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Article

# Artificial Intelligence and the Future of Legal Accountability

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Abstract: AI is quickly changing legal systems all throughout the world, which raises big challenges about who is responsible, who is liable, and how to govern ethically. As AI-driven tools have more and more of an effect on things like contract analysis, compliance monitoring, and predictive policing, the conventional bases of legal accountability are facing new and difficult problems. It is harder to find errors in advanced AI systems since they are independent and flexible. This is especially true when results come from methods that aren't clear, machine-learning biases, or decentralized data processes. This new landscape calls for a re-evaluation of liability frameworks to figure out who should be responsible: the creators, the users, the deployers, or the AI systems themselves. The incorporation of AI necessitates strong legislative frameworks that guarantee transparency, elucidation, and equity. This study examines the changing ways that countries around the world are dealing with these issues, suggests models for shared responsibility, and stresses the necessity for consistent legal norms to find a balance between innovation and justice. In the end, the future of AI in law hinges on how well it fits with ideas of accountability and human rights.

**Keywords**: Artificial Intelligence, Legal Accountability, Liability Frameworks, Algorithmic Transparency, Regulatory Governance

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# INTRODUCTION

AI, or artificial intelligence, is one of the most important technologies of the 21st century. It has had an impact on many areas, including healthcare, finance, government, and the administration of justice. AI is quickly becoming more common in the legal field thanks to tools like predictive analytics, automated contract drafting, ediscovery tools, risk assessment algorithms, and judicial decision-support systems. These new technologies promise to make things more efficient, accurate, and accessible, but they also make things more complicated

when it comes to who is responsible and accountable. Conventional legal frameworks are predicated on the notions of human agency, intention, and control principles that become progressively challenging to implement when decisions are influenced or rendered by autonomous AI systems.

Machine-learning models are hard to understand, can be biased, and self-learning algorithms are hard to forecast. This makes for a unique lack of responsibility. When an AI system makes mistakes, treats people unfairly, or causes injury, it might be hard to figure out who is responsible:

the developers, programmers, data producers, deployers, or end users. Also, the fact that AI technologies are used in more than one country makes it harder to set up jurisdictional boundaries and regulatory control.

As legal systems try to deal with these new problems, more and more people throughout the world are calling for the creation of clear rules for how AI can be used that make sure it is fair, open, and follows the rules. This means changing how liability works, making rules for how algorithms should be able to explain themselves, and putting in place safety measures for governance. So, it's important to think about the future of legal accountability in the age of AI, not just to defend people's rights but also to keep people's faith in the courts and government. This introduction lays the groundwork for examining the necessary evolution of legal accountability in light of swiftly evolving AI technologies.

#### Statement of the Research Problem

The swift incorporation of Artificial Intelligence (AI) into legal processes spanning judicial decision-making, predictive policing, contract management, and compliance systems has engendered a substantial disparity between technology progress and current legal accountability frameworks. Conventional legal principles of culpability, based on human intention, control, and foreseeability, are progressively insufficient for tackling harms inflicted by autonomous or semi-autonomous AI systems. When an AIpowered tool gives an incorrect, biased, or dangerous result, it is still not obvious who should be legally responsible: the developers, data trainers, system deployers, end users, or the AI system itself. There is a lot of disagreement about this. This lack of clarity creates a situation where no one is responsible, which puts people's rights at risk, erodes public trust, and puts legal institutions at risk of ethical and procedural problems. Consequently, the research subject focuses on the identification of suitable, flexible, and ethically sound accountability frameworks capable of effectively tackling the challenges presented by AI within contemporary legal systems.

# Significance of the Study

This study is very important in today's legal and technical world. As AI becomes a bigger part of legal work, like helping judges, automating contracts, making sure laws are followed, and predicting crime, the need for strong and transparent accountability systems is more important than ever. The study helps close the gap between law and new technology by looking at how AI changes our ideas about accountability, liability, and transparency.

The study offers significant insights for policymakers, legal professionals, technologists, and academics by suggesting organized frameworks of responsibility that adhere to ethical norms and human rights criteria. These findings can help create new rules that can deal with the hazards of AI while still encouraging fresh ideas. The study stresses the need for algorithmic fairness and explainability in judicial systems to stop bias and make sure that justice is done.

