K-Means Based Recommendation Engine

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DSC680

Project 3

Project Overview

- Document Database and Pre-processing
- Reccomendation engine
 - Pyspark Pipelines
 - RegexTokenizer
 - StopWordRemover
 - Word2Vec
 - K-means Model

Project Objective

- Create a Basic Recommendation Engine Using a K-Means Model
 - Intake Document From User
 - 2. Process Text
 - 3. Predict Cluster Label with K-Means Model
 - 4. Return a Set of Similar Documents

Project Description

- Aggregate data
 - Light Novel EPUBs > Text Files
- Build Pyspark Pipeline Model and Prediction Database
 - 1. Arrange Pipeline Stages
 - 2. Fit Pipeline to Document Database to Train Model
- Transform Document Database with Pipeline Model to Pair Sample Data and Predictions
- Intake Documents From User
- Return Recommended documents

Document Database

- The Document Database used in this project was sourced from Light Novel Crawler a Python program that retrieves Light Novels from Light Novel Aggregator Websites
- The Documents started out in EPUB format and were converted into text files
 - 1. Unzip Epub files into HTML files
 - 2. BeautifulSoup to extract text within tags
 - 3. Preprocess Text to remove symbols
 - 4. Read Documents into a Pyspark DF

Create Pipeline

- Pyspark Pipeline
 - Pyspark.ml.Pipeline
- Pull Text Into a Format for Processing
 - Pyspark.ml.feature.RegexTokenizer
- Remove Uninformative Words
 - Pyspark.ml.feature.StopWordRemover
- Change Tokens into Computer processable Vectors
 - Pyspark.ml.feature.Word2Vec
- Process Document Vectors to Predict Similar Documents
 - Pyspark.ml.clustering.KMeans

Train and Fit Pyspark PipelineModel

- Once the Document Database and Pipeline are created the Document Database Must First be .fit() to the Pipeline To Train the Model
 - Unsupervised Model
- To Record the Predictions of Similar Documents (Predicted Cluster Labels) the Document Database must be .transform()ed with the new PipelineModel

Use Model to Recommend Documents

- Once the Model Has Been Fit and the Original Document Database Has Been Classified, It Can Then Be Used to Recommend Similar Documents
- User's Document is Submitted
- The Existing Pipeline is used to .transform() the User's Document
- The Transformation Returns a Cluster Label
- Filter Original Document Database to Return Documents Predicted to be in the Same Cluster

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Validity of Model

- How can we tell if the model's predictions are valid?
 - We cannot definitively say this
- We can infer the accuracy of the model by looking at the predicted data
- Same author, same book > prediction is probably valid