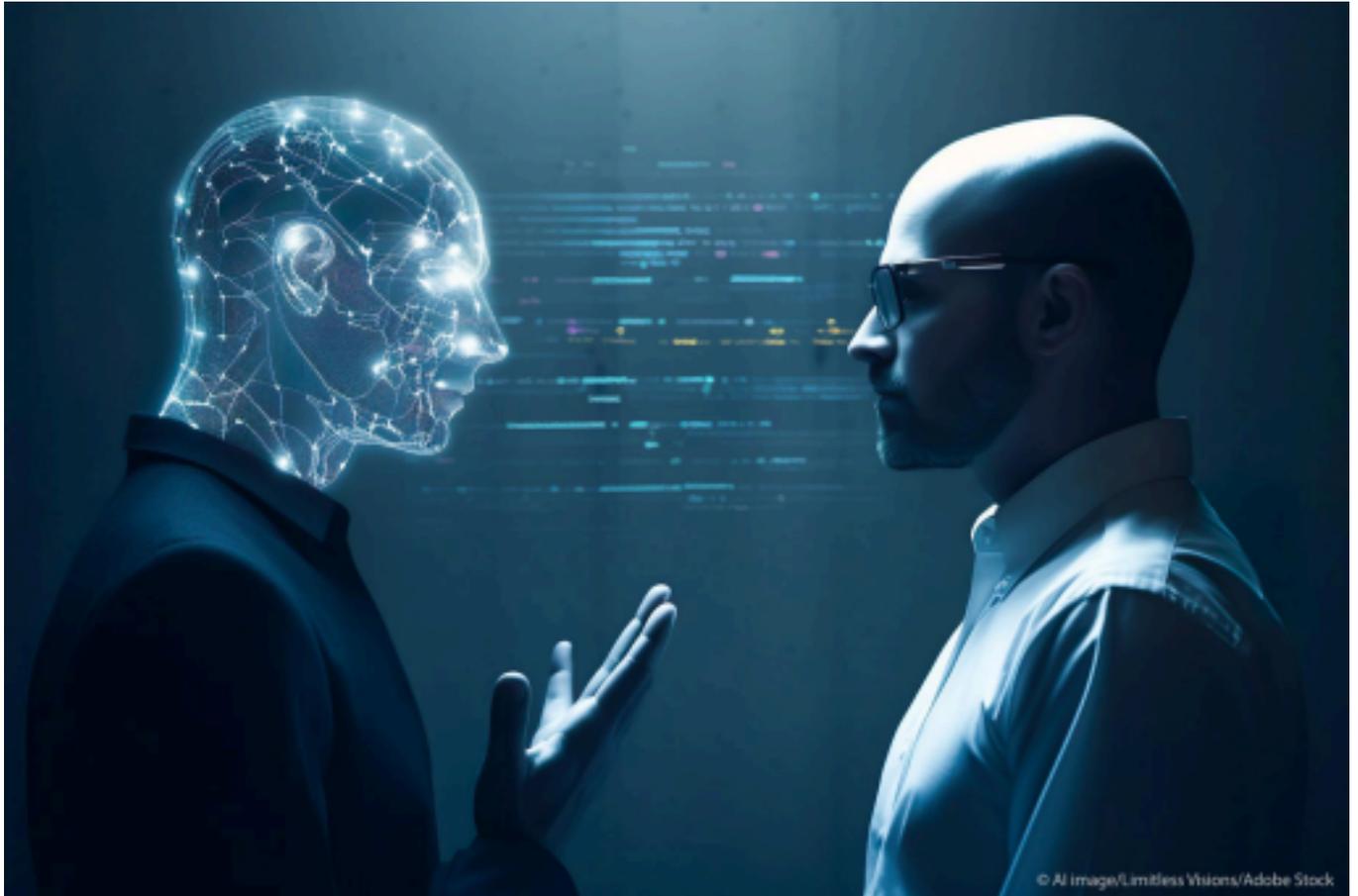


Algorithmic Bias and AI Ethics: Balancing Innovation with Social Justice



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Abstract

This report examines the ethical challenges raised by artificial intelligence (AI) in data science, particularly how biased algorithms can lead to discrimination in high-stakes areas like hiring, criminal justice, and finance. It explores the dangers of treating AI as neutral; however, in reality, it often reflects and amplifies human and historical biases. Further, the report includes research into algorithmic bias through detailed case studies, as well as creative explorations through rhetorical poetry and dialogue that together provide emotional and personal perspectives on algorithmic injustice and highlight the human consequences of biased AI decision-making. By blending academic analysis with poetic and narrative lenses, the reader gains a deeper understanding of how ethical flaws in AI impact real lives and amplify historical inequalities while being perceived as objective. Overall, it argues that while AI has the power to transform society, it must be developed and used with transparency, fairness, and human oversight to prevent reinforcing systemic inequalities.

Introduction

Mechanical eyes scanning for resumes, the AI hiring tool instantly rejects a candidate, not for lack of skill, but because their name “sounds female.” Imagine a world where algorithms decide who gets a job, who qualifies for a mortgage, and even who is more likely to commit a crime, all without human intervention. Now, imagine that these decisions are filled with bias that disproportionately harms marginalized communities. These are not dystopian nightmares. These are real consequences of artificial intelligence (AI) powered decision-making in data science today. AI systems are being integrated and utilized rapidly in high-stakes areas such as criminal justice, hiring, and finance. Yet, beneath their promise lie serious ethical issues. While AI has revolutionized data science by increasing efficiency and objectivity, it has created ethical

vulnerabilities and poses a significant risk of perpetuating and amplifying societal bias.

Consequently, their reliance on biased and flawed training data and lack of accountability have produced discriminatory outcomes in key decision-making areas. Rather than eliminating bias, AI often amplifies and reinforces historical inequalities while being seen as neutral and objective. Cases like the COMPAS recidivism algorithm, Amazon's hiring tool, and biased mortgage lending reveal how algorithmic bias reinforces existing inequalities (Larson et al. sec. 2). To ensure that AI serves as a tool for equity rather than oppression, ethical frameworks must prioritize transparency through explainable