Amazon Sales Chatbot

Phase 2: Progress Report

Presented by

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Project Overview (summary)

- Our Amazon Sale Chatbot is a virtual assistant designed to enhance the shopping experience on the Amazon platform.
- Objectives
 - o to increase user engagement, and satisfaction on the Amazon platform using NLP.
- Target audience
 - Amazon customers

Our Data Source

Dataset information: to mention which dataset is used for which task???

 Amazon product dataset (extracted from meta data) -https://huggingface.co/datasets/McAuley-Lab/Amazon-Reviews-2023

Train dataset - https://github.com/amazonqa/amazonqa?tab=readme-ov-file

Classification Model

- We want to classify the question into 17 classes (Toys_and_Games, Health_and_Personal_Care, Cell_Phones_and_Accessories, Home_and_Kitchen, Musical_Instruments, Baby, Sports_and_Outdoors, Patio_Lawn_and_Garden, Video_Games, Pet_Supplies, Tools_and_Home_Improvement, Beauty, Electronics ,Grocery_and_Gourmet_Food, Automotive, Office_Products, Clothing_Shoes_and_Jewelry)
- We have trained 2 models (biLSTM and CNN) on different parameter with small (500 sample questions per product category) and large dataset (1000 sample questions per product category). Moreover, we saved the model when the validation loss is improve. Therefore, we select the best model with specific parameter based on minimum validation loss

biLSTM Model with small dataset

Run Name	Created <u>=</u> ↑	Duration	min_val_loss
biLSTM500-5-epochs-128-hidden dim-2-num layers	② 12 hours ago	1.2min	2.191
biLSTM500-5-epochs-128-hidden dim-4-num layers		1.9min	2.543
biLSTM500-5-epochs-128-hidden dim-6-num layers	② 12 hours ago	2.6min	2.71
biLSTM500-5-epochs-256-hidden dim-2-num layers		1.7min	2.835
biLSTM500-5-epochs-256-hidden dim-4-num layers		3.2min	2.978
biLSTM500-5-epochs-256-hidden dim-6-num layers		4.8min	3.262
biLSTM500-5-epochs-512-hidden dim-2-num layers		4.1min	3.328
biLSTM500-5-epochs-512-hidden dim-4-num layers		8.4min	3.626
biLSTM500-5-epochs-512-hidden dim-6-num layers		12.6min	3.029
biLSTM500-10-epochs-128-hidden dim-2-num layers		1.8min	3.255

biLSTM500-10-epochs-128-hidden dim-4-num layers		3.4min	3.36
biLSTM500-10-epochs-128-hidden dim-6-num layers		4.8min	3.165
biLSTM500-10-epochs-256-hidden dim-2-num layers		3.0min	4.14
biLSTM500-10-epochs-256-hidden dim-4-num layers		5.9min	3.917
biLSTM500-10-epochs-256-hidden dim-6-num layers		8.6min	3.696
biLSTM500-10-epochs-512-hidden dim-2-num layers		7.8min	4.771
biLSTM500-10-epochs-512-hidden dim-4-num layers		16.2min	4.34
biLSTM500-10-epochs-512-hidden dim-6-num layers	⊙ 10 hours ago	25.2min	3.966

For biLSTM, when we increase the epoch, hidden dimension, and number of layers, the model is worse.

biLSTM Model with large dataset

Run Name	Created <u>=</u> ↑	Duration	min_val_loss
biLSTM1000-5-epochs-128-hidden dim-2-num layers	⊗ 8 hours ago	2.6min	1.977
biLSTM1000-5-epochs-128-hidden dim-4-num layers	⊗ 8 hours ago	4.4min	2.257
biLSTM1000-5-epochs-128-hidden dim-6-num layers	Ø 7 hours ago	6.1min	2.55
biLSTM1000-5-epochs-256-hidden dim-2-num layers	⊘ 7 hours ago	4.0min	2.748
biLSTM1000-5-epochs-256-hidden dim-4-num layers	Ø 7 hours ago	7.4min	2.739
biLSTM1000-5-epochs-256-hidden dim-6-num layers	⊘ 7 hours ago	11.7min	2.887
biLSTM1000-5-epochs-512-hidden dim-2-num layers	7 hours ago	10.6min	3.19
biLSTM1000-5-epochs-512-hidden dim-4-num layers	⊘ 7 hours ago	21.9min	3.326
biLSTM1000-5-epochs-512-hidden dim-6-num layers	∅ 6 hours ago	35.8min	2.977
biLSTM1000-10-epochs-128-hidden dim-2-num layers	∅ 6 hours ago	4.4min	3.062

