

Driving Innovation Through GenAl & Data Science

Summer Internship May' 25 – Aug' 25

> Kunal Sachdev GenAl Engineer Intern

AGENDA



- GenAl-powered Business Insights with Visualization
- Grievance Prediction Modelling Enhancement
 - Integrating QRC data
 - TOF Feature Analysis
- Mis-selling Prediction Modelling
- Acknowledgements





Data Genie – GenAl Powered Business Insights

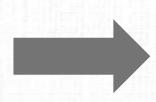
GenAI POC

GenAl-powered Business Insights with Visualization



To empower business leaders with cutting-edge GenAl tools for decision intelligence.

- 200+ BI Reports published daily
- C-suite Executives have to view the detailed reports and extract required insights manually.



A unified **GenAl-powered** chat interface to provide data insights in **natural language** from enterprise Data Lake (Dataverse).

Newly created Dataverse is the single source of truth for this project.

Converting **Hourly Login** report, as **Proof-of-concept** (POC), into conversational **GenAl-powered** insights – in collaboration with **AWS** and **Lumiq**.

User Inputs
Query

Question Intent
Decoding

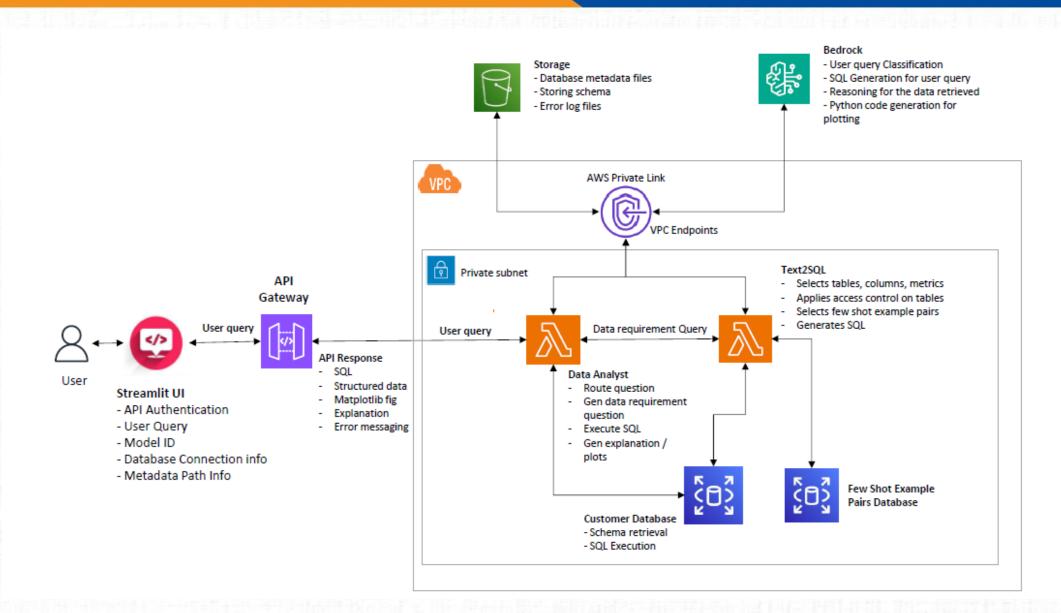
SQL Generation

Text/Graph
Output

Conversational Analytics with GenAl – Technical Architecture

Internal





Conversational Analytics with GenAl – Technical Flow



User Inputs Query

Question Intent Decoding

SQL Generation

Text/Graph
Output

- Streamlit Frontend
- API Authentication
- API Gateway
- Redshift DB configuration
- Model ID Claude 3.0Sonnet
- Metadata Info

- Validate User Input
- Decode Question Intent:

SQL or SQL + Plot

Establish DB connection

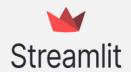
with SQLAlchemy

Engine

Extract Table Schema

- AWS Lambda QueryBot
- OpenSearch similarity search
- Prompt assembly –
 system prompt, SQL
 prompt, few-shot
 examples
- Bedrock Model Invoke

- Code rectification
- Row-level security (RLS)
- Generate natural language answer
- Display answer,SQL, plot















GenAl-powered Business Insights with Visualization – First Principles



- 1. Interface and Login:
 - Chat-bot interface
 - API-based interface
 - Mobile App/POS/Website In-app authentication and authorization
- Data privacy and security
 - Row-level security (Eg: A zonal head can access data only for that zone)
 - Masking/Removing access to PII
- 3. Simple, Intuitive UI
 - SSO enabled
 - User-friendly
 - Error handling for invalid prompts
- 4. Model Training and Accuracy

 - Feedback loop implementation auto and manual
 - Model Drift Detection over time

GenAl – Accomplishment & Way Forward



What is the channel-wise sales for the product Edelweiss Life - Wealth Ultima V0.2 yesterday?

According to the SQL query and the result table provided, the channel-wise sales for the 'Edelweiss Life - Wealth Ultima V0.2' product yesterday is as follows:

■ Data View ⇔						
	main_channel	subchannel	policy_count	premium_amount		
0	Agency	Organic Agency	2	160000		
1	Bancassurance	CSB	3	200000		
2	Bancassurance	Csb	1	300000		
3	Bancassurance	ESFB	1	50000		
4	Bancassurance	Esfb	1	100000		
5	Direct Sales	Direct Sales (PAN India)	1	100000		

 Demonstrated a GenAl-based chat interface to provide business insights for Hourly Logins with RLS.

Way Forward:

- Integration of Speech-to-Text for user input.
- User feedback loop for generated responses.
- Display frequently-used queries for bookmarking/favorites.
- Follow-up question suggestions.
- Next report for implementation Operational Reporting (Underwriting, Pending Inventory, etc.)





