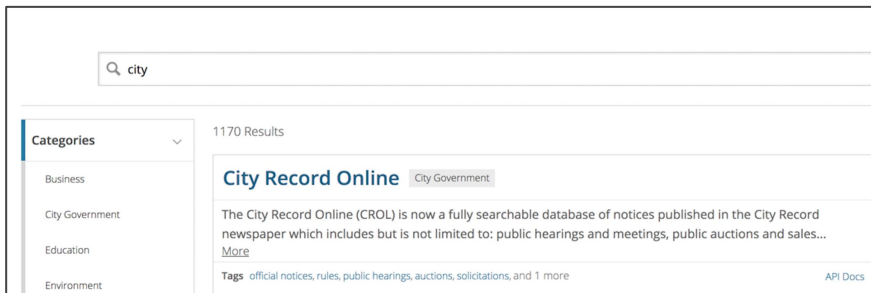


Search Engine in PySpark

Big Data Term Project
Spring 2018

Project Goal: search engine for structured data

NYC OpenData



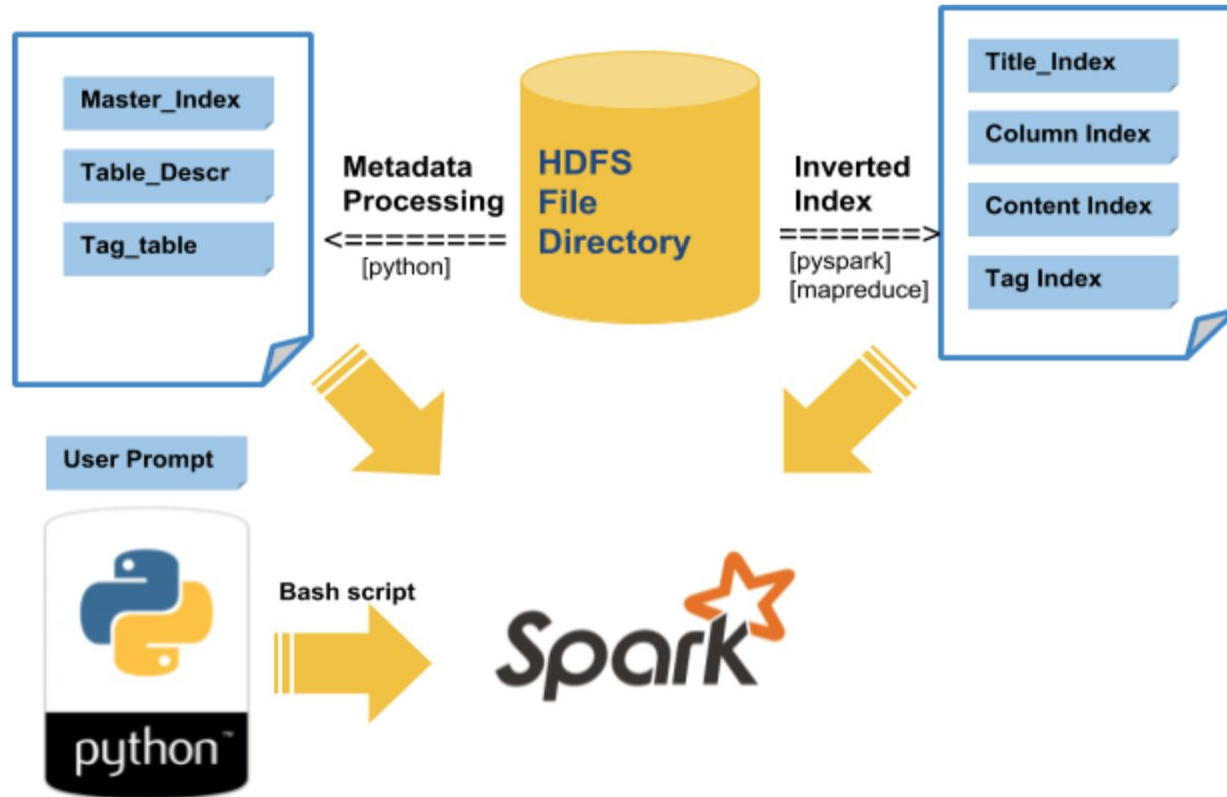
**Only Title Search
Available!**

We will offer search by:

