· CASE STUDY

HOME DEPOT

+

PROBLEM STATEMENT



Developing a content recommendation system that delivers highly relevant content (articles) by measuring the similarity between user search terms and content titles, ensuring precise and meaningful suggestions.

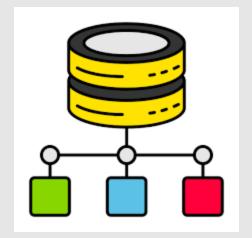
INFORMATION PROVIDED

Dataset

- content-data.csv: set of content titles and slug
- Label.csv: search term and article pairs (manually evaluated for relevancy)
- Test.csv: a set of search terms for which you are required to provide the content recommendations

Additional Info

We ignore user's past search behavior



METHODOLOGY USED

Core Objective

- Match user search terms with relevant content titles effectively.
- System returns highly relevant articles and minimizes irrelevant results.

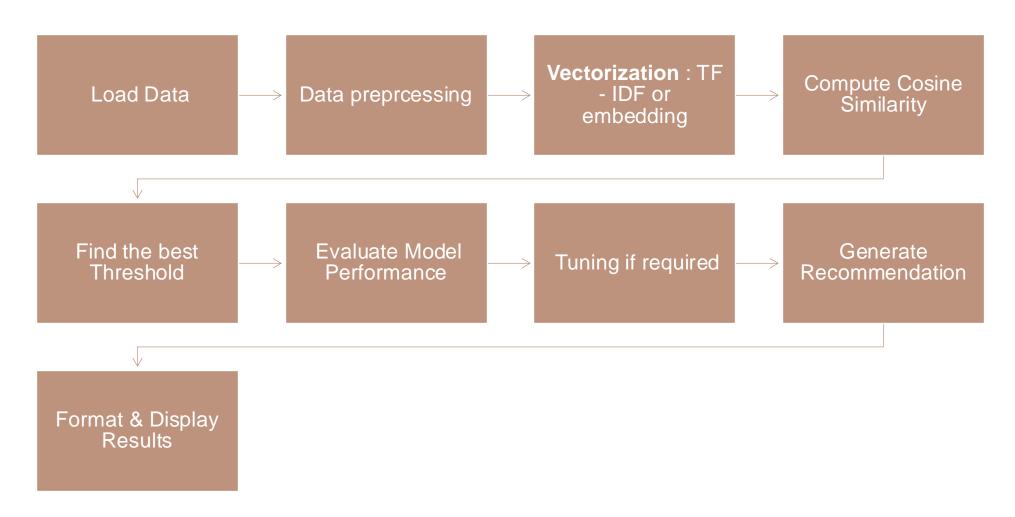
Key Components

- User search term
- Compute similarity between the search term and content titles.

Approach for Similarity Computation

- TF-IDF + Cosine Similarity
- Sentence Embeddings + Cosine Similarity
- Hybrid Approach : Combine both scores (e.g., 0.3 * TF-IDF + 0.7 * Embeddings Similarity).
- Sentence Embeddings+ FAISS + Rag





DETAILED FLOW

Data Cleaning

- 1.Missing values
- 2.Renamed columns
- 3. Merged content.csv and labels.csv

Data Preprocessing

- 1.Convert lowercase
- 2. Remove stop-words (spaces, punctuations)
- 3. Merged content.csv and

Vectorization

- 1.TF-IDF: converts text into numeric values Used for keyword search
- Sentence Transformer:
 Uses transformer all-MiniLM-L6-v2.
 Semantic meaning vectors

COSINE SIMILARIT

Term	Search Query	Title 1	Title 2
how	0.3	0.2	0.0
install	0.4	0.5	0.0
kitche n	0.5	0.5	0.4

Title 1 Similarity: 0.92 (Highly Relevant)

How Cosine Similarity is used in Recommendation System

- 1. After text is converted into Vector Representation
- Compute Cosine Similarity: Calculates similarity between user search from test and each content tile in labels data
- 3. Apply a Similarity Threshold : If **cosine similarity > threshold**, consider it **relevant**.

$$\cos(heta) = rac{A \cdot B}{||A|| imes ||B||}$$

THRESHOD SELECTION FOR CLASSIFICATION

Why a threshold?

- Classifies search -terms & article pairs as Relevant/ Not relevant
- Tested multiple threshold (0.0 1.0)
- Evaluated using F1 score for optimal performance

MODEL EVALUATION METRICS

TF-IDF

→	Best threshol	d: 0.35000000	000000003	with macr	o F1: 0.6078	363583981561
_		precision	recall	f1-score	support	
	NOT RELEVANT	0.55	0.28	0.37	43	
	RELEVANT	0.78	0.92	0.85	123	
	accuracy			0.75	166	
	macro avg	0.67	0.60	0.61	166	
	weighted avg	0.72	0.75	0.72	166	

Sentence Embedding

Best threshol	d: 0.4 with r precision		0.73507819 f1-score	98084903 support
NOT RELEVANT RELEVANT	0.58 0.87	0.65 0.84	0.62 0.85	43 123
accuracy macro avg weighted avg	0.73 0.80	0.74 0.79	0.79 0.74 0.79	166 166 166

MODEL EVALUALTION METRICS

Hybrid- Model

→ Best threshol	d: 0.33 with precision	macro F1: recall		883908622 support	
NOT RELEVANT RELEVANT	0.60 0.88	0.67 0.85	0.64 0.86	43 123	
accuracy macro avg weighted avg	0.74 0.81	0.76 0.80	0.80 0.75 0.80	166 166 166	
Search Term:	**34 in. to 3	36 in. x 72	2 in. show	er door**	

RECOMMENDATION TF-IDF



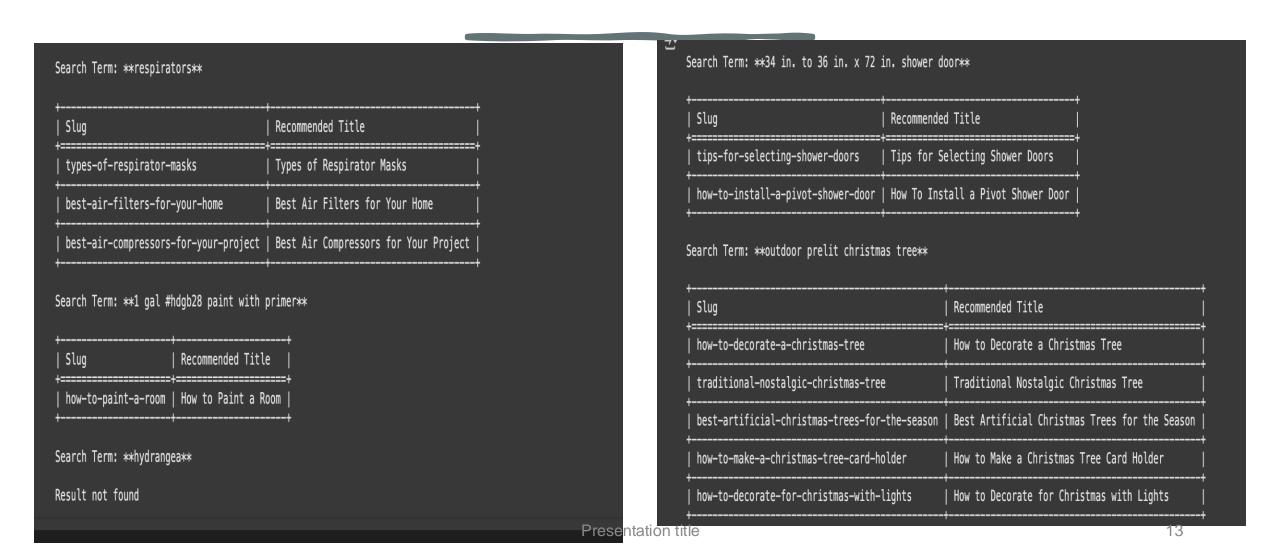
Search Term: **outdoor prelit christmas tree** Recommended Title how-to-decorate-a-christmas-tree How to Decorate a Christmas Tree traditional-nostalgic-christmas-tree Traditional Nostalgic Christmas Tree how-to-make-a-christmas-tree-card-holder | How to Make a Christmas Tree Card Holder how-to-fix-christmas-tree-lights How to Fix Christmas Lights outdoor-decor-ideas Outdoor Decor Ideas Search Term: **respirators** Result not found Search Term: **1 gal #hdgb28 paint with primer** how-to-paint-a-room | How to Paint a Room Search Term: **hydrangea** Presentation title found

how-to-repair-cracks-in-a-concrete-driveway | How to Repair Cracks in a Concrete Driveway

