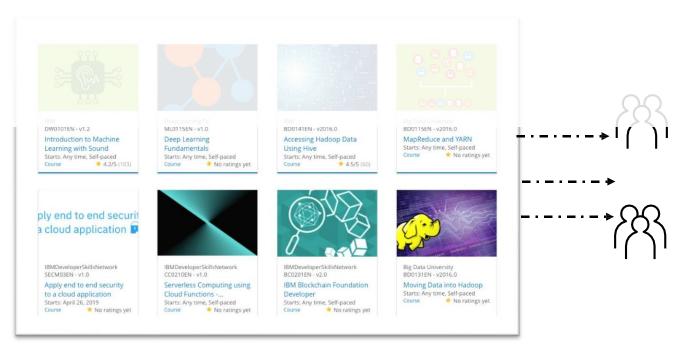
## Build a Personalized Online Course Recommender System with Machine Learning

Amenallah Bouhachem 09-05-2023



#### Outline

- Introduction and Background
- Exploratory Data Analysis
- Content-based Recommender System using Unsupervised Learning
- Collaborative-filtering based Recommender System using Supervised learning
- Conclusion
- Appendix

#### Introduction

- A course recommendation system will help in:
  - Finding better courses
  - Finding courses that well suits each person's interests
  - We aim to find the best courses to recommend to users based on their interests, their friend's interests, and the courses they are enrolled in.

