DPR

Insurance Premium Prediction

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Abstract

This Low-Level Design Document encapsulates .Insurance premium prediction for low-level documents is the use of data analysis and modeling to estimate insurance costs based on factors like age, health, lifestyle, etc. This helps insurers set fair premiums for life, health, or disability coverage, ensuring pricing matches individual needs and risk levels. These data repositories, securely stored within a data warehouse, serve as reservoirs of opportunity. Our system embarks on a transformative journey through the realms of data, expertly harnessing its power. The journey unfolds through meticulously orchestrated steps: data ingestion from Kaggle datasets, data transformation for cleanliness and relevance, model building to extract meaningful patterns, and the establishment of an efficient batch prediction pipeline. We don't stop there; we extend this journey to the end-users with a well-crafted, user lifriendly interface, bridging

INTRODUCTION

Why this DPR Documentation?

The main purpose of this DPR documentation is to add the necessary details of the project and provide the description of the machine learning model and the written code. This also provides the detailed description on how the entire project has been designed end-to-end.

Key points:

Describes the design flow
Implementations
Software requirements
Architecture of the project
Non-functional attributes like:
Reusability
Portability
Resource utilization

1 Description

1.1 Problem Perspective

- 1. Data Quality and Generalization: Data limitations and potential overfitting could affect the accuracy and generalizability of predictive models.
- 2. Privacy and Regulations: Ethical concerns and evolving insurance regulations may pose challenges in handling personal data and ensuring compliance.

- Model Transparency and Adoption: Complex models might lack transparency, hindering user understanding and acceptance.
- External Factors: Economic and environmental influences on insurance markets may not be fully considered in predictions.

1.2 Problem Statement

Project The goal of this project is to give people an estimate of how much they need based on their individual health situation. After that, customers can work with any health insurance carrier and its plans and perks while keeping the projected cost from our study in mind. This can assist a person in concentrating on the health side of an insurance policy rather han the ineffective part.

1.3 Proposed Solution

Our solution involves developing a personalized premium prediction model that utilizes advanced machine learning techniques. It will consider individual factors like age, smoker, sex, and pregnency to generate precise premium estimates. We will ensure data quality and privacy compliance, create a user-friendly interface, and continuously update the model to adapt to changing regulations and market conditions. Additionally, educational efforts will promote awareness and adoption among consumers and industry stakeholders..

1.4 Solution Improvements

Improvements the solution for leveraging previous insurance data, fostering cross-functional collaboration, integrating external data sources, continuous monitoring, machine learning models, and customer feedback. By implementing these enhancements, the solution becomes more comprehensive, adaptable, and capable of addressing evolving challenges, enabling indivisuals to stay informative and responsive about the premiums.

2 Technical Requirements

There are not any hardware needs needed for victimization this application, the user should have an interactive device that has access to the web and should have the fundamental understanding of providing the input. And for the backend half the server should run all the package that's needed for the process and provided information to show the results.

2.1 Tools Used

□ Python 3.8 is employed because the programming language and frame works like numpy, pandas, sklearn,flask, streamlit and alternative modules for building the model.
 □ Visual Studio Code is employed as IDE.
 □ Front end development is completed victimization HTML/CSS
 □ Flask is employed for each information and backend readying
 □ GitHub is employed for version management
 □ Streamlit Cloud and localhost is used for Deployment



3 Data Requirements

The info demand is totally supported the matter statement. and also, the information set is accessible on the Kaggle within the type of standout sheet(.xlsx), because the main theme of the project is to induce the expertise of real time issues, we have a tendency to once more mercantilism {the information into the prophetess data base and commerce it into csv format.

3.1 Data Gathering from Main Source

The data for the current project is being gathered from the Kaggle dataset, which is available at the following link: Insurance Premium Prediction | Kaggle. This dataset serves as the primary source for our project's data analysis and premium prediction tasks.

3.2 Data Description

We have train (1070) and test (268) data set, train data set has both input and output

Columns Are:

Age: Age of the insured individuals, a key factor in premium calculation due to its influence on health risk.

Sex: Gender of the insured individuals, impacting premiums based on gender-specific health risks. BMI (Body Mass Index): Measure of body weight relative to height, influencing premiums based on health implications.

Children: Number of dependents covered, affecting policy costs due to family size.

Smoker: Smoking status (yes/no), a significant factor in premium pricing due to health risks associated with smoking.

Region: Geographic location of the insured, which can affect healthcare costs and insurance pricing.

Expenses: Actual medical expenses incurred, providing insight into healthcare costs and utilization for premium calculation.

age	sex	bmi	children	smoker	region	expenses
19	female	27.9	0	yes	southwest	16884.92
18	male	33.8	1	no	southeast	1725.55
28	male	33	3	no	southeast	4449.46
33	male	22.7	0	no	northwest	21984.47
32	male	28.9	0	no	northwest	3866.86
31	female	25.7	0	no	southeast	3756.62
46	female	33.4	1	no	southeast	8240.59
37	female	27.7	3	no	northwest	7281.51
37	male	29.8	2	no	northeast	6406.41
60	female	25.8	0	no	northwest	28923.14
25	male	26.2	0	no	northeast	2721.32
62	female	26.3	0	yes	southeast	27808.73
23	male	34.4	0	no	southwest	1826.84
56	female	39.8	0	no	southeast	11090.72
27	male	42.1	0	yes	southeast	39611.76
19	male	24.6	1	no	southwest	1837.24
52	female	30.8	1	no	northeast	10797.34
23	male	23.8	0	no	northeast	2395.17
56	male	40.3	0	no	southwest	10602.39
30	male	35.3	0	yes	southwest	36837.47
60	female	36	0	no	northeast	13228.85
30	female	32.4	1	no	southwest	4149.74

