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CS 8674 – Independent Master's Project

Summer Term 2016

Review of Project

Supervisor: Dr. Ian Gorton, Professor and Director of Computer Science, Seattle Campus

*Create, Host, Deploy in Python-ese*

*Creating a mySQL database web application*

*Hosted by Flask & Publicly Deployed on Amazon Web Services*

## *Introduction*

As a neophyte utilizing the numerous packages available to the ever-increasing industry used Python language, I pursued the following project in order to have the opportunity to explore and implement tools and skills prized by the Computer Science industry, in particular Data Science.

As a note there are currently two versions of the requested Flaskapp for the Scientific Abstracts Database. One runs on a local machine, one is publicly deployed on Amazon Web Services. Each are available at public Github repositories for code searching, improvements, and expansion.

Local : <https://github.com/noveroa/DataBases/>

AWS : <https://github.com/noveroa/DataBaseAWS>

### **Objective:**

To build a Web-based data capture, storage and analysis system for academic conference papers. The initial data set was based on a collection of text files from three software architecture conferences over a span of the years 1999 to 2014. Each included paper titles, authors, abstracts, affiliations and keywords. The aim is to provide a user with a collection of analysis methods supported by Web-based visualization to explore the abstracts in interesting ways.

### **Learning Objectives:**

The project was originally designed to include three main learning objectives. Two can be coalesced into the second topic, with the third, the inclusion of the public deployment added.

- Design a database for storing textual data and associated metadata:
  - o Python Relational Database
    - Python, Python Pandas, sqlite3
    - Cleaning and parsing complex text files
    - Creation of a database (with interest in relational)
- Implement a Web-based system
  - o Enable a user to submit queries and search based on specific attributes of the data set (i.e. year(s) of publication, author, institution, topics)
  - o Visualize the results using Web-based visualization
    - Python visualization packages : matplotlib, Seaborn, wordcloud
    - Hosting the database in a web-application useable on local browsers
  - o Hosting the database in a web-application useable on local browsers
    - Flask – Hosted Web Application
    - Python, Flask, jQuery, apache, Bootstrap
- Implement and Deploy a public server for the web-based application.
  - o Amazon Web Services
    - Deploying the Flask application useable to customers
    - AWS, Ubuntu, apache, mod-wsgi

```
if __name__ == '__main__':
```

```
    app.debug=True
```

