



OVERVIEW

At Comcast, customer support can cause major costs — especially when issues go unresolved.

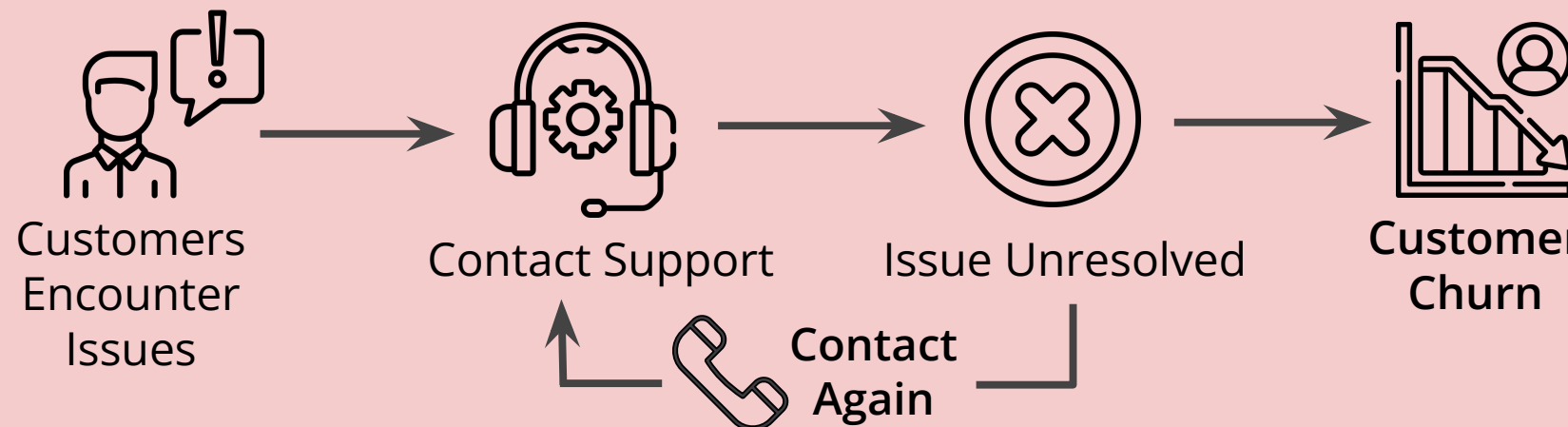
\$100M

Annualized cost of handling customer issue calls

16.7%

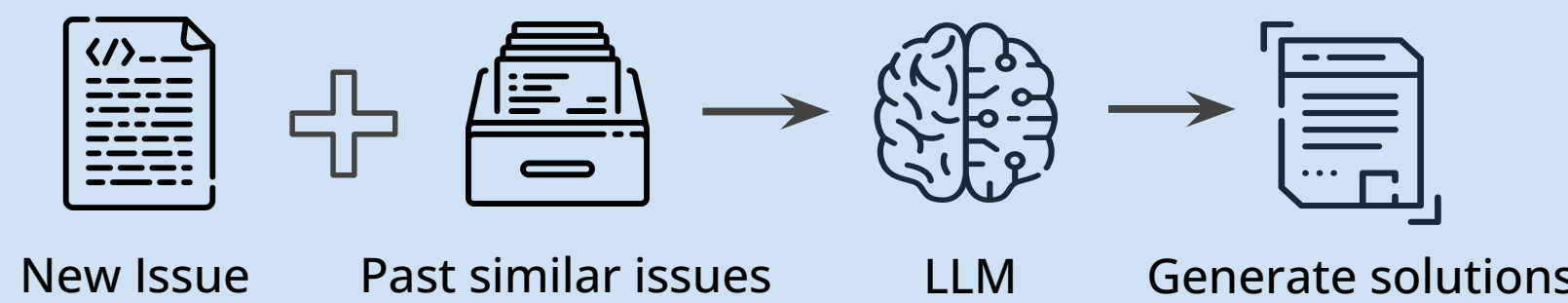
Share of total churn attributable to unresolved issues

Not resolving issues means more calls and lost customers.