The findings can also help build public trust by encouraging the appropriate use of AI and protecting people from any misuse or unfair results. The study eventually enhances the worldwide dialogue on responsible AI governance, acting as a fundamental

reference for forthcoming legal reforms and transdisciplinary investigations.

#### Research Gap

Even if there is more and more writing about AI and the law, we still don't have a clear and complete grasp of how legal responsibility works with autonomous and semi-autonomous AI systems. Current research predominantly emphasizes technical obstacles, ethical considerations, or overarching regulatory frameworks, frequently neglecting the intricate legal issues that emerge when AI-generated actions or judgments cause harm. There is a dearth of scholarly work that methodically investigates the adaptation or expansion of conventional liability doctrines such as negligence, strict liability, and vicarious liability to suit AI-driven processes.

Additionally, limited research provides comparative analyses of global accountability models or suggests realistic, implementable frameworks that reconcile technological advancement with legal protections. Current research also doesn't look into the roles of algorithmic opacity, bias, data ownership, and cross-border regulatory conflicts enough. Moreover, empirical evidence concerning actual legal issues utilizing AI is limited, resulting in a disconnect between theoretical discourse and practical implementation.

Consequently, this study fills a significant research void by amalgamating legal theory, technological realities, and policy viewpoints to provide structured accountability frameworks adapted to the dynamic characteristics of AI.

#### **Objectives:**

- 1. To examine how Artificial Intelligence challenges traditional legal accountability frameworks and identify gaps in existing doctrines of liability, responsibility, and regulatory oversight.
- **2.** To propose structured, ethically grounded, and practically implementable models for ensuring transparency, fairness, and accountability in AI-driven legal processes.

# **Hypotheses:**

- 1. Alternate Hypothesis (H<sub>1</sub>): The integration of Artificial Intelligence into legal processes significantly disrupts traditional accountability frameworks, creating identifiable gaps that require new, structured models of liability and regulatory oversight.
- 2. Null Hypothesis ( $H_0$ ): The integration of Artificial Intelligence into legal processes does not disrupt traditional accountability frameworks, and existing legal doctrines of liability and responsibility remain sufficient to address AI-driven outcomes.

#### **Review of Literature:**

1. Calo (2015) examines the ways in which robotics contests conventional legal doctrines through the integration of autonomous decision-making and erratic machine behavior. He contends that current cyberlaw frameworks are inadequate for resolving liability, privacy, and accountability concerns associated with advanced robotic systems. The paper stresses the necessity for new legal doctrines that take into account how technology works, with a focus on openness and rules. Calo's research is a basic study that helps us understand how new

technologies change the way we think about the law and how we need new ways to govern them.<sup>1</sup>

- 2. Bryson and Winfield (2017) stress how important it is to include moral rules in the creation of AI and autonomous systems. Their work makes the case for standardized frameworks that make sure the AI development process is open, secure, and responsible. They talk about the dangers of AI being able to do whatever it wants and suggest that ethical design guidelines could help stop these problems. This paper is important because it moves the focus from regulating AI after harm has happened to proactive ethical engineering. This gives a structured way to manage AI responsibly.<sup>2</sup>
- **3.** Kroll et al. (2017) talk on the growing need for algorithmic accountability as automated systems have more and more of an effect on legal and administrative judgments. They say that algorithms that aren't clear make things less fair, less open, and less fair. The authors suggest technological and legal ways to make sure that algorithmic decision-making is accountable, like auditability, verification methods, and procedural safeguards. This study is essential for comprehending the intersection of law and technology in developing frameworks that safeguard individuals against prejudiced or unjust AI-driven results.<sup>3</sup>
- 4. Selbst and Powles (2017) conduct a critical examination of the notion of "meaningful information" in data protection legislation, with specific emphasis on the right to explanation in automated decision-making. They say that current laws, like those in the GDPR, don't really provide people a good understanding of how algorithms work because they are too complicated and the rules are too unclear. Their work shows how the law and technology don't always match up, and it stresses the need for explicit, enforceable norms for transparency in AI-driven systems.<sup>4</sup>
- 5. Wirtz, Weyerer, and Geyer (2019) examine the effects of Artificial Intelligence on public sector performance, emphasizing advantages such as enhanced efficiency, precision, and service delivery. However, they stress that there are big problems, such as gaps in accountability, ethical issues, and the possibility of algorithmic bias in government operations. The authors put out a thorough study plan that focuses on governance, transparency, and regulatory frameworks. Their work helps us understand how public institutions need to have balanced responsibility and strong oversight systems in place when they use AI.<sup>5</sup>
- 6. Aletras et al. (2016) present one of the earliest empirical studies using machine learning to predict judicial decisions of the European Court of Human Rights. Their research demonstrates that AI models can identify patterns from case texts with nearly 79% accuracy, highlighting AI's potential in legal analytics and decision-support. However, the study also underscores concerns about transparency

