

# Technical Report: Insurance Charges Prediction

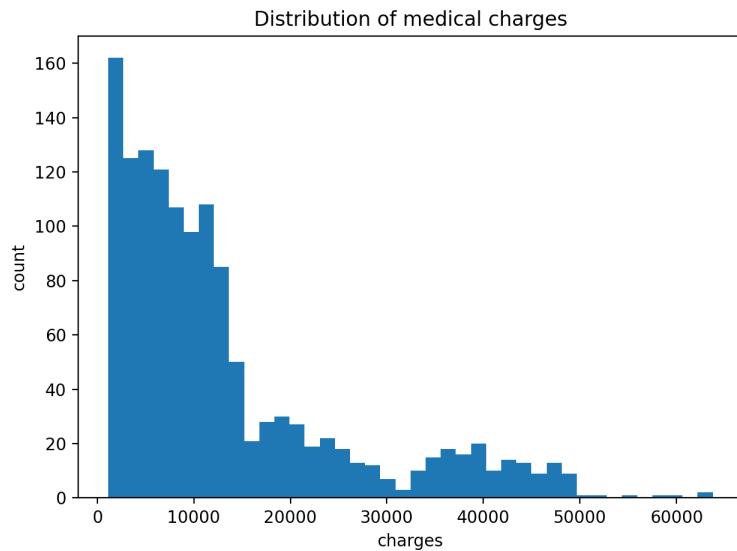
Goal: build an end-to-end Machine Learning pipeline that predicts medical insurance charges from demographic and lifestyle factors, then deploy the best model as a FastAPI service and a simple Streamlit UI.

## Dataset

Rows: 1338 | Columns: 7 | Target: charges (continuous). Features: age, sex, bmi, children, smoker, region.

## Quick EDA findings

- No missing values in this dataset (0 missing across all columns).
- Charges are right-skewed: min=1121.87, median=9382.03, max=63770.43.
- Smoker status is a strong driver of charges (smokers have much higher median charges).



# Pipeline, Models, and Evaluation

## Data Validation (Schema)

We enforce a schema: required columns, allowed categorical values, and valid ranges for numeric fields. The same validation is reused in training and in the API to protect the model from bad inputs.

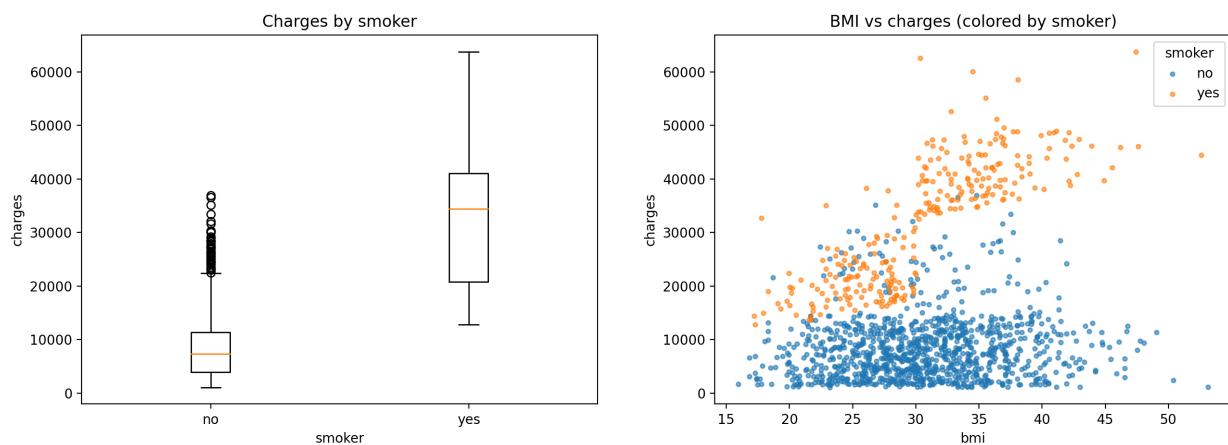
## Preprocessing & Feature Engineering

Categorical variables are One-Hot Encoded. Numeric variables are imputed (median) and scaled with StandardScaler. Feature engineering adds BMI category, age group, a smoker indicator, and interaction terms (BMI $\times$ smoker, age $\times$ smoker).

## Model comparison on test set

| Model             | MAE     | MSE         | R2    |
|-------------------|---------|-------------|-------|
| random_forest     | 2430.35 | 19754043.29 | 0.873 |
| ridge             | 2767.23 | 20624971.86 | 0.867 |
| linear_regression | 2762.71 | 20700442.41 | 0.867 |

Selected model for deployment: random\_forest.



# Serving & Product Demo

## FastAPI

The service exposes GET /health and POST /predict (JSON). It loads the saved sklearn Pipeline (feature engineering + preprocessing + model) and returns predicted charges.

## Streamlit UI

The Streamlit app collects user inputs and calls the API endpoint. This demonstrates a complete workflow from user interface to backend model inference.

## How to run

- pip install -r requirements.txt
- python -m src.train
- uvicorn app.api:app --reload
- streamlit run app/web\_app.py

## Future improvements

- Add cross-validation and hyperparameter tuning (GridSearchCV).
- Experiment with additional models (Gradient Boosting) or a small neural network (bonus).
- Add logging/monitoring and a more robust validation library (e.g., pandera) if allowed.