AI, diverse and representative datasets, and human oversight ("Data Ethics in AI: 6 Key Principles for Responsible Machine Learning" par. 5). Achieving this balance requires solving the tension between technological innovation and ethical responsibility while addressing the limitations of current governance structures in keeping pace with AI's rapid evolution before its invisible roots deepen the very inequalities it was meant to solve. Ensuring fairness in AI decision-making is critical for creating responsible and trustworthy technology.

Data Science and AI: Transforming the Digital World

Moreover, the rise of data science and artificial intelligence has profoundly reshaped industries, driving innovation, efficiency, and automation. In today's digital world, businesses and governments rely on AI to make decisions, predict trends, and optimize resources and actions. AI is capable of processing vast amounts of data faster and more accurately than humans, making it an essential tool across various sectors. As Washington State University notes, "It would appear that we are now entering the fourth industrial revolution, which is driven by AI, robotics, quantum computing, and other innovations. This Fourth Industrial Revolution promises to disrupt our world once again" ("The Fourth Industrial Revolution" par. 1). However,

with this rapid advancement comes a growing concern about ethical implications, such as data privacy, algorithmic bias, and accountability (IBM par. 4). As IBM states, “AI ethics is a multidisciplinary field that studies how to optimize the beneficial impact of artificial intelligence while reducing risks and adverse outcomes” (IBM par. 5). These concerns highlight the importance of understanding how AI is transforming industries and why ethical considerations must evolve alongside technological advancements.

Furthermore, one of the most significant ways data science and AI have revolutionized industries is through their ability to analyze and extract meaningful insights from massive datasets. According to UC Berkeley’s School of Information, “The ability to take data—to be able to understand it, to process it, to extract value from it, to visualize it, to communicate it—that’s going to be a hugely important skill in the next decades” (Varian par. 4). Companies utilize AI to improve customer experiences, streamline operations, and make data-driven decisions. For example, AI-powered recommendation systems personalize the content on platforms such as Netflix and Amazon, whereas financial companies use machine learning to detect fraudulent transactions in real time (IBM par. 5). The growing reliance on AI demonstrates its potential to reshape industries as the fourth revolution, but it also raises ethical concerns about transparency and fairness in decision-making.