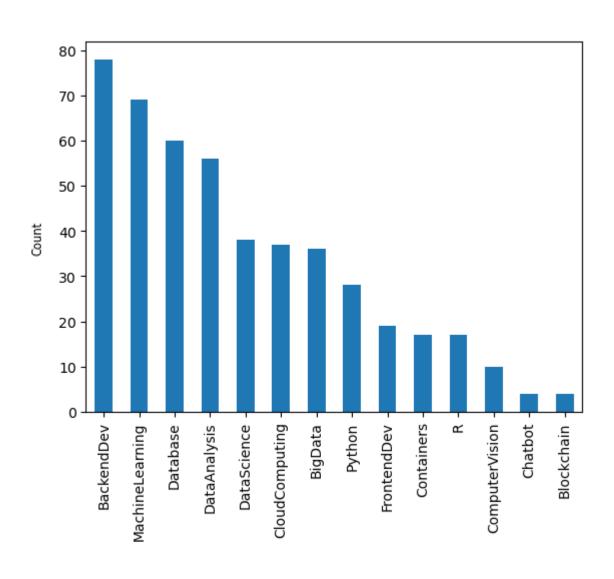
#### Obstacles

- We have many approaches
- Each approach has different assumptions

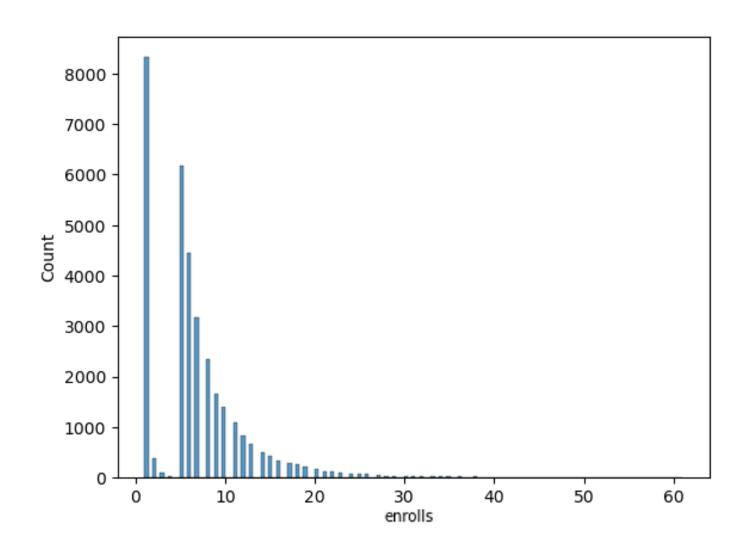
## **Exploratory Data Analysis**



### Course counts per genre



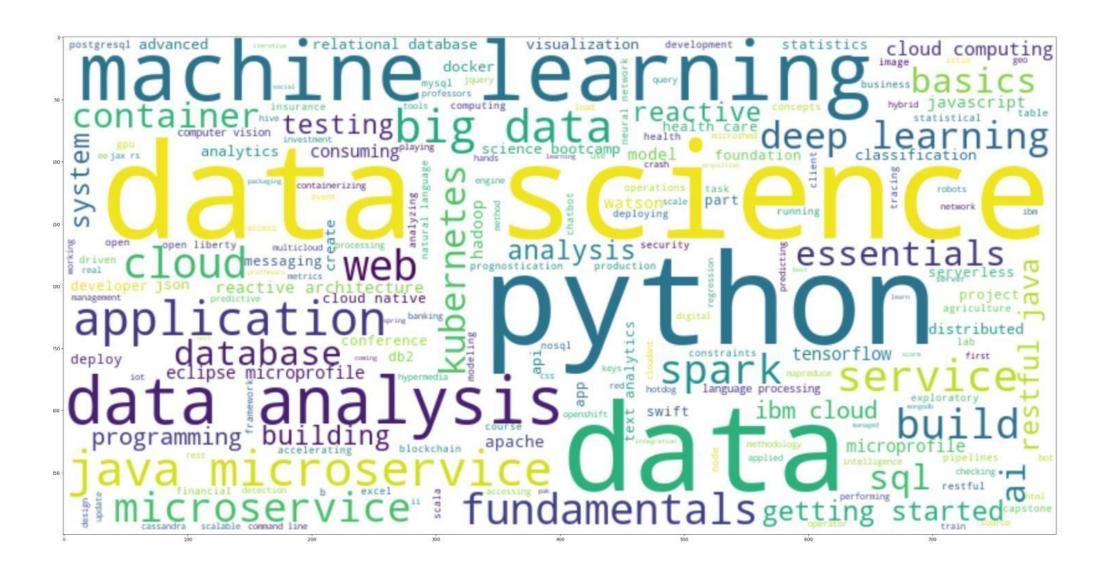
#### Course enrollment distribution



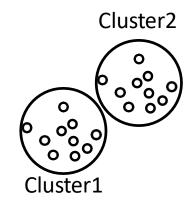
## 20 most popular courses

	TITLE	Enrolls
0	python for data science	14936
1	introduction to data science	14477
2	big data 101	13291
3	hadoop 101	10599
4	data analysis with python	8303
5	data science methodology	7719
6	machine learning with python	7644
7	spark fundamentals i	7551
8	data science hands on with open source tools	7199
9	blockchain essentials	6719
10	data visualization with python	6709
11	deep learning 101	6323
12	build your own chatbot	5512
13	r for data science	5237
14	statistics 101	5015
15	introduction to cloud	4983
16	docker essentials a developer introduction	4480
17	sql and relational databases 101	3697
18	mapreduce and yarn	3670
19	data privacy fundamentals	3624

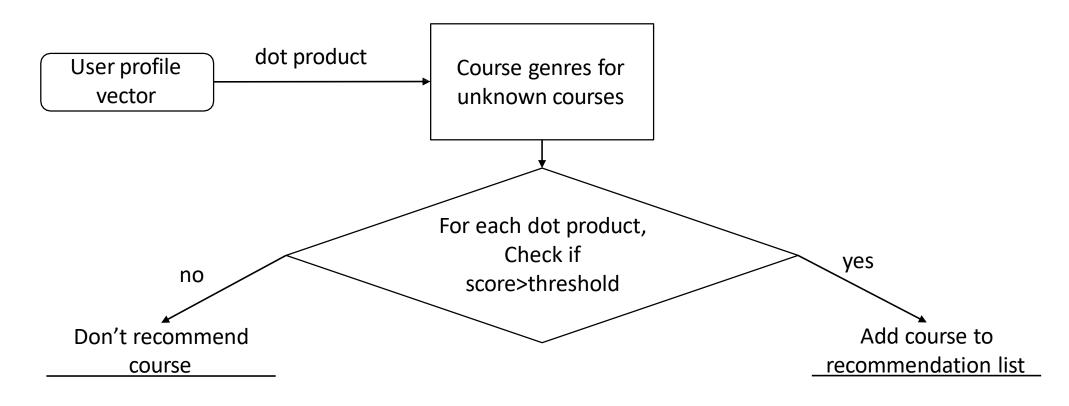
#### Word cloud of course titles



## Content-based Recommender System using Unsupervised Learning



# Flowchart of content-based recommender system using user profile and course genres



## Evaluation results of user profile-based recommender system

Score\_threshold = 10.0

On average, how many new/unseen courses have been recommended per user (in the test user dataset)

