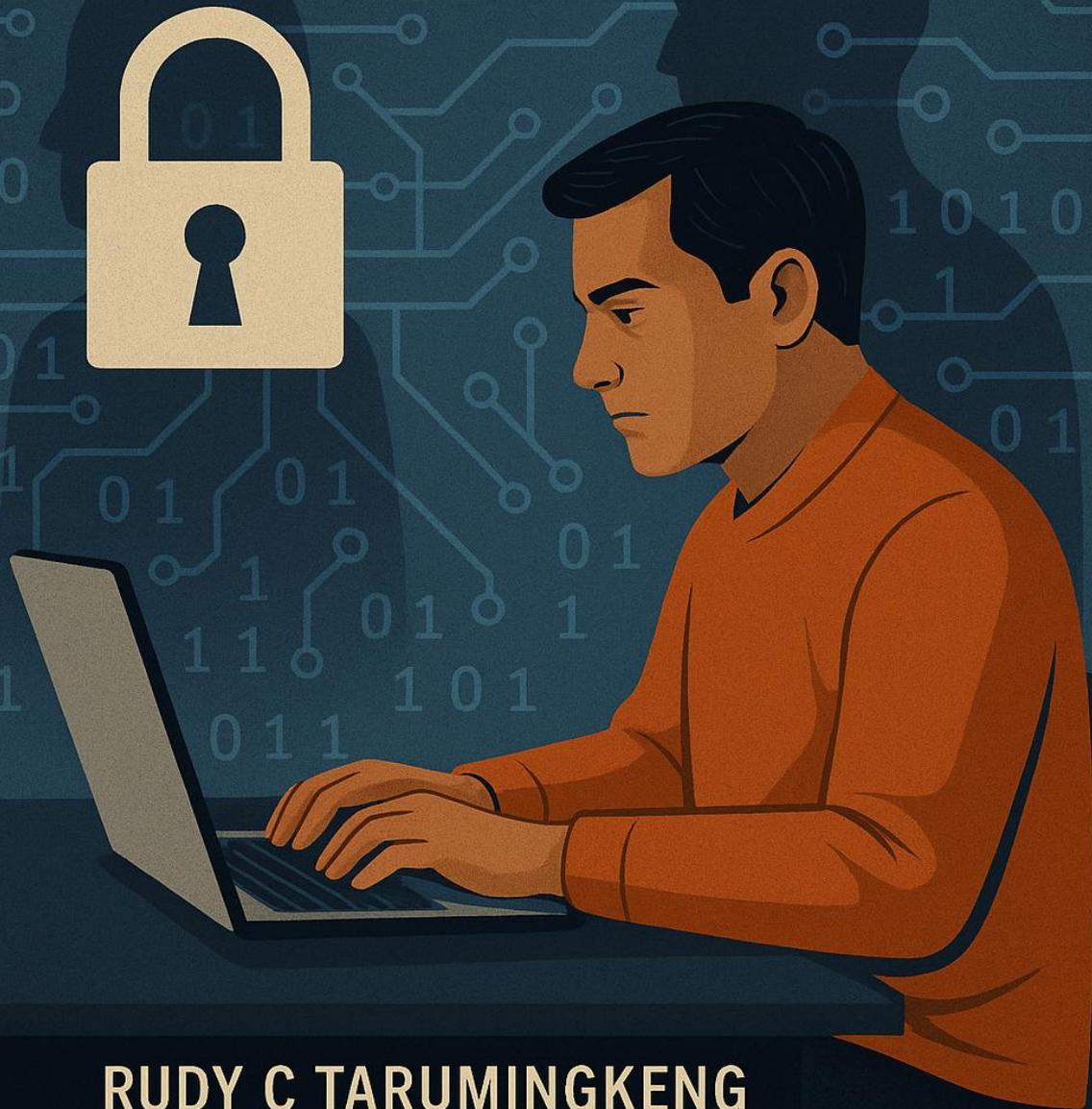


DATA ETHICS IN DATA SCIENCE PRIVACY AND SECURITY CHALLENGES



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Rudy C Tarumingkeng: **Data Ethics in Data Science - Privacy and Security Challenges**

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12 October 2025

DATA ETHICS IN DATA SCIENCE: PRIVACY AND SECURITY CHALLENGES

By Rudy C. Tarumingkeng

1. Introduction: The Moral Dimension of Data

The digital age has transformed data into the new currency of knowledge, power, and economic growth. Every click, transaction, and movement of an individual in cyberspace generates a traceable data footprint. Data science—an interdisciplinary field that combines statistics, computer science, and domain expertise—enables organizations to extract value from this vast information universe. However, with such immense power comes a proportional ethical responsibility. Data ethics emerges as a crucial discipline that guides how data is collected, analyzed, and used in ways that respect human dignity, protect privacy, and ensure social justice.

