

CS235: DATA MINING TECHNIQUES

FALL 2016

ASSIGNMENT REPORT

SUBMITTED BY: Abhishek Kumar Srivastava #ID – 861307778 INSTRUCTOR: Vagelis Papalexakis University of California Riverside

Part 1: Data Crawling

I have written the crawler in Python. In the crawler for each section i.e Data Mining, Databases, Machine Learning & Artificial Intelligence I have called a general function *data_from_url(target_url)* passing appropriate URL for each section.

In this function, page number are added incrementally and using request API the web-page is requested. Then using beautiful soup I have extracted the URL for each conference and then requested again the web-page for that Conference from there I extracted Conference Name and Conference Location. Appropriate time limiter was added of 10 seconds between each page requests.

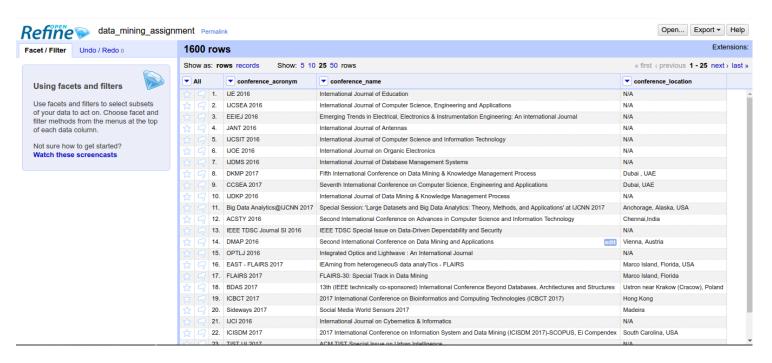
Finally after getting all the required data i.e Conference_Acronym,Conference_Name and Conference_Location for all 20 pages and for all 4 categories I added those values to the file in the tab separated manner in appending manner using CSV Writer.

Please check **crawler.py** for the crawler code and **data_mining_original.tsv** for the tab separate values crawled using the crawler.

Part 2: Data Cleaning

I have used **OpenRefine** for the data cleaning task as recommended by the Instructor. Following steps were taken in data cleaning with the screen shots for few of them:

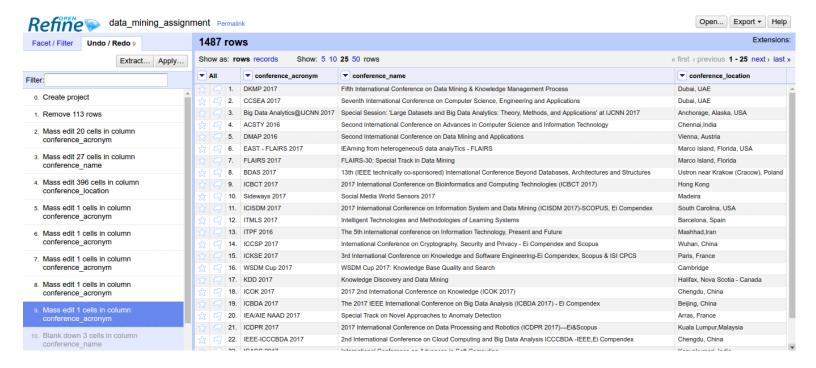
1. Importing Data



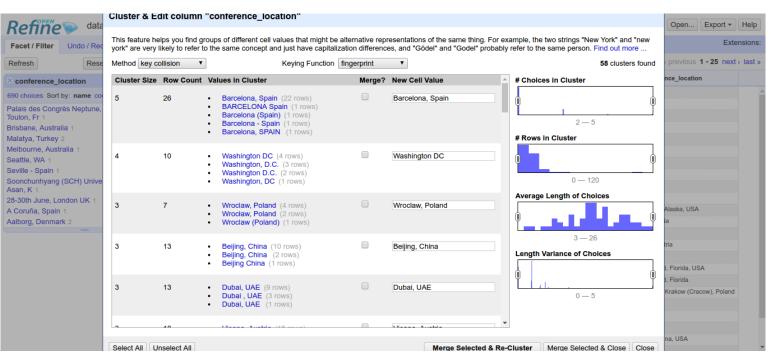
After importing the tsv file directly in the OpenRefine we can see that there are 1600 rows i.e 400 rows for each category (20 pages with 20 conference list each) which validated the correctness of the crawler that it has correctly crawled all the required pages needed to be crawled.