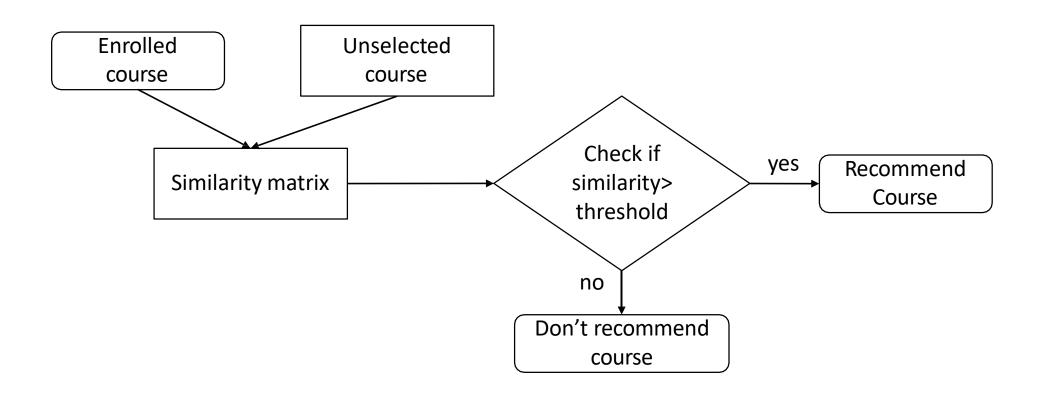
```
res_df['SCORE'].mean()
```

18.62679972290352

What are the most frequently recommended courses? Return the top-10 commonly recommended courses across all users

```
COURSE ID
TA0106EN
              608
GPXX0IBEN
              548
excourse22
              547
              547
excourse21
ML0122EN
              544
              533
excourse06
              533
excourse04
              533
GPXX0TY1EN
              524
excourse31
              516
excourse73
```

# Flowchart of content-based recommender system using course similarity



## Evaluation results of course similarity based recommender system

Threshold = 0.6

On average, how many new/unseen courses have been recommended per user (in the test user dataset)

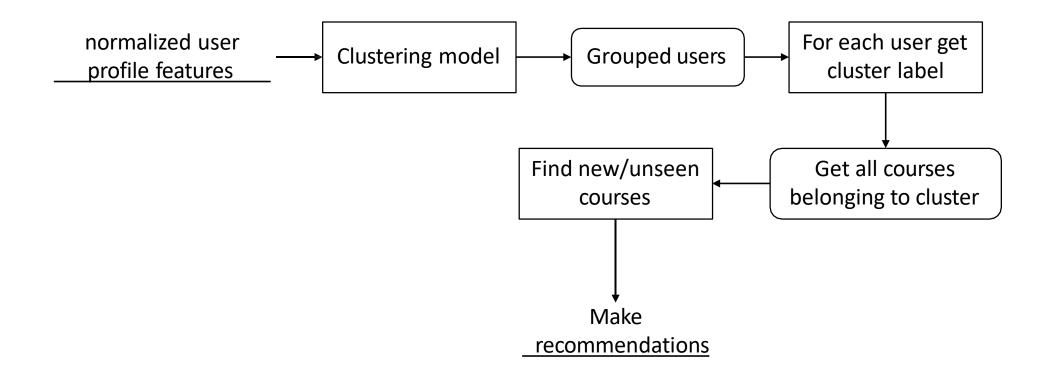
```
s = 0
for i in range(len(res_df['COURSE_ID'])):
    s+=len(res_df['COURSE_ID'].iloc[i])
avg = s/len(res_df['COURSE_ID'])
```

```
avg
11.377
```

What are the most frequently recommended courses? Return the top-10 commonly recommended courses

```
579
excourse22
excourse62
              579
DS0110EN
              562
excourse65
              555
excourse63
              555
excourse72
              551
              550
excourse68
              539
excourse67
excourse74
              539
              506
BD0145EN
```

## Flowchart of clustering-based recommender system



## Evaluation results of clustering-based recommender system

Number of clusters = 20

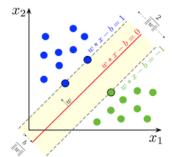
On average, how many new/unseen courses have been recommended per user (in the test user dataset)

```
s = 0
for r in user_recommendations.value
    s+=r[1:].sum()
avg=s/len(user_recommendations)
print(avg)
5.733
```

