



DECODING AI BIAS IN MEDICINE: INTEGRATING AYURVEDA WITH ARTIFICIAL INTELLIGENCE - CASE STUDIES AND IMPLICATIONS

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ABSTRACT This paper investigates biases in artificial intelligence (AI) applications in medicine, focusing on the marginalization of traditional, indigenous, and holistic systems like Ayurveda. Drawing from *Decoding AI Bias in Medicine*, it examines how AI overlooks non-Western healing practices and proposes strategies for integration. The introduction outlines AI's evolution in healthcare and its biases against holistic approaches. Three case studies demonstrate Ayurveda's integration with AI: (1) combining Ayurveda with Western oncology for improved outcomes; (2) AI-enhanced predictive analytics in Ayurvedic diagnostics; and (3) AI-driven personalization in chronic disease management. These cases highlight opportunities for culturally sensitive AI to promote equitable healthcare. Conclusions advocate for diverse datasets, ethical AI design, and policy reforms to harmonize technology with traditional wisdom, ensuring inclusive global health solutions.

KEYWORDS : AI Bias, Ayurveda Integration, Holistic Healing, Traditional Medicine, Case Studies**INTRODUCTION**

Artificial intelligence (AI) has emerged as a transformative force in modern medicine, revolutionizing diagnostics, treatment planning, and patient care delivery. By leveraging machine learning, deep learning, and predictive analytics, AI enables unprecedented precision in analyzing vast datasets, from electronic health records to genomic profiles [2]. Applications range from AI-driven imaging for early cancer detection to predictive models for chronic disease management, significantly enhancing healthcare efficiency. However, the integration of AI into medicine is not without flaws. A critical issue, as highlighted in *Decoding AI Bias in Medicine* by James and Tano (2025), is the systemic bias embedded in AI systems, particularly their neglect of traditional, indigenous, and holistic healing practices such as Ayurveda. This paper explores these biases, their origins, and the potential for integrating Ayurveda with AI to foster equitable healthcare, using case studies to illustrate practical applications.

AI's biases stem primarily from the datasets used to train algorithms, which are often skewed toward Western biomedical paradigms. These datasets predominantly reflect clinical trials, hospital records, and research from developed nations, where allopathic medicine dominates [3]. Consequently, AI models may undervalue or misinterpret non-Western systems like Ayurveda, a 5,000-year-old Indian holistic practice emphasizing balance among body, mind, and spirit through diet, herbal remedies, yoga, and lifestyle adjustments. Ayurveda's focus on preventive care and individualized treatment—based on doshas (Vata, Pitta, Kapha)—contrasts with the standardized, intervention-heavy approaches of Western medicine. This discrepancy leads to AI algorithms that prioritize pharmaceutical solutions over herbal or lifestyle-based therapies, not due to inefficacy but due to data scarcity [1]. For example, an AI system trained on Western datasets might dismiss turmeric's anti-inflammatory properties, widely validated in Ayurvedic practice, because such data is underrepresented.

The historical marginalization of non-Western healing systems exacerbates this bias. Colonial medical campaigns and funding disparities have long sidelined traditional practices, labeling them as folklore or superstition [1]. During the 19th and 20th centuries, Western medical frameworks were imposed on colonized regions, suppressing indigenous knowledge systems. This legacy persists in modern medical research, where funding prioritizes biomedicine over holistic approaches, limiting the documentation of Ayurveda's efficacy. As a result, AI systems trained on these skewed datasets perpetuate health inequities, particularly for populations relying on traditional medicine. In India, where over 60% of the population uses Ayurveda alongside allopathy, AI's bias risks alienating millions, undermining trust in healthcare technology [4].

Moreover, AI's standardization challenges extend to cultural and spiritual dimensions of healing. Ayurveda integrates physical, psychological, and spiritual health, often using rituals or energy-

based practices like meditation. These are difficult to quantify in data-driven models, leading to AI misinterpretations. For instance, an AI system might misclassify a patient's response to yoga as a placebo effect, ignoring its documented benefits in stress reduction and chronic pain management [1]. This reflects a broader issue: AI's reliance on empirical, Western-centric metrics struggles to capture the holistic efficacy of systems like Ayurveda, which prioritize patient-centered, preventive care.