and interpretability, as predictive models may not fully explain underlying legal reasoning. This work significantly contributes to debates on AI in judicial processes.<sup>6</sup>

- 7. Allen, Smit, and Wallach (2005) examine how artificial systems can be designed to exhibit moral behavior through top-down, bottom-up, and hybrid approaches. Their study explores whether ethical principles should be programmed explicitly or learned autonomously through machine interactions. The authors argue that hybrid models offer the greatest potential by combining structured moral rules with adaptive learning. This foundational work highlights important implications for accountability, as morally aware AI systems could reduce risks while raising new regulatory and ethical challenges.<sup>7</sup>
- 8. Barfield and Pagallo (2018) provide a comprehensive examination of the evolving relationship between law and artificial intelligence. Their handbook covers key themes such as liability, data protection, autonomous systems, and regulatory governance. The authors highlight significant gaps in existing legal frameworks and emphasize the urgency of developing AI-specific accountability mechanisms. By integrating perspectives from law, ethics, and technology, this work serves as a foundational reference for understanding how legal systems must adapt to address challenges arising from advanced AI deployment.<sup>8</sup>
- 9. Bathaee (2018) critically examines how the "black box" nature of advanced AI systems undermines traditional legal concepts of intent and causation. He argues that machine-learning models operate through opaque processes that make it difficult to identify responsibility when harm occurs. The article highlights limitations in negligence and strict liability doctrines when applied to autonomous systems. Bathaee's work is a significant contribution, emphasizing the need for new liability frameworks that address the unique challenges posed by non-transparent AI technologies.<sup>9</sup>
- 10. Binns (2018) explores the concept of fairness in machine learning by drawing parallels with political philosophy, particularly theories of justice and equality. He argues that technical definitions of fairness often overlook deeper ethical principles, resulting in systems that may satisfy mathematical fairness criteria but still produce unjust outcomes. The study highlights the need for interdisciplinary approaches combining philosophy, ethics, and computer science. Binns's work is crucial in shaping discussions on algorithmic accountability and responsible AI governance.<sup>10</sup>
- 11. Burrell (2016) analyzes the different forms of opacity inherent in machine-learning systems, identifying three types: intentional secrecy, technical complexity, and the inherent opacity of algorithmic processes. She argues that these factors make it difficult for users, regulators, and even developers to fully understand how AI models

<sup>&</sup>lt;sup>1</sup> Calo (2015)

<sup>&</sup>lt;sup>2</sup> Bryson and Winfield (2017)

<sup>&</sup>lt;sup>3</sup> Kroll et al. (2017)

<sup>&</sup>lt;sup>4</sup> Selbst and Powles (2017)

<sup>&</sup>lt;sup>5</sup> Wirtz, Weyerer, and Geyer (2019)

<sup>&</sup>lt;sup>6</sup> Aletras et al. (2016)

<sup>&</sup>lt;sup>7</sup> Allen, Smit, and Wallach (2005)

<sup>&</sup>lt;sup>8</sup> Barfield and Pagallo (2018)

<sup>&</sup>lt;sup>9</sup> Bathaee (2018)

<sup>&</sup>lt;sup>10</sup> Binns (2018)

generate outcomes. Burrell's work is central to debates on transparency and explainability, highlighting how opaque AI systems complicate accountability, oversight, and public trust in automated decision-making.<sup>11</sup>