The tension between innovation and protection defines the core of data ethics. On one hand, data-driven insights improve health care, governance, and business efficiency. On the other, irresponsible use of data can lead to manipulation, surveillance, discrimination, and erosion of trust. Ethical dilemmas multiply as artificial intelligence (AI) systems become more autonomous, and as the boundaries between personal and public data blur.

This essay explores the complex interplay between **data ethics, privacy, and security**, drawing from philosophical traditions, legal frameworks, and practical case studies. It aims to demonstrate that the future of data science depends not merely on algorithmic sophistication but on ethical maturity.

2. Understanding Data Ethics: Conceptual Foundations

2.1 Definition and Scope

Data ethics refers to the moral principles and values that govern the acquisition, storage, analysis, sharing, and application of data. It extends traditional research ethics to the digital realm, where the subjects are often unaware that their data is being collected or processed. While professional codes such as those by ACM (Association for Computing Machinery) and IEEE provide formal guidelines, data ethics involves broader questions: What counts as informed consent in the age of big data? Who owns the data produced by an individual's digital life? How can algorithms be made accountable?

2.2 Philosophical Roots

The moral foundation of data ethics rests on three major traditions:

- **Deontological ethics (Kantian tradition):** Data practitioners have duties toward individuals whose data they handle—duties of respect, transparency, and honesty.
- **Consequentialism (Utilitarianism):** Data use should maximize overall well-being. For instance, sharing anonymized health data for epidemiological research can benefit society.
- **Virtue ethics:** Scientists and organizations must cultivate virtues such as integrity, humility, and responsibility when working with data.

These traditions intersect in modern frameworks such as *Responsible Data Science* and *Ethical AI*, which emphasize fairness, accountability, transparency, and explainability (FATE).

3. The Privacy Challenge

3.1 What is Data Privacy?

Privacy is the right of individuals to control how their personal information is collected, used, and shared. In data science, this right is

often compromised by large-scale data aggregation, behavioral tracking, and predictive modeling. The **Cambridge Analytica scandal** (2018), where millions of Facebook users' data were harvested for political manipulation, remains a defining moment in public consciousness regarding privacy loss.

3.2 Dimensions of Privacy in Data Science

1. **Informational privacy**: Protection of data from unauthorized access.
2. **Decisional privacy**: Freedom from external manipulation in personal choices.
3. **Psychological privacy**: Preservation of mental space and autonomy against algorithmic persuasion.

When algorithms infer emotions, habits, or political preferences, they intrude into the inner life of individuals, challenging the ethical limits of technological curiosity.

3.3 Informed Consent and Transparency

Traditional research ethics rely on informed consent—participants know what data are collected and why. However, in big-data ecosystems, data are often repurposed without explicit consent. Users rarely read privacy policies, and even when they do, these documents are too complex to ensure meaningful understanding. Therefore, **transparency** and **data literacy** become essential pillars. Data controllers should provide clear explanations of collection purposes, potential risks, and rights to deletion or correction.

3.4 Differential Privacy and Anonymization

From a technical perspective, privacy protection involves **de-identification** and **differential privacy**—mathematical methods that add noise to datasets to prevent re-identification of individuals. Yet, perfect anonymity is almost impossible. Combining multiple anonymized datasets often reveals identities. Thus, ethical practice demands a continuous balance between data utility and privacy assurance.

4. The Security Challenge

4.1 Defining Data Security

Data security refers to the protection of data from unauthorized access, alteration, or destruction. In the context of data science, where massive datasets are transferred across platforms and cloud infrastructures, breaches are not only technological but ethical failures. A breach of security can result in financial loss, identity theft, and emotional harm.

4.2 Key Principles of Data Security

1. **Confidentiality** – Ensuring that only authorized entities access the data.
2. **Integrity** – Preventing unauthorized modification or corruption.
3. **Availability** – Guaranteeing that data remain accessible to legitimate users when needed.

These principles are operationalized through encryption, authentication, and auditing protocols, but the ethical dimension lies in accountability: who takes responsibility when systems fail?

4.3 Case Studies of Ethical Breaches

- **Equifax Data Breach (2017)**: The exposure of personal information of 147 million people highlighted corporate negligence in cybersecurity.
- **Singapore's Health Data Leak (2018)**: Confidential medical records, including those of government leaders, were compromised—raising questions about public trust.
- **AI Model Inversion Attacks**: Researchers demonstrated that machine-learning models can be reverse-engineered to reveal sensitive training data, showing that even anonymized models can leak private information.

