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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

Background:

- Many people across the world take courses online.
- Just like Netflix or amazon, these courses can be recommended using recommender systems
- Using machine learning in python, we are able to make a solid predictor at guessing ratings or guessing courses that users might be interested in

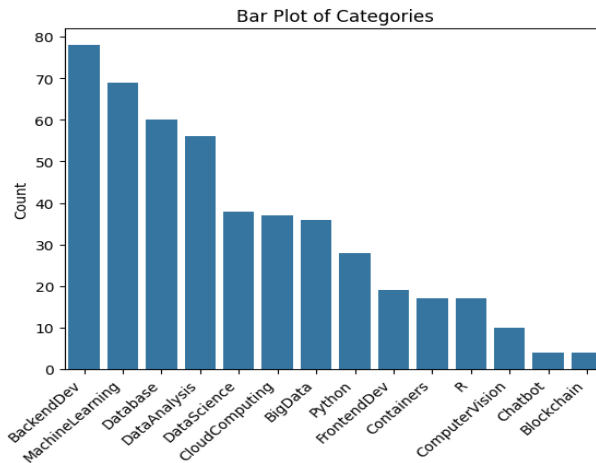
Problem statement:

- In this analysis, we will use multiple unsupervised learning algorithms to predict which courses a user might be interested in based on multiple factors
- And we will then use supervised learning models like KNN and neural networks to try and predict ratings of courses that a user has not taken

Exploratory Data Analysis

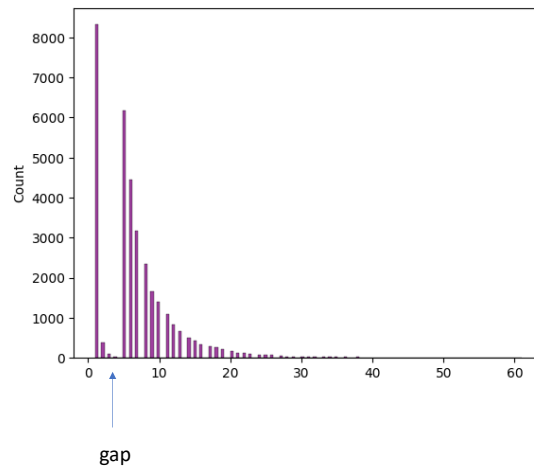


Course counts per genre



From this, we see that Back end Developing is number 1 with almost 80 course names its in, followed my machine learning and database. On the other end, chatbot and blockchain each have only 4 occurrences.

Course enrollment distribution



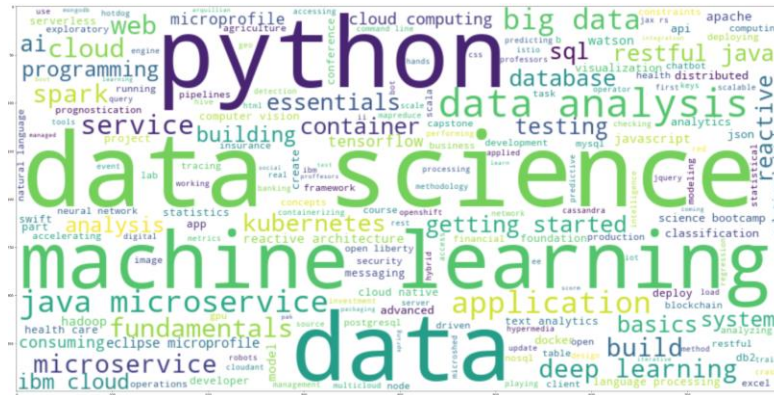
There is an evident left skew here with a big gap at around 2-4. This means that 1 enrollment was very common, but 2-4 were not, but then 5 and onward were back to having a fair amount of people. Almost all people either take 1 course, or many, it seems.

20 most popular courses

TITLE		
mapreduce and yarn	3670	← 20 th MOST ENROLLED
sql and relational databases 101	3697	
deep learning with tensorflow	3914	
docker essentials a developer introduction	4480	
introduction to cloud	4983	
statistics 101	5015	
r for data science	5237	
build your own chatbot	5512	
deep learning 101	6323	
data visualization with python	6709	
blockchain essentials	6719	
data science hands on with open source tools	7199	
spark fundamentals i	7551	
data science methodology	7719	
data analysis with python	8303	
machine learning with python	9394	
hadoop 101	10599	
big data 101	13291	
introduction to data science	14477	
python for data science	14936	← MOST ENROLLED
dtype: int64		

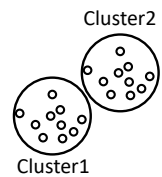
Here, we see python for data science is the most taken course. This would make sense, because it sounds like it is a very introductory course to data analysis, a very popular field.