2. Removing whitespaces and conference location as N/A

First for all 3 Columns trimmed leading and trailing whitespaces and since Conference Location is important requirement for the task to be done further I have removed the data rows in which location is not mentioned. After doing so 1487 rows were left useful.



3. Data Clustering

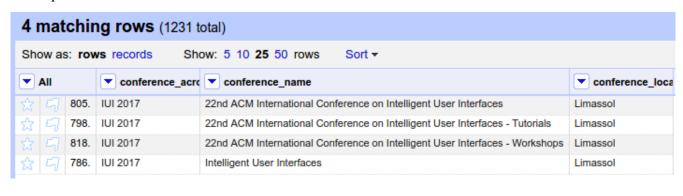


There were many mismatches for the same name so they were clustered using the clustering feature of OpenRefine. Clustering applied to all Conference name and rows. One example as a snapshot.

4. Anomalous City and Conferences:

Even after clustering there were many mismatched data set were there. They were fixed manually.

Multiple kind of Anomalies are attached:









5. Deletion of Repeated Values

There were many data rows which were repeated multiple times because they may be crawled because same conference may belong to multiple categories like Machine Learning and Artificial Intelligence.



There were some repetitions that were not so obvious

6 matching rows (1487 total)								
Show as: rows records Show: 5 10 25 50 rows								
▼/	▼ AII		▼ conference_acro	▼ conference_name	▼ conference_locat			
☆	9	233.	ICDM 2015	ICDM 2015 Student Travel Grants	Atlantic City, NJ, USA			
☆	9	293.	ICDM 2015	International Conference on Data Mining	Atlantic City, NJ, USA			
盘	9	406.	ICDM 2015	ICDM 2015 Student Travel Grants	Atlantic City, NJ, USA			
☆	9	409.	ICDM 2015	International Conference on Data Mining	Atlantic City, NJ, USA			
	4	1387.	ICDM 2015	ICDM 2015 Student Travel Grants	Atlantic City, NJ, USA			
☆	q	1442.	ICDM 2015	International Conference on Data Mining	Atlantic City, NJ, USA			

3 matching rows (1487 total)							
Show as: rows records Show: 5 10 25 50 rows							
▼ AII			▼ conference_acro	▼ conference_name	▼ conference_location		
☆		975.	ECAI 2016	8th International Conference on Electronics, Computers and Artificial Intelligence	Ploiesti, Romania		
ឋ₃	П	1024.	ECAI 2016	European Conference on Artificial Intelligence	The Hague, the Netherlands		
☆		1323.	ECAI 2016	European Conference on Artificial Intelligence	The Hague, the Netherlands		

These case I handled using markdown feature. I marked down similar rows as blank leaving only 1 of the possible data entries among the repeated ones and finally selecting the all blank rows and deleting them.

6. Splitting city from location and Year from acronym

I split City from the conference_location column using "," delimiter and retained only first column and deleted the rest of the column which could have been States and Countries column. After that I replaced space in location by "_" to make them a single word.

I also split Year from the conference_acronym to get Conference acronym and Conference Year. Which were helpful for many different purposes. I used "<space>" as a delimiter and kept only last column and joined rest of them together.

I also replaced "," as "<space>" in the name stop its interference with the csv values which I have used as input file for evaluation of MapReduce functions.

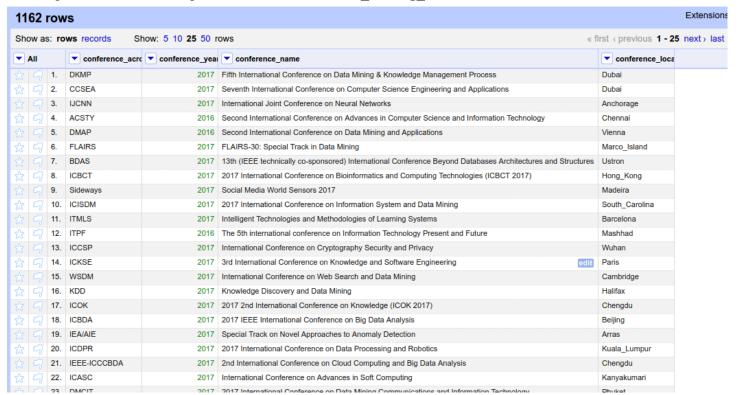
I again trimmed leading and trailing whitespace in the data set.