These examples underline that **data security is not only a technical issue but a moral obligation**—to prevent harm and maintain trust between citizens, governments, and corporations.

5. The Ethical Tension: Innovation versus Protection

5.1 The Paradox of Big Data

The promise of big data lies in its predictive power—health diagnoses, crime prevention, personalized learning. However, achieving this potential often requires extensive data sharing, which conflicts with privacy preservation. This creates an ethical paradox: progress demands openness, but openness threatens protection.

5.2 The "Privacy–Utility Trade-off"

Organizations often justify data collection by claiming benefits to consumers or society. Yet, without robust ethical guidelines, such justifications slide into *data paternalism*—assuming that users should surrender privacy for the greater good. A just ethical framework must minimize the privacy–utility trade-off through *privacy-by-design* approaches that integrate protection mechanisms from the outset.

5.3 Algorithmic Transparency and Accountability

Algorithms shape human decisions—from credit scoring to hiring to criminal sentencing. Their opacity (“black box problem”) challenges ethical governance. Algorithmic transparency requires explainability—users must understand how and why a decision was made. Accountability mechanisms include impact assessments, auditing systems, and oversight boards to ensure that algorithms comply with ethical norms.

6. Regulatory and Legal Frameworks

6.1 The Global Landscape

Several regulatory frameworks attempt to protect privacy and security in data science:

- **European Union’s GDPR (General Data Protection Regulation):** Sets strict rules on consent, data minimization, and the right to be forgotten.
- **California Consumer Privacy Act (CCPA):** Grants consumers control over personal data collection and sales.
- **Indonesia’s Personal Data Protection Law (PDP Law, 2022):** Aligns with global standards, emphasizing lawful processing, purpose limitation, and individual rights.

These laws provide a baseline but cannot replace ethical reasoning. Compliance is necessary but not sufficient for moral responsibility.

6.2 Corporate Data Ethics Policies

Leading companies such as Google, Microsoft, and IBM have developed internal data-ethics boards to oversee algorithmic impacts. However, critics argue that self-regulation often becomes a public-relations tool rather than a true ethical commitment. Genuine ethical governance requires independent oversight, transparency reports, and stakeholder participation.

7. Data Bias and Discrimination

7.1 The Hidden Bias in Data

Data reflect historical inequalities. When algorithms learn from biased datasets, they reproduce and even amplify social injustices. For example:

- Predictive policing tools disproportionately target minority communities.
- Facial-recognition systems perform poorly on darker skin tones.
- Credit algorithms penalize applicants from low-income areas.

7.2 Ethical Implications

Bias transforms technical problems into moral issues. It challenges the principle of justice—treating individuals fairly regardless of race,

gender, or background. Ethical data science demands **bias detection**, **algorithmic auditing**, and **diverse data representation** to prevent systemic harm.

8. Data Ownership and Control

8.1 Who Owns the Data?

Ownership is ambiguous in the digital ecosystem. Users generate data, but platforms monetize it. This asymmetry creates what Shoshana Zuboff calls *surveillance capitalism*—an economic system built on extracting behavioral data. The ethical question is whether users should receive compensation or at least control over how their data is commercialized.

8.2 Data Sovereignty and Localization

Nations increasingly assert sovereignty over their citizens' data. Data localization laws require that personal data be stored within national borders to protect against foreign surveillance. Yet, these laws also raise concerns about censorship and government overreach. Ethical governance must balance **national security**, **economic openness**, and **individual freedom**.

9. Emerging Frontiers: AI, IoT, and Genomic Data

9.1 Artificial Intelligence and Ethical Automation

AI systems analyze personal data to predict behavior. When decisions are automated—such as in recruitment or lending—the ethical stakes rise. AI ethics emphasizes *human-in-the-loop* models to maintain human oversight, prevent unfair discrimination, and ensure that machines serve human welfare, not replace moral judgment.

9.2 Internet of Things (IoT)

Connected devices continuously collect data from homes, cars, and wearables. While they enhance convenience, they also create continuous surveillance environments. Ethical design must include

explicit consent, secure communication protocols, and options for users to deactivate data collection.

9.3 Genomic and Health Data

Genetic data present unique ethical challenges: they reveal information not only about individuals but about families and ethnic groups. Unauthorized access or misuse can lead to stigmatization. Ethical management requires strict anonymization, secure biobanks, and participatory governance involving communities whose data are used.

10. Toward an Ethical Framework for Data Science

10.1 The Principles of FAIR and CARE

Ethical data management aligns with the **FAIR principles**—Findable, Accessible, Interoperable, and Reusable—and the **CARE principles**—Collective Benefit, Authority to Control, Responsibility, and Ethics—especially for indigenous data. Together, they balance openness with justice.

