

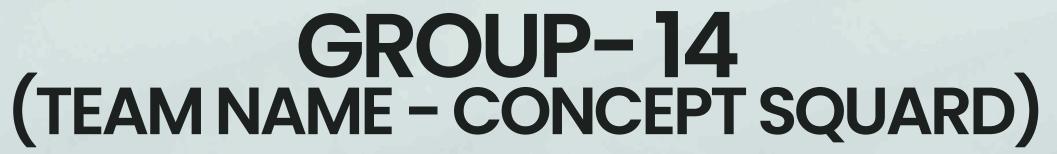








LEARNATHON 4.0



Title Name

Al-Powered Fraud Detection in Auto Insurance: Predictive Modelling for Smarter

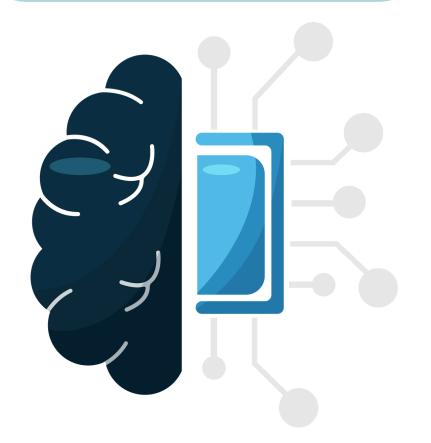
Claims Management

PRESENTED BY

Mr. Ayushman Panigrahy Mr. Debasish Mishra Mr. A. Ajit Mr. Mihir Ranjan Aich

Problem Statements §

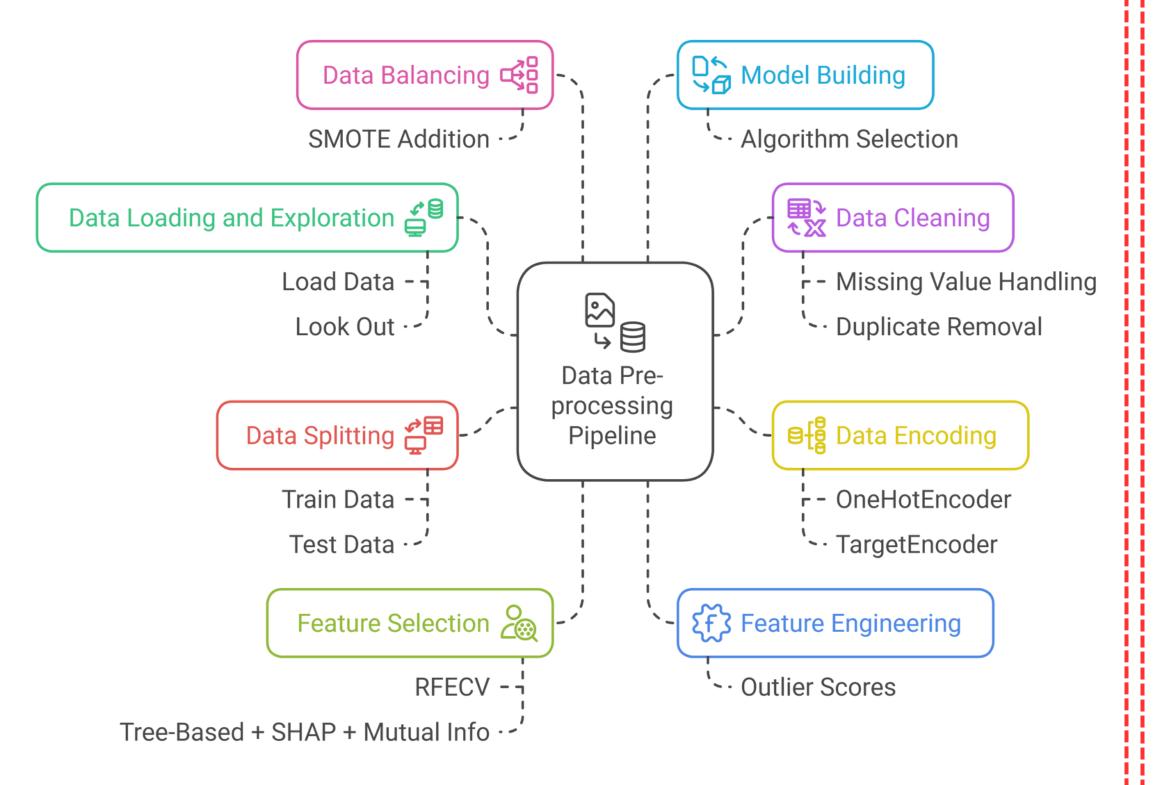
Al-Powered Fraud Detection in Auto Insurance: Predictive Modeling for Smarter Claims Management