7. Removing Random Unwanted Words from Conference Name

There were many mistyped words which I removed from the Conference_Name column few examples of the "Deadline Extended", "Extension", "Ei Compendex", "Scopus"

I removed these kind of words as many as possible by searching them and replacing them using OpenRefine replace methods to make more sense from the conference names.

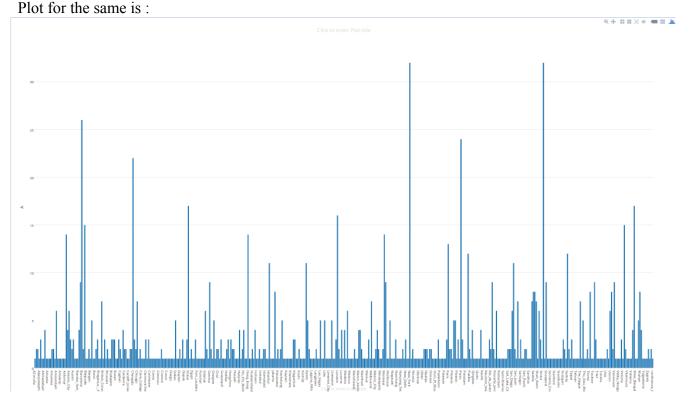
After cleaning as much as possible in my opinion this was the my final data set of 1162 rows. Cleaned up data in a comma separate value is in the **data mining refined.csv** file.



Part 3: Hadoop

part a) To calculate the number of conferences per city I first split the row using "," as splitting point then from separate values I used conference city and count 1 as the value.

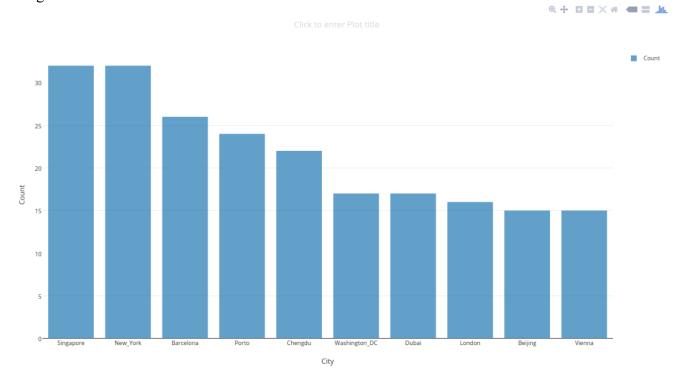
Reducer added up all the values for a particular city resulting in total number of conference per city. Result generated is in the **conf_per_city.txt** file and the code is in **conf_per_city.java**



The Top 10 locations from my cleaned up data set is:

New_York:32, Singapore:32, Barcelona:26, Porto:24, Chengdu:22, Dubai:17, Washington_DC:17, London:16, Beijing:15, Vienna:15

Histogram for the same:



part b) To calculate the list of conferences per city I did the same thing as before by spliting the row using "," as splitting point then from separate values I used conference_city as the key and conference acronym as the value.

In Reducer I concatenated all the value I get for a particular key which resulted in list of conference per city. But it can happen that if a particular conference happened in a particular city in different year it will result into repeated values so I also added the check in the Reducer that if a particular conference is already added then I wont add it again which will result only in unique conference occurred in the city. Although there were not many cases there this were frequent only for those cities where conferences occurs frequently.

Conference per city with repeated value is present in **list_conf_per_city.txt** and without repetition one is present in list **conf per city without repetition.txt** and the code is in **list conf per city.java**

Output example:

Acoruña: JISBD, Aalborg: CSE SPLINE, Aberdeen: EANN SoMePeAS, Aberystwyth: SAB, Adelaide: DPBA APCCM DICTA, Agra: HTC, Ahmedabad: COMAD, Aizu-Wakamatsu: DNIS MLSLP, Aksaray: SIN, Alicante: Big Data

part c) To calculate the list of cities per conference I did the same thing as above only change was I used conference acronym as the key and conference city as the value.

In Reducer I concatenated all the value I get for a particular key which will result in list of cities per conference. Same approach was taken to reduce the redundancy of cities per conference.

Cities per conference with repeated value is present in list_city_per_conf.txt and without repetition one is present in list_city_per_conf_without_repetition.txt and the code is in list_city_per_conf.java

Output example:

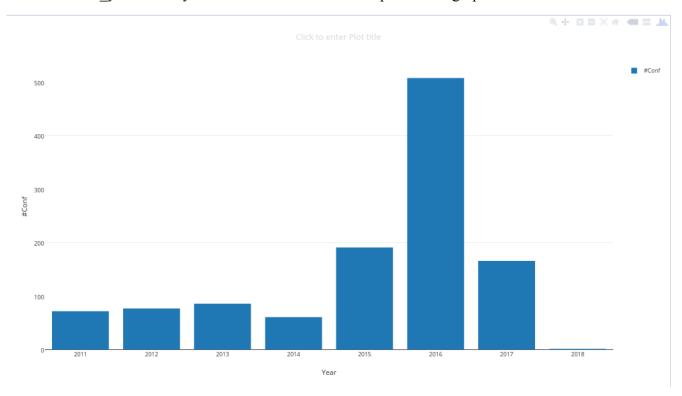
4DML: USA, **AAIA**: Gdansk, **AAMAS**: Sao_Paulo Singapore, **ACALCI**: Canberra, **ACCV**: Taipai, **ACIIDS**: Kaohsiung Kuala_Lumpur, **ACIRS**: Tokyo Wuhan, **ACIS**: Krabi, **ACML**: Hong Kong Hamilton, **ACMME**: Kuala Lumpur

part d) To calculate the number of year wise conference for a particular I used conference_acronym + ":" + conference_year as the key and count '1' as the value.

Reason for using such key is it wont mix all the count for a particular city and wont mix count for a particular year. Thats why it will give me count of conference occurred in a particular city for a particular year.

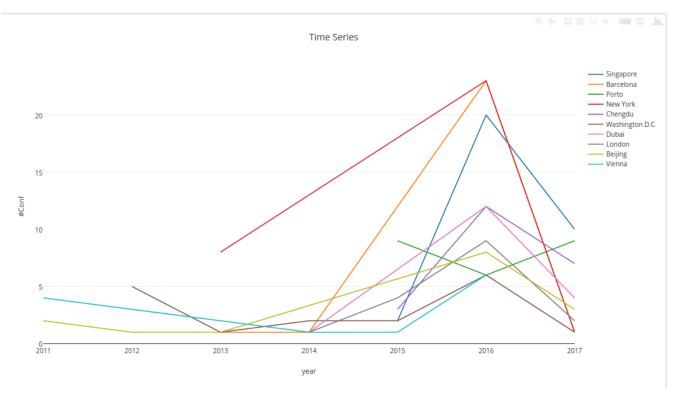
In Reducer I just summed up all the value I get for a the key which will result in time series of conference per city.

This is the graph of number of conference occurred per year I calculate it just by using conference year as a key and count 1 as the value and plotted the graph.



For Plotting the time series of a particular city I did it by again feeding the collected data into the OpenRefine and splitting conference acronym and for a particular city I extracted the year and count for top 10 cities and downloaded csv for each city.

Using online graph plotting tool I uploaded all csv's and plotted a line curve for each of the cities as year vs number of conferences and used different colored line to represent a different city.



Cities per conference with repeated value is present in **year_wise_conf.txt** and the code is in **year wise conf.java**

Output example:

ACoruña:2011	1
Aalborg:2013	1
Aalborg:2016	1
Aberdeen:2016	1
Aberdeen:2017	1
Aberystwyth:2016	1
Adelaide:2013	1
Adelaide:2015	2
Agra:2016	1
Ahmedabad:2013	1

References

- [1] For All Python Related Stuff: http://docs.python-guide.org/en/latest/
- [2] For Python Crawler: https://www.crummy.com/software/BeautifulSoup/bs4/doc/
- [3] For Fixing Compilation issues & Other Stuffs: https://stackoverflow.com
- [4] For Graph Plotting: https://plot.ly/