

# From Data to Drugs a Review: Harnessing AI for Accelerated Pharmaceutical Development

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**Abstract:** Drug development accelerates discovery. AI changed medication development. AI accelerates pharmaceutical research from data analysis to medicine development. To fulfil global healthcare requirements, pharmaceutical development must be speedy. AI accelerates and improves medication development decisions. AI impacts medication discovery. AI validates medications quicker. AI-based virtual screening and drug discovery may quickly find therapeutic candidates with high target molecule affinity. Predictive modelling accelerates drug discovery. The review examines preclinical AI development. AI evaluates huge biological and chemical databases for medication safety and effectiveness. AI-driven in silico toxicity and safety evaluations reduce risks and enhance preclinical research. AI may improve pharmaceutical formulation and delivery. AI enhances clinical trial design and recruiting. Real-time data analysis and clinical trial monitoring provide unmatched insights into medication effectiveness and safety, expediting decision-making and trial length. Predictive AI may improve trial results and drug development. The research examines AI's involvement in regulatory and commercial approval. AI-prepared data speeds acceptance. AI improves post-marketing pharmacovigilance and safety. Market entry and health economics are explored. AI in pharmaceutical research faces data quality, integration, ethical, and regulatory issues. Discussed are pharmaceutical AI implementation options. Finally, AI will change pharmaceuticals. Precision and personalised medicine using AI suggests patient-specific therapy. AI may expedite pharmaceutical development and improve patient outcomes, highlighting the need for ongoing research and cooperation to employ AI in global healthcare.

**Keywords:** Pharmaceutical development, AI, drug discovery, preclinical, clinical trials, regulatory approval  
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