Indeed, AI’s impact is evident in critical fields such as healthcare, finance, and education. In healthcare, AI-driven predictive models assist in diagnosing diseases, personalizing treatments, and even forecasting disease outcomes (IBM par. 7). IBM showcases an example of AI’s role in medicine, where a platform analyzes medical records to categorize patients based on their risk of stroke and predicts the success of treatment plans (par. 8). In finance, AI enables algorithmic trading, fraud detection, and risk assessment, improving both security and efficiency

(UChicago par. 2). Meanwhile, in education, AI-driven tools personalize learning experiences, helping students grasp concepts at their own pace (IBM par. 6). While these advancements offer immense benefits and efficiency, they also introduce ethical issues. Biased training data can lead to unfair medical diagnoses, financial discrimination, or inequalities in educational resources, which requires the need for ethical oversight in AI development.

Despite AI's transformative potential, its unchecked growth raises concerns that cannot be ignored. The increasing use of AI in decision-making demands greater transparency and accountability from companies and policymakers. As IBM notes, "Lack of diligence in this area can result in reputational, regulatory, and legal exposure, resulting in costly penalties" (par. 10). While AI continues to revolutionize industries, addressing its ethical challenges is crucial to ensuring it remains a force for good rather than a tool for exploitation. The balance between innovation and ethical responsibility will define the future of AI in our digital world (Rahwan et al. p. 485).

The Ethical Challenges of AI and Data Science

Significantly, the increasing use of AI in data science has exposed serious ethical concerns, particularly regarding privacy, surveillance, and bias in decision-making. AI systems often rely on vast amounts of personal data, raising issues about how that data is collected, stored, and used (IABAC® sec. 5). While companies claim AI enhances objectivity, the reality is that many models operate as "black boxes," making it difficult to understand how decisions are made (IABAC® sec. 6). The lack of transparency in AI decision-making processes means that users often do not know why they were denied a loan, flagged for fraud, or rejected from a job. This secrecy weakens public trust and raises concerns about accountability, especially when AI is used in high-stakes areas such as law enforcement and healthcare. As Nick Bilton, a tech columnist

from *The New York Times*, warns, “Imagine how a medical robot, originally programmed to rid cancer, could conclude that the best way to obliterate cancer is to exterminate humans who are genetically prone to the disease” (Gesikowski par. 5). Similarly, without clear guidelines on data

ethics and accountability, AI has the potential to reinforce systemic biases and violate fundamental rights, making it crucial to address these concerns now (IBM par. 13).

Additionally, one of the most pressing ethical issues in AI is its impact on privacy and surveillance. AI-driven data collection has become widespread, with companies and governments using AI to track user behavior, monitor online activity, and even predict future actions. This level of surveillance raises significant privacy concerns, especially when sensitive data such as health records, financial information, or biometric data is involved (IABAC® sec. 9). In many cases, AI collects data without explicit consent or repurposes existing data for new applications without informing users (IABAC® sec. 10). For instance, a healthcare AI system initially used for diagnosing illnesses may later expand to mental health predictions, but

without

new consent from patients. This expansion becomes an ethical violation (“Data Ethics in AI: 6 Key Principles for Responsible Machine Learning” par. 7). The European Union’s General Data Protection Regulation (GDPR) has attempted to address some of these issues by enforcing strict data protection laws, but many countries lack similar regulations, leaving users vulnerable to exploitation (IABAC® sec. 11).

Beyond privacy, AI also plays a significant role in spreading misinformation and amplifying biases. Many AI algorithms, particularly those used in content recommendation and news aggregation, prioritize engagement over accuracy (IBM par. 15). This has led to the widespread sharing of fake news and propaganda, influencing public opinion and even election

outcomes (IBM par. 16). AI models trained on biased datasets reinforce those biases, leading to discriminatory outcomes in hiring, policing, and lending practices. For example, facial recognition software has been shown to misidentify individuals from minority groups at significantly higher rates than white individuals due to biased training data (“Data Ethics in AI: 6 Key Principles for Responsible Machine Learning” par. 8). Furthermore, AI models used in hiring processes often favor male candidates over equally qualified female applicants because they are trained on historical hiring data that reflects past discrimination. These examples demonstrate how AI, rather than eliminating human bias, often amplifies and validates existing inequalities, raising urgent ethical concerns.

