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Study on Predicting Disease Outcomes with Machine Learning in Healthcare Analytics

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Abstract: *Healthcare is undergoing a transformative paradigm shift driven by advancements in data analytics and machine learning. Predictive healthcare analytics has emerged as a promising tool for early diagnosis, prognosis, and treatment recommendation. This paper provides an in-depth exploration of the application of machine learning techniques to predict disease outcomes in healthcare settings. It discusses the challenges, opportunities, and real-world applications of predictive analytics, highlighting the potential to revolutionize patient care and improve healthcare outcomes.*

Keywords: Machine learning, Healthcare Analytics, Predictive Modeling.

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