# IT-562 Recommendation Systems and Engines

### **Team- 404**

## **Article Recommendation System**

### Result Report of Hybrid Model:

## 1. Existing User:

Past Interactions:

title	eventStrength
6 reasons why I like KeystoneML	3.735522
Auto-scaling scikit-learn with Spark	3.726831
5 reasons your employees aren't sharing their	3.475085
At eBay, Machine Learning is Driving Innovativ	3.475085
Algorithms and architecture for job recommenda	3.356144
10 Stats About Artificial Intelligence That Wi	3.339137
Al Is Here to Help You Write Emails People Wil	3.269033
Deep Learning for Chatbots, Part 1 - Introduction	3.195348
Graph Capabilities with the Elastic Stack	3.104337
Being A Developer After 40 - Free Code Camp	3.007196
Building with Watson Technical Web Series	2.887525
Worldwide Ops in Minutes with DataStax & Cloud	2.855990
5 Unique Features Of Google Compute Engine Tha	2.782409
How to choose algorithms for Microsoft Azure M	2.687061
Bad Writing Is Destroying Your Company's Produ	2.632268
Creative Applications of Deep Learning with Te	2.608809
How Netflix does A/B Testing - uxdesign.cc - U	2.594549
Machine Learning Is No Longer Just for Experts	2.536053
How Google is Remaking Itself as a "Machine Le	2.464668
Text summarization with TensorFlow	2.454176

### New Recommendations based on Hybrid Model:

titl	contentId
The barbell effect of machine learning	3269302169678465882
Power to the People: How One Unknown Group of	5092635400707338872
How Google is Remaking Itself as a "Machine Le	-7126520323752764957
The AI business landscap	7395435905985567130
Machine Learning Is Redefining The Enterprise	1415230502586719648
Being A Developer After 40 - Free Code Cam	-5756697018315640725
The AI business landscap	7395435905985567130
Graph Capabilities with the Elastic Stac	-8085935119790093311
How Google is Remaking Itself as a "Machine Le.	5250363310227021277
Machine Learning for Designer	638282658987724754

#### 2. New User:

Checks whether User already exists or not. If user doesn't exist, the system assigns the user an unique UserID and recommends the most popular articles from each Genre till 5 interactions. The system takes in the User Interaction after each Recommendation.

#### 3. Accuracy Results:

- The Training and Testing data is split on the based of a reference date. The ratio of Training to Testing Split is 4:1.
- We have used Recall-5 and Recall-10 method for testing the Accuracy of our model. (Recall at k is the proportion of relevant items found in the top-k recommendations).
- We have Compared our Hybrid Interactive Model (which uses Collaborative filtering and Content-based Filtering along with User Interactions and Time-based Weights) to pure Popularity based, Collaborative-based and Content-based approaches. The accuracy results are as shown below:

## recall@10 recall@5

#### modelName

Popularity	0.341984	0.219637
Collaborative Filtering	0.415750	0.290335
Content-Based	0.510611	0.397852
Final	0.520583	0.394656