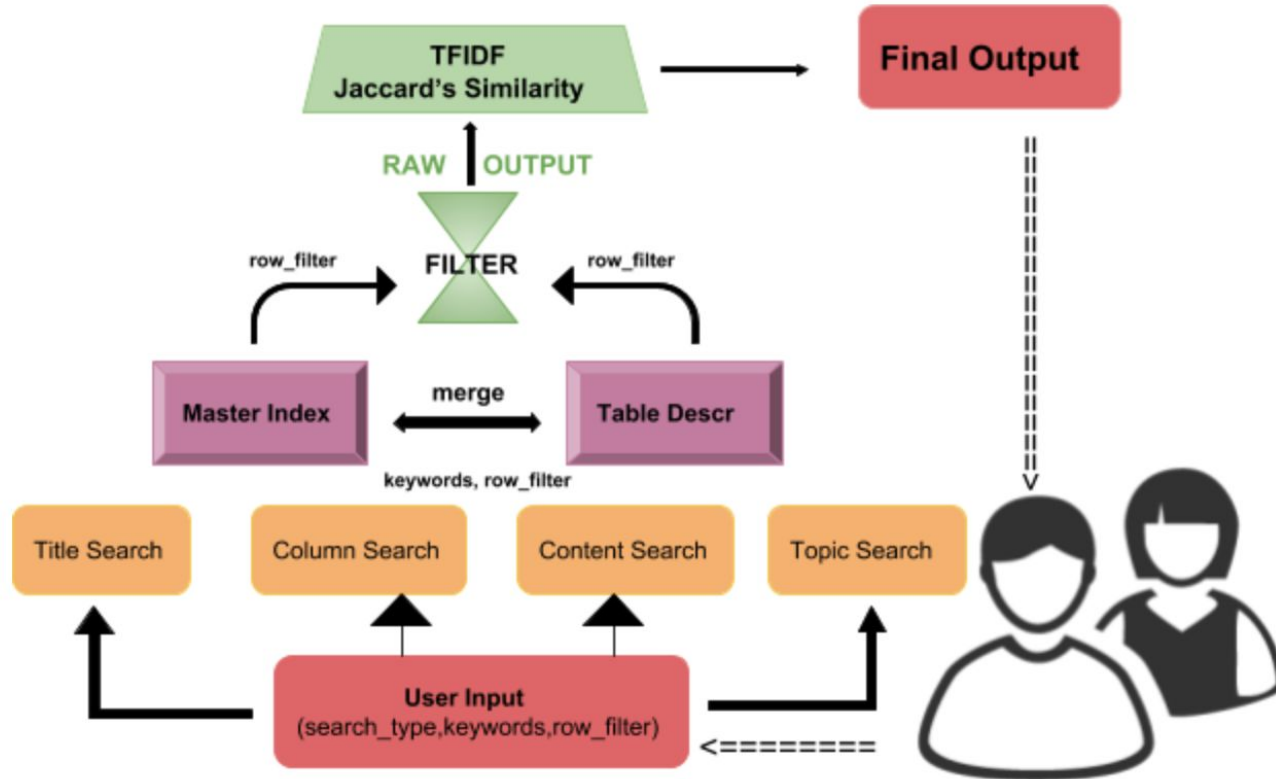
- **Title**
- **Column**
- **Content**
- **Topic**
- **Filter by table length**



Architecture & Design: Method Design



Architecture & Design: Application Design



Json Processing and Other Tables

metadata

Tag_table (Doc_ID, tags) (generate tag_index)

Table_Desc (Doc_ID, Category, Description)

Table_Schema (used to generate column_index)
(Doc_ID, Schema)

**HDFS
Directory**

Master_Index (used to generate title_index)
(Doc_ID, Table_Name, Table_Length)

data

All tables in .csv format for content index

Inverted Indices: title, column, context, tags

1. **Collect** the data
2. Parse collection of tables
 - a. **Lowercase** all words: “SPARK” -> “spark”
 - b. **Tokenize**: “search+engine” -> “search”, “engine”
 - c. Filter out **stop words**: “the”, “a”, “an”, etc.
 - d. Stem each token using **Porter Stemmer Algorithm**:
“fisher”, “fishing”, “fishes”, “fished” -> “fish”



Inverted Index

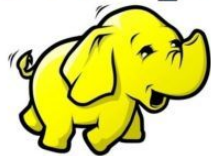


- Building the index using **MapReduce**
 - **Key**: vocabulary term
 - **Value**: postings list (tables where the term appears & in-table term frequency)

We compared two methods for constructing indices

- Built **two versions** of inverted indices
 - PySpark
 - Hadoop
- Decided to go with **PySpark**
 - Built-in functionality for querying

hadoop



PySpark

Querying

User Queries

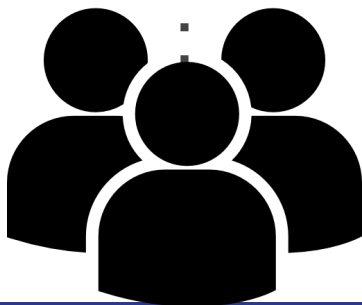
Spark.SQL Queries

Python

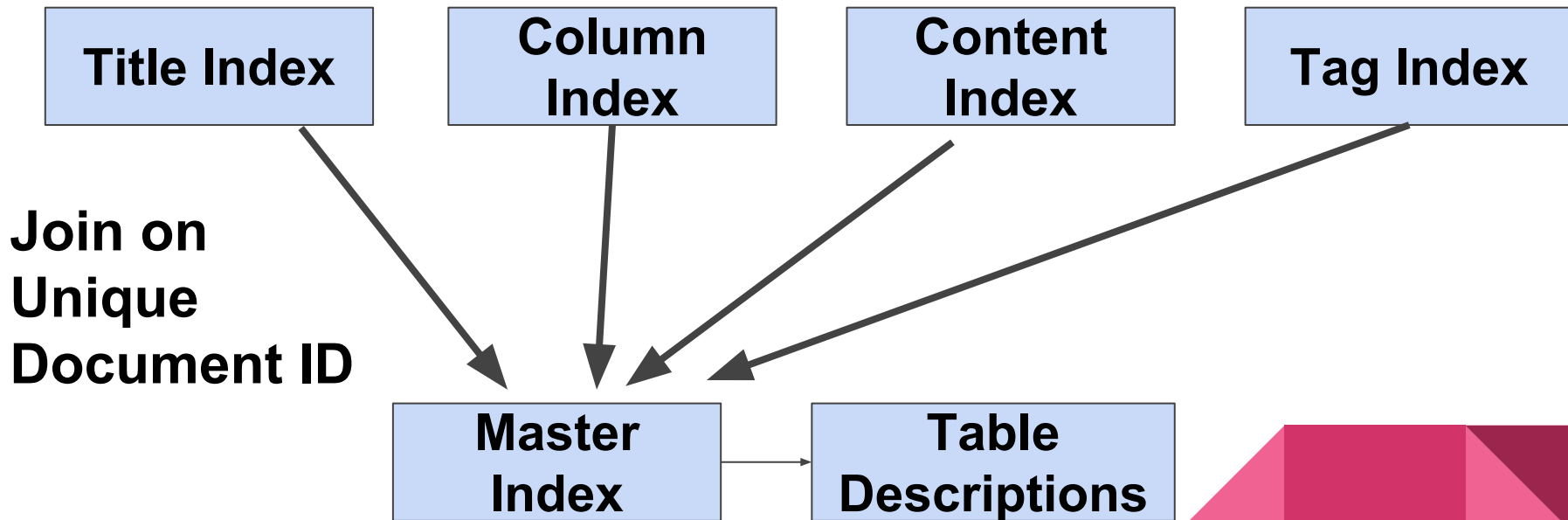
- Search_Type:
---1,2
- Keywords:
---cab,taxi,city
- Row_Filter:
---100000

Spark

```
title_search([cab,taxi,city  
,100000)  
column_search([cab,taxi  
,city],100000)
```



Processing Queries



Search Ranking: TF-IDF

- Term Frequency & Document Frequency

$$TF(t, d)$$

$$DF(t, D)$$

- TF.IDF Scoring

$$IDF(t, D) = \log \frac{|D| + 1}{DF(t, D) + 1}$$

$$TFIDF(t, d, D) = TF(t, d) \cdot IDF(t, D).$$

Search Ranking: Vector Space Model

- Represent each document as a vector, with each entry being the TF.IDF weight of the corresponding term t in document d
- Represent each query as a vector in the same vector space as the documents
- Cosine similarity: the relevance between each document and the query

$$sim(q, d) = \frac{\vec{V}(q) \cdot \vec{V}(d)}{|\vec{V}(q)| |\vec{V}(d)|}$$

Search Ranking

Preprocessing the tables

- Calculate TF.IDF scores from the inverted index files
- 4 inverted index files -> 4 TF.IDF models
- Save the models into files

Searching in run-time

- Get the query & Parse
- Convert into vector
- Read the TF.IDF models
- Calculate the cosine similarities
- Rank & Output

Demo

Here is your content search result

Table_Name	Category	Description
[2015-2016 Student...	Education	[Student Disciplin...
[Capital Project S...	Housing & Develop...	[List of capital p...
[Mayor's Office to...	Social Services	[The dataset conta...
[Capital Commitmen...	City Government	[This dataset cont...
[LinkNYC New Site ...	Social Services	[A map of proposed...
[Benefits and Prog...	Social Services	[This dataset pro...
[2017 Diversity Re...	Education	[Missing]
[Parking Violation...	City Government	[Parking Violation...
[Full-Time And Ful...	City Government	[This dataset cont...
[Verified Location...	City Government	[An agency-verifie...
[2005 Street Tree ...	Environment	[Citywide street t...
[2015-16 Student D...	Education	[Missing]
[FHV Base Aggregat...	Transportation	[Monthly report in...
[2015-16 Demograph...	Education	[Missing]
[Deaths by Year an...	Public Safety	[This is a breakdo...
[2015 - 2016 Audit...	Education	[Official audited ...
[FY17 MMR Agency P...	City Government	[NYC agency perfor...
[Doing Business Se...	City Government	[The Doing Business...
[2016-17 Physical ...	Education	[Missing]
[NYPD Complaint Ma...	Public Safety	[This dataset incl...

Here is your topic search result

Table_Name	Category	Description
[Most Popular Baby...	Health	[The most popular ...]
[NYCHA Application...	Housing & Develop...	[Priority codes as...]



Results



- Improved **Search Capabilities**
 - Title, Column, Content, Tags, Row Filter
- **Efficient Ranking**
 - TF-IDF using MLlib
- **Scalable** Implementation
 - Can built additional applications on top