PROBLEM

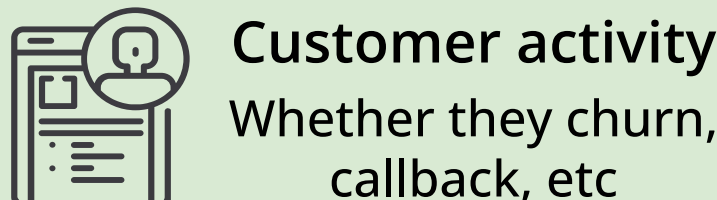
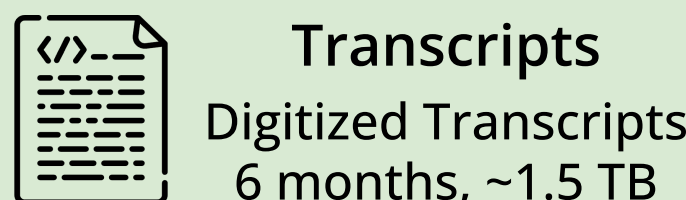
We build an AI engine (LLM + RAG) to generate better solutions.



GOAL

CONTEXT

So by leveraging historical data on how we solved issues:

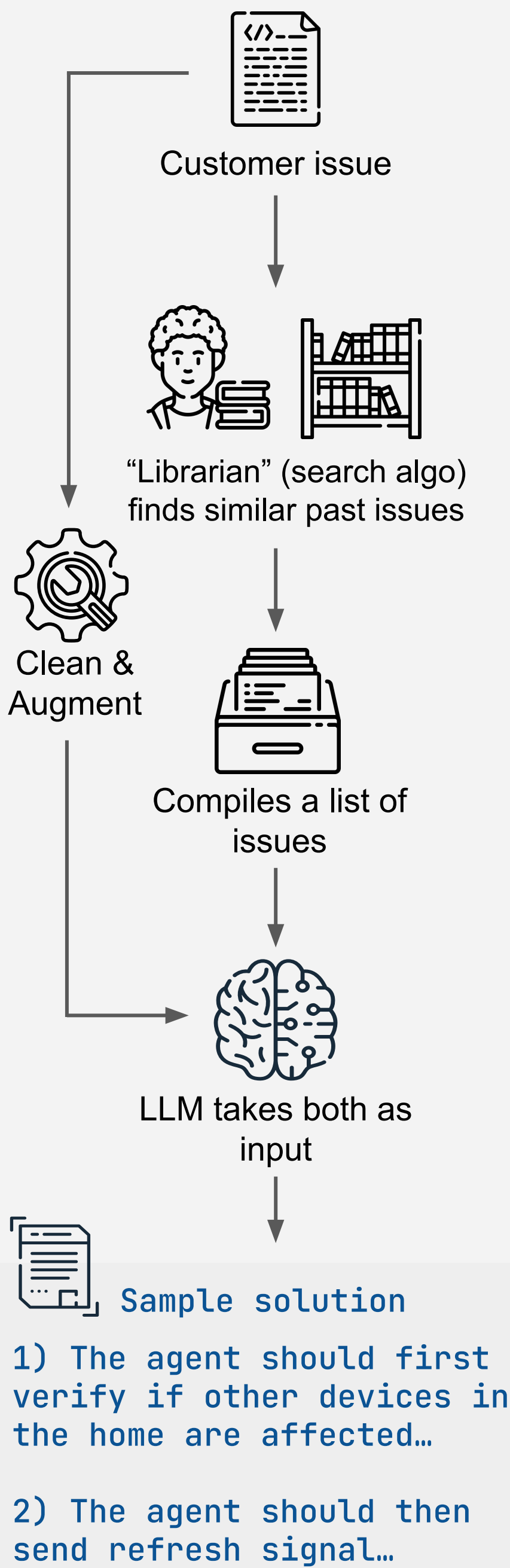


DATA

APPROACH

LLMs are powerful “learners” — they generate best when given relevant, high-quality information. So, we build them a “library” of past cases and train a smart “librarian” to retrieve best ones at runtime.

