Carnegie Mellon University

Productionizing a Recommendation System for a Large Scale use

Team4: The-land-of-harry-potters-amazing-endgame-part-2

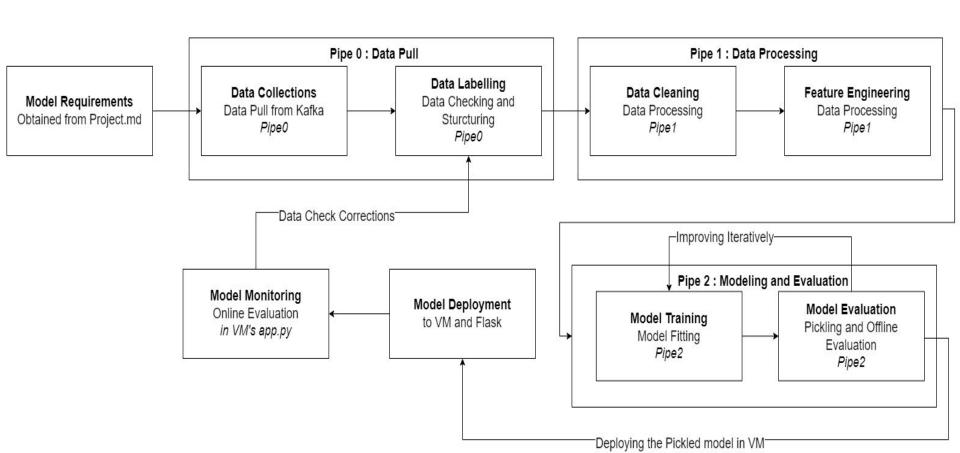
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17-645 Machine Learning in Production Fall 2022

Agenda

- Design and Reflection
 - Infrastructure
 - Software
 - Model
 - o Teamwork

Pipeline



Asynchronicity

- Some code for telemetry gathering is executed when a recommendation request is received
- Asynchronous reading/writing speeds up response time, prevents blocking

```
import asyncio

async def main():
    print('Hello ...')
    await asyncio.sleep(1)
    print('... World!')

asyncio.run(main())
```

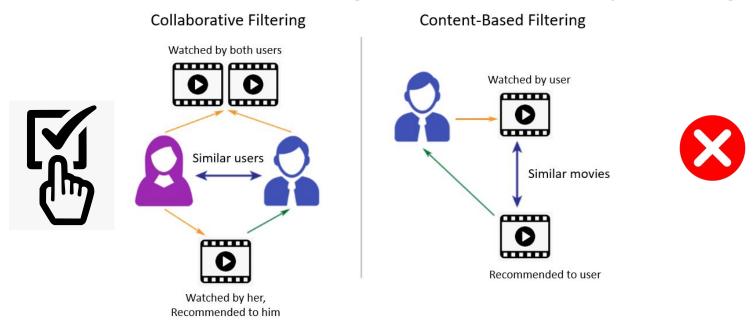
Infrastructure challenges and learnings

- Single point of failure Load balancer
- Log file disaster separate data and application
 - Scalability
 - Maintainability
 - Security



Model Design Choices

We choose Collaborative Filtering as the recommender system design



Model Design benefits

Benefits of Collaborative Filtering as compared to Content Based



Personalization



Scalability



Improved diversity

So many options of Collaborative Filtering! SVD?KNN?SVDPP?NMF?!!



Model Design Challenges



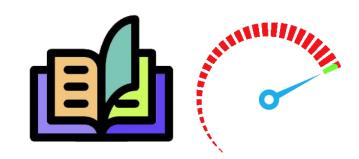
Dilemma of Time Complexity vs Space Complexity

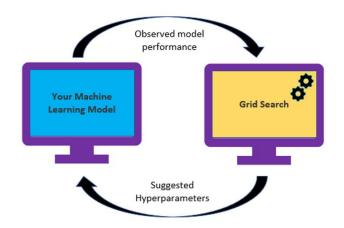
Movielens 1M	RMSE	MAE	Time
SVD	0.873	0.686	0:01:07
SVD++ (cache_ratings=False)	0.862	0.672	0:41:06
SVD++ (cache_ratings=True)	0.862	0.672	0:34:55
NMF	0.916	0.723	0:01:39
Slope One	0.907	0.715	0:02:31
k-NN	0.923	0.727	0:05:27

SVD has the best time and space complexity for 1M+ Data

Interesting Modelling Design Choice Outcomes

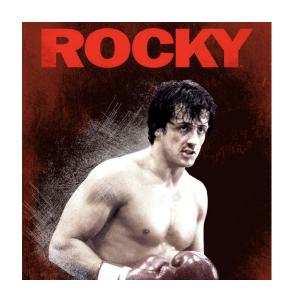
Hashmaps/Dictionary based storage retrieval can **drastically improve model inference times**





Grid Search for Hyperparameter tuning helped **improve Model Performance substantially**

Teamwork





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Thank you
Open to Questions