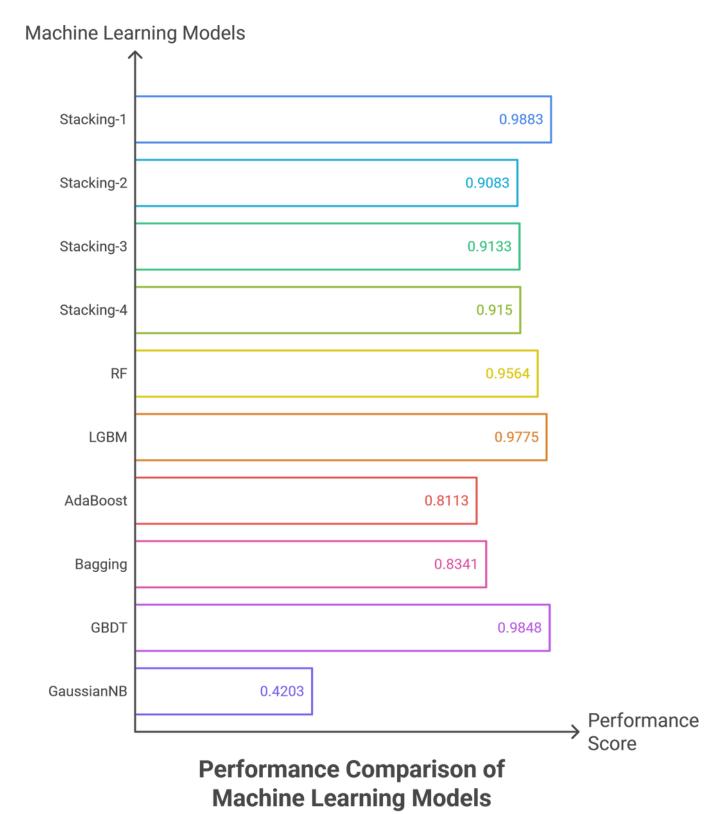
What We Are Solving (Our Al-Based Approach)

- Automatically predicts the likelihood of a claim being fraudulent using machine learning models.
- Analyzes critical features such as customer age, policy number, vehicle price, accident area, and more.
- Provides a real-time fraud prediction system for smarter and faster decision-making.
- Helps to reduce manual effort, enabling investigators to focus only on high-risk claims.
- Supports insurers in minimizing financial losses by flagging fraud early in the process.
- M Delivers interactive dashboards and visual analytics for deeper insight into fraud patterns.
- Improves operational efficiency, risk management, and overall claims processing.

