

use of patient data in AI applications. Moreover, algorithmic biases and disparities in data representation may introduce unintended consequences, exacerbating existing healthcare disparities. Standardization of AI models and rigorous validation in real-world clinical settings are essential to ensure the reliability and generalizability of AI-driven predictions. Additionally, fostering interdisciplinary collaboration between clinicians, data scientists, and regulatory bodies is critical for harnessing the full potential of AI in breast cancer care [6].

Applications: Early detection: AI-powered predictive models can analyze mammograms and other imaging scans to detect breast cancer at an early stage, leading to timely interventions and improved outcomes.

Risk assessment: AI algorithms can assess individual patient data, including genetic, demographic, and lifestyle factors, to accurately estimate breast cancer risk and tailor screening and prevention strategies accordingly [7].

Treatment planning: AI-based predictive modeling can analyze diverse data sources, including genomic profiles, pathology reports, and clinical histories, to recommend personalized treatment plans optimized for each patient's unique characteristics and preferences [8].

Drug discovery: AI techniques such as deep learning can expedite the discovery of novel therapeutic targets and drug candidates by analyzing vast amounts of molecular data and predicting their efficacy in combating breast cancer.

Clinical decision support: AI systems can provide clinicians with real-time decision support by synthesizing patient data, evidence-based guidelines, and expert knowledge to assist in diagnosis, treatment selection, and monitoring [9].

Response prediction: AI algorithms can predict individual patient responses to specific treatments based on their molecular profiles, imaging features, and clinical parameters, enabling the selection of the most effective therapies while minimizing adverse effects.

Adaptive therapy: AI-driven predictive modeling can dynamically adjust treatment regimens based on evolving patient characteristics, disease progression, and treatment responses, optimizing therapeutic efficacy and minimizing toxicity over time.

Outcome prediction: AI models trained on large-scale patient datasets can accurately predict long-term outcomes such as recurrence risk, survival probabilities, and quality of life metrics, facilitating informed decision-making and patient counseling [10].

Resource allocation: AI-powered predictive analytics can optimize resource allocation within healthcare systems by forecasting future demand for diagnostic tests, treatments, and supportive services based on population trends and disease prevalence.

Clinical trials optimization: AI algorithms can accelerate the design and execution of clinical trials by identifying eligible patients, predicting trial outcomes, and optimizing trial protocols, leading to faster drug development and regulatory approval processes.

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