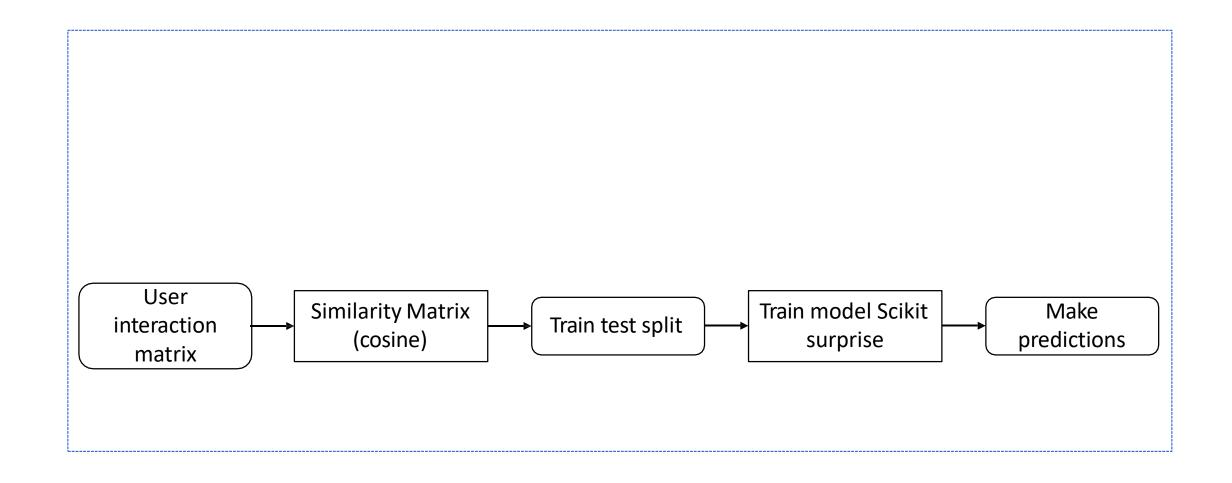
What are the most frequently recommended courses? Return the top-10 commonly recommended courses

```
579
DS0103EN
DA0101EN
              532
BD0111EN
              456
DS0101EN
              444
BD0101EN
              428
              386
PY0101EN
DS0105EN
              319
ML0101ENv3
              299
              296
BC0101EN
              286
ML0115EN
```

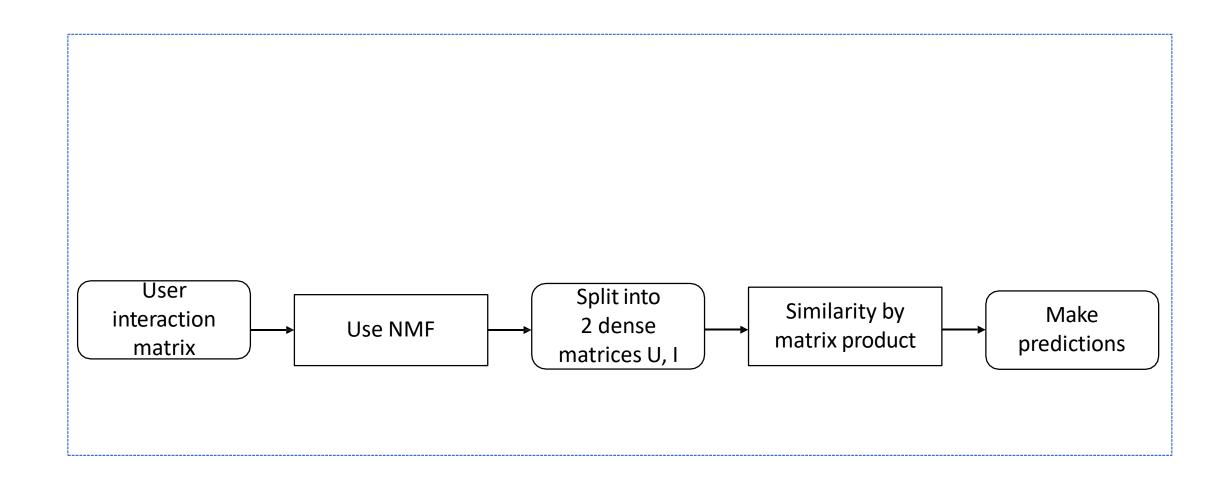
# Collaborative-filtering Recommender System using Supervised Learning