Furthermore, the lack of accountability in AI decision-making further complicates these ethical dilemmas. Many AI models operate in ways that even their creators do not fully understand. The

“black box” nature of machine learning makes it difficult to determine how decisions are reached, making it nearly impossible to hold anyone accountable when AI systems cause harm (IABAC® sec. 14). If an AI-driven hiring tool discriminates against certain applicants or an autonomous vehicle makes a fatal error, who is responsible? The company that developed the AI? The engineers who trained it? The users who deployed it? These unanswered questions highlight the need for clearer accountability frameworks in AI governance. Without them, companies can deflect blame onto AI, avoiding responsibility for unethical decisions. This lack of accountability has real-world consequences, as seen in cases where AI-driven sentencing algorithms in the criminal justice system disproportionately assign harsher sentences to minority defendants (Larson et al. sec. 4). As AI continues to be integrated into critical aspects of life, ensuring transparency and accountability is more important than ever (IBM par. 18).

Addressing the ethical challenges of AI and data science is no longer optional. It is an

urgent necessity. The rapid pace of AI innovation has far outstripped regulation, leaving society to deal with the consequences of biased decision-making, mass surveillance, and misinformation (IABAC® sec. 15). Ethical AI development requires a commitment to fairness, transparency, and user protection. This means adopting measures such as explainable AI models, stronger data privacy laws, and diverse training datasets to prevent bias. Companies and policymakers must take immediate steps to establish ethical guidelines that ensure AI serves humanity rather than magnifies existing inequalities. If these issues remain unaddressed, the risks of AI outweigh its benefits, making it a threat rather than a tool for progress.

Biased Algorithms: The Root of AI Discrimination

Importantly, artificial intelligence is often perceived as a neutral and objective decision-making tool, yet the reality is that AI systems are only as fair as the data they are trained on. When machine learning models rely on biased datasets, they produce skewed outcomes that reinforce historical inequalities (Larson et al. sec. 5). This bias originates from training data that reflects existing societal prejudices, including racial, gender, and socioeconomic disparities. For example, an investigation by ProPublica into the COMPAS recidivism algorithm found that Black defendants were twice as likely as white defendants to be misclassified as high-risk offenders, leading to harsher sentencing outcomes (Larson et al. sec. 7). Because the algorithm was trained on historical arrest and conviction data, which is already influenced by systemic racial disparities, it learned to associate race with criminality rather than evaluating defendants based on behavioral factors. This demonstrates how AI, rather than eliminating human bias, can amplify and validate discrimination in critical areas such as the criminal justice system.

Additionally, the consequences of biased AI can affect job recruitment, financial services, and even access to housing. A striking example is Amazon's automated hiring tool, which systematically discriminated against women because it was trained on resumes from past hires who were predominantly male software engineers (Goodman par. 5). The algorithm downgraded resumes that included words like "women's," such as "women's rugby team," and favored candidates who used action verbs more commonly found in men's resumes (Goodman par. 6). Similarly, in the financial sector, an AI mortgage system disproportionately charged Black and Latinx borrowers higher interest rates than white borrowers for identical loans (UChicago par. 3). These examples illustrate how AI systems inherit and perpetuate existing social inequalities, making it harder for marginalized groups to access opportunities and financial stability. Because AI decisions often appear objective and data-driven, they can be more difficult to challenge than human biases.

Moreover, the biggest challenge in creating unbiased AI is that bias is deeply embedded in both the data and the development process. AI systems do not inherently understand fairness; they merely identify statistical patterns in the data they are given. When datasets reflect historical inequalities, AI models learn and reinforce those patterns (IBM par. 21). Additionally, biases can arise from algorithmic design choices, such as how certain variables are weighted in decision-making processes (Jonker and Rogers par. 10). Even when developers attempt to correct for bias, they may unintentionally introduce new forms of discrimination, as seen in attempts to "de-bias" hiring algorithms by removing gender-based data, which can still leave areas like education history or work experience that correlate with gender disparities (IBM par. 22). Furthermore, AI bias is compounded by a lack of transparency and accountability. Not having an

understanding of how AI systems reach decisions makes it difficult to detect or correct biased outcomes (“Data Ethics in AI” par. 10). Without comprehensive oversight and diverse representation in AI development teams, ensuring fairness in machine learning remains a persistent and complex challenge.