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biLSTM1000-10-epochs-128-hidden dim-4-num layers	∅ 6 hours ago	8.0min	3.212
biLSTM1000-10-epochs-128-hidden dim-6-num layers	∅ 6 hours ago	11.3min	2.975
biLSTM1000-10-epochs-256-hidden dim-2-num layers		7.5min	3.818
biLSTM1000-10-epochs-256-hidden dim-4-num layers		14.4min	3.736
biLSTM1000-10-epochs-256-hidden dim-6-num layers		21.6min	3.11
biLSTM1000-10-epochs-512-hidden dim-2-num layers		21.2min	4.36
biLSTM1000-10-epochs-512-hidden dim-4-num layers		41.1min	3.891
biLSTM1000-10-epochs-512-hidden dim-6-num layers	✓ 4 hours ago	1.0h	3.758

The biLSTM model is better when the dataset is larger

CNN Model

Small dataset

Large dataset

Run Name	Created ∓↓	Duration	min_val_loss
CNN500-30-epochs-150-n_filters	⊗ 8 hours ago	7.0min	2.099
CNN500-30-epochs-100-n_filters	⊗ 8 hours ago	6.4min	2.073
ONN500-30-epochs-50-n_filters	Ø hours ago	4.8min	2.076
ONN500-20-epochs-150-n_filters	Ø hours ago	4.9min	2.1
CNN500-20-epochs-100-n_filters	9 hours ago	4.5min	2.066
CNN500-20-epochs-50-n_filters	Ø hours ago	3.4min	2.111
CNN500-10-epochs-150-n_filters	Ø hours ago	2.5min	2.191
CNN500-10-epochs-100-n_filters	Ø hours ago	2.3min	2.202
CNN500-10-epochs-50-n_filters	Ø hours ago	1.8min	2.231

Run Name	Created =	Duration	min_val_loss
CNN1000-30-epochs-150-n_filters	Ø 7 hours ago	13.3min	1.871
ONN1000-30-epochs-100-n_filters	⊘ 7 hours ago	12.1min	1.867
ONN1000-30-epochs-50-n_filters	⊗ 8 hours ago	8.9min	1.898
ONN1000-20-epochs-150-n_filters	⊗ 8 hours ago	9.3min	1.886
ONN1000-20-epochs-100-n_filters	⊗ 8 hours ago	8.3min	1.897
CNN1000-20-epochs-50-n_filters	⊗ 8 hours ago	6.3min	1.927
ONN1000-10-epochs-150-n_filters	⊗ 8 hours ago	4.7min	2.007
ONN1000-10-epochs-100-n_filters	⊗ 8 hours ago	4.2min	2.028
ONN1000-10-epochs-50-n_filters	⊗ 8 hours ago	3.3min	2.042

For CNN, we can conclude that when the number of filters, epoch, amount of data increase, the model is better

Classification Model

Based on minimum validation loss, CNN with 30 epochs, 150 number of filters is our best model and we trained the model with a larger dataset (2000 dataset per category)

```
test_loss, test_acc = evaluate(model, test_loader, criterion, test_loader_length)

print(f'Test Loss: {test_loss:.3f} | Test Acc: {test_acc*100:.2f}%')

✓ 1.1s

Test Loss: 1.880 | Test Acc: 44.22%
```

We have tested model with test set and the model got around 44% accuracy

Language Model

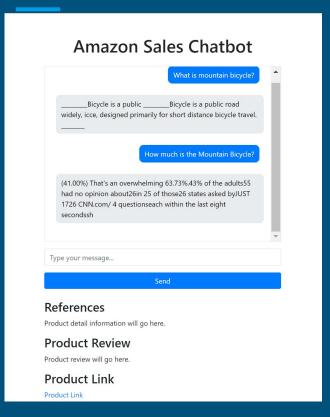
The product information from the meta datafiles are extracted and split into sample sizes of 100, 500 and 1000 products per category for experimentation.

The data samples are then stored as separate vector stores, and the application will determine which vector store the langchain should retrieve from based on the result of the classification model

Design test cases

- Basic functionality testing
 - Verify that the chatbot can understand and respond to basic user queries.
- Product information retrieval testing
 - Ensure that the chatbot can provide accurate information of products from different categories.
- Recommendations testing
 - ensure that the chatbot can provide the recommendation based on different categories of products
- Error handling testing
 - To test that chatbot can handle error and unexpected question

Web Design



Amazon Sales Chatbot

Chat Interface:

- Functionality:
 - Allows users to type messages (Type your message...), send them, and receive responses from the chatbot.
 - Messages are displayed in a visually distinct format, with user messages on the right and bot responses on the left.

Additional Sections:

- References:
 - Contains detailed information about the product.
- Product Review:
 - Displays product review content for user reference.
- Product Link:
 - Provides a link for users to access more detailed product information.

Our Project GitHub Repository

https://github.com/minnbanya/NLP-Project

Thank you For your attention