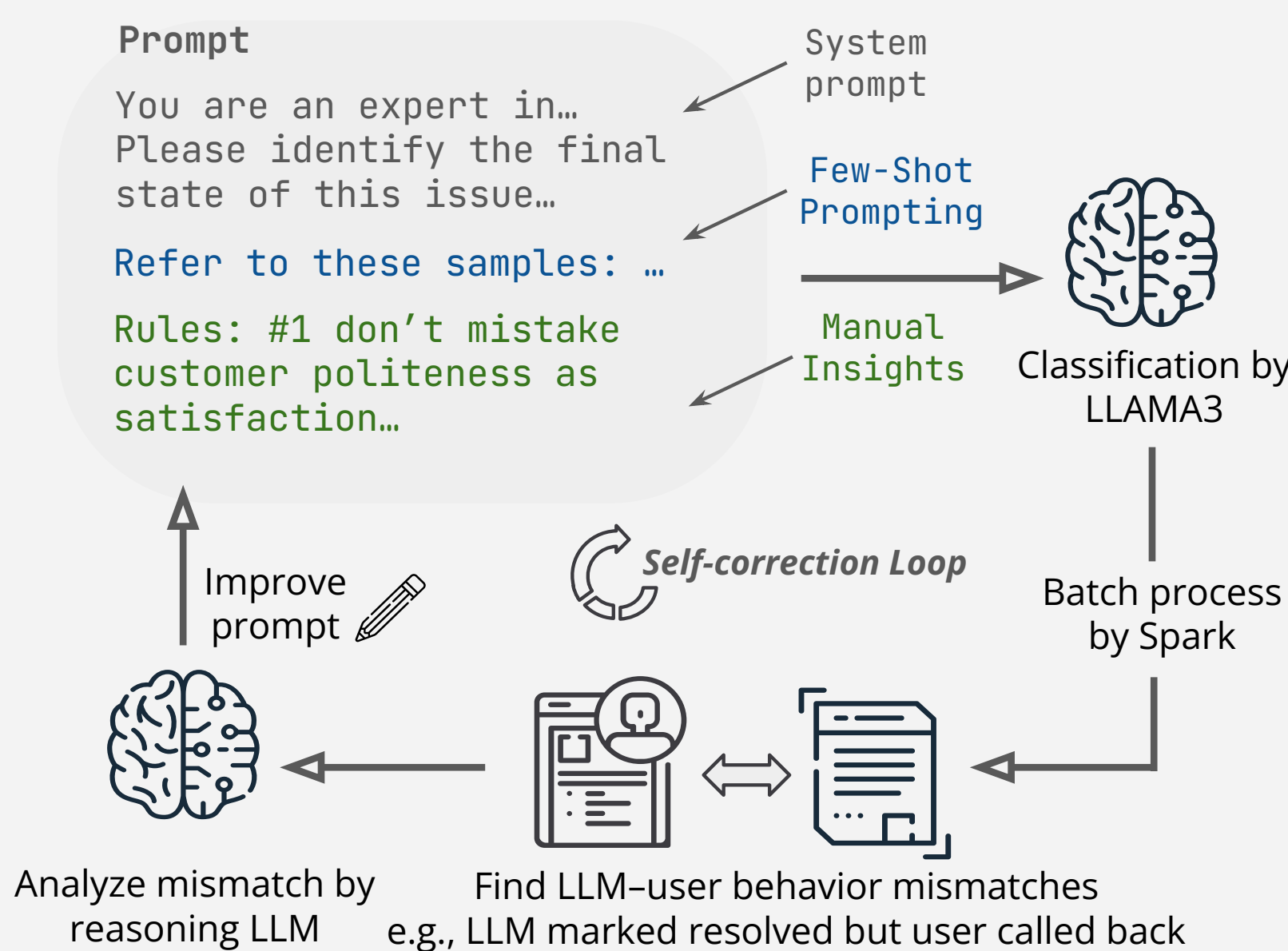
Pipeline^①



Build A Library^②

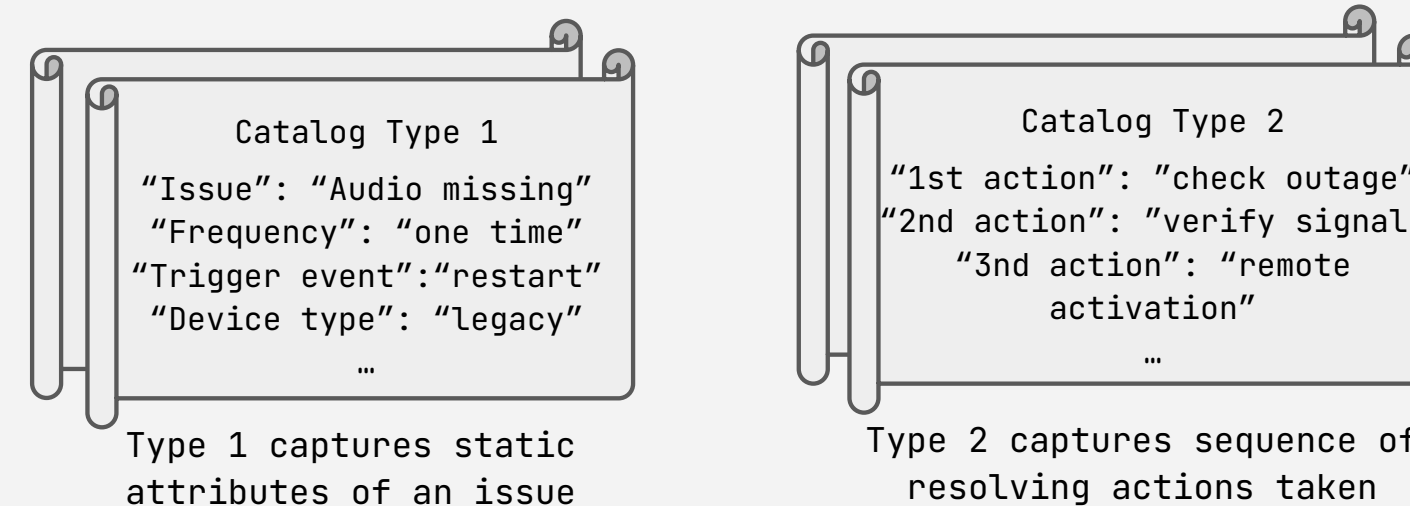
Curate the Shelves

Use LLM to distinguish between resolved and unresolved issues; retain mostly resolved ones to ensure quality.