Grievance & Mis-selling Prediction Models

QRC Integration, TOF Feature Analysis, Mis-selling Prediction

Grievance Prediction – Integrating QRC Data



Integrating 49L+ customer queries, requests, and complaints (QRC) with existing Grievance model

- New Data Integrated:
 - A policy's holder's number of queries, and number of requests before their first complaint date.
 - QRC Data Sourced from D365 mid-2022 onwards.
- Machine Learning Model:
 - Training Data Period: Apr 2017 March 2024
 - Validation Data period: April 2024 July 2025
 - Model Trained:
 - 5 LightGBM Ensemble with different random samples
 - Average of top-3 predictions of the 5 models.
 - Model Predictions shared every week with Grievance Team





Top Decile
Coverage
increase from
76% to 97%

Grievance Prediction – Performance Comparison



New Grievance Model - Top Decile Coverage increase from 76% to 97%

Current Grievance Model: (Till 20-July-2025)

Decile	GRV#	Non GRV #	Total Policies	Population	GRV Cum
1	1691	82732	84423	10%	76.34%
2	168	84255	84423	10%	83.83%
3	126	84296	84422	10%	89.62%

Performance shown on all active policies (NCD/NBMIS records as of Jan' 25)

New Grievance Model: (Live from 21-July-2025 Onwards)

Decile	GRV#	Non GRV #	Total Policies	Population	GRV Cum
1	2759	80318	83077	10%	96.94%
2	35	83042	83077	10%	98.17%
3	19	83058	83077	10%	98.84%

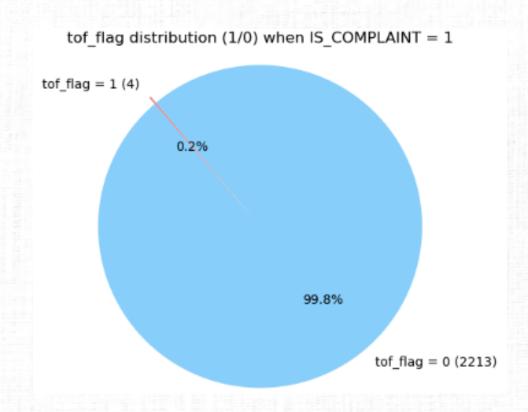
Performance shown on all active policies (NCD/NBMIS records as of July' 25)

Grievance Prediction – Transfer-of-Funds (TOF) Feature Analysis



Hypothesis: Transfer-of-funds (TOF) is an important predictor for grievance prediction

- Analysis done on Pure TOF cases.
- Chi-Square Test:
 - There is a statistically significant association between TOF and Grievance.
- Cramer's V Test:
 - The strength of said association is negligible it's too weak to be meaningful



Conclusion: TOF is not an important predictor for Grievance (at least for now)

Mis-selling Prediction – New Model



Training a new model to predict policy Mis-selling

Target variable:

- Whether a policy was mis-sold to a particular customer
- Includes QRC data in training.
- Machine Learning Model:
 - Training Data Period: Apr 2017 March 2024
 - Validation Data period: Apr 2024 July 2025
 - Model Trained:
 - 5 LightGBM Ensemble with different random samples
 - Average of top-3 predictions of the 5 models.
 - Model Predictions shared every week with Grievance Team





Model output:
Top Decile
Coverage of

90.56%

Mis-selling Prediction – Performance



Mis-selling Model – Top Decile Coverage of 90.56%

New Mis-selling Model: (Live from 21-July-2025 Onwards)

Decile	GRV#	Non GRV #	Total Policies	Population	GRV Cum
1	163	6781	6944	10%	90.56%
2	11	6933	6944	10%	96.67%
3	4	6940	6944	10%	98.89%

Performance shown on validation data (NCD/NBMIS records April' 24 – July' 25)

Significant Variables

Base Sum Assured

Premium (wo Tax)

Proposer Age, Policy Term

Query Count, Request Count

Significant Derived Variables

Historic Mean Age of Life Insured for LA Occupation Type for Grievance policies

Historic Mean Annual Premium for an Occupation Type

Historic Product Policy Count for Grievance Policies

Historic Region Grievance Rate





Acknowledgements

Acknowledgements



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- Mr. Nilanjan Chowdhury (Data Scientist II)
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HR

Rohini Shetty

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thank you!

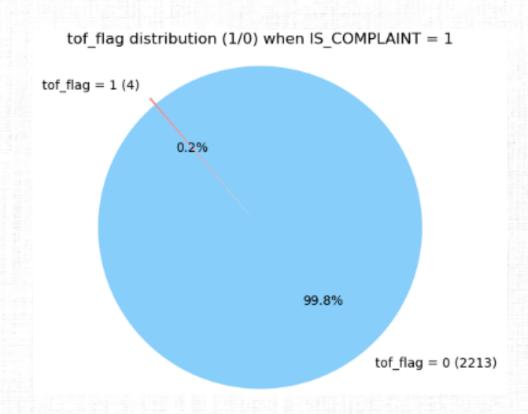
wish you a zindagi unlimited!

Grievance Prediction – Transfer-of-Funds (TOF) Feature Analysis



Hypothesis: Pure Transfer-of-funds (TOF) is an important predictor for grievance prediction

- Chi-Square Test: To test the association b/w two variables
 - P-value: 0.016 (< 0.05)
 - There is a statistically significant association between tof_flag and IS_COMPLAINT
- Cramer's V Test: Measures the strength of said association
 - Value: 0.0087 (very close to 0)
 - Extremely weak association between tof_flag and IS_COMPLAINT.



Conclusion: Pure TOF is not an important predictor for Grievance (at least for now)