- 12. Crawford and Paglen (2019) examine the political and ethical implications of image datasets used to train machine-learning systems. They argue that training sets often contain biased, culturally loaded, or improperly labeled images, which can embed discriminatory assumptions into AI models. Their analysis reveals how invisibly curated datasets shape algorithmic behavior and societal outcomes. This study is significant in highlighting dataset accountability and the urgent need for transparency and ethical standards in AI training practices. <sup>12</sup>
- 13. Dani (2021) explores the regulatory challenges and opportunities India faces in governing Artificial Intelligence. The study evaluates gaps in existing legal frameworks, emphasizing issues related to data protection, algorithmic bias, and accountability in AI deployment. Dani argues for a comprehensive, India-specific regulatory model that balances innovation with public safety and ethical safeguards. This work contributes to policy discourse by highlighting India's unique socio-legal context and the urgent need for adaptive AI governance frameworks.<sup>13</sup>
- 14. Doshi-Velez and Kim (2017) emphasize the necessity of developing a rigorous scientific foundation for interpretable machine learning. They argue that explainability is essential for accountability, trust, and ethical deployment of AI, especially in high-stakes legal and medical domains. The authors propose frameworks for evaluating interpretability and highlight the limitations of black-box models. Their work significantly advances discussions on transparent AI, reinforcing the importance of explainable algorithms in ensuring meaningful oversight and fair decision-making.<sup>14</sup>
- 15. Eubanks (2018) provides a critical examination of how automated systems disproportionately target and disadvantage marginalized communities. Through real-world case studies, she demonstrates how algorithmic tools in welfare, policing, and public administration can reinforce inequality rather than reduce it. The book highlights systemic biases embedded in data and decision-making systems, raising concerns about fairness, transparency, and accountability. Eubanks's work is pivotal in understanding the social justice implications of AI and the need for ethical, equitable governance frameworks.<sup>15</sup>
- 16. Floridi and Taddeo (2018) explore the emerging field of data ethics, emphasizing the moral responsibilities associated with data collection, processing, and algorithmic decision-making. They argue that ethical challenges arise from issues such as privacy, data ownership, and the potential misuse of large-scale datasets. The authors propose foundational principles for ethical data governance. Their work is highly relevant in discussions of AI accountability, as responsible data practices form the basis for transparency, fairness, and

trustworthy artificial intelligence systems.<sup>16</sup>

- 17. Gasser and Almeida (2017) propose a layered model for AI governance that integrates technical, legal, ethical, and institutional dimensions. They argue that effective AI regulation must operate across multiple levels, from infrastructure and algorithms to societal norms and policy frameworks. The authors highlight the need for adaptive, flexible governance systems capable of responding to rapid technological change. Their work is significant for shaping holistic approaches to AI accountability, balancing innovation with safeguards that protect public interest. <sup>17</sup>
  18. Goodman and Flaxman (2017) analyze the European Union's regulatory approach to algorithmic decision.
- Union's regulatory approach to algorithmic decision-making, particularly focusing on the debated "right to explanation" under the GDPR. They highlight the challenges of implementing meaningful transparency in complex AI systems and discuss legal ambiguities surrounding automated decisions. The authors argue that stronger interpretability standards are necessary to ensure fairness and accountability. Their work provides critical insights into how regulatory frameworks can address the opacity of AI while protecting individual rights. <sup>18</sup>
- 19. Kuner (2020) examines the intersection of data protection law and artificial intelligence, arguing that traditional legal frameworks face significant limitations in regulating rapidly evolving technologies. He highlights issues related to algorithmic opacity, cross-border data flows, and the difficulty of enforcing compliance in decentralized systems. Kuner emphasizes the need for a more dynamic and adaptive regulatory approach. His work contributes to understanding how existing data protection laws struggle to address AI's complex accountability challenges. <sup>19</sup>
- 20. Mittelstadt et al. (2016) provide a comprehensive mapping of ethical concerns surrounding algorithmic decision-making. They identify key issues including bias, opacity, autonomy, responsibility, and fairness, emphasizing how algorithms can produce harmful or discriminatory outcomes. The authors argue that existing ethical frameworks are insufficient to address the complexity of modern AI systems. Their work is foundational in highlighting the multidimensional nature of algorithmic accountability and the need for stronger ethical and regulatory safeguards in AI deployment.<sup>20</sup>
- 21. Narayanan and Chen (2018) explore how the rise of artificial intelligence is reshaping the 21st-century legal system. They analyze challenges involving liability, transparency, due process, and the reliability of algorithmic decision-making. The authors argue that existing legal doctrines are inadequate for addressing harms caused by autonomous systems and emphasize the need for AI-specific legal reforms. Their work provides valuable insights into the structural adjustments required within judicial and regulatory institutions to ensure

<sup>&</sup>lt;sup>11</sup> Burrell (2016)

<sup>12</sup> Crawford and Paglen (2019)

<sup>&</sup>lt;sup>13</sup> Dani (2021)

<sup>&</sup>lt;sup>14</sup> Doshi-Velez and Kim (2017)

<sup>&</sup>lt;sup>15</sup> Eubanks (2018)

<sup>&</sup>lt;sup>16</sup> Floridi and Taddeo (2018)

<sup>17</sup> Gasser and Almeida (2017)

<sup>&</sup>lt;sup>18</sup> Goodman and Flaxman (2017)

<sup>&</sup>lt;sup>19</sup> Kuner (2020)

<sup>&</sup>lt;sup>20</sup> Mittelstadt et al. (2016)

accountability in an AI-driven legal landscape.21

- 22. Pasquale (2015) provides a seminal critique of opaque algorithmic systems that increasingly shape decisions in finance, healthcare, policing, and public administration. He argues that the "black box" nature of these systems concentrates power, reduces transparency, and undermines accountability. Pasquale highlights the dangers of unregulated algorithmic influence and calls for stronger oversight, transparency mandates, and legal safeguards. His work is foundational in understanding how secrecy in AI decision-making threatens fairness, due process, and democratic governance.<sup>22</sup>
- 23. Russell and Norvig (2016) present one of the most comprehensive foundational texts on artificial intelligence, covering algorithms, machine learning, robotics, and intelligent agents. Although primarily technical, the book also addresses ethical and societal implications of AI systems, including issues of autonomy, control, and decision-making reliability. Their work provides an essential background for understanding how AI systems function and why their complexity creates challenges for legal accountability, transparency, and regulatory oversight in high-stakes environments.<sup>23</sup>
- 24. Sandeen and Rowe (2020) explore emerging legal and ethical issues arising from the widespread adoption of artificial intelligence in business and governance. They discuss concerns related to data misuse, algorithmic bias, intellectual property, and liability for autonomous decision-making. The authors argue that traditional legal frameworks struggle to regulate AI effectively and call for updated policies that ensure fairness, transparency, and accountability. Their work provides important insights into how legal systems must evolve to address AI-driven challenges.<sup>24</sup>
- 25. Wachter, Mittelstadt, and Floridi (2017) emphasize the need for transparent, explainable, and accountable AI systems, particularly in robotics where autonomous actions can lead to significant ethical and legal consequences. They argue that explainability is crucial for ensuring fairness, preventing bias, and enabling regulatory oversight. The authors propose frameworks for developing accountable AI that align with human values and legal norms. Their work is influential in shaping discussions on responsible AI design and governance.<sup>25</sup>

# 1. RESEARCH METHODOLOGY

# 1. Research Design

This study adopts a **mixed-method research design**, combining **doctrinal analysis** with **empirical primary data**. The descriptive component examines existing legal frameworks, judicial interpretations, and scholarly debates on AI accountability, while the empirical component assesses public perception and stakeholder awareness through structured primary responses.

# 2. Nature of the Study

The research uses a qualitative-quantitative hybrid approach.

• Qualitative: Legal doctrines, case laws, ethical

- frameworks, and academic literature are analyzed.
- Quantitative: Statistical analysis of responses from selected participants is conducted to understand contemporary perceptions of AI and accountability.

#### 3. Sources of Data

#### a. Primary Data

Primary data was collected using a **structured Google Form questionnaire** designed to assess awareness, concerns, and opinions regarding AI-driven legal decision-making and accountability mechanisms.

- Sample Size: 125 respondents
- Target Group: Law students, legal practitioners, academicians, and technologically aware users
- **Sampling Technique:** Convenient and purposive sampling

#### b. Secondary Data

Secondary data was obtained from:

- Peer-reviewed journals, books, and law review articles
- Government reports and international policy documents (GDPR, OECD, EU AI Act)
- Judicial decisions and regulatory guidelines
- Digital databases: JSTOR, HeinOnline, LexisNexis, Google Scholar

#### 4. Data Collection Method

The **Google Form** consisted of close-ended, Likert scale, and multiple-choice questions focusing on:

- Knowledge of AI applications
- Perceived risks and benefits
- Views on algorithmic transparency
- Opinions on legal responsibility and liability
- Ethical and regulatory concerns

The form was distributed online via email, WhatsApp, and academic networks to ensure diverse participation.

# 5. Method of Data Analysis

- Quantitative Analysis: Responses were tabulated and analyzed using percentage analysis, frequency distribution, and graphical representation (pie charts, bar graphs). Basic statistical tools such as **chi-square tests** were used to examine associations and validate hypotheses.
- Qualitative Analysis: Doctrinal method was used to interpret legal standards, accountability gaps, and comparative international perspectives.

# 6. Scope of the Study

The study incorporates both legal theory and public perception, enabling a holistic understanding of accountability challenges in AI integration. While global policies are referenced, the primary focus remains on legal systems applicable to democratic nations, including India.

# 7. Ethical Considerations

 Participation in the Google Form was voluntary and anonymous.

<sup>&</sup>lt;sup>21</sup> Narayanan and Chen (2018)

<sup>&</sup>lt;sup>22</sup> Pasquale (2015)

<sup>&</sup>lt;sup>23</sup> Russell and Norvig (2016)

<sup>&</sup>lt;sup>24</sup> Sandeen and Rowe (2020)

<sup>&</sup>lt;sup>25</sup> Wachter, Mittelstadt, and Floridi (2017)

- No personal or sensitive information was collected.
- All secondary sources are cited appropriately to maintain academic integrity.

Here are the **Variables** for the study "Artificial Intelligence and the Future of Legal Accountability", aligned with your objectives and mixed-method research design:

#### Variables of the Study

# 1. Independent Variables (IVs)

These variables influence perceptions, accountability concerns, or legal implications of AI.

- 1. Level of Awareness about AI
- 2. Nature of AI System Used (Predictive, Autonomous, Analytical, Decision-Support)
- 3. Perceived Algorithmic Transparency
- 4. Data Quality and Bias in AI Models
- 5. Regulatory Knowledge of AI Laws
- 6. User Experience with AI-Based Tools
- 7. Technological Literacy of Respondents

### 2. Dependent Variables (DVs)

These reflect outcomes affected by independent variables.

- 1. Perceived Legal Accountability in AI Decisions
- 2. Trust in AI-Based Legal Systems
- 3. Acceptance of AI in Judicial/Legal Processes
- 4. Concern about AI Bias and Fairness
- 5. Support for New AI-Specific Liability Laws
- 6. Perceived Risk of Harm or Error in AI Outcomes

#### 3. Control Variables (CVs)

These variables remain constant or are used to maintain uniformity in analysis.

- 1. Geographical Location of Respondents
- 2. Mode of Data Collection (Google Form Only)
- 3. Structure of Questionnaire Items
- 4. Fixed Sample Size (125 respondents)

#### 4. Demographic Variables

These help in comparing responses across different groups.

- 1. Age Group
- 2. Gender
- 3. Educational Qualification
- 4. Profession (Student, Lawyer, Academic, Tech Professional, etc.)
- 5. Experience in Using AI Tools (Low/Medium/High)

# 5. Qualitative Variables

Essential for doctrinal and conceptual analysis.

- 1. Legal Interpretations of Accountability
- 2. Ethical Perspectives on AI Decision-Making
- 3. Perceived Fairness and Due Process

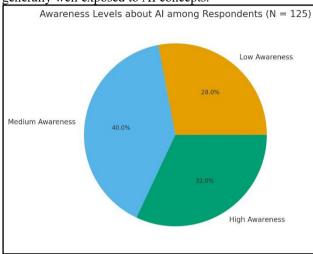
### DATA ANALYSIS

## 1.1 Awareness Level about AI

Awareness Level	Frequency	Percentage
Low Awareness	35	28.0%
Medium Awareness	50	40.0%

High Awareness	40	32.0%
Total	125	100%

Most respondents (40%) have **medium awareness** about AI, followed by **high awareness** (32%). Only 28% reported low awareness, indicating that the sample is generally well exposed to AI concepts.

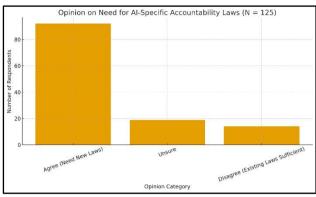


**Pie chart shown:** "Awareness Levels about AI among Respondents (N = 125)"

# 1.2 Opinion on Need for AI-Specific Accountability Laws

Opinion Category	Frequency	Percentage
Agree (Need New Laws)	92	73.6%
Unsure	19	15.2%
Disagree (Existing Laws Sufficient)	14	11.2%
Total	125	100%

A very high proportion (73.6%) **agree that new AI-specific accountability laws are required**, whereas only 11.2% feel existing laws are sufficient. 15.2% are undecided, showing some uncertainty but overall strong support for reform.



