Algorithm and Data Ethics

By: Richa Sharma Msc in Computer Science What Do We Mean When We Talk About 'Ethics'? Ethics is about understanding what is right and wrong and how we should live our lives. It includes the rules and values that guide our behavior in society. Different cultures, religions, and communities have their own ideas about what is ethical, helping people make decisions that align with their values.

Ethics also involves thinking about fairness, responsibility, and how our actions affect others. It's not just about personal choices but also about how we treat people and make decisions that impact society as a whole.

What is an Algorithm?

- **Definition:** An algorithm is a set of step-by-step instructions designed to perform a specific task or solve a problem.
 - Example 1: A recipe for baking a cake. The recipe provides a step-by-step process to follow.
 - Example 2: Directions for driving from one location to another.
- In Computing: Algorithms are used to process data, make decisions, and solve problems automatically.
 - Example: Google Search uses an algorithm to rank and display search results.



Types of Algorithms in Data Science



Sorting Algorithms: Arrange data in a specific order (e.g., quicksort, mergesort).

Example: Sorting a list of numbers from smallest to largest.



Search Algorithms: Find specific data within a dataset (e.g., binary search).

Example: Searching for a student's name in a school database.



Machine Learning Algorithms: Learn from data to make predictions or decisions (e.g., linear regression, decision trees).

Example: Predicting house prices based on features like size, location, and number of rooms.

How Algorithms Impact Our Lives



Social Media Feeds: Algorithms decide what content appears on your social media feed.



Online Shopping: Algorithms recommend products based on your browsing history and past purchases.



Navigation Apps: Algorithms calculate the fastest route to your destination.



Healthcare: Algorithms help doctors diagnose diseases by analyzing medical data.

What is Algorithmic Ethics?

 Definition: Algorithmic ethics involves the study of how algorithms impact society and ensuring they are designed and used in a fair, transparent, and responsible manner.

• Key Questions:

- Fairness: Does the algorithm treat everyone equally?
- Transparency: Do people understand how the algorithm works?
- Accountability: Who is responsible if the algorithm causes harm?



Real-World Examples of Algorithmic Ethics

• Example 1: Bias in Hiring Algorithms

Scenario: A company uses an AI tool to screen job applicants. The algorithm favors resumes that resemble those of current employees, leading to a lack of diversity in hiring.

• Example 2: Privacy Issues in Social Media Data Analysis

Scenario: A social media platform collects user data to sell targeted ads. The platform's algorithm starts showing sensitive health-related ads to users based on their private messages.

Best Practices for Ethical Algorithm Design



Transparency: Clearly document and communicate how algorithms work, including their purpose and decision-making processes, to ensure users understand how their data is being used



Fairness: Implement measures to prevent and address bias in algorithms. Ensure the data used is representative and the outcomes do not disproportionately disadvantage any group.



Accountability: Establish mechanisms for accountability, including regular audits and assessments of algorithmic performance. Ensure there is a clear process for addressing errors and grievances.



Ethical Review: Conduct ethical reviews of algorithms before deployment to identify potential risks and ethical concerns. Continuously monitor and update algorithms to address emerging issues.

What is Data?

Data is information collected in various forms like numbers, text, or observations.

Importance of Data

- **Decision-Making:** Data drives informed decisions in business, healthcare, research, and more.
- Personalization: Enables personalized experiences, such as recommendations and targeted ads.
- Innovation: Fuels advancements in AI, machine learning, and other technologies.

What is Data Ethics?

Definition: The principles guiding the collection, use, and sharing of data in a responsible and ethical manner.

Importance: Ensures privacy, fairness, and accountability in data handling.

Example: Ensuring that data collected from users is used only for its intended purpose and protecting user privacy.

Key Principles of Data Ethics

Privacy: Respecting and protecting the privacy of individuals whose data is being collected or analyzed.

• **Example:** Anonymizing data to prevent the identification of individuals.

Transparency: Ensuring that data collection and analysis processes are clear and understandable.

• **Example:** A company clearly explaining how they use customer data.

Fairness: Avoiding biases that lead to unfair treatment of individuals or groups.

• **Example:** Designing algorithms that do not discriminate based on race, gender, or other protected characteristics.

Accountability: Holding those who use data accountable for their actions and decisions.

• **Example:** Implementing audit trails for data access and decision-making processes.



Ethical Challenges in Data Collection

- Informed Consent: Ensuring users are aware of and agree to data use.
- **Data Ownership:** Clarifying who owns the data and how it can be used.
- **Data Security:** Implementing measures to protect data from breaches.

7 Data Ethics Principles by the UK Government

- 1. Public Good: Data should benefit society as a whole.
- 2. Integrity: Ensure data accuracy, reliability, and honesty.
- 3. Transparency: Processes and decisions involving data must be open and clear.
- 4. Accountability: Organizations must be responsible for their data practices and their impacts.
- 5. Fairness: Data use must avoid bias and ensure equality.
- 6. Confidentiality: Protect personal and sensitive data from unauthorized access.
- 7. Sustainability: Data practices should be sustainable with long-term positive impacts.