Equally important, regulatory frameworks, such as the European Union’s AI Act, have begun imposing strict penalties on companies that fail to reduce algorithmic bias, recognizing that biased AI erodes public trust in technology (IBM par. 23). While AI has revolutionized data science, it has also revealed deep ethical vulnerabilities, making it crucial to approach AI development with accountability and fairness at the forefront. Without proactive intervention, biased AI will continue to amplify historical injustices, reinforcing discrimination under the facade of technological objectivity.

The Role of Regulation and Ethical AI Development

Consequently, the rapid growth of AI in data science has raised serious ethical concerns regarding personal data collection and usage. Companies and governments gather vast amounts of personal data from users, often without explicit consent or understanding of its implications (IABAC® sec. 18). For instance, biometric data, health records, and financial information are frequently stored in centralized databases, making them vulnerable to breaches and misuse (IABAC® sec. 19). Governments also use AI for mass surveillance, raising concerns about privacy and civil liberties (IABAC® sec. 20). Without clear regulations, companies can exploit user data for profit.

To address these concerns, various laws and regulations should be implemented to govern AI and data science. For example, the EU AI Act imposes severe penalties on companies that fail

to comply with its ethical AI guidelines, with fines reaching up to EUR 35 million or 7% of a company's global revenue (IBM par. 24). Tech companies bear significant responsibility for ensuring fairness and transparency in AI development. Many corporations prioritize efficiency and profit over ethical considerations, leading to rushed AI deployments that perpetuate bias. Companies must implement internal oversight mechanisms, such as third-party audits and bias-detection tools, to identify and remove discriminatory patterns in AI models (Bandy 20). Ethical AI should not be an afterthought but a fundamental aspect of development.

The Future of AI: Balancing Innovation and Ethics

Looking ahead, AI development must be approached responsibly to ensure that innovation does not come at the cost of ethical integrity. Even Elon Musk, a tech genius and CEO of Tesla and SpaceX, remarks, "With artificial intelligence, we're summoning the demon" (Gesikowski par. 6). However, a key solution is integrating fairness-focused regulations with AI research rather than imposing restrictions that hinder innovation (Floridi and Cowls par. 13). For example, policymakers can implement "regulatory sandboxes" where AI companies can experiment with new models under ethical oversight before full-scale deployment (Floridi and Cowls par. 14). Additionally, interdisciplinary collaboration between ethicists, technologists, and social scientists can help create AI systems that align with human values (Rahwan et al. 492).

Instead of treating AI as an entirely autonomous decision-maker, developers should design AI as an assistive tool that complements human judgment, particularly in high-stakes fields. This approach allows AI to enhance efficiency without amplifying discriminatory patterns.

Additionally, third-party audits and bias-detection algorithms should be standard practice in AI deployment, ensuring companies remain accountable for their systems' decisions (Bandy 21).

Moreover, striking a balance between progress and ethical concerns is crucial because AI's impact extends beyond technology. It shapes economies, societies, and human rights. Ethical AI ensures that advancements in automation, healthcare, and data science benefit everyone. On the other hand, overly restrictive regulations could prevent innovation, slowing advancements in fields like medical research and climate modeling. The solution lies in adaptive governance with regulations that evolve alongside AI technology while maintaining ethical safeguards (Rahwan et al. 496). By prioritizing transparency, fairness, and human oversight, society can harness AI's potential without sacrificing justice.

Conclusion

In a nutshell, while artificial intelligence may seem like an impartial force driving progress in data science, it has deepened ethical concerns by reinforcing systemic biases under the illusion of neutrality. AI's dependence on flawed historical data and lack of accountability have led to discriminatory outcomes, proving that technology does not eliminate human prejudice but often magnifies it. Addressing these ethical challenges is not just about improving technology, rather, it is about protecting fundamental human rights and ensuring fairness in an increasingly automated society. Without transparency, accountability, and diverse representation in AI development, these biases will only become more deeply rooted. This reality demands urgent attention because unchecked AI threatens to enhance inequality, erode public trust, and undermine the very progress it promises. As historian Yuval Noah Harari warns, "Technology is not good or bad. It is up to us to decide how to use it" (115). The path forward requires diverse data representation and robust regulatory frameworks to ensure technology serves humanity and not the other way around. Prioritizing ethical AI development will define not just the future of technology but the future of equality itself. The future of AI is still unwritten, and it is up to us to

decide what kind of world we want it to create.

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The Invisible Judge

A line of code, so neat, so pure,

Make silent decisions dressed as secure.

No heart to feel, no eyes to see,

Yet it shapes the fate of you and me.

A name on paper, turned into vapor,
No value to the maker, graded by a faker.
4 years of college and experience stripped of skin,
A resume lost, before life could begin.
A loan denied with numbers cold,
A future traded, a story untold.
The judge who wears neither robe nor face,
It lives in wires and holds no place.
Yet verdicts fall without a sound,
Lost the purpose of AI that seems to be found.
Bias hidden in the data's breath,
An algorithm that signs your death.
History's weight in a digital world,
Discrimination recycled, the harm has been blurred.
Oh fair machine, you seem so wise,
But the truth is hidden in your disguise.
Behind each choice, an invisible hand,
Of flawed designers across the land.
So check the code like a pencil, not a pen,

Before injustice strikes again.

Explanation: This poem, “The Invisible Judge,” explores the emotional reality behind algorithmic bias and the illusion of fairness that AI creates. It uses personification to give the algorithm human-like power, calling it a “judge” that lacks eyes or a heart. Yet it holds real-world authority over human futures. Further, the poem uses repetition, such as “No heart to feel, no eyes to see,” to emphasize the indifferent nature of AI decision-making and its emotional disconnect from the lives it affects. Moreover, the use of metaphors such as “invisible hand” highlights the human designers whose unconscious biases enter into the algorithms, amplifying and turning machines into silent enforcers of historic discrimination. Ultimately, the poem aims to raise awareness about the unseen but powerful consequences of data-driven decisions, forcing readers to consider how easily technology can become a modern-day destroyer of opportunity and fairness.

A Mirror Made of Math

In numbers they trust, in patterns they see,

A mirror of us, but who holds the key?

It calculates worth, it measures your face,

But fairness is lost in the algorithm’s race.

Data whispers from decades past,

Teaching machines truths that don’t last.

Old wounds coded in endless loops,

Carves the future hollow like a scoop.
From job hiring to courtrooms wide,
The algorithm stands by the judge's side.
Its choices hidden in logic and fact,
But justice and empathy are it lacks.
A mirror made of math reflects our sins,
While power smiles and progress begins.
The future runs on lines of code,
But whose stories were erased remains untold.
Rewrite the code, rethink the plan,
Bias will linger unless checked by man.
Machines can assist, but not replace,
The human heart and moral grace.

Explanation: The second poem, "A Mirror Made of Math," builds on the first by deepening the idea that AI is not an unbiased authority but rather a mirror reflecting human flaws. The central metaphor of the mirror symbolizes how AI replicates human biases instead of erasing them. In addition, it also uses rhyme and rhythm to create a flow that imitates the patterns found in both algorithms and societal prejudice. Phrases like "Old wounds coded in endless loops" reflect the repeating nature of bias, especially when historical data is used to train AI models without critical oversight. In a nutshell, the poem suggests that while AI systems may appear logical and objective, they lack empathy and accountability, which are essential for justice. This connects

with the first poem by emphasizing that ethical AI is about human responsibility and conscious moral design.

The Interview That Wasn't Fair

Setting: A quiet In-N-Out shop in downtown San Francisco. It is early evening. Jade, a recent college graduate, sips coffee across from Dr. Parshv Patel, a Chief Data Science Officer (CDO) and longtime family friend.

Jade (sighing as she puts her phone down): I've got to be honest, Dr. Patel. I don't know what went wrong. I nailed the interview. My degree is solid, my experience lines up, and even the recruiter said I was a top contender. But then the rejection email came, cold and short, like I was never considered in the first place.

