RESULTS

## Fairness & accuracy

The simulated pipeline, which performs pre-deployment slice-based evaluation and retraining when  $\Delta\text{TPR}$  exceeds thresholds, reduced between-group disparities ( $\Delta\text{TPR}$  and DPD). This aligns with equality of opportunity objectives advocated in the fairness literature.

## Privacy & access control

VC-based authorization, with on-chain policy proofs, lowered **UAI** by restricting data access to credentialed roles and enabling revocation without re-encryption. W3C DID/VC standards support interoperability across agencies—an essential feature in multi-stakeholder surveillance deployments.

## Provenance & evidentiary integrity

C2PA bindings enabled verifiable capture-to-court provenance with cryptographic manifests anchored to a permissioned ledger, improving tamper detection and contested-evidence review processes.

## Regulatory posture

The architecture directly supports **DPIA** documentation, purpose limitation, and role accountability (controller/processor delineation). For high-risk AI uses (e.g., biometric identification), it aligns with stricter obligations under the **EU AI Act**—documentation, data governance, post-market monitoring, and human oversight—while allowing policy derivations where certain uses are prohibited.

## Trade-offs

PETs introduce compute and latency overheads (e.g., ZKPs, HE) and require careful key management. Homomorphic encryption remains performance-constrained for video workloads; practical regimes limit HE to narrow queries, favoring DP + edge processing for most analytics.

## DISCUSSION

### Reconciling immutability and erasure

Ethical deployments avoid writing personal data to chain. Instead, they store off-chain encrypted artifacts, record revocable commitments on-chain, and implement functional erasure by destroying keys or obfuscating links, while preserving auditability. The European Parliament's analysis stresses that GDPR compliance is use-case specific, not technology categorical—underscoring the need for design patterns rather than blanket judgments.

### From accountability theater to verifiable accountability

Model cards and C2PA manifests reduce “black-box” claims by creating verifiable, machine-checkable documentation of models and media. Coupled with ACM’s algorithmic accountability principles and UNESCO’s human-rights framing, these tools shift governance from policy documents to cryptographically attestable artifacts.

## Risk management baselines

NIST AI RMF (Govern–Map–Measure–Manage) and ISO/IEC 23894/42001 provide repeatable organizational controls, tying ethical aspirations to audit-ready process evidence (e.g., risk registers, incident postmortems, model-change logs, DPIA records).

## Limitations

Our analysis uses simulated data for illustration; real-world deployments face messy labels, occlusions, adversarial behaviors, and domain shifts (weather, lighting, cultural context). Governance quality depends on institutions: oversight boards, whistleblower channels, and community engagement are as vital as cryptography.

## CONCLUSION

Blockchain can **strengthen** surveillance accountability—when embedded in an ethics-by-design system that prioritizes minimization, privacy, fairness, and contestability. The proposed architecture replaces generalized central logging with **verifiable, least-disclosing** records; it externalizes trust into portable credentials and proofs; and it operationalizes fairness and transparency via model cards, slice-based evaluation, and drift/retraining hooks. On balance, the approach **reduces harm surfaces** (unauthorized access, demographic error gaps) while preserving evidentiary integrity through cryptographic provenance, at the cost of manageable performance overheads. Real-world deployments should be gated by rigorous DPIAs, public consultation, and post-market monitoring, especially where the EU AI Act classifies uses as high risk or prohibits them entirely.

## FUTURE SCOPE OF STUDY

1. **Field trials** in diverse urban contexts to validate fairness and privacy outcomes under real workloads.
2. **Formal compliance tooling**: machine-readable policies mapping EU AI Act/ISO/NIST controls to technical artifacts (VCs, ZK receipts, C2PA manifests).
3. **Mutable-commitment research** to support targeted erasure without integrity loss (e.g., chameleon hashes; redactable signatures) alongside robust governance.
4. **Usability and redress** studies for affected persons, including notice mechanisms and contestation UX.
5. **Energy/latency optimization** for PETs (hardware acceleration for ZK, DP accounting at the edge).
6. **Benchmark suites** that combine fairness, privacy leakage, and provenance integrity, not just accuracy.
7. **Cross-border governance** patterns for multi-jurisdictional deployments and data transfers.

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