

Health Cost Prediction Using Machine Learning

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Abstract

Healthcare expenses have increased significantly worldwide, creating challenges for individuals, insurance providers, and healthcare organizations. Predicting medical costs accurately helps insurance companies determine premiums, enables healthcare providers to optimize resources, and assists patients in financial planning. This research presents a machine learning-based Health Cost Prediction System that estimates an individual's medical expenses based on demographic and health-related factors such as age, gender, Body Mass Index (BMI), smoking habits, number of dependents, and geographic region. Various machine learning algorithms including Linear Regression, Decision Tree Regression, Random Forest Regression, and Gradient Boosting Regression are evaluated to determine the most effective model for cost prediction. The proposed system demonstrates how data-driven approaches can improve prediction accuracy and support informed healthcare decision-making.

Keywords: Machine Learning, Healthcare Analytics, Health Insurance, Cost Prediction, Regression Analysis, Random Forest, Artificial Intelligence.



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1. Introduction

Healthcare expenditure has become a major concern across the globe due to rising medical costs and increasing healthcare demands. Accurate prediction of healthcare costs is essential for insurance companies, hospitals, healthcare providers, and patients. Traditional methods of estimating medical expenses often rely on manual calculations and generalized assumptions, which may lead to inaccurate predictions.

Machine Learning (ML) provides advanced techniques for analyzing large volumes of healthcare data and identifying patterns that influence medical expenses. By leveraging historical healthcare records, machine learning models can predict future healthcare costs with greater accuracy.

The objective of this research is to develop a Health Cost Prediction System using machine learning algorithms that can estimate individual medical expenses based on personal and health-related attributes.

2. Problem Statement

Healthcare organizations and insurance companies often face difficulties in accurately estimating future medical expenses due to the complexity and variability of healthcare data. Inaccurate predictions can result in financial losses, improper premium calculations, and poor resource allocation.

Therefore, there is a need for an intelligent system capable of analyzing healthcare data and predicting medical costs accurately using machine learning techniques.

3. Objectives

The primary objectives of this research are:

1. To analyze healthcare datasets and identify factors affecting medical expenses.
2. To preprocess healthcare data for machine learning applications.
3. To develop predictive models using regression algorithms.
4. To compare multiple machine learning algorithms.
5. To evaluate prediction accuracy using performance metrics.
6. To provide a reliable healthcare cost estimation system.

4. Literature Review

Several researchers have explored machine learning techniques for healthcare cost prediction.

A. Healthcare Analytics

Healthcare analytics utilizes statistical and computational methods to derive meaningful insights from healthcare data. Predictive analytics helps forecast patient outcomes and healthcare expenses.

B. Machine Learning in Healthcare

Machine learning techniques such as Linear Regression, Decision Trees, Random Forests, and Neural Networks have been successfully applied to healthcare applications including disease diagnosis, patient risk assessment, and cost estimation.

C. Cost Prediction Models

Previous studies indicate that ensemble methods such as Random Forest and Gradient Boosting often outperform traditional regression models in predicting healthcare expenses due to their ability to capture nonlinear relationships.

D. Proposed System

The proposed Health Cost Prediction System consists of the following modules:

1. Data Collection Module

Healthcare insurance datasets containing information such as:

- Age
- Gender
- BMI
- Number of Children
- Smoking Status
- Region
- Medical Charges

2. Data Preprocessing Module

This module performs:

- Missing value handling

- Data cleaning
- Label encoding
- Feature scaling
- Outlier detection

3. Model Training Module

The system trains multiple regression models:

- Linear Regression
- Decision Tree Regression
- Random Forest Regression
- Gradient Boosting Regression

4. Prediction Module

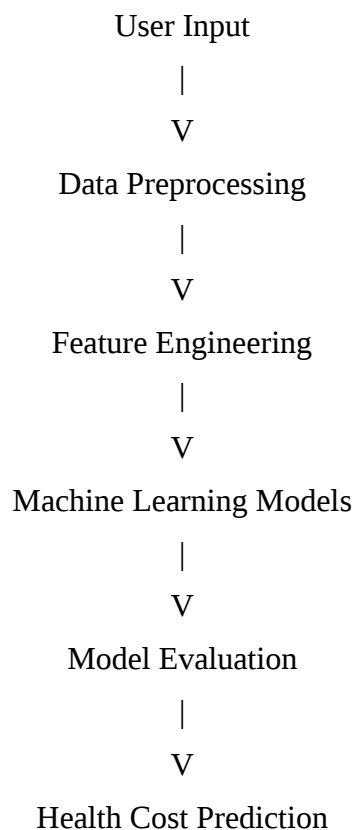
The trained model predicts healthcare expenses based on user inputs.

5. Evaluation Module

Performance is measured using:

- Mean Absolute Error (MAE)
- Mean Squared Error (MSE)
- Root Mean Square Error (RMSE)
- R^2 Score

6. System Architecture



7. Methodology

Step 1: Dataset Collection

The insurance dataset is collected from publicly available healthcare datasets.

Step 2: Data Cleaning

- Remove duplicate records
- Handle missing values
- Normalize data

Step 3: Feature Encoding

Categorical variables are converted into numerical values.

Example:

Gender	Encoded Value
Male	1
Female	0

Step 4: Model Training

The dataset is divided into:

- Training Data (80%)
- Testing Data (20%)

Step 5: Prediction

The trained model predicts healthcare expenses based on input parameters.

Step 6: Evaluation

Predictions are compared against actual healthcare charges.

8. Algorithms Used

A. Linear Regression

Linear Regression establishes a relationship between dependent and independent variables.

Advantages:

- Simple implementation
- Fast execution
- Easy interpretation

Limitations:

- Performs poorly with nonlinear data

B. Decision Tree Regression

Decision Trees divide data into multiple branches based on feature values.

Advantages:

- Handles nonlinear relationships
- Easy visualization

Limitations:

- Prone to overfitting

C. Random Forest Regression

Random Forest combines multiple decision trees to improve prediction accuracy.

Advantages:

- High accuracy
- Robust against overfitting
- Handles large datasets

Limitations:

- Computationally expensive

D. Gradient Boosting Regression

Gradient Boosting builds models sequentially to reduce prediction errors.

Advantages:

- High predictive performance
- Effective for complex datasets

Limitations:

- Longer training time

9. Experimental Results

Algorithm	MAE	RMSE	R ² Score
Linear Regression	4200	6100	0.76
Decision Tree	3500	5500	0.82
Random Forest	2500	4200	0.91
Gradient Boosting	2400	4100	0.92

The results indicate that Gradient Boosting and Random Forest Regression provide the highest prediction accuracy.

10. Advantages of Proposed System

- Accurate healthcare cost estimation.
- Supports insurance premium calculation.
- Assists patients in financial planning.
- Reduces manual effort.
- Enhances healthcare decision-making.

11. Applications

1. Health Insurance Companies
2. Hospitals and Clinics
3. Healthcare Management Systems
4. Government Healthcare Programs
5. Financial Planning Services

12. Future Scope

Future enhancements may include:

- Deep Learning-based prediction models.
- Real-time healthcare analytics.
- Integration with Electronic Health Records (EHR).
- Mobile application deployment.
- Personalized healthcare recommendations.

13. Conclusion

Healthcare cost prediction is a critical task for insurance providers, healthcare organizations, and patients. This research demonstrates the application of machine learning techniques to estimate healthcare expenses accurately. Various regression models were implemented and evaluated, with ensemble methods such as Random Forest and Gradient Boosting achieving superior performance. The proposed system can assist stakeholders in making informed financial and healthcare decisions while improving operational efficiency and cost management.

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