Ethical considerations are paramount in addressing AI bias. The lack of diverse datasets raises questions of data sovereignty, particularly for indigenous communities whose knowledge risks exploitation without consent [5]. Incorporating Ayurvedic data requires collaboration with traditional practitioners to ensure cultural sensitivity and ethical use. Transparency in AI decision-making—explaining why certain treatments are recommended—is also critical to build trust among patients and providers. Regulatory frameworks, such as those proposed by the World Health Organization, advocate for inclusive data practices and fairness audits to mitigate bias [4].

The potential for AI to enhance Ayurveda is significant. By digitizing ancient texts like the *Charaka Samhita* and integrating them with modern datasets, AI can validate Ayurvedic remedies, predict treatment outcomes, and personalize care plans. Case studies from *Decoding AI Bias in Medicine* demonstrate successful integrations, such as AI-driven dosha assessments improving oncology outcomes or diabetes management. These examples underscore the need for interdisciplinary collaboration among data scientists, clinicians, and Ayurvedic practitioners to design culturally competent AI systems.

However, challenges remain. Technical hurdles include harmonizing disparate data sources, such as biometric readings and qualitative Ayurvedic assessments. Cultural resistance from both Western-trained clinicians and traditional healers can impede adoption. Ethical AI deployment demands continuous monitoring, robust consent processes, and protection against commercial misuse of traditional knowledge. Policy advocacy is essential to establish global standards for equitable AI, ensuring that technology respects diverse medical traditions [1].

This paper uses three case studies to illustrate how AI can bridge Ayurveda with modern medicine, mitigating bias while enhancing outcomes. By addressing data inclusivity, ethical design, and cultural sensitivity, AI can empower Ayurveda, fostering a healthcare ecosystem that values all healing paradigms. The following sections detail these case studies, drawing conclusions for future AI integration in holistic medicine.

Case Study 1: Integrating Ayurveda Into Western Oncology Care

A U.S. oncology center integrated Ayurveda with chemotherapy for 50 patients with advanced cancer [1]. AI analyzed electronic health records to predict dosha imbalances, recommending herbal mixtures (e.g., turmeric-based compounds), yoga, and dietary adjustments.

Challenges included regulatory barriers and clinician skepticism. Outcomes showed a 25% improvement in quality-of-life scores, 30% reduction in side effects, and enhanced adherence. AI-enhanced imaging validated synergistic effects, demonstrating AI's role in bridging Ayurveda with Western protocols.

Case Study 2: AI-Enhanced Predictive Analytics in Ayurvedic Medicine

An Indian clinic used AI to analyze data from 200 patients with digestive disorders, correlating symptoms with dosha profiles and herbal efficacy [1]. Machine learning processed digitized Ayurvedic texts and biometric data to recommend regimens like triphala. Initial biases from Western-skewed data were corrected, improving accuracy by 40%. Results included faster symptom relief (15 days average) and lower relapse rates, highlighting AI's potential to modernize Ayurveda through inclusive data.

Case Study 3: AI-Driven Personalization in Ayurvedic Chronic Disease Management

In Nepal, AI personalized Ayurvedic treatment for 150 diabetic patients, integrating glucose data with prakriti assessments [1]. Wearables provided real-time feedback, adjusting herbal and yoga plans. Community engagement addressed cultural resistance and privacy concerns. Outcomes included a 20% HbA1c reduction and high patient satisfaction, showing AI's ability to respect traditional preferences while enhancing chronic care.

CONCLUSIONS

These case studies demonstrate that integrating Ayurveda with AI can mitigate biases, enhancing holistic care. Diverse datasets, ethical AI design, and interdisciplinary collaboration are critical to success. Future efforts should focus on global standards for AI fairness, protecting cultural sovereignty, and advocating inclusive policies. AI can empower Ayurveda, ensuring equitable healthcare that respects all traditions (approx. 150 words).

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