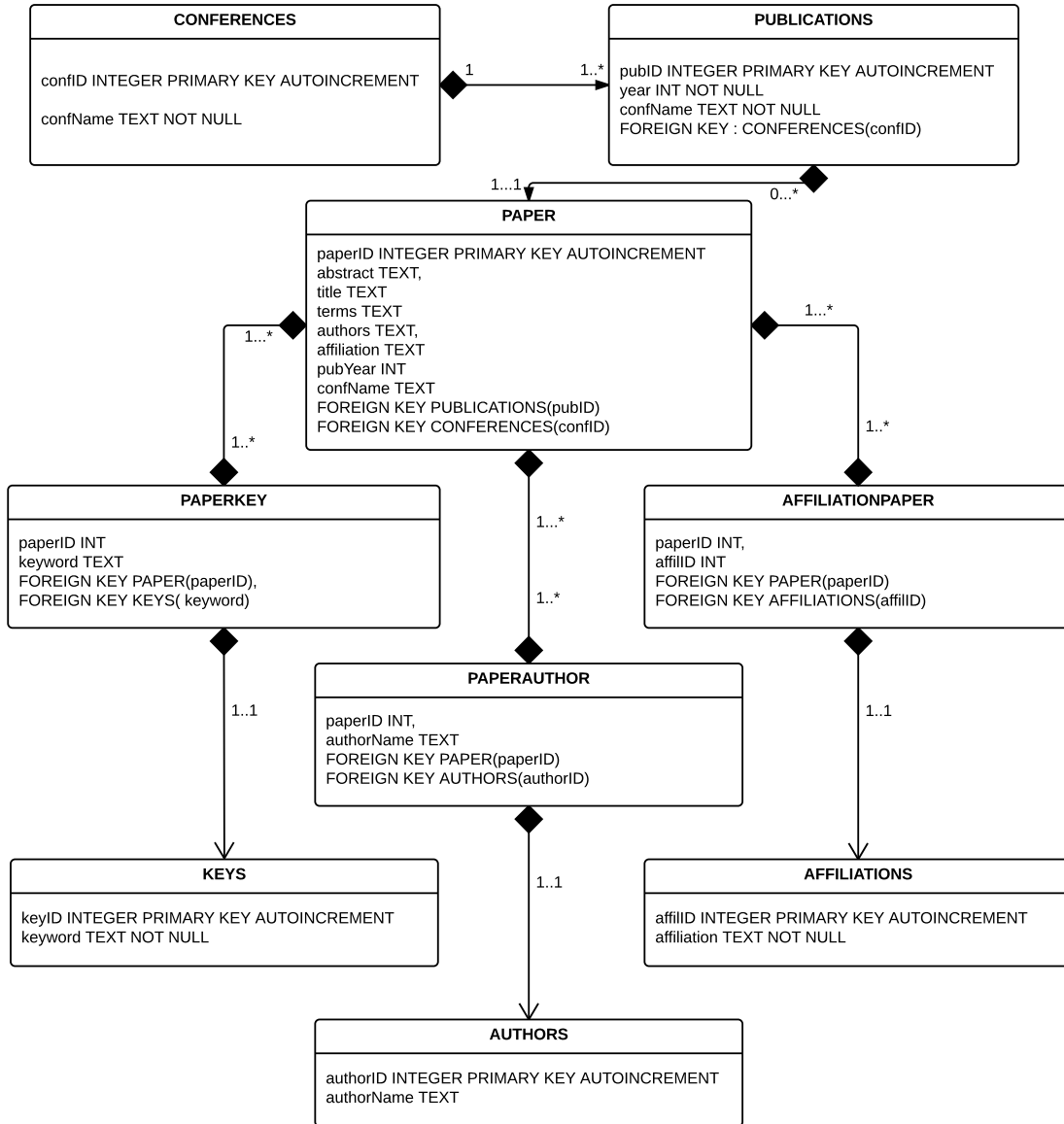
```
    app.run(debug = True)
```

---

## The Database

The database was created implementing the packages for Python utilizing Python Pandas DataFrames and sqlite3, a syntax similar to MySQL. Sqlite was chosen due to its inherent compatibility with python and pandas. There are limits to utilizing sqlite (including no full joins, foreign key issues).

The database schema:



I had inherently begun this process determined to make a relational database using python and sqlite3. Python being a language used more and more often, rather than Java. Also, environments can be created using Anaconda and conda – and maintain updated packages as necessary. Further, Jupyter notebooks are a great learning and sharing tool for testing codes without the command line or writing scripts first. It supports nice markup to also explain what you are working on.

Sample Notebook (with a Pandas DataFrame included)

The screenshot shows a Jupyter Notebook interface with the following content:

### Setting Up Basic Sqlite with Abs Data

- Single Table

```
In [1]: import sqlite3
import pandas as pd
```

### Load the sqlite scripts

```
In [2]: import sqlcommands as cmd
cmd = reload(cmd)
```

### Get Data.

get the pandas Abstracts DataFrame from hd5 file and shape

```
In [3]: rowtotal = cmd.parseDataFrame()
rowtotal[0]
(1548, 8)
```

```
Out[3]: Pandas(Index=1, Abstract='Empirical software engineering focuses on the evaluation of so
ftware engineering technologies, such as processes and tools, by comparing related sets
of data. It has contributed a valuable body of knowledge in several areas such as softw
are economics and software quality, which in turn drove important advances in related to
ols and techniques. Unfortunately this is not yet the case for software architecture,
where empirical studies are still few. Such a condition demands for further empirical r
esearch efforts on the topic of software architecture and suggests specific areas of imp
rovement. In this paper we discuss several essential, innovative, and maybe provocative
questions such as: Why do we have so few applications of empirical software engineering
on software architecture? Which are the main difficulties? What can we do? 21 refs', _
2='1 DISP, Univ. of Rome, Rome, Italy', Authors=['Falessi, D', 'Kruchten, P', 'Canto
ne, G']), _4='C6110B Software engineering techniques C0310F Software management', Conf
='ECSA', Title='Issues in applying empirical software engineering to software architectu
re', terms=['software architecture - software quality', 'empirical software engineering
- software architecture - software economics - software quality'], year='2007')
```

### Create Abstracts Total DB

- can already query on this with pandas

```
In [4]: absDF = cmd.createTOTALTable(rowtotal)
x = list(absDF.columns)
absDF.head()
```

```
Out[4]:
```