10.2 The “Three-Layer” Ethical Model

1. **Technical layer:** Data minimization, encryption, and anonymization.
2. **Organizational layer:** Governance structures, data stewardship roles, and ethical audits.
3. **Societal layer:** Public engagement, education, and dialogue on digital rights.

Ethical data science must operate across these layers to sustain trust.

10.3 Responsible Innovation

The concept of *Responsible Research and Innovation (RRI)* promotes anticipation, reflection, inclusion, and responsiveness. Data scientists should anticipate potential harms, reflect on values guiding their work, include diverse perspectives, and respond to public concerns.

11. Ethical Leadership and Culture in Data Organizations

Ethical data governance requires not only policies but cultures of integrity. Leadership must articulate ethical visions, provide training, and reward ethical behavior. Ethical data culture resembles *safety culture* in engineering: everyone shares responsibility for preventing harm.

Examples of good practice include:

- Ethical impact assessments before deploying new models.
 - Regular transparency reports on data use.
 - Cross-disciplinary ethics committees involving legal, social, and technical experts.
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12. Educational Imperatives: Teaching Data Ethics

Data-science curricula must include ethics as a core component, not an elective. Students should learn to identify bias, reason morally about trade-offs, and engage with real-world dilemmas. Case-based learning—analyzing Facebook’s or Google’s ethical failures—cultivates critical reflection. Universities must act as laboratories of ethical innovation, training professionals who understand both the power and the limits of data.

13. Cross-Cultural Perspectives

Ethical values differ across societies. Western frameworks emphasize individual autonomy and consent, while Asian traditions highlight communal harmony and collective responsibility. In Indonesia, *Pancasila* provides a moral compass that values humanity, social justice, and unity—principles that can enrich global discussions on data ethics. Integrating cultural wisdom into data-ethics discourse ensures inclusivity and legitimacy.

14. Ethical Risks of Emerging Technologies

14.1 Deep Learning and Surveillance

Facial-recognition technologies can enhance security but also enable authoritarian surveillance. Ethical regulation should prohibit indiscriminate biometric monitoring and ensure proportional use aligned with human rights.

14.2 Synthetic Data and Deepfakes

Synthetic data help protect privacy but can also fabricate false realities. Deepfakes threaten truth and trust in media. Ethical responses include authenticity verification technologies and legal deterrents against malicious creation of fake content.

14.3 Quantum Computing and Data Security

Quantum computing may eventually break classical encryption, creating new vulnerabilities. Ethical foresight demands investment in post-quantum cryptography and international collaboration to prevent cyber imbalance.

15. The Role of International Cooperation

Data flow transcends national borders; hence, ethics must be global. Initiatives such as the OECD Principles on AI, UNESCO's *Recommendation on the Ethics of Artificial Intelligence* (2021), and ASEAN's digital-governance frameworks aim to harmonize standards. Global data ethics must prevent a digital divide where powerful nations exploit data from developing countries without reciprocity.

16. Reflections: From Data to Humanity

The deeper ethical question is not only how data are used but what kind of society data create. A surveillance society normalizes control, while an ethical data society nurtures trust and participation. Data should serve *human flourishing*—the realization of human potential in freedom and dignity.

Ethical data science thus moves beyond compliance toward *moral imagination*: the capacity to foresee unintended consequences and act with compassion. As technology evolves faster than regulation, ethics becomes the compass guiding innovation toward humane ends.

17. Recommendations for Ethical Practice

1. **Adopt Privacy-by-Design:** Embed privacy protections in system architecture from the beginning.
 2. **Perform Data Impact Assessments:** Evaluate risks before data collection and model deployment.
 3. **Ensure Algorithmic Explainability:** Make models interpretable to stakeholders.
 4. **Promote Diversity in Data Teams:** Prevent bias through multidisciplinary collaboration.
 5. **Foster Public Dialogue:** Engage citizens in decisions about data governance.
 6. **Implement Continuous Ethical Audits:** Regularly review compliance with evolving norms.
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18. Conclusion: Building a Just Data Future

Data ethics stands at the crossroads of science, morality, and humanity. As data become the infrastructure of modern life, ethical responsibility becomes a civic duty. Privacy and security are not technical afterthoughts but foundations of digital trust. Without them, even the most advanced algorithms lose legitimacy.

A just data future requires three virtues: **transparency**, **accountability**, and **empathy**. Transparency ensures openness; accountability ensures responsibility; empathy ensures humanity. When data scientists act with these virtues, they transform data science from a mere analytical discipline into a moral endeavor.

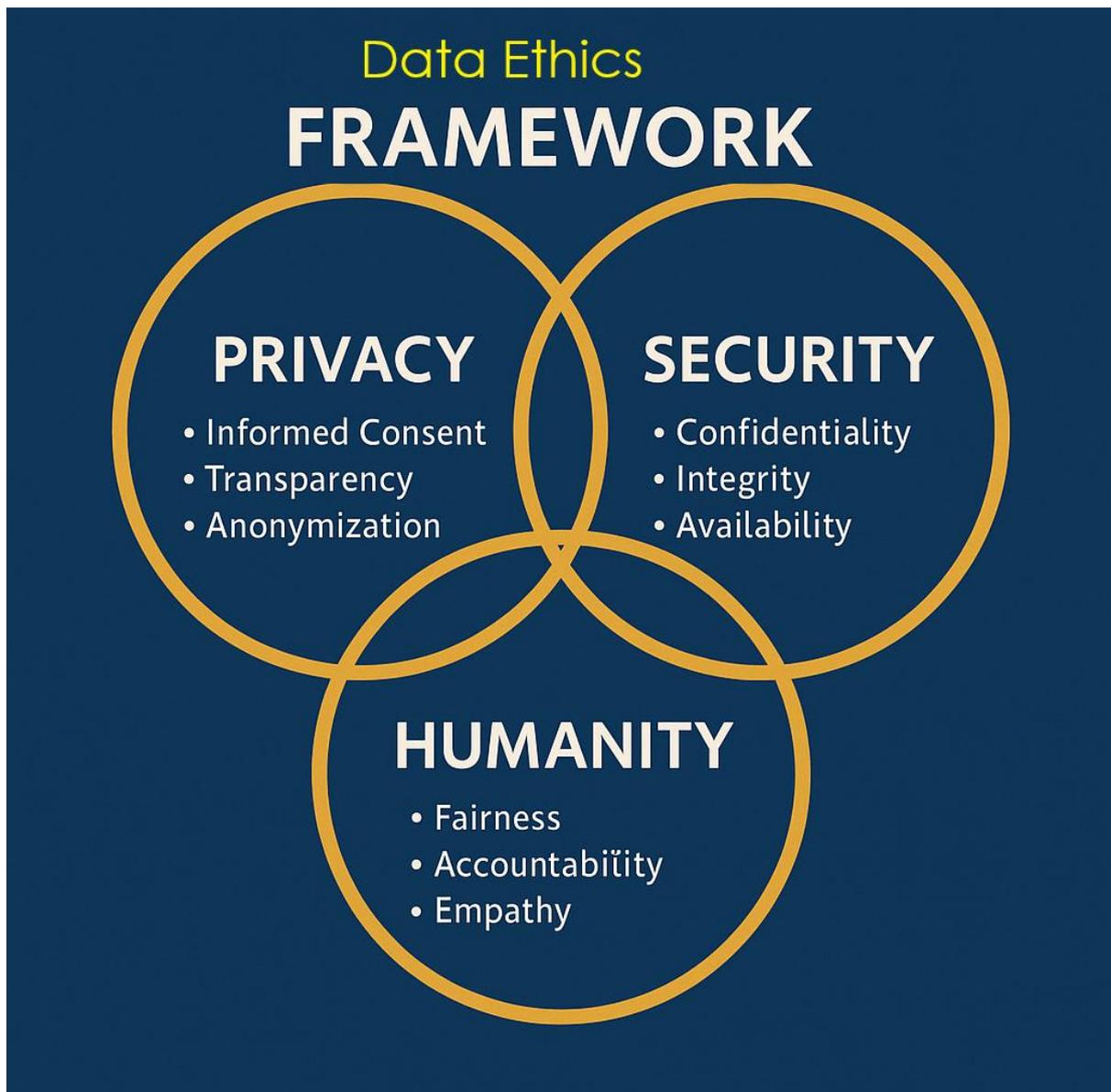
The ultimate goal of data ethics is not to limit innovation but to **humanize it**—to ensure that progress in data science aligns with the enduring values of justice, respect, and the common good. Only by embedding ethics into every algorithm and database can humanity build a digital civilization that protects both knowledge and conscience.

Glossary

- **Data Ethics:** The moral principles guiding data handling and analysis.
 - **Differential Privacy:** A mathematical method for preserving privacy by adding noise to data.
 - **Algorithmic Bias:** Systematic and unfair discrimination embedded in machine-learning models.
 - **Data Sovereignty:** The concept that data are subject to the laws of the nation where collected.
 - **Privacy-by-Design:** Integrating privacy principles into the architecture of technology systems.
 - **Surveillance Capitalism:** Economic system where personal data are commodified for profit.
 - **FAIR & CARE Principles:** Frameworks promoting responsible data stewardship.
 - **Deepfake:** Artificially generated video or audio mimicking real persons.
 - **Post-Quantum Cryptography:** Security techniques resistant to quantum attacks.
 - **Ethical AI:** AI designed to be fair, transparent, and accountable.
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This section deepens the philosophical, socio-political, and managerial discourse behind data ethics and connects theory to real-world implications.

Reflection and Further Discussions

Extending the Ethical Conversation on Data, Privacy, and Security

1. The Moral Paradox of Progress

Data science is often portrayed as a triumph of human ingenuity—our ability to convert complexity into comprehension. Yet every technological leap simultaneously exposes our moral fragility. The more we can know, the more we must decide what *not* to know. This paradox defines the ethical landscape of data science: the collision between **epistemic ambition** (the desire to know everything) and **moral restraint** (the duty to protect human dignity).

Historically, the Enlightenment promised liberation through knowledge. Today, the “data enlightenment” risks turning knowledge into surveillance. The same data analytics that predict disease outbreaks can also predict consumer weakness, political tendencies, or even emotional vulnerability. Hence, the question is no longer whether we can analyze data, but whether we *should*—and under what moral constraints.

This moral tension calls for an ethical framework that tempers curiosity with compassion, and innovation with humility. The goal is to transform the data revolution into a **moral renaissance**, where technological intelligence serves, rather than subverts, the human condition.

2. From Ethics to Trust: The Social Capital of Data

Ethics in data science is not an abstract philosophy—it is a prerequisite for **trust**, the social capital of the digital age. Without trust, data sharing collapses, citizens resist digital governance, and companies lose legitimacy.

Trust emerges from three pillars:

1. **Competence**: Organizations must demonstrate that they can safeguard data technically.
2. **Integrity**: They must act transparently and uphold commitments.
3. **Benevolence**: They must use data for purposes aligned with social good.

When these pillars are weak, scandals erupt—as seen in Cambridge Analytica, Equifax, and numerous health-data leaks. Trust, once broken, is almost irreparable. Rebuilding it requires not only technological reinforcement but also **ethical transparency**, where users are treated not as data points but as moral subjects.

Governments, too, rely on trust to implement digital policies. Smart cities, electronic IDs, and predictive policing depend on citizens' willingness to share personal data. If data ethics is ignored, public resistance can derail innovation. Therefore, **ethical legitimacy** is as crucial as legal compliance.

3. The Political Economy of Data

Ethics cannot be divorced from economics. Data have become a commodity—a resource mined, processed, and sold in global markets. The political economy of data reveals power asymmetries: Big Tech firms possess disproportionate control over the flow of information, shaping not only markets but also minds.

3.1 Surveillance Capitalism Revisited

Shoshana Zuboff's notion of *surveillance capitalism* describes a new economic logic in which human experience becomes raw material for behavioral prediction and monetization. Every online activity—scrolling, clicking, pausing—is harvested for predictive analytics. This system erodes the boundary between consumer and product: users become the *source* of value rather than its beneficiary.

The ethical question is whether such extraction constitutes a form of exploitation. While users enjoy “free” services, they pay with their autonomy. The moral economy of data thus requires a shift from **data extraction** to **data stewardship**, where individuals retain agency and benefit from their informational contributions.

3.2 Data Colonialism

A related concept, *data colonialism*, critiques how global corporations extract data from developing countries without equitable return. Digital infrastructures built by foreign companies often transfer local data abroad, reinforcing dependency. Indonesia and other ASEAN nations face this challenge as they strive for digital sovereignty. Ethical global data governance must recognize information as a shared resource, not a tool of domination.

4. The Ontology of Data: Are We Reducing Humans to Numbers?

Philosophically, the ethics of data science also involves the question of **human ontology**—what it means to be a person in a quantified world. Datafication reduces experience to measurable variables: emotions become sentiment scores; relationships become network nodes; identity becomes metadata. This quantification, though powerful, risks obscuring the qualitative dimensions of human life—love, pain, spirituality, and freedom.

Hannah Arendt warned that when humans are viewed merely as data objects, political tyranny becomes easier. The “banality of evil,” she argued, arises from treating persons as numbers to be managed. Similarly, algorithmic governance can unintentionally reproduce this dehumanization when decision-making is outsourced to code.

Thus, ethical data science must re-center the **human subject**. Data should describe people, not define them. Quantitative insight must be balanced by qualitative understanding, ensuring that the narrative of humanity is not lost in the statistics of efficiency.

5. Ethical Design: Embedding Values into Technology

5.1 From Code to Conscience

Ethical data design begins at the level of architecture—how systems are conceived, coded, and deployed. The principle of **privacy-by-design** transforms ethics from an afterthought into a foundation. Similarly, **security-by-design** ensures resilience against cyber threats.

The concept of **value-sensitive design (VSD)** goes further, proposing that moral values such as fairness and autonomy be built into the very logic of algorithms. For example:

- Recommender systems can be tuned to diversify content exposure rather than amplify bias.
- Predictive models can be calibrated to avoid historical discrimination.
- Data dashboards can visualize uncertainty rather than conceal it.

Such design choices reflect moral agency. Every line of code becomes a moral act, shaping the kind of society we inhabit.

5.2 Ethical Usability and Human Experience

Ethical design also concerns user experience. Consent forms, for instance, should be understandable—not buried in legal jargon. Interfaces can empower users with real-time control over their data, fostering autonomy. Ethical usability aligns functionality with dignity: technology should respect users’ cognitive limits and moral rights simultaneously.

6. Cultural Dimensions of Data Ethics

Ethical principles are universal in aspiration but contextual in application. In Western societies, individual consent is central. In Asian and African traditions, community well-being often outweighs individual autonomy. A pluralistic ethics of data must integrate both perspectives.

In Indonesia, the philosophy of **Pancasila**—anchored in divinity, humanity, unity, democracy, and justice—offers a normative compass. It invites a *communitarian* view of data ethics, where privacy and security are pursued not in isolation but as expressions of mutual respect (*gotong royong*). For instance, collective consent models can allow communities to decide how indigenous data or cultural heritage are used.

Global frameworks should thus respect cultural diversity while maintaining universal principles of justice and human rights. This **intercultural dialogue** can prevent the imposition of one-size-fits-all ethics from technologically dominant regions.

7. The Future of Privacy: From Ownership to Stewardship

Traditional notions of privacy treat data as personal property—something to own or sell. However, as data become relational and networked, ownership becomes inadequate. My health data, for example, reveal information about my family; my social media posts affect others’ reputations.

A more nuanced model is **data stewardship**, where entities manage data on behalf of individuals under strict fiduciary duties. Like doctors or lawyers, data stewards must act in the best interest of data subjects. This model combines individual rights with collective responsibility, ensuring ethical governance beyond market logic.

Emerging ideas such as *data trusts*, *data cooperatives*, and *digital commons* illustrate this paradigm shift. In the future, users may collectively negotiate with companies about how their data are used—transforming passive consumers into active data citizens.

8. Security as a Moral Responsibility

While privacy protects autonomy, security protects existence. A data breach is not merely a technical glitch—it can destroy lives through identity theft, financial loss, or reputational harm. Thus, data security is a matter of **justice and care**.

Ethical responsibility in security extends beyond compliance checklists. It demands:

- **Proactive vigilance:** anticipating threats before they occur.
- **Transparency after breaches:** admitting faults and compensating victims.
- **Global solidarity:** sharing cybersecurity knowledge across nations and sectors.

The rise of ransomware attacks and state-sponsored cyberwarfare underscores the moral urgency of digital peacekeeping. Ethical leadership in cybersecurity means balancing national defense with civil liberty—protecting citizens without turning societies into digital fortresses.

9. Education and Ethical Literacy

The next generation of data scientists must be fluent not only in Python and statistics but also in **moral reasoning**. Universities should integrate ethics into every stage of data-science education, from data collection to algorithmic interpretation. Case-based pedagogy—examining dilemmas such as biased facial recognition or health-data monetization—encourages reflective judgment.

Moreover, ethical literacy should extend to policymakers, journalists, and citizens. Society as a whole must develop a **civic understanding of data**—what it reveals, what it conceals, and what it costs. Only through collective awareness can ethical norms gain social traction.

In Indonesia and Southeast Asia, ethical literacy programs can be embedded in digital-transformation roadmaps, ensuring that inclusivity and morality accompany technological growth.

10. AI Ethics as the Next Frontier

As AI systems increasingly interpret data autonomously, they raise new questions: Can machines be moral agents? Who is responsible when an algorithm causes harm—the coder, the company, or the code itself?

Current consensus holds that moral agency remains human; algorithms lack consciousness and intention. Yet ethical responsibility can be *distributed*: programmers must design responsibly, organizations must govern transparently, and users must act critically. This networked accountability forms the backbone of **AI ethics**.

AI amplifies both potential and peril. It can democratize knowledge or deepen inequality. Ethical governance must ensure that AI serves human values, not market efficiency alone. This includes designing *explainable AI*, auditing datasets for bias, and establishing oversight boards that include ethicists and affected communities.

11. Ethical Governance in the Age of Data Abundance

Governance mechanisms must evolve from reactive regulation to **anticipatory ethics**—policies that foresee dilemmas before they materialize. Key strategies include:

- **Ethical Impact Assessments (EIA):** Analogous to environmental assessments, these evaluate the societal risks of data projects.
- **Data Ethics Boards:** Independent bodies that review algorithms and ensure alignment with public values.
- **Transparency Reports:** Regular disclosures on data collection and algorithmic decision-making.
- **Multi-stakeholder Dialogues:** Involving academia, government, industry, and civil society.