**Bar graph shown:** "Opinion on Need for AI-Specific Accountability Laws (N = 125)"

# 2. Chi-Square Test of Association

# 2.1 Objective Tested

Objective 1: To examine how Artificial Intelligence challenges traditional legal accountability frameworks. Operationalised as: Is there an association between AI awareness and opinion on the need for new AI-specific accountability laws?

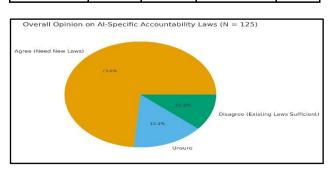
# 2.2 Hypotheses

 H<sub>0</sub> (Null Hypothesis): There is no significant association between level of awareness about AI and opinion on the need for new AI-specific accountability laws.

H<sub>1</sub> (Alternate Hypothesis): There is a significant association between level of awareness about AI and opinion on the need for new AI-specific accountability laws.

2.3 Observed Frequency Table (O)

Awareness \ Opinion	Agree (Need New Laws)	Unsure	Disagree (Existing Laws Sufficient)	Row Total
Low Awareness	20	8	7	35
Medium Awareness	38	7	5	50
High Awareness	34	4	2	40
Column Total	92	19	14	125



#### 2.4 Expected Frequencies (E)

(Expected = (Row Total × Column Total) / Grand Total) (rounded to two decimals)

Awareness \ Opinion	Agree	Unsure	Disagree
Low Awareness	25.76	5.32	3.92
Medium Awareness	36.80	7.60	5.60
High Awareness	29.44	6.08	4.48
Low Americas - Expected Distribution  Agree  Guile  Table  State  State  State	1525	Ages Chages State	Awareness - Expected Distribution

# 2.5 Chi-Square Calculation

$$\chi^2 = \sum \frac{(O-E)^2}{E}$$

Using all 9 cells:

$$\chi^2 \approx 7.9996$$

Degrees of freedom:

$$df = (r-1)(c-1) = (3-1)(3-1) = 4$$

Using the chi-square distribution:

• 
$$\chi^2$$
 (calculated)  $\approx$  8.00 df = 4 p-value  $\approx$  0.092

# 2.6 Interpretation

• At 5% significance level ( $\alpha = 0.05$ ), critical  $\chi^2$  for df = 4  $\approx$  9.49.

• Since 7.9996 < 9.49 and  $p \approx 0.092 > 0.05$ , the result is **not statistically significant** at 5% level. Therefore, we **fail to reject H<sub>0</sub>** at 5% significance level.

#### **Conclusion of Test:**

There is **no strong statistical evidence at the 5% level** to say that awareness level and opinion on the need for new AI-specific accountability laws are associated, although the pattern shows that respondents across all awareness levels largely support new laws (descriptively 73.6% agree).

At 10% significance level ( $\alpha = 0.10$ ) the association would be considered weakly significant (since  $p \approx 0.092$ ), indicating a marginal relationship between awareness and support for new legislation.

# Challenges

# 1. Lack of Clear Liability Frameworks

AI systems work on their own and typically make choices without any help from people. When something goes wrong, it's hard to hold developers, data trainers, deployers, or end users responsible. Current legal doctrines, like negligence and vicarious liability, inadequately include autonomous decision-making.

# 2. Algorithmic Opacity (Black Box Problem)

Deep learning systems and other advanced AI models work in ways that are hard to understand. Judges, lawyers, and regulators can't always figure out how an algorithm got to a certain conclusion, which makes it hard to hold people accountable and look at evidence.

#### 3. Bias and Discrimination

AI models that are trained on biased or incomplete datasets might lead to unfair results in sectors like recruiting, punishment, surveillance, or credit scoring. When bias emerges from more than one place, it might be hard to figure out who is responsible for discriminating effects.

# 4. Cross-Border Regulatory Conflicts

AI technology works across different legal systems, which can cause problems with data protection regulations, liability standards, and following the rules. Bringing global accountability standards into line with each other is still a big problem.

# 5. Absence of Robust AI-Specific Legislation

A lot of countries don't have complete AI laws. Current rules, which were made for people, don't work for autonomous systems, which creates gaps in responsibility.

## 6. Rapid Technological Evolution

AI is growing quicker than the law can keep up with it. New algorithms, use cases, and hazards come up by the time rules are put into place. This lag in regulation makes it harder to enforce and hold people accountable.

### 7. Ethical and Human Rights Concerns

Problems like privacy abuses, mass spying, and unfair stereotyping make it harder to hold people accountable. Under current legal frameworks, it is often unclear how to hold organizations accountable for ethical consequences.

#### 8. Data Ownership and Quality Issues

Liability issues come up when AI makes bad conclusions because it uses wrong, old, or illegal data. It is still not obvious who is responsible among data collectors, processors, and algorithm designers.

# Remedies Available

# 1. Development of Comprehensive AI-Specific Legislation

Governments need to make specific AI rules that make it clear who is responsible when autonomous systems are involved. These regulations can fill up the gaps in accountability by making clear the responsibilities of developers, deployers, and end-users, which will help keep people from misusing the technology.

# 2. Mandatory Algorithmic Transparency and Explainability

By making explainable AI (XAI) a legal obligation, judges and regulatory agencies can look into how an AI system came up with its results. To stop people from making decisions that aren't clear, you can undertake transparency audits, write down models, and create "explainability reports."

# 3. Implementation of Strict Data Governance Standards

Standardized frameworks for data quality, accuracy, and permission mitigate biased or erroneous AI outputs. Penalties for misusing personal data, obligatory data

audits, and stricter compliance with data protection legislation like GDPR or India's DPDP Act are some of the solutions.

#### 4. Establishment of AI Liability Insurance

Specialized insurance coverage for AI developers, organizations, and institutions can make sure that people who are hurt by AI are paid. This strategy is similar to product liability insurance and helps deal with AI outputs that are unforeseen or risky.

# 5. Creation of Regulatory Oversight Bodies

Independent AI regulators can check to see if rules are being followed, provide risk assessments, hand out punishments, and require fixes. These groups make sure that there is always oversight and fill in any holes in accountability.

### 6. Human-in-the-Loop (HITL) Mandates

Legal requirements for human oversight in high-stakes AI decisions like punishment, medical diagnosis, or surveillance act as preventative measures. HITL makes sure that a person is ultimately in charge.

#### 7. Standardized Ethical Guidelines and Certification

Certification systems for safe and ethical use of AI can help keep things safe. Before being put into use, systems may need to be tested, checked for fairness, audited for bias, and re-certified on a regular basis.

#### 8. Robust Grievance Redressal Mechanisms

Making complaint forums, digital grievance portals, and ombudsman systems easy to use ensures that people who are hurt by AI choices may get help quickly. This makes procedural justice stronger.

# 9. Judicial Review and Legal Recourse

Injunctions, punitive damages, compensatory orders, and corrective directions are among types of remedies that courts can give. Judicial monitoring is a strong way to stop the illegal or dangerous use of AI.

#### Conclusion

There is little doubt that artificial intelligence is changing the way modern judicial systems work. It is making them more efficient, accurate, and creative than ever before. But its quick integration also puts long-held ideas about accountability, responsibility, and ethical governance to the test. As AI systems increasingly affect judicial reasoning, administrative decision-making, and regulatory enforcement, conventional accountability frameworks are insufficient to mitigate harms resulting from opaque algorithms, autonomous decision routes, and data-driven biases.

This study underscores the pressing necessity for a reenvisioned legal framework capable of addressing the intricacies of AI. To avoid unfair results and keep people's trust in the legal system, it is important to create AI models that are open, understandable, and based on ethics. Also, developers, deployers, and end-users need to have clear roles and responsibilities, which can only be done with thorough laws, strong regulatory agencies, and structured accountability systems.

The results show that the future of legal responsibility depends on establishing a balance between new technology and basic ideas of justice and human rights. To make strong frameworks that protect people while still encouraging new ideas, we need to bring together law, technology, ethics, and policy. As AI keeps becoming better, legal institutions need to be just as flexible,

adaptable, and forward-thinking to make sure that accountability stays at the heart of technological advancement.

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