Code.Black()

NM396: SENTIMENT ANALYSIS AND CUSTOMER REVIEW RATING

Team Structure

FrontEnd

Akansha Singh

Paridhi Gupta

BackEnd



Kanishk Singhal

Business Logic / ML Model



Udit Singh

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Kushagra Shukla

(Team Leader)

Basic Functionality

INPUT TYPES

Single Review

Bulk Review (csv file)

OUTPUT TYPES

Rating of 1 to 5

Tabular + Graphical O/P

User Rating, Review Text, Model Sentiment Rating, Avg. Product Rating

Note:

1 -> Worst (High Negative)

5 -> Best (High Positive)

Rating output by model on the scale of 1 to 5 represents degree of the sentiment

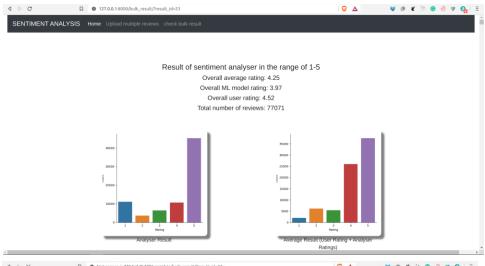
Processing

Input

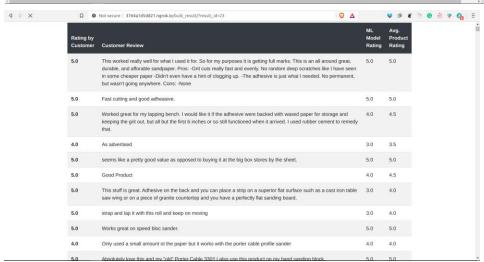
Input



csv file (with 77k reviews



O/P



Write something...

Rtx2080 with Intel i9 processor and on top of that 144hz display panel. Only down side with this machine is it's speakers and design. Apart from that, on performance it's a beast. I would totally recommend it if you are not into aesthetics that much.

Result of sentiment analyser in the range of 1-5

Sentence provided: Rtx2080 with Intel i9 processor and on top of that 144hz display panel. Only down side with this machine is it's speakers and design. Apart from that, on performance it's a beast. I would totally recommend it if you are not into aesthetics that much.

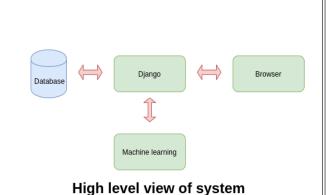
Overall average rating: [4]

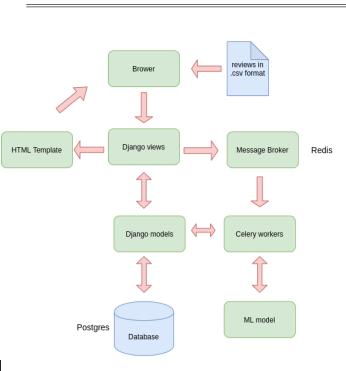
Note:

csv file structure
Row 1 (headers):
review_text , user_ratings
All other rows:
<text>, <digit in [1,5]>

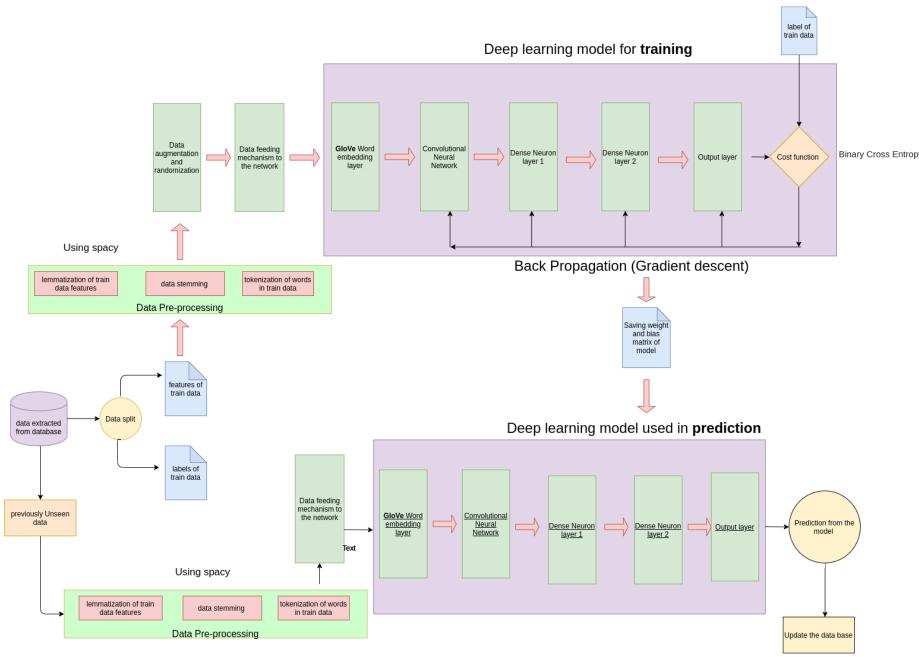
	A	В	С	D	Е
1	review_body	star_rating			
2	Excellent!!!	5			
3	Great quality wooden track (better than some other	5			
4	Cards are not as big as pictured.	2			
5	my daughter loved it and i liked the price and it cam	5			
6	Do not buy these! They break very fast I spun then f	1			
7	Great item. Pictures pop thru and add detail as "	5			
8	To keep together, had to use crazy glue.	3			
9	I was pleased with the product.	5			
10	Children like it	5			
11	Showed up not how it's shown . Was someone's old	1			
12	Really liked these. They were a little larger than I tho	5			
13	Nice huge balloon! Had my local grocery store fill it	5			
14	Great deal	5			
15	As Advertised	4			
16	Comes w a 15\$ servo so expect to spend 150 more of	3			
17	awesome ! Thanks!	5			
18	I got this item for me and my son to play around wit	5			
19	It was a birthday present for my grandson and he LC	5			
20	Got a wrong product from Amazon Vine and unable	3			
21	You need expansion packs 3-5 if you want access to	1			
22	Awesome customer service and a cool little drone! E	5			
23	I got these for my daughters for plane trip. I liked the	4			