Word cloud of course titles

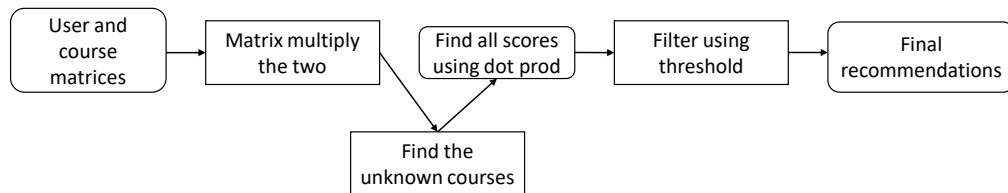


This is the world cloud of most common phrases in all the course list. I was kind of confused for this part because I did not have to code anything. The code for this word cloud was already given, so I am just pasting the image here.

Content-based Recommender System using Unsupervised Learning



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



After we multiply the matrices for user and the course, we need to create a matrix for each user's unknown courses. This will allow us to recommend courses only that the user has not taken. After we create this matrix, we need to find all scores for each user's unknown courses from the matrix. We find this score by dotting the unknown course matrix's vector for the user with the test user's vector. Once we have a score for each unknown course, we filter them out with threshold and recommend all courses with a score above the threshold.

Evaluation results of user profile-based recommender system

Hyperparameters: score threshold = 40

Most recommended courses

10th MOST
RECOMMENDED

Avg # of recommended courses

```
9]: np.float64(28.943518182916517)
```

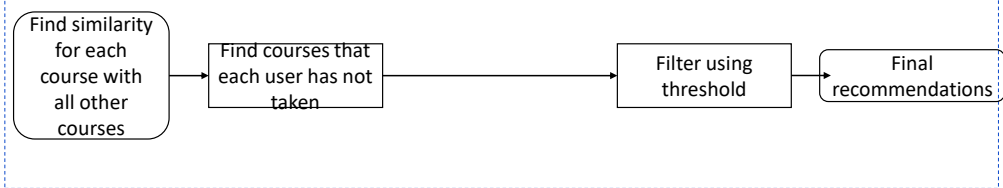
MOST RECOMMENDED

	Count	Code
270	7203	spark fundamentals ii
271	7633	accelerating deep learning with gpu
272	7671	introduction to data science in python
273	7671	applied machine learning in python
274	7853	cloud computing applications part 2 big data...
275	7970	spark overview for scala analytics
276	8769	analyzing big data in r using apache spark
277	8954	getting started with the data apache spark ma...
278	9138	foundations for big data analysis with sql
279	9138	analyzing big data with sql

1. Average number of unseen courses: 28.94. This is quite a lot. This is good because it means we are recommending new stuff
2. Top 10 courses: At the number 1 spot (at the bottom), we see analyzing big data with SQL with 9138 recommendations. In 10th place, we see spark fundamentals ii with 7203 recommendations

Flowchart of content-based recommender system using course similarity

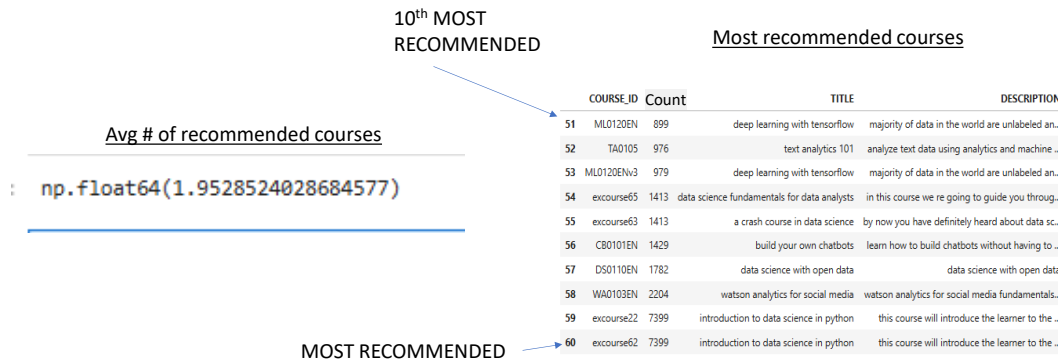
- Plot a flowchart which should clearly illustrate how you implemented the course similarity based recommender system



Firstly we need to establish each course's similarity with all other courses. This value ranges from 0 to 1. Similar to the recommender using content and course genres, we again need to find the unknown courses. This is much simpler than the previous recommender, because we already have scores, which is just the similarity value between courses. We will filter these courses by this similarity measure and recommend all those that are above the threshold.

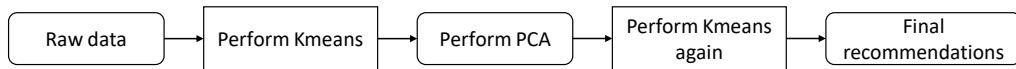
Evaluation results of course similarity based recommender system

similarity threshold: .6 is the similarity threshold I chose



1. For the average number of new courses recommended per person, we see that it is just under two. This means that our algorithm and our threshold is really strict on the similarity we are allowing between recommendations. If we wanted more average recommendations, we could decrease the threshold to solve this.
2. At number 1, we see introduction to data science in python with 7399 recommendations. AT 10th place is deep learning with TensorFlow at 899 recommendations

Flowchart of clustering-based recommender system



First, we use the Kmeans clustering algorithm on the user profile feature vectors in order to generate clusters for each user vector. Then we apply PCA on this in order to reduce the dimensionality. Then, Kmeans again. Finally, we have our recommendations.

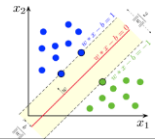
Evaluation results of clustering-based recommender system

Hyperparameters: $n_clu = 7$ (for Kmeans), $n_components = 9$ (for PCA)

Avg # of recommended courses		Most recommended courses			
<code>np.float64(103.2084009321259)</code>		10 th MOST RECOMMENDED			
		COURSE_ID	count	TITLE	DESCRIPTION
		115	RP0151EN 33629	r 101	in this introduction to r you will master the...
		116	SECM03EN 33659	apply end to end security to a cloud application	this mini course walks you through key securit...
		117	DP0101EN 33676	openrefine 101	this introduction course is for a less technic...
		118	COM001EN 33692	scalable web applications on kubernetes	this mini course walks you through how to scaf...
		119	PHPM002EN 33717	php web application on a lamp stack	this tutorial walks you through the creation o...
		120	ML0122ENV3 33728	accelerating deep learning with gpus	training complex deep learning models with lar...
		121	BD0151EN 33740	text analytics 101	the analysis of emails blogs tweets forums ...
		122	HCC104EN 33763	hybrid cloud conference serverless lab	hybrid cloud conference serverless lab
		123	HCC105EN 33767	hybrid cloud conference ai pipelines lab	hybrid cloud conference ai pipelines lab
		124	OS0101EN 33841	introduction to open source	this course introduces you to open source soft...

1. Holy cow that's a lot of courses recommended per user. This could be because I only used 7 clusters, meaning there are a ton of courses within each cluster. If we wanted less recommendations, we could increase the amount of clusters
2. At number 1, we see introduction to open source with 33,841 recommendations. This is kind of weird because this is getting recommended to almost everybody. Im pretty sure I did the code right though, so maybe its just really similar to a lot of courses.

Collaborative-filtering Recommender System using Supervised Learning

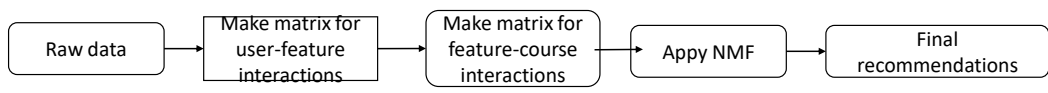


Flowchart of KNN based recommender system

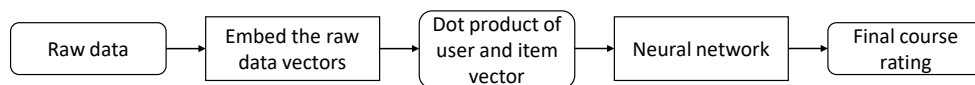


With KNN, there isn't really much to do on our part. We could convert the data into a sparse matrix, but we don't have to. The KNN library just takes care of everything for us.

Flowchart of NMF based recommender system

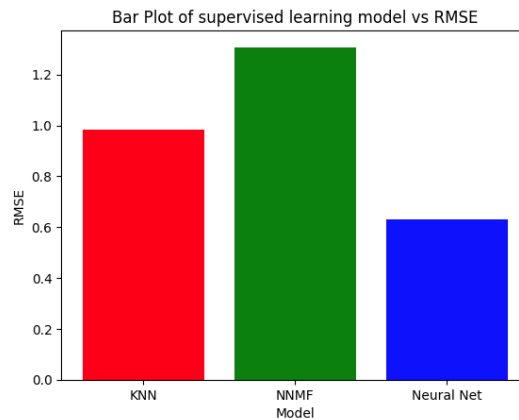


Flowchart of Neural Network Embedding based recommender system



With our raw data, we first need to embed the vectors into the embedded vectors, which will make the neural network compute more accurately. Once we have 2 embedded vectors, one for the user and one for the item, we then dot product them and feed this vector into the activation function for the neural network. I'm pretty sure that Relu was the function that was set in the notebook. Then, we fit and evaluate our model on our test data, and we have a matrix of course ratings for each user.

Compare the performance of collaborative-filtering models



As one might expect, the neural network has the lowest RMSE out of all of them. However, I would like to point out that the KNN model I used had the test size as .9, because anything less would cause my computer to crash. So, if I had more memory, maybe the KNN would do better. But for the resources I have, the neural network was the best by quite a bit.

Conclusions

- For predicting user ratings of an unknown course, the neural network has the lowest RMSE by a fair amount.
- This would lead me to recommend the neural network model over the NMF or KNN model to businesses.
- For course recommendations, I noticed that there were quite a range of values in recommended courses. We saw just under 2 all the way to over 100.
- This leads me to recommend [the user profile-based recommender system](#), because its average courses recommended is about 28, which is not too little but not too big
- This means that a customer will not be overwhelmed by a heap of possible courses, but also exposes them to enough where they might find something interesting.

Thank you!

- Thank you for taking time to review my presentation!

Thanks! Have a nice day!