Build a Personalized Online Course Recommender System with Machine Learning

William Ruffu 11/14/2024



IBMDeveloperSkillsNetwork

Apply end to end security

to a cloud application

Starts: April 26, 2019

SECM03EN - v1.0



Starts: Any time, Self-paced



BD0141EN - v2016.0







BD0131EN - v2016.0 Moving Data into Hadoop Starts: Any time, Self-paced



Outline

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

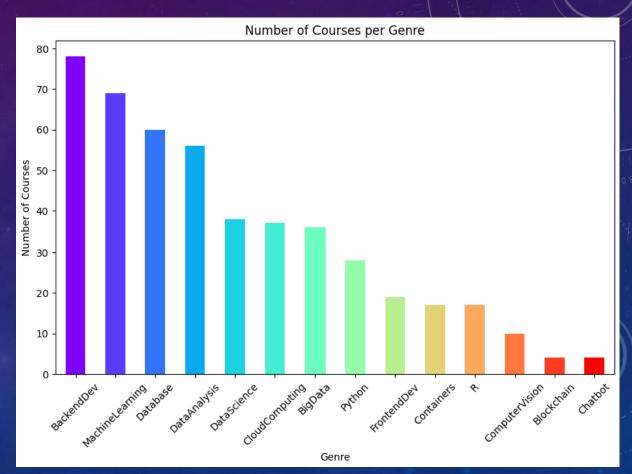
Introduction

- Project Context: Build a ML based course recommender system to improve user satisfaction and retention for online learning platform.
- Goal: Increase Customer Lifespan
- Hypothesis: Personalized recommendations will enhance user engagement & business value.
- Approach: Use content-based collaborative filtering & clustering methods with neural networks to develop recommender system(s).

EXPLORATORY DATA ANALYSIS

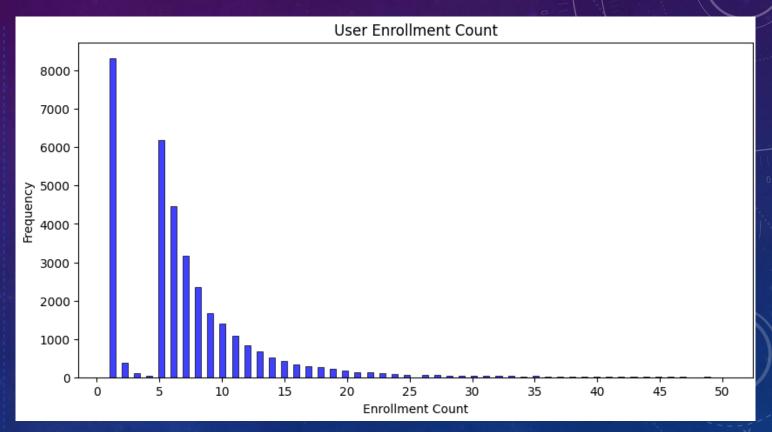
COURSE COUNTS PER GENRE

- Distribution of courses across genres
- Determined Genres Features Using the Course Catalog
- Used to Group Subclasses of Customers for Recommendation



COURSE ENROLLMENT DISTRIBUTION

- As the histogram shows, many customers only enroll in one course.
- An engaging recommendation can drive significant business value!



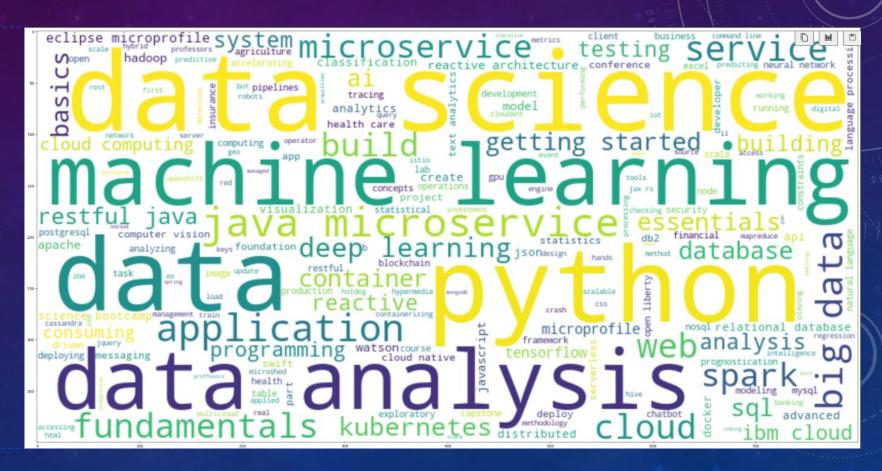
20 MOST POPULAR COURSE

- List the most popular 20 courses
- Python and Data Science are very popular courses!

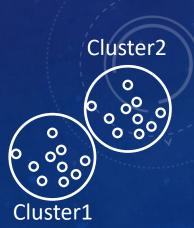
TITLE	Enrollment	
python for data science	14936	
introduction to data science	14477	
big data 101	13291	
hadoop 101	10599	
machine learning with python	9394	
data analysis with python	8303	
data science methodology	7719	
spark fundamentals i	7551	
data science hands on with open source tools	7199	
blockchain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

WORD CLOUD OF COURSE TITLES

WordCloudVisualization of OurCourse Offerings



CONTENT-BASED RECOMMENDER SYSTEM USING UNSUPERVISED LEARNING



FLOWCHART OF CONTENT-BASED RECOMMENDER SYSTEM USING USER PROFILE AND COURSE GENRES

- User profiles matched with course genres
- Using Provided csv Data: Course Genre, User Profile, Ratings

Sample Process:

Raw data

Convert data into vector arrays and account for already enrolled courses.

Dot Product
Multiplication
on User and
Genre Vectors

Feature
Engineered
Course and
Genre Vectors

Predicted
Interest Score
for Each
Course

EVALUATION RESULTS OF <u>USER PROFILE</u> BASED RECOMMENDER SYSTEM

Recommendation Score Minimum Set to 20

Average of 14 new courses recommended per user based on the currently enrolled course.

Top-10 commonly recommended courses across all users:

	COURSE_ID	count	TITLE
0	excourse72	9138	foundations for big data analysis with sql
1	excourse73	9138	analyzing big data with sql
2	TMP0105EN	8954	getting started with the data apache spark ma
3	RP0105EN	8769	analyzing big data in r using apache spark
4	SC0103EN	7970	spark overview for scala analytics
5	excourse31	7853	cloud computing applications part 2 big data
6	excourse21	7671	applied machine learning in python
7	excourse22	7671	introduction to data science in python
8	ML0122EN	7633	accelerating deep learning with gpu
9	BD0212EN	7203	spark fundamentals ii

FLOWCHART OF <u>CONTENT-BASED</u> RECOMMENDER SYSTEM USING <u>COURSE</u> <u>SIMILARITY</u>

- From looking at similar courses from the users enrolled course we can make different predictions.
- User profiles matched with course genres.

Example Process:

Raw data:
User, Course,
& Genre
Datasets as
Base

Data Processing:
Used BOW &
Course Similarity
Matrices from
Previous
Recommendation
Exercises

Used Enrolled
Courses to
Run Similarity
Scores per
User

Applied
Threshold &
Custom
Function

Generate Multiple Recommendations Per User

EVALUATION RESULTS OF <u>COURSE</u> <u>SIMILARITY</u> BASED RECOMMENDER SYSTEM

Similarity Threshold = .6 or 60% Similarity Score

On average, this method produced 10 new/unseen courses recommendations per user (in the test user dataset).

Top-10 commonly recommended courses:

	COURSE_ID	count	TITLE
0	DS0110EN	15003	data science with open data
1	excourse22	14937	introduction to data science in python
2	excourse62	14937	introduction to data science in python
3	excourse63	14641	a crash course in data science
4	excourse65	14641	data science fundamentals for data analysts
5	excourse68	13551	big data modeling and management systems
6	excourse72	13512	foundations for big data analysis with sql
7	excourse74	13291	fundamentals of big data
8	excourse67	13291	introduction to big data
9	BD0145EN	12497	sql access for hadoop

FLOWCHART OF <u>CLUSTERING-BASED</u> RECOMMENDER SYSTEM

- User segments created for targeted recommendations.
- K-Means clustering with PCA for dimensionality reduction.

General Process:

Using User Profile Dataset Only

Extract Features and Normalize with StandardScalar

Perform KMeans
Clustering &
PCA to Reduce
Dimensionality

PCA Transformed Feature Vectors Used to Cluster Users

Recommendations Based on Similarity

EVALUATION RESULTS OF <u>CLUSTERING</u>-BASED RECOMMENDER SYSTEM

PCA Similarity Threshold of .9

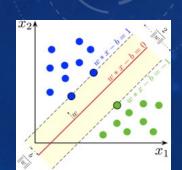
Enrollment Threshold = 1500

Using a enrollment threshold of 1500 in the test_users group, we can recommend 16 courses on average per user.

Top-10 Commonly Recommended Courses

recommended_	_course
ST0101EN	25176
CC0101EN	25162
RP0101EN	24762
ML0115EN	24123
BC0101EN	24089
CNSC02EN	23691
DV0101EN	23529
DS0301EN	23512
BD0115EN	23471
DB0101EN	23152

COLLABORATIVE-FILTERING RECOMMENDER SYSTEM USING SUPERVISED LEARNING



FLOWCHART OF KNN BASED RECOMMENDER SYSTEM

- Enrollment History & User Ratings csv Data
- User-Item Interaction Matrix Used
- Scikit-learn Surprise Library

General Process:

Raw data converted using .pivot to convert to a sparse matrix

knn model2 =
KNNBasic(sim
options={'nam
e': 'cosine',
'user_based':
False})

knn_model2 .fit(train set) predictions2
=
knn model2.te
st(Testset)

accuracy.rm
se(predicti
 ons2)
= Results
in a RMSE
 of 1.286

FLOWCHART OF NMF BASED RECOMMENDER SYSTEM

- Non-Negative Matrix Factorization
- Surprise Library

General Process:

Used ratings.csv

Converted data using .pivot

model =
NMF(n_factors
=32,
biased=False,
verbose=True,
init_low=.5,
init_high=5)

model.fit(tra
inset)

predictions =
model.test(te
stset)

accuracy.r mse(predic tions)

Results in 1.288 RMSE

FLOWCHART OF NEURAL NETWORK EMBEDDING BASED RECOMMENDER SYSTEM

- TensorFlow Keras Model SuperClass RecommenderNet
- One-Hot Encoded w/ Adam Optimizer = .001
- Rating Dataset

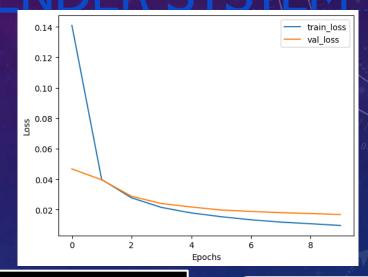
General Process:

Raw data split into item and user sets.

One-Hot Encoding of each Set Embedding Layer(s)

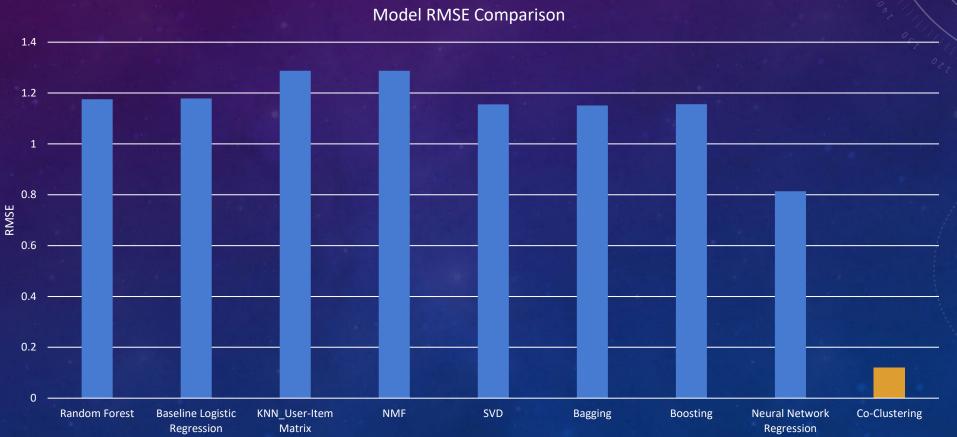
Smaller Embedded matrice(s) of 16 x 1

Rating Result of Dot Product



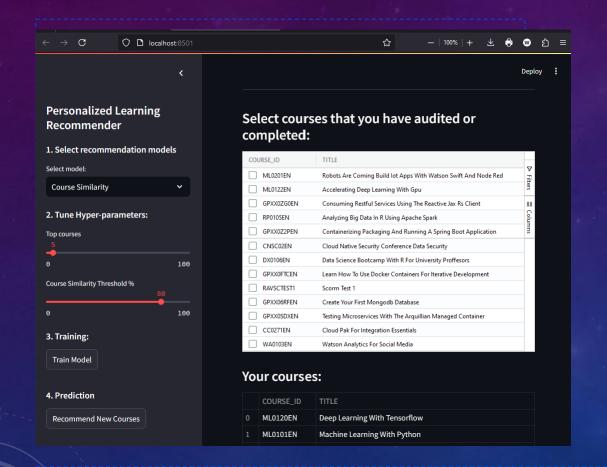
COMPARE THE PERFORMANCE OF COLLABORATIVE-FILTERING MODELS

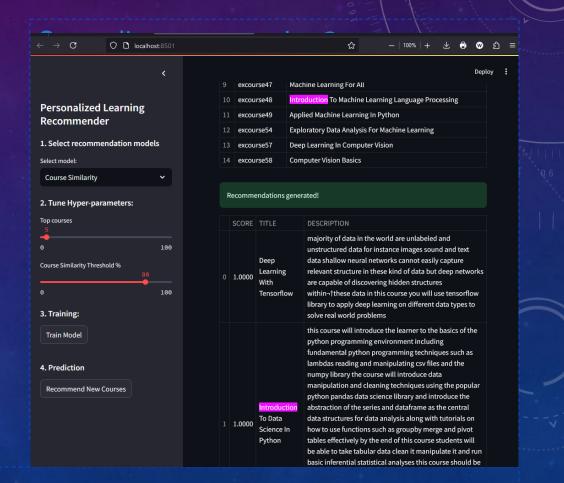
- Summary Stats of RMSE Performance Metrics Across Collaborative-filtering Models



Embedding

DEPLOYED STREAMLIT APP





Conclusions

Main Insights:

- Machine learning significantly enhances accuracy in recommendations.
- Neural network models outperform other methods in accuracy and scalability.
- Clustering-based approaches effectively segment users for targeted suggestions.

Strategic Value:

- Improved user engagement drives platform growth and retention.
- Tailored recommendations increase trust and customer satisfaction.
- Scalable systems provide adaptability for future datasets and user bases