Under the Hood





Django Backend Architecture



Data flow through our ML model

We do things differently

We use CNN. Why?

- Less runtime after training
- Less weights to be trained
- Model size is relatively small for same accuracy
- Work on low power machines.

We use Job Queueing. How & Why?

- Celery for an asynchronous task queue based on distributed message passing
- Redis as blazing-fast message broker
- We need not to wait because somebody else uploaded a csv file with 1 lakh entries.

Result?

93 % accuracy



We can handle multiple requests simultaneously

Model in the making

Training Metrics

```
de + Text
     print(| '\t vat. Loss: {vatiq_toss:.3|} | vat. Acc: {vatiq_acc*ivo:.2|}%')
 Epoch: 01 | Epoch Time: 1m 48s
         Train Loss: 0.217 | Train Acc: 86.53%
          Val. Loss: 0.170 | Val. Acc: 82.25%
 Epoch: 02 | Epoch Time: 1m 47s
         Train Loss: 0.130 | Train Acc: 89.65%
          Val. Loss: 0.164 | Val. Acc: 88.62%
 Epoch: 03 | Epoch Time: 1m 47s
         Train Loss: 0.112 | Train Acc: 92.43%
          Val. Loss: 0.150 | Val. Acc: 90.32%
 Epoch: 04 | Epoch Time: 1m 47s
         Train Loss: 0.096 | Train Acc: 94.64%
          Val. Loss: 0.135 | Val. Acc: 92.29%
 Epoch: 05 | Epoch Time: 1m 46s
         Train Loss: 0.079 | Train Acc: 96.60%
          Val. Loss: 0.123 | Val. Acc: 93.93%
```

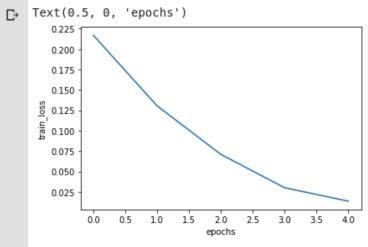
Testing Metrics

```
[53] model.load_state_dict(torch.load('tut5-model2.pt'))
    test_loss, test_acc = evaluate(model, test_iterator, criterion)
    print(f'Test Loss: {test_loss:.3f} | Test Acc: {test_acc*100:.2f}%')

Test Loss: 0.119 | Test Acc: 93.18%
```

Loss Graph

```
[52] import matplotlib.pyplot as plt
   plt.plot(loss_list)
   plt.ylabel('train_loss')
   plt.xlabel('epochs')
```



It was not always at 93 %

- We changed the training data size from 80k to 4 lakh
- Introduced learning rate decay
- Switched from 100 dimension embedding to 200 dimension of GloVe so that we have more trainable paramteres.
- Introduced dropout regularization
- Increased embedding vocabulary size from 25k to 1 lakh.

And of course hours of hard work by the team:)

Tech Stack

ML Model

- Pytorch
- GloVe
- Spacy
- Torchtext
- Pandas

O PyTorch

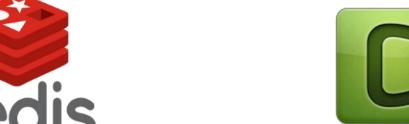
BackEnd

- Django
- Celery
- Redis
- PostgreSQL

FrontEnd

- Bootstrap
- HTML
- CSS
- JavaScript







Summary

- 93 % accuracy with ML model.
- Bulk reviews handled via job queueing using Celery + Redis
- Tested 77k reviews under 600 seconds of CPU time using ngrok.
- Compatible with both GPU and CPU
- Can be easily integrated with the BHUVAN database.
- Can be used via Mobile devices as well. (responsive web design)
- API support can also be provided easily using Django REST