DATA PER-PROCESSING PIPRLINE OF OUR APPROACH



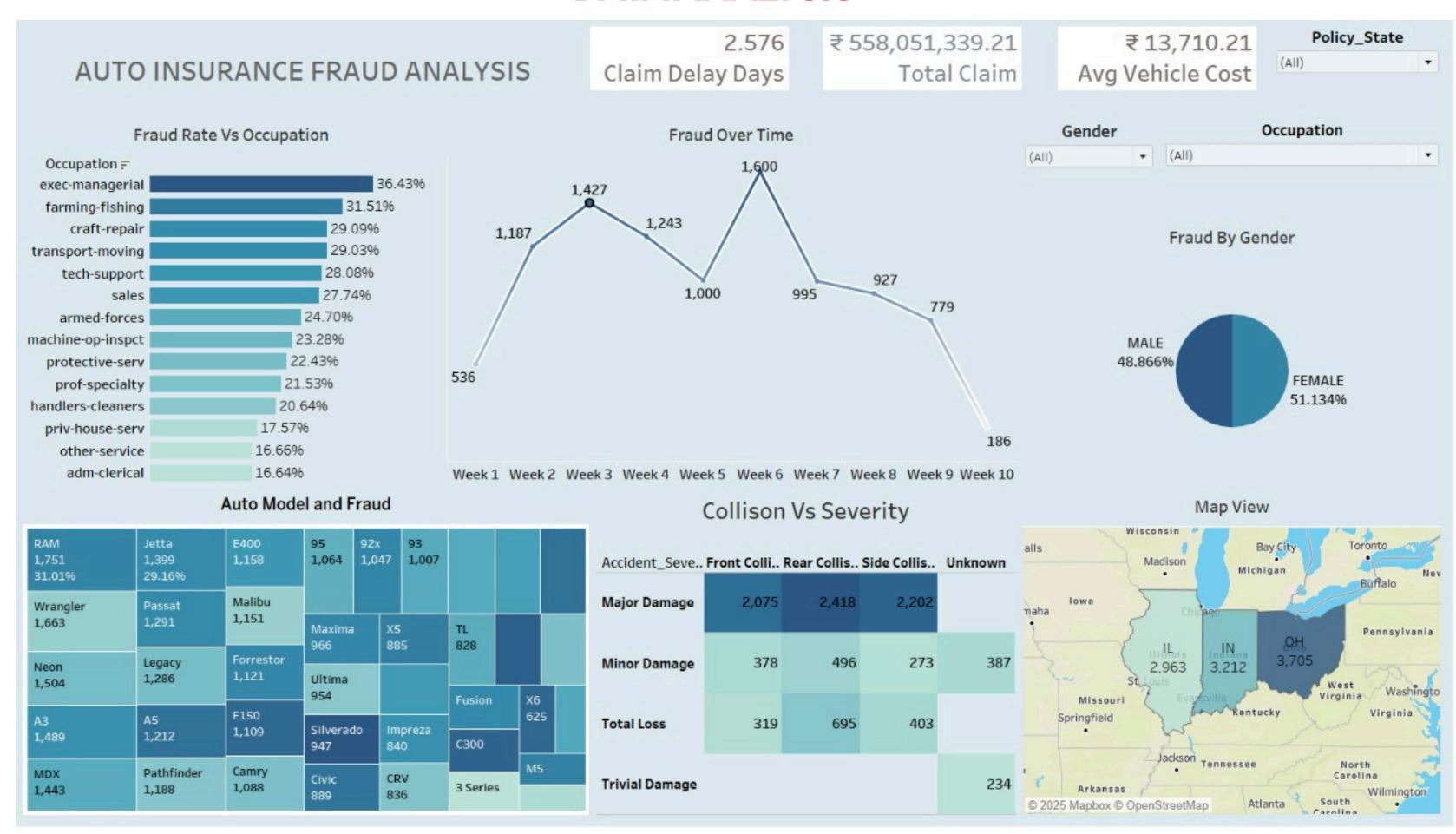
DIFFERNT MODEL USED IN OUR APPROACH



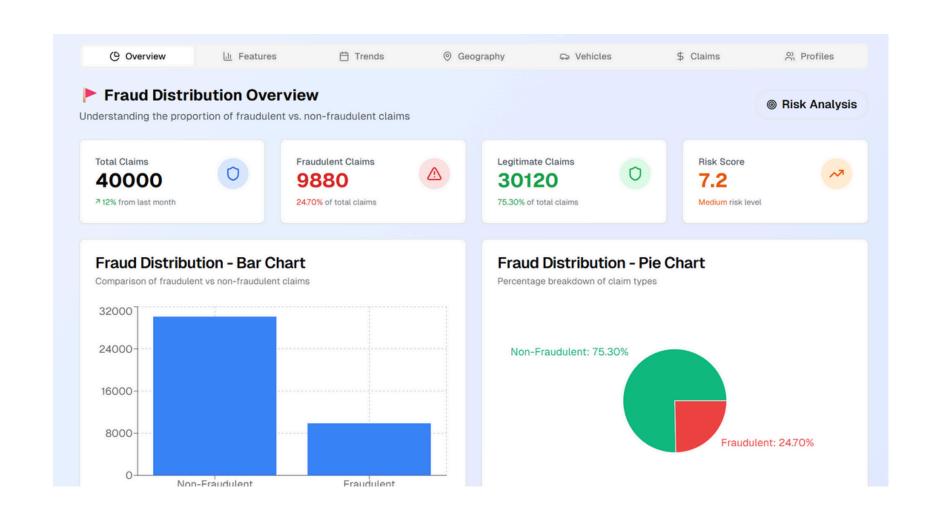
COMPREHENSIVE EVALUATION OF MULTIPLE MACHINE LEARNING MODELS

Metric	Stacking-1 (RF+GBDT	Stacking-2 (GBDT+LG	Stacking-3 (RF+GBDT	Stacking-4 (RF+LGBM	RF	LGBM	AdaBoost	Bagging	GBDT	GaussianN B
Accuracy	0.9883	0.9083	0.9133	0.915	0.9564	0.9775	0.8113	0.8341	0.9848	0.4203
Precision	0.9916	0.8972	0.8874	0.8986	0.85	0.9165	0.6068	0.627	0.9422	0.2907
Recall (Sensitivit	0.914	0.9068	0.9319	0.9211	1	1	0.6705	0.8102	1	0.9349
Specificity	0.9034	0.9097	0.8972	0.9097	0.9421	0.9701	0.8574	0.8419	0.9799	0.2515
F1 Score	0.9027	0.902	0.9091	0.9097	0.9189	0.9564	0.637	0.707	0.9702	0.4434
ROC AUC	0.9687	0.9725	0.9708	0.9756	0.9991	0.9998	0.766	0.8992	0.9999	0.7897
Balanced	0.9087	0.9082	0.9145	0.9154	0.9711	0.9851	0.764	0.8261	0.9899	0.5932
Cohen's Kanna	0.8161	0.8159	0.8264	0.8294	0.8894	0.9413	0.5099	0.5938	0.9601	0.1068
Log Loss	0.3433	0.2309	0.2312	0.2305	0.2113	0.1271	0.6107	0.4058	0.1124	1.6994
MSE	0.0917	0.0917	0.0867	0.085	0.0436	0.0225	0.1888	0.1659	0.0152	0.5797
RMSE	0.0028	0.3028	0.2944	0.2915	0.2088	0.15	0.4345	0.4074	0.1231	0.7614
MAE	0.0917	0.0917	0.0867	0.085	0.0436	0.0225	0.1888	0.1659	0.0152	0.5797
R ²	0.9315	0.6315	0.6516	0.6583	0.7656	0.879	0.0147	0.1079	0.9185	-2.1165

DATA ANALYSIS



WEBSITE FOR REALWOLD PROBLEM SOLUTION





The dashboard is organized into several tabs for streamlined fraud analysis. The Overview tab summarizes total claims, fraud percentages, and key indicators through dynamic KPIs and charts. The Features tab displays the most influential variables affecting fraud predictions, using importance plots and correlation visuals. In Trends, users can analyze how fraud patterns evolve over time, with timelines and distribution plots. The Geography tab visualizes the regional distribution of fraud cases using interactive maps and location filters. The Vehicles tab breaks down fraud statistics by vehicle type, age, and price to highlight high-risk categories. The Claims section provides insights into fraud based on claim reasons, amounts, and durations, helping spot abnormal behaviors. Lastly, the Profiles tab presents customer demographics like age, occupation, and income level, showing how different profiles correlate with fraud likelihood.

Thankyou very much!