These measures turn ethical aspiration into institutional practice. They embody what philosopher Hans Jonas called the *imperative of responsibility*—acting so that the consequences of technology remain compatible with the permanence of human life.

12. The Role of Religion and Moral Philosophy

Ethical reflection cannot ignore the spiritual dimensions of data life. Religious traditions offer enduring insights into the stewardship of knowledge. In Christian ethics, the principle of *imago Dei* (humans created in the image of God) affirms the sanctity of personal identity and autonomy—values violated when data are exploited. In Islamic ethics, the concept of *amanah* (trust) emphasizes accountability for what one holds, including information. In Eastern philosophies, the pursuit of harmony and balance parallels the ethical equilibrium between privacy, security, and collective good.

These traditions remind us that ethics is not merely about rules but about **character**—the formation of virtuous agents who

handle data with wisdom. As technology evolves faster than law, moral character becomes the most reliable firewall.

13. Human Dignity as the Ultimate Principle

At the heart of data ethics lies **human dignity**. This concept transcends culture and religion: every person deserves respect, privacy, and protection from exploitation. When algorithms predict behavior, they must not pre-judge worth; when data are analyzed, they must not erase individuality.

Human dignity requires:

- **Right to explanation** – understanding algorithmic decisions.
- **Right to consent** – deciding how personal data are used.
- **Right to be forgotten** – reclaiming autonomy over digital traces.

Upholding dignity transforms data science from a discipline of prediction into a vocation of service. It restores the moral meaning of knowledge: to know is to care.

14. The Ethics of Uncertainty

Ethics often deals with certainty—clear rules, clear outcomes. Yet data science thrives in **uncertainty**. Algorithms operate probabilistically; predictions are never absolute. Ethical humility demands acknowledging this uncertainty. Overconfidence in data can lead to technocratic arrogance—believing that numbers can replace judgment.

Transparency about uncertainty—displaying confidence intervals, error margins, and limitations—enhances rather than diminishes trust. It communicates honesty and invites dialogue. In this sense, uncertainty is not a weakness but a virtue of ethical science.

15. Reflections on the Future: From Data to Wisdom

The data revolution represents a new phase of human evolution—what some call the *infosphere*. Yet the ultimate question remains: will humanity ascend from data to wisdom?

The **DIKW hierarchy** (Data → Information → Knowledge → Wisdom) reminds us that data alone are meaningless without context, interpretation, and ethical reflection. The addition of *Humanity* to this continuum completes it:

Data → Insight → Intelligence → Wisdom → Humanity.

True progress is measured not by how much data we collect, but by how wisely we use it. Wisdom entails restraint, empathy, and foresight—the capacity to align information with moral purpose.

As data pervade every sphere of life, our ethical evolution must keep pace. Just as the Industrial Revolution demanded labor rights, the Data Revolution demands **digital rights**—the moral infrastructure of the information age.

16. Concluding Discussion: Toward an Ethical Civilization of Data

The ethics of data privacy and security is not a peripheral concern—it is the foundation of civilization in the digital century. The way societies handle data will determine the texture of human freedom in the decades to come.

We stand at a crossroads:

- One path leads to *data totalitarianism*, where surveillance replaces citizenship, and humans become predictable cogs in algorithmic systems.
- The other path leads to *data humanism*, where information serves enlightenment, solidarity, and flourishing.

The choice is moral, not technical. It requires collective wisdom, ethical education, and compassionate governance.

In the Indonesian context, this vision resonates with the spirit of **Pancasila Humanism**: advancing technology while safeguarding humanity. Ethical data science becomes part of the nation’s moral architecture—integrating knowledge, integrity, and care.

If the twentieth century was the age of industrial power, the twenty-first must be the age of **ethical intelligence**. Data, in the end, are not just facts—they are fragments of human stories. To treat them ethically is to honor the people behind them.

 **Visual Reflection Suggestion**

A complementary infographic (which can follow your previous blue–gold theme) could be titled:

“From Data to Humanity: The Ethical Evolution of Knowledge.”

It would illustrate a vertical flow:

Data → Information → Knowledge → Wisdom → Ethics → Humanity

Each level labeled with key virtues: *Accuracy, Context, Understanding, Reflection, Responsibility, Compassion.*



Rudy C Tarumingkeng: **Data Ethics in Data Science - Privacy and Security Challenges**

Copilot for this article - Chatgpt 5, Access date: 12 October 2025.
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