GenAl Insights from Insurance Call Transcripts

Sentiment & Call Outcome Detection using Cohere LLM

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Project Overview

Goal:

Analyze member transcripts to determine:

- Sentiment (Positive, Neutral, Negative)
- Call Outcome (Issue Resolved, Follow-Up Needed)

Scope:

- 200 call transcripts
- Focus on **customer ("Member:")** lines only

Why It Matters

Business Value:

- Track customer satisfaction at scale Compared to Questionnaire (high cost, inefficient)
- Spot unresolved issues needing escalation Compared to manually logged and follow-up (slow & high missing rate)
- Improve agent training with real feedback Compared to standalone performance evaluation system (high cost)

Benefits:

- Data-driven insights
- Proactive issue resolution
- Better customer experience
- Reduce operational cost

Before and After using GenAl

<u>Use Case</u>	Before GenAl	After GenAl
Customer complaints	Detected manually (random QA)	Automatically flagged in real time
Agent coaching	Based on occasional reviews	Continuous feedback from every call
Monthly reporting	High effort, slow	Automated summaries, weekly insights
Escalation Tracking	Manual logs	NLP-driven resolution tagging

Data Source & Assumptions

Data:

- 200 . txt files, each a full call transcript
- Only "Member:" lines used for analysis

Assumptions:

- Transcripts are accurate and representative
- Speaker tags are consistent
- LLM outputs reflect customer sentiment

Tech Stack

Environment: (easy to handle)

- Python (in Google Colab) Excellent for DS to experiment
- Cohere LLM API (command-r-plus) abundant of GenAI model for selection, without the need to hold the models, which will be expensive

Libraries: (generic toolkits)

• pandas, cohere, matplotlib, seaborn, tqdm

Pipeline Overview:

- Load transcripts
- 2. Extract member lines
- 3. Use GenAl models for sentiment analysis & call-outcome classification
- 4. Aggregate and visualize

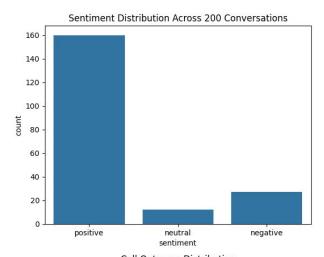
Key Results

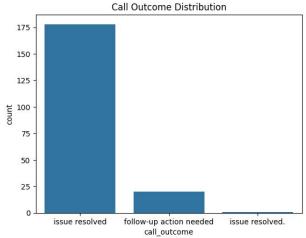
- Sentiment Distribution
- **Call Outcome Distribution**
- Cross Tab:
 - 89.5% issues resolved on-spot Mostly with positive sentiment
 - 10.5% issues need follow-up actions mostly with negative sentiment
 - 80% issues have positive sentiment mostly with issue resolved on-spot

90% vs 10% & 80% vs 20%

— Imbalanced and hiding the under-representatives

call-outcomes	negative	neutral	positive	Grand Total
follow-up action needed	8.00%	1.50%	1.00%	10.50%
issue resolved	5.50%	5.00%	79.00%	89.50%
Grand Total	13.50%	6.50%	80.00%	100.00%





Even Out

call-outcomes	negative	neutral	positive	Grand Total
follow-up action needed	76.19%	14.29%	9.52%	100.00%
issue resolved	6.15%	5.59%	88.27%	100.00%
Grand Total	13.50%	6.50%	80.00%	100.00%

Agent skills, processing time, tones, etc.

– Further investigation needed to improve sentiments.

•	For cases need for a follow-up:
	14.29% 'neutral' can be improved
	to 'Positive'

 For issue solved cases: it make sense to investigate the 6% 'negative' and 5.6% 'neutral

Sentiment	follow-up action needed	issue res	olved	Grand Total
positive	1.25%	98.75%		100.00%
neutral	23.08%	76.929	%	100.00%
negative	59.26%	40.749	%	100.00%
Grand Total	10.50%	89.509	6	100.00%

Issues contents, key concerning points, procedures, etc.

Further content investigation is needed.

- For 'neutral' clients: investigate
 why 76% with issue-resolved but
 still not happy ('positive')
- For 'negative' clients: investigate why 40% with issue-resolved but still angry ('negative')

Risks & Limitations

Model Risks:

- Hallucinations / misclassification
- Ambiguity in customer tone

Data Risks:

- Inconsistent formatting
- Missing speaker labels
- Potential transcription errors

Mitigation:

- Manual review sampling
- Flag uncertain outputs

Future Improvements

Next Steps:

- Fine-tune Cohere model for higher accuracy
- Include agent lines for fuller context
- Automate real-time monitoring of calls
- Include mega data for call transcript to understand more information, such as handling time

Advanced Ideas:

- Emotion tracking
- Conversation summarization
- Escalation prediction

Thank You