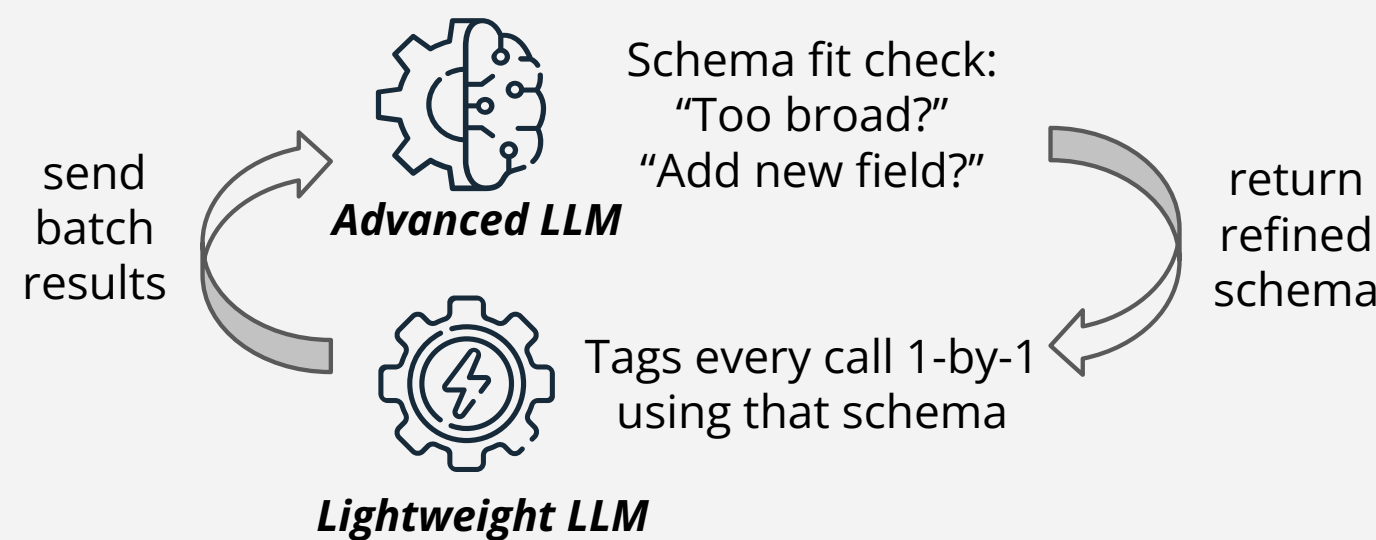


Design the Catalogs

Extract structured attributes from raw text. This structured catalog serves as a map for locating similar cases — which supports more precise retrieval later.



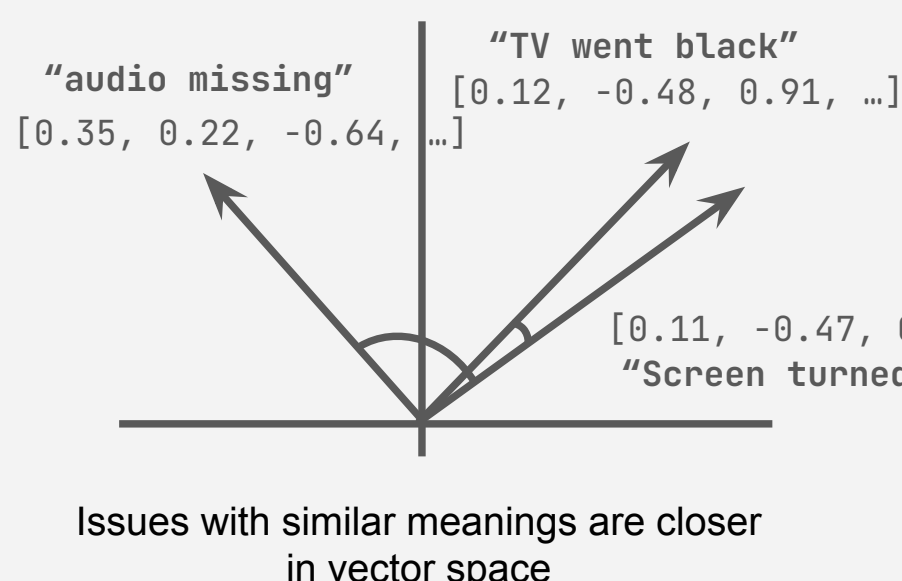
The schema of the catalog is generated in an iterative way:



Train A Smart Librarian^③

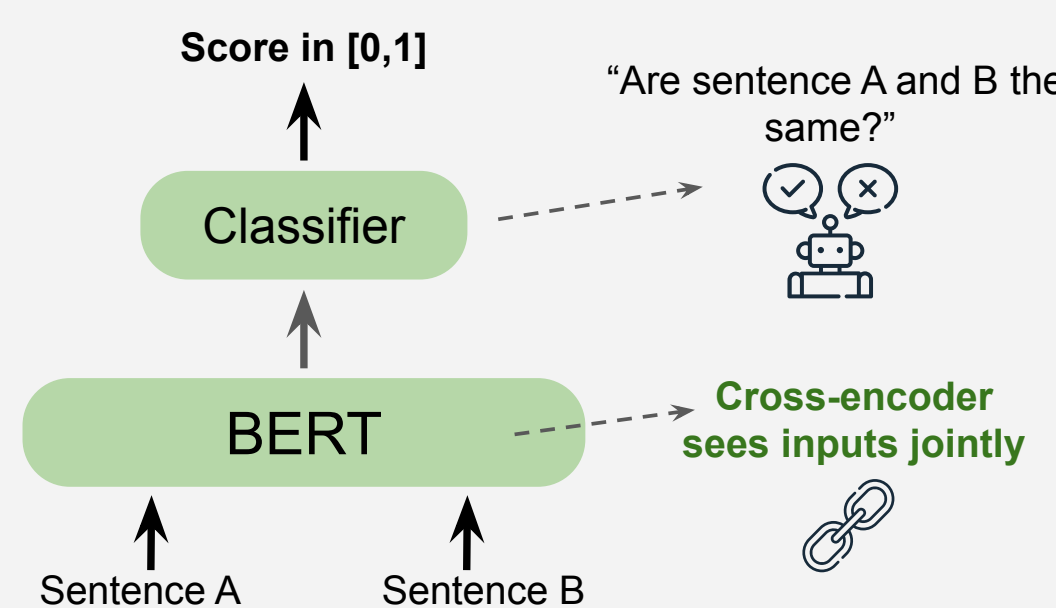
Utilize Vector Retrieval

Embed cases in vector space for a fast, first-pass retrieval based on meaning.



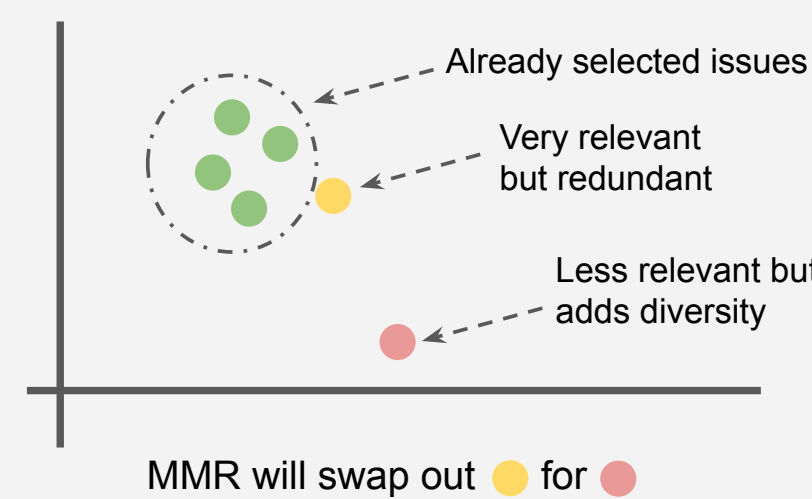
Rerank with Precision

Use cross-encoder and structured catalog matching together to surface the most relevant cases.



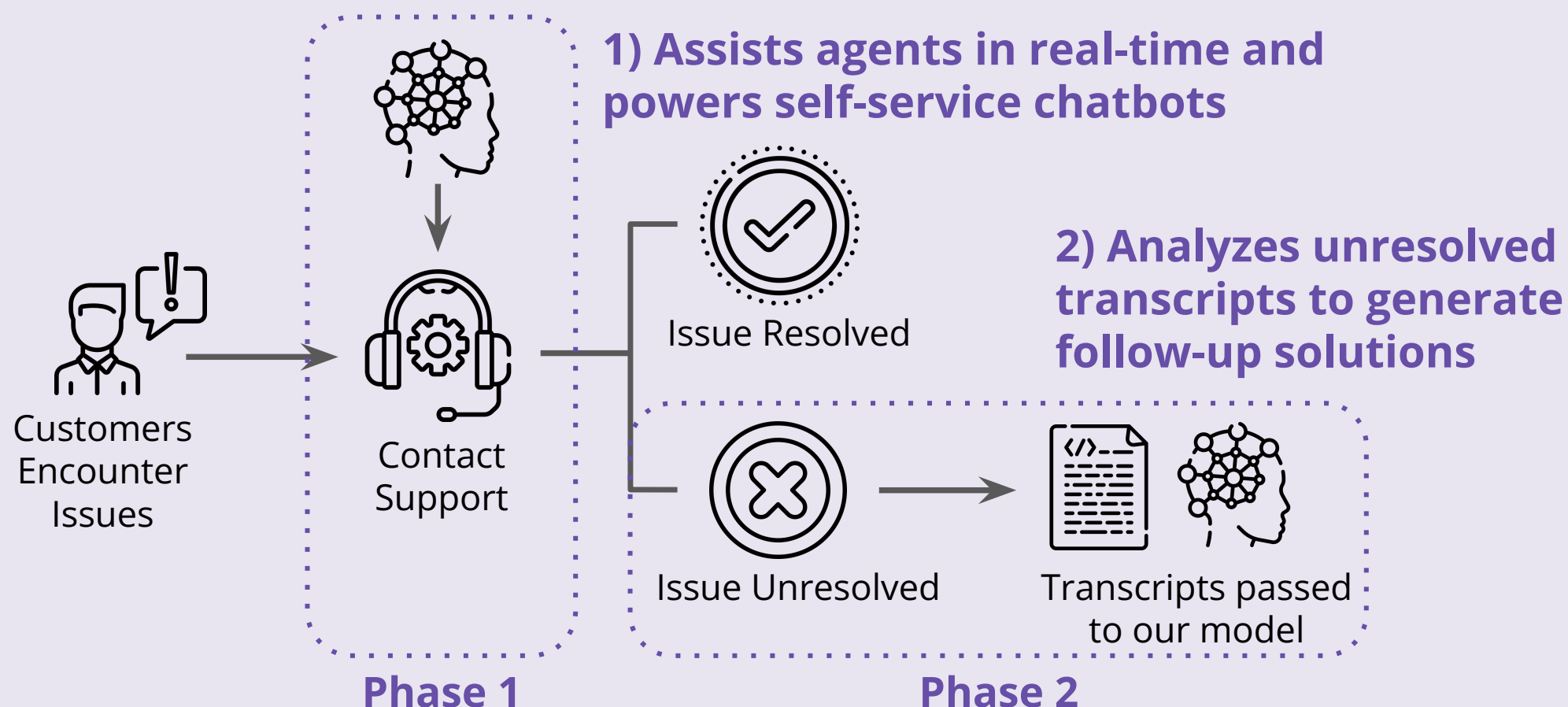
Diversify the Results

Add few unresolved cases for contrast and apply MMR (Maximal Marginal Relevance) to reduce redundancy.



USE CASES

Our AI engine can apply at two critical phases in the support flow:



RESULTS & IMPACT

By powering more accurate resolutions, our AI engine drives up retention and cuts costs:

+10pp

Increase in accuracy / resolution rate

+\$30M

Annualized savings from less calls and churns

We also created ecosystem values for the data science team:

