PandaRec - A Recipe-Based Recommendation System for the Pandas Library

Frederick Vandermoeten

Overview

- Introduction
- Approach
- Evaluation
- Conclusion

Introduction: Pandas

Pandas

- Python library for data manipulation and analysis
- Works with DataFrames
- Existing recommendation systems don't work well with Pandas

Code Recommendation Systems

- Help developers to write code
- Design dimensions:
 - Input
 - Recommendation Engine
 - Output
 - User Feedback

Existing Code Recommendation Systems

- Code completion
 - Part of most IDEs
 - not that useful for Pandas
- Code generation
 - Boilerplate automation
 - Generative Al
 - ChatGPT
 - Copilot

Overview

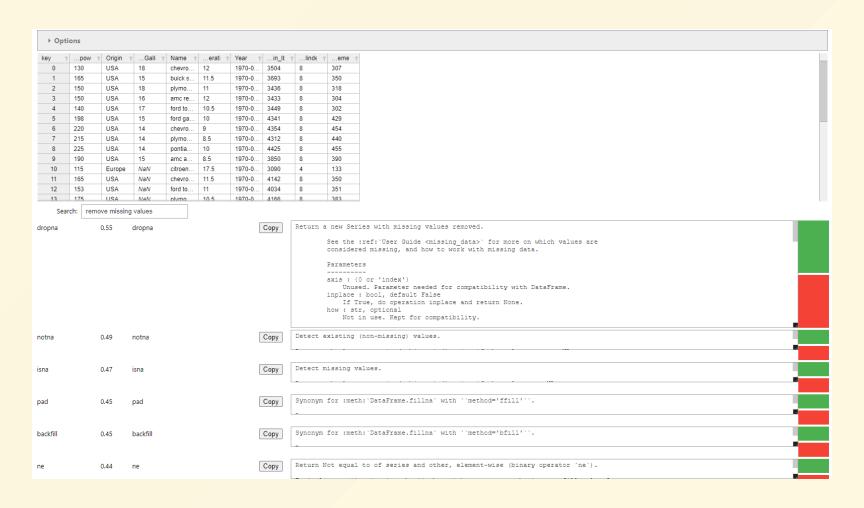
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Approach: My Solution

My Solution

- Jupyter Notebook
- Recipe-based recommendation system
- Swappable recommendation engine

Jupyter Notebook



Approach: Recipes

Recipes

- A data structure that describes a task
- Contains:
 - Name
 - Code snippet
 - Description
- saved in JSON format
- Generated from: dir() function, existing code snippets, handwritten

The Recommendation Engine

- Strategy pattern
- Gets current context
- Search function that returns a ranked list of recipes

Ranking Strategies: Lexical Search

- Name Search
- Fuzzy Search
 - Levenshtein distance
- Index Search
 - Lemmatize words
 - Build an inverted index

Ranking Strategies: Semantic Search

- Use a NLP model to calculate the similarity between the query and the recipe descriptions
- BERT: Bidirectional Encoder Representations from Transformers
- Sentence-BERT: Sentence Embeddings using Siamese BERT-Networks

Ranking Strategies: Other

- OpenAl Embeddings
- OpenAl Chat Completion
- Websocket

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Evaluation: Metrics

Metrics

- Speed:
 - Setup and search time
 - Under 100ms for search time is acceptable
- Accuracy:
 - NDCG

Evaluation: Metrics

NDCG

- Normalized Discounted Cumulative Gain
- Measures the ranking quality

$$DCG = \sum_{i=1}^k rac{gains}{log_2(i+1)}$$

$$NDCG = rac{DCG}{IDCG}$$

Last rank k is important: NDCG@k

Evaluation: Speed

Speed

Name	Setup Delay	Search Delay	
Name Search	68ns \pm 0.2ns	6.53μs \pm 0.08μs	
Fuzzy Name Search	68.4 ns \pm 0.5 ns	756μs \pm 1.3μs	
Fuzzy Description Search	276ns \pm 0.8ns	10ms \pm 0.05ms	
Index Search	7.87 s \pm 0.08 s	1.74ms \pm 0.01ms	
Semantic Search	85s \pm 1s	38.2ms \pm 0.9ms	
OpenAl Embedding	3 s ± 1 s	$668 \mathrm{ms} \pm 949 \mathrm{ms}$	
OpenAl Chat Completion	530 ns \pm 14.9ns	$27.4s\pm3.7s$	
Saved Index/ Embedding			
Index Search	1.5 ms \pm 0.03 ms	-	
Semantic Search	$682 \mathrm{ms} \pm 10 \mathrm{ms}$	-	
OpenAl Embedding	$27.5 \mathrm{ms} \pm 0.25 \mathrm{ms}$	-	

Evaluation: Accuracy

Accuracy Setup

Generate Test Dataset

Evaluation: Accuracy

Accuracy

Name	NDCG@5
Name Search	0.0
Fuzzy Name Search	0.41
Fuzzy Description Search	0.27
Index Search	0.22
Semantic Search	0.64
OpenAl Embedding	0.57

Evaluation: Accuracy

Accuracy on different recipe datasets

	docstring	cookbook	snippets
NameSearch	0.00	0.00	0.00
FuzzySearchName	0.41	0.45	0.53
FuzzySearchDescription	0.27	0.31	0.48
IndexSearch	0.22	0.40	0.29
SemanticSearch	0.64	0.71	0.80
OpenAIEmbeddings	0.57	0.44	0.88

Evaluation: Other

Other

- Jupyter Notebooks
 - Different Websites use different versions
 - Widget library doesn't support DataFrame traitlets
 - Easy to set up and use
- Usefulness of the recipes
 - Fast
 - All-in-one solution
 - Restricted by quality of recipes

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Conclusion: Summary

Summary

- Lexical search is not accurate
- Using recipes together with semantic search is the best approach
 - fast
 - accurate
 - can still be improved
- Jupyter Notebooks good for prototyping

Conclusion: Future Work

Future Work

- Improve the semantic search by fine-tuning the model
- Improve the recipe quality
- Get rid of most algorithms
- Port to standalone application or VSCode extension
- Easy to implement more ranking strategies