Dr. Parshv Patel (nodding slowly): I am sorry to hear that, Jade. Did they mention anything about why?

Jade (sipping her coffee with tired eyes): Just the usual line: "We've selected a candidate whose qualifications align more closely with the role." It didn't even sound like a human wrote it. Dr.

Parshv Patel (looking concerned): I looked into it out of curiosity. Companies use an AI screening system before human eyes even see the applications. You know, resumes go through an algorithm first.

Jade (leaning forward): That explains a lot. So a machine tossed me out before anyone even read my name. Have you ever wondered what the AI was trained on?

Dr. Parshv Patel: I found a developer forum online. Apparently, their system was trained on past successful employee data. However, most of their selected candidates, especially for this role, have been white men from big-name universities like UC Berkeley, from where I graduated, and

also Princeton, Carnegie Mellon, Cornell, UCLA, USC, etc. The AI just “learned” to prefer more of the same.

Jade (worried): That is the real hidden danger, Dr. Patel. AI systems aren’t inherently fair, no matter how smart they sound.

Dr. Parshv Patel: Correct! They reflect the data they’re fed, which often carries human prejudice and bias. If a company has always hired a particular type, the AI will keep repeating the cycle.

Jade (voice cracking slightly): It hit me hard. I thought data was neutral. I believed in merit and spent countless hours to get here. I stayed up late nights, took extra shifts, and juggled family responsibilities, which means I earned that chance. But now I’m starting to see the flaw. The algorithm didn’t even see me. It saw statistics. I wasn’t rejected for who I am; I was rejected for not fitting a biased mold.

Dr. Parshv Patel (softly): Exactly. AI often wears the mask of objectivity. Companies say, “Oh, the algorithm chose the best candidate,” when really it’s just automating past discrimination. It’s like an old bias dressed up in new code.

Jade (looking down at her hands): So we’re trusting machines to be gatekeepers for jobs, for loans, and even for court decisions, and they could be just as flawed as the people who programmed them?

Dr. Parshv Patel: Sometimes even worse. At least with people, you can challenge a decision face-to-face. Algorithms are black boxes. No explanations, no appeals. The more society leans on them, the more invisible the injustice becomes, which would definitely be hard to fix. Jade

(frustrated): It is just so disheartening. I worked so hard, thinking I would finally break through and crack a job. But it feels like the future is being written by past mistakes. Dr.

Parshv Patel (firmly): It doesn’t have to be. That’s why we need people like you who are

young, aware, and ready to question the system. AI doesn't write its own rules. Humans do. If we change the rules, the algorithms will follow.

Jade (after a long pause): I hope you're right. But right now, I feel like the world is letting machines decide our fate.

Dr. Parshv Patel: That is why ethics has to stand shoulder to shoulder with technology. Without it, we're not advancing; rather, we're just automating inequality.

The coffee shop gets darker as they finish their drinks. The conversation continues, heavy and unresolved, much like the ethical future of AI.

Explanation: This dialogue, "The Interview That Wasn't Fair," explores the emotional and personal impact of algorithmic bias in AI-driven hiring processes. Through the conversation between Jade, a recent job seeker, and Dr. Parshv Patel, a Chief Data Science Officer (CDO), the dialogue humanizes the abstract issue of data discrimination. The dialogue uses rhetorical strategies like ethos by presenting the CDO's expert insight, pathos through Jade's disappointment and frustration, and logos by utilizing logical explanations of how AI bias develops to present the complex reality of algorithmic injustice in a relatable form. The use of figurative language, such as "an old bias dressed up in new code," captures the misleading nature of AI's objectivity and highlights how technological systems often amplify human prejudice under the illusion of fairness. The dialogue also contrasts the optimism of a hardworking graduate with the cold, automated rejection to highlight the gap between human potential and machine judgment. By making the conversation casual, the dialogue allows readers to see the human cost of AI bias, urging them to think critically about how algorithms affect real people's futures. The dialogue underscores the research report's claim that AI, without ethical oversight,

does not erase discrimination but magnifies it.