

Modeling Proposal and Procedures:

1. Generate labeled sample from Best Buy Employee subreddit data
 - a. Chunked individual comments and submissions into a list of sentences/contexts
 - b. Hand-wrote 3 questions, sampled 30 random contexts and 10 contexts close in embedding space (embedding model: multi-qa-mpnet-base-dot-v1)
 - c. Manually (human) labeled 40 contexts per question as relevant (True) or irrelevant (False)
 - d. Constructed a ranking based on the sum of relevant labels (Treat as our ground truth)
 - e. Resample contexts for each question
 - i. Take top 10 according to sum of relevant labels (Use this as our consensus True)
 - ii. Take 20 whose sum are 0 (use as our consensus False)
2. Evaluate retriever pipelines for quality across embedding models, similarity metrics, pipeline methodology
 - a. Procedure:
 - i. Pick an embedding
 - ii. Fill the retrieval system (database) with 30 contexts for each question
 - iii. Pick a similarity metric
 - iv. Retrieve 10 documents
 - v. Calculate precision, recall, f1 scores
 - b. Pipelines:
 - i. Retriever Class
 - ii. Chroma vector db implementation
 - iii. Qdrant vector db implementation
3. Evaluate pipeline for retrieval time
 - a. Procedure:
 - i. Fill pipeline with all chunked contexts for a given subreddit (Best Buy)
 1. Embed each context
 2. Add to database for retrieval
 - ii. For a given list of questions
 1. Run retrieval on question
 2. Calculate time to retrieve
4. Evaluate quality on advanced pipelines/procedures
 - a. Does clustering of embeddings offer any improvement?
 - b. Does self-querying retrieval (using an llm to split an input query into a filtering query for metadata and an embedded query) offer significant improvement?
 - c. How does multiquerying impact performance?
 - i. Classification of question itself
 - d. How does parent doc retrieval impact performance?
 - e. Does embedding on the summary of threads improve performance?
 - f. Can we account for temporal questions?