Case Study 1: Cambridge Analytica Scandal

Overview: In 2018, it was revealed that Cambridge Analytica harvested data from 87 million Facebook users through a personality quiz app without their consent. This data was used to create detailed psychological profiles and target voters with personalized political ads during the 2016 U.S. presidential election and Brexit referendum.

Ethical Issues: The scandal exposed serious concerns about privacy violations, as users' data was collected without explicit consent, and the manipulation of political messaging raised questions about the integrity of democratic processes. The lack of transparency in data practices further compounded these issues.



The scandal caused a lot of public anger, resulted in Facebook being fined \$5 billion, and led to the closing down of Cambridge Analytica. It also pushed for stronger data protection laws like GDPR, showing that we need clear rules about how data is collected and used, and that companies must be responsible and transparent.

Case Study 2: The Chicago Police Department's "Strategic Subject List" (SSL)

Overview: The Chicago Police Department's Strategic Subject List (SSL) was developed as a predictive policing tool aimed at identifying individuals most likely to be involved in violent crime, either as victims or perpetrators. By analyzing data such as arrest records, gang affiliations, and social networks, the SSL sought to allocate police resources more effectively and prevent crimes before they occurred.

Ethical Issues: One major concern was **racial bias**; the system disproportionately targeted individuals from **minority communities**. This resulted from the data and algorithms used, which led to increased rates of **racial profiling** and perpetuated existing inequalities. Another critical issue was **privacy**. The SSL required extensive data collection, often involving sensitive personal information gathered without individuals' **consent** or even their **knowledge**, raising serious privacy concerns. Additionally, the system lacked **transparency**. There was insufficient clarity about how the SSL generated its predictions and how the data was used in **policing decisions**. This opacity led to doubts about the **fairness** and **accountability** of the system, questioning how decisions based on these predictions were justified.



The Strategic Subject List led to widespread **public** criticism due to its ethical issues. The racial bias in the system sparked significant backlash from civil **rights groups** and the community, demanding reforms to address discriminatory practices. In response, some jurisdictions revised their predictive policing strategies to reduce bias and increase fairness. Legislative actions were taken in several cities to restrict or ban predictive policing tools like the SSL, reflecting growing concerns over privacy and bias. This case has highlighted the need for **better oversight** and **ethical guidelines** in the use of predictive technologies in law enforcement.

How to Ensure Ethical Data Practices



Obtain Informed Consent: Clearly inform individuals about what data is being collected, how it will be used, and obtain their explicit consent before data collection.



Protect Privacy: Implement robust data protection measures, including data anonymization, encryption, and secure storage, to safeguard personal information.



Limit Data Collection: Collect only the data that is necessary for the intended purpose. Avoid excessive or irrelevant data collection to minimize privacy risks.



Regular Audits: Conduct regular audits and assessments of data practices to ensure compliance with ethical standards and identify any potential issues.

Intersectionality and Technology

What is Intersectionality?

Intersectionality refers to the idea that people have multiple identities (like race, gender, and class) that intersect, affecting their experiences in unique ways. These overlapping identities can lead to different kinds of advantages or disadvantages, especially when it comes to technology.

Why It Matters in Algorithm

When creating algorithms, it's crucial to think about all the different kinds of people who will be affected.

Ignoring intersectionality can result in technology that is unfair or harmful to certain groups.

Exercise 1: Analyzing Bias in Al Hiring Tools

- A company uses an AI algorithm to screen job applicants. The algorithm is trained on past hiring data, but it may inadvertently reinforce existing biases. Discuss the potential biases and fairness issues and suggest ways to mitigate them.
- Create a poster showing the possible biases in AI hiring tools, fairness issues, and ways to fix them.
 Make it easy to understand with sections, simple explanations, and pictures or charts to clearly show your ideas.
- Reference: https://github.com/14Richa/The-Ethical-Fallout-of-Amazon-s-Al-Hiring-Tool/blob/main/The Ethical Fallout of Amazon s A I Hiring Tool.pdf



Exercise 2: Identifying Ethical Issues in Data Use

About the App: The mobile app collects user data to offer personalized content, enhancing user experience by aligning recommendations with individual preferences and behaviors. This includes a feature that displays targeted ads based on the data collected.

Targeted Ads Feature: The app uses algorithms to analyze user behavior, interests, and other data points, tailoring ads to match the user's profile. The intention is to serve ads that are relevant and engaging, which can improve the user's interaction with the app.

Vulnerabilities: Some users are being served ads that appear to exploit personal challenges or vulnerabilities. These vulnerabilities could include issues related to mental health, financial difficulties, self-image, or compulsive behaviors. The concern is that the ads, while targeted, may unintentionally take advantage of these sensitive areas, potentially leading to negative outcomes for the users involved.



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