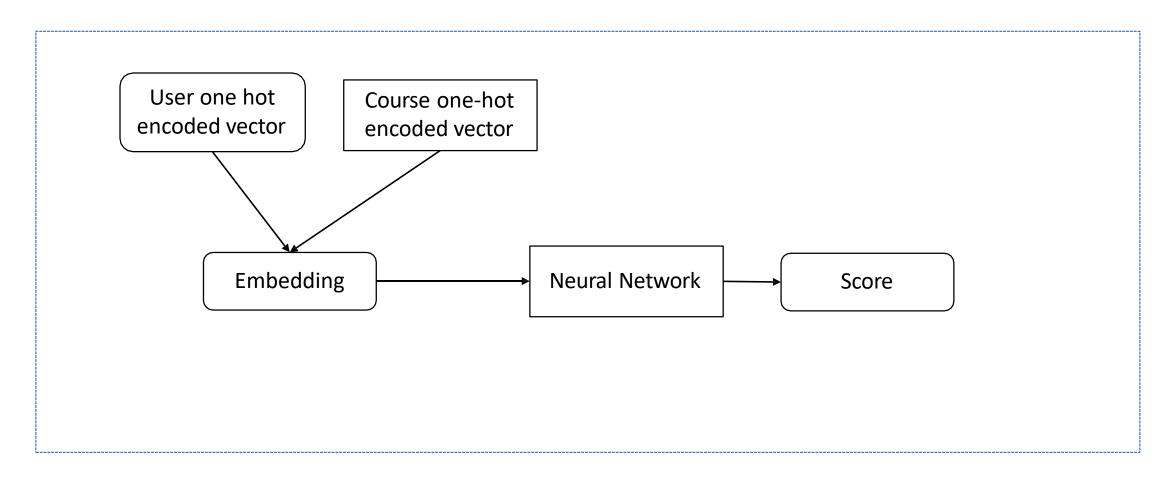
## Flowchart of KNN based recommender system



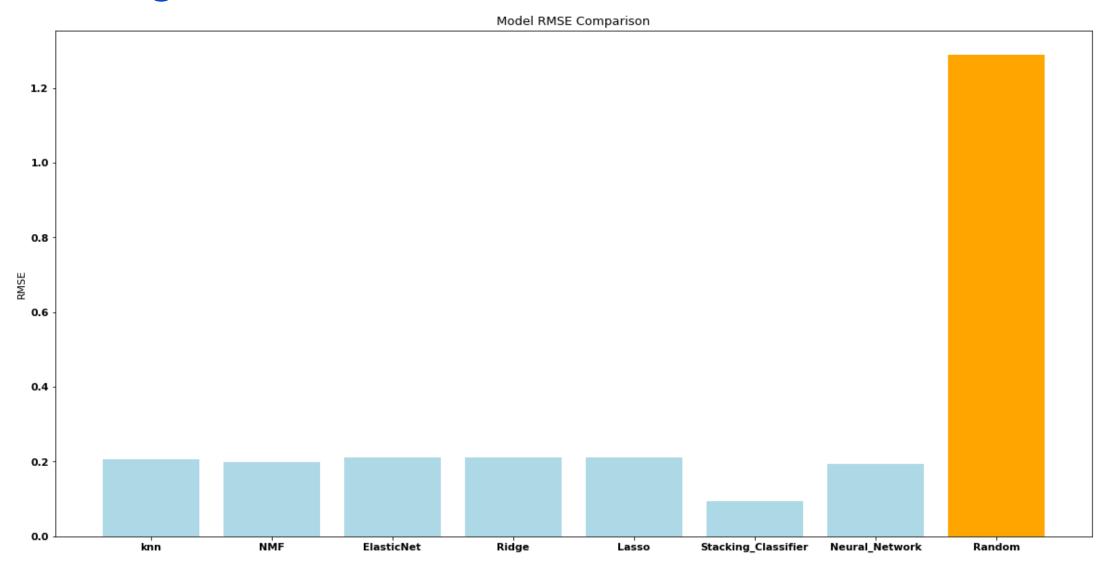
### Flowchart of NMF based recommender system



# Flowchart of Neural Network Embedding based recommender system



## Compare the performance of collaborativefiltering models



#### Conclusions

- Similar performance of models
- User profile based has highest number of recommendations
- Stacking Classifier has best performance
- Similarity matrix's high complexity
- NMF as a solution

## **Appendix**

•All materials link

https://drive.google.com/drive/folders/1qmcCH5yeUCmEXe1\_N9FHdXIYcR8ErT91?usp=sharing