3.3 Data Ingestion

The cornerstone of our data-driven project was established through a systematic process of data acquisition and ingestion. Utilizing Kaggle, a reputable platform renowned for its high-quality datasets, we identified and acquired the crucial data required for our insurance price prediction project. This dataset, integral to our goal of accurate price forecasting, was meticulously downloaded and securely stored within our local system infrastructure. Subsequently, we initiated the data ingestion phase, where the dataset seamlessly integrated into our project's data pipeline. This meticulous approach ensures that our project is built upon a solid foundation, setting the

approach ensures that our project is built upon a solid foundation, setting the stage for robust and precise insurance price prediction models and analysis

4 Data Transformation

Steps performed in pre-processing are:

- · First read data from Artifact folder
- Checking unnecessary columns
- One column has product id which is unique for every product so I deleted that column.
- Checked for null values
- there are too many null values are present in two columns that's why I deleted them
- Performed one-hot encoder on categorical columns.
- Scaling is performed for needed information.
- And, the info is prepared for passing to the machine learning formula

5 Design Flow

5.1Modelling

The pre-processed information is then envisioned and every one the specified insights are being drawn. though from the drawn insights, the info is at

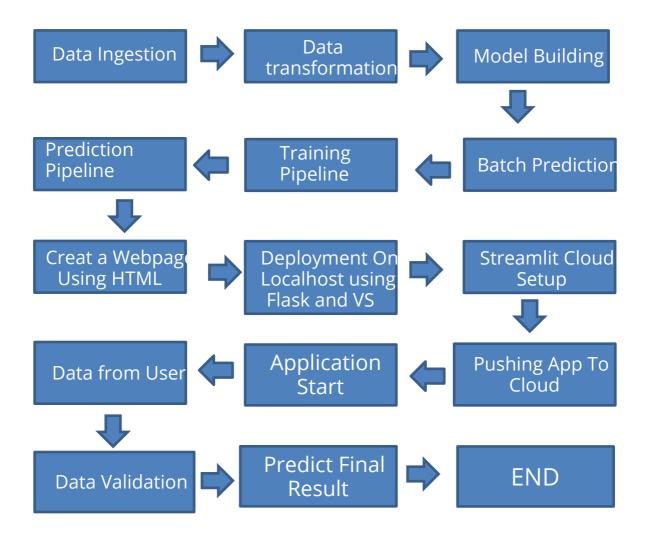
randomunfold however still modelling is performed with completely different machinelearning algorithms to form positive we tend to cowl all the chances. and eventually, Gradient Boosting performed well .

5.2UI Integration

Both CSS and HTML files are being created and are being integrated with the created machine learning model. All the required files are then integrated to theapp.py file and tested locally

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5.3 Modelling Process & 5.4 Deployment Process



6Data from User

The data from the user is retrieved from the created HTML web page.

7Data Validation

The data provided by the user is then being processed by app.py or application.py file and validated. The validated data is then sent for the prediction.

8Rendering the Results

The data sent for the prediction is then rendered to the web page.

9Deployment

The tested model is then deployed to Streamlit Cloud. So, users can access the project from any internet devices.

Conclusion

Insurance Premium Prediction represents a crucial step towards informed insurance purchases. By harnessing data analysis and machine learning, it brings transparency and fairness to premium calculation, empowering users to make confident, personalized choices. As this journey continues, it promises to reshape the insurance landscape, ensuring individuals and businesses can purchase premiums with clarity and confidence.

Q & A:

Q1) What's the source of data?

Ans-The data for training is provided by the client in multiple batches and each batch contain multiple files.

Q 2) What was the type of data?

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Q 3) What's the complete flow you followed in this Project?

Ans-Refer Page no 11 for better Understanding.

Q 4) After the File validation what you do with incompatible file or files which didn't pass the validation?

Ans-Files like these are moved to the Achieve Folder and a list of these files has been shared with the client and we removed the bad data folder.

Q 5) How logs are managed?

Ans- We are using different logs as per the steps that we follow in validation and modeling like File validation log, Data Insertion, Model Training log, prediction log etc.

Q 6) What techniques were you using for data pre-processing?

Ans-Removing unwanted attributes Cleaning data and imputing if null values are present. Converting categorical data into numeric values.

Q 7) How training was done or what models were used?

Ans-Before dividing the data in training and validation set, we performed preprocessing over the data set and made the final dataset. As per the dataset training and validation data were divided. Algorithms like SVR,

Decision Tree, Random Forest, Gradient Boosting were used based on the recall, final model was used on the dataset and we saved that model.

Q 8) How Prediction was done?

Ans-The testing files are shared by the client. We Performed the same life cycle on the provided dataset. Then, on the basis of dataset, model is loaded and prediction is performed. In the end we get the accumulated data of predictions.

Q 9) What are the different stages of deployment?

Ans-First, the scripts are stored on GitHub as a storage interface.

The model is first tested in the local environment.

After successful testing, it is deployed on Streamlit Cloud.