RECOMMENDATION SENTENCE EMBEDDING



RECOMMENDATION HYBRID MODEL



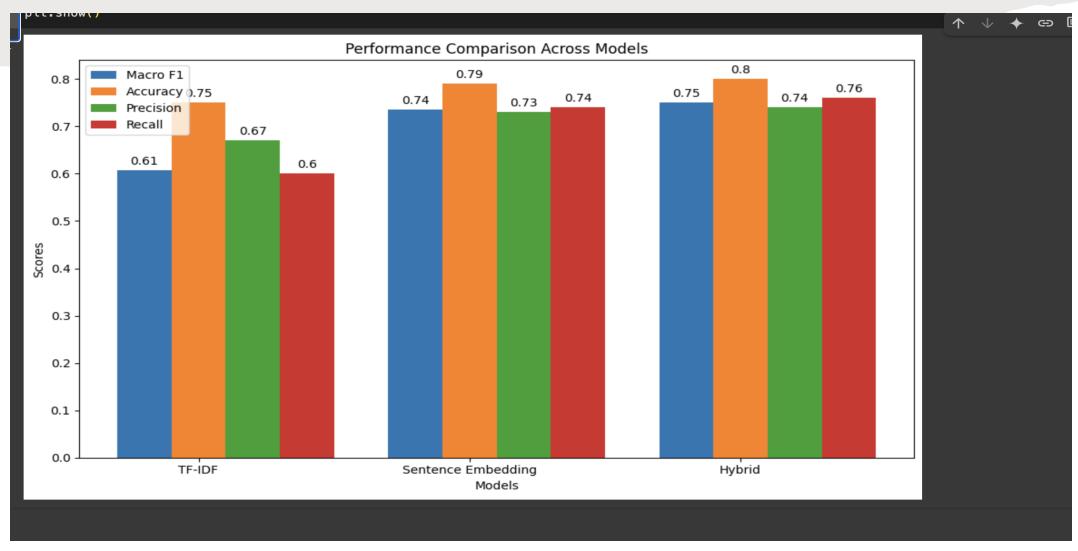
RECOMMENDATION FAISS+ RAG+ LLM



COMPARISION

Feature	TF-IDF	Sentence Embedding	Hybrid (TF-IDF + Embedding)	FAISS + LLM
Speed	Fastest	Slower (due to model inference)	Moderate	Fast (FAISS Optimized)
Semantic Understanding	Limited	Stronger semantic similarity	Best	Best with LLM support
Threshold Optimization	Required	Required	Required	Not needed
Performance	Decent for exact matches	Better for paraphrased quieries	Best overall	Best with contextual reasoning

VISUAL



KEY FINDINGS

Key Findings:

- TF-IDF is efficient but lacks deep semantic understanding.
- Sentence Embeddings provide more context-aware recommendations.
- Hybrid Model balances performance and accuracy.
- FAISS-Based Retrieval + LLM achieves the best combination of speed, accuracy, and relevance.

• Future Improvements:

- Experiment with more advanced transformer models (e.g., BERT, SBERT).
- Implement dynamic weight tuning for hybrid models.
- Enhance real-time search performance with optimized embeddings and FAISS clustering.
- Improve Data volume



1. Duplicate Results

• **Issue:** The same article (identified by its slug) appears multiple times in the recommendations.

Solution:

- Normalize slugs (lowercase, remove extra spaces).
- Use a set (seen_slugs) to track and filter out duplicates.

2. Weak Matches (Low Similarity Scores)

 Issue: Some recommendations have low similarity scores, leading to irrelevant or weak matches.

Solution:

- Apply a stricter similarity threshold (e.g., > 0.6).
- Improve preprocessing and model tuning to enhance relevance.

3. Low Recall (No Results Found)

• **Issue:** Some search terms return no recommendations due to insufficient matches.

Solution:

- Increase top_k (number of recommendations retrieved).
- Use query expansion techniques (e.g., synonyms) to broaden the search.

CHALLENGES

4. Dynamic Weight Tuning

- Issue: Hybrid models require manual tuning of weights for combining TF-IDF and embeddings.
- Solution:
 - Implement dynamic weight tuning based on dataset characteristics.
 - Experiment with different weighting schemes for optimal performance.

THANK YOU

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