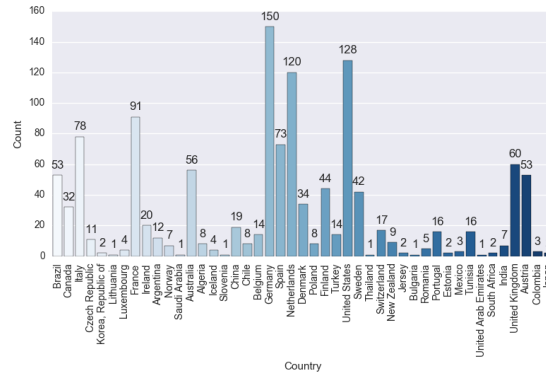
	ID	Abstract	Author affiliation	Authors	Classification Code	Conf	Title	terms	year
0	1	Empirical software engineering focuses on the ...	1 DISP, Univ. of Rome, Rome, Italy	['Falessi, D', 'Kruchten, P', 'Cantone, G']	C6110B Software engineering techniques C0310...	ECSA	Issues in applying empirical software engineer...	['software architecture - software quality', '...	2007
1	2	Software architecture description	1 LSTS ENIT, Tunis, Tunisia	['Jerad, C', 'Barkaoui, K', '...	C6110B Software engineering	ECSA	Hierarchical verification in Maude of	['formal verification - rewriting	2007

Concerning the relational database aspect, although, SQLAlchemy can also be utilized for more manageable migration and foreign key manipulation. I did ultimately start and continued to use sqlite3 alone for the small database. It is interesting to note, though, that Python DataFrames are also an easy and intuitive way to build a database as well, though foreign keys are not supported. The limits here seem to be the amount of data that can be handled and creating well structured normal forms. Yet, many of the manipulations that are done via the creation of normal form tables could be done with a creation of a single DataFrame. Merging, selecting, combining, aggregating, and querying are quite easy. Further one can easily drop duplicates and find counts of unique entries per column or selected columns. Again, speed and memory may be impacted by this approach. Thus, the sql database was used. As the project moved closer to the final stages, more sql-like statements, rather than pandas, were used. I had thought that future collaborators may find it more useful – in particular if the database expands. A good review of the comparisons between pandas and sql:

[http://pandas.pydata.org/pandas-docs/stable/comparison\\_with\\_sql.html](http://pandas.pydata.org/pandas-docs/stable/comparison_with_sql.html)

One can still inspect and manipulate the sql database using pythonic tables; which I do employ throughout the code of the flask. It is a good way to create visualizations and sub/merged tables without using explicit sql syntax which is stricter than pythonic ways. Packages such as wordcloud, scipy, scikit-learn, matplotlib, seaborn are all available to the DataFrame for visual manipulation. If this database had more numeric data, one can format the DataFrames

in such a way to utilize the statistical packages and scikit-learn for data mining and even machine learning which I have done in my past work.



In the creation of the database, itself, the most time-consuming and thus most-likely candidate for improvement would be the parsing of the text files to add to the database. The original files supplied for the database followed no patterns and thus proved difficult to parse correctly.

The code to do so, that is create a DataFrame with basic columns which are further parsed and formatted before being entered into the database and the files for the Abstracts are located in both the local and AWS instances:

- <https://github.com/noveroa/DataBases/DataBaseParsing/>
- [https://github.com/noveroa/DataBaseAWS/Abstracts\\_Gorton](https://github.com/noveroa/DataBaseAWS/Abstracts_Gorton)

To initialize:

- `python intialiazeDBstore.py /<AbstractsFolder>/*.txt`
- You will have a .hdf5 file from which to import to create the sql database.

The code `parsefiles.py` is a script that begins the manipulation of the text files into a single DataFrame saved in an hdf5 file. As the database expands, one may consider writing to a csv file instead. There is much room for refactoring and improvement.

The next step was to create a sqlite database from this python DataFrame. Again the Pandas documentation offers a great path to begin to connect the two. It is nice to have the chance to create a sql database while being comfortable with python and not needing to concentrate on the correct sql syntax, thus allowing for quicker inspection and manipulation (`scripts/sqlcommands.py`). One can recreate the database running `python load_createSqlDB.py` found in `scripts/`.

[http://pandas.pydata.org/pandas-docs/stable/generated/pandas.DataFrame.to\\_sql.html](http://pandas.pydata.org/pandas-docs/stable/generated/pandas.DataFrame.to_sql.html)

Other than the great Stack Overflow, another great site is from Sebastian Raschka is the author of the bestselling book “Python Machine Learning,” whose work I have looked at before.

[http://sebastianraschka.com/Articles/2014\\_sqlite\\_in\\_python\\_tutorial.html](http://sebastianraschka.com/Articles/2014_sqlite_in_python_tutorial.html)

Once you have the columns of the DataFrame as you want them, you can simply read them into Sql. Sql does support foreign key and primary key with auto-increments. For example:

```

with sqlite3.connect(db) as con:
    print ("Opened %s database successfully" %db);
    cur = con.cursor()
    #drops the Abstracts table if it exist in db already
    cur.execute("DROP TABLE IF EXISTS " + table)
    print ('table dropped')

    #create the table
    cur.execute("CREATE TABLE " + table +
                "(pubID INTEGER PRIMARY KEY AUTOINCREMENT, \
                 year INT, \
                 confName TEXT NOT NULL, \
                 FOREIGN KEY(confName) REFERENCES 'hs'(confID))" % foreignKey);

    print('Created %s table' %table);
    data = list(data_frame.groupby(DfCol).count().index.get_values())
    dfPubs = pd.DataFrame(data, columns = tableCol)

    #insert into the table
    dfPubs.to_sql(table, con, flavor='sqlite',
                  schema=None, if_exists='append',
                  index=False, index_label=None,
                  chunksize=None, dtype=None)
    print("Records %s created successfully"%table);

sql = "SELECT * FROM " + table
dfPubs = pd.read_sql_query(sql, con)

#return table as Pandas DataFrame for inspection
return dfPubs

```

It must be noted, now, that the current iteration of the database DOES NOT utilize foreign keys appropriately. This was only discovered in the final week of the project. For sqlite, each time the connection is made to the database the foreign keys must be explicitly turned on. There was a start to this (explicitly calling con.execute("PRAGMA foreign\_keys = ON")) with each connection; however, key errors and bugs are persistent and thus being not necessarily of high priority (One can do a cascade delete the long way – which is employed currently (f\_deletionbyPaperID.py). BUT SHOULD be used for the next iteration of the design.

For each individual table of the relational database the tables are created by first parsing the given columns and reformatting before using. Again this was the hold up of the project. Due to the nature of the given text files, Unicode must be respected which is not a default for python – regex can be used in python and is utilized thusly. This is where many bugs still persist in the database. Much code is spent between Unicode and strings and attempting to resolve such issues. These, however, were inherent from the method chosen to parse the original abstract files – admittedly the file parsing should be reinvestigated, json files may be the key.

#### Notebook to show inserting. deleting a json file into the database

```

import pandas as pd
import sqlite3 as sq
import f_astFull as f
r = reload(f)
#import deletionbyPaperID as delP
delP = reload(delP)

```

```

mydb = 'scripts/Abstracts.sqlite.db'
fFile2 = 'static/data/jsont.json'
fFile2df = r.jsonDF(fFile2)
fFile2df

```

	Abstract	Author affiliation	Authors	Conf	Title	terms	year
abstract	My IQ is one of the highest – and you all know...	NBC and the apocalypse	'The Donald', 'Drumh'	QoSA	this is the USA	'nuclear', 'Software', 'Research', 'four'	2010

```

r.getContent(mydb)
t = r.sqlCMDToPD('ABSTRACTSTOTAL', mydb)
c = r.sqlCMDToPD('CONFERENCES', mydb)
pub = r.sqlCMDToPD('PUBLICATIONS', mydb)
au = r.sqlCMDToPD('AUTHORS', mydb)
k = r.sqlCMDToPD('KEYS', mydb)
pap = r.sqlCMDToPD('PAPER', mydb)
af = r.sqlCMDToPD('AFFILIATIONS', mydb)
pk = r.sqlCMDToPD('PAPERKEY', mydb)
ap = r.sqlCMDToPD('AFFILIATIONPAPER', mydb)
pa = r.sqlCMDToPD('PAPERAUTHOR', mydb)
au.head()

```

```

(1140, 7)
(3, 2)
(27, 3)
(1949, 2)
(5557, 2)
(1102, 8)
(830, 2)
(17436, 2)
(1102, 2)
(2432, 2)

```

```

t = r.sqlCMDToPD('ABSTRACTSTOTAL', mydb)
c = r.sqlCMDToPD('CONFERENCES', mydb)
pub = r.sqlCMDToPD('PUBLICATIONS', mydb)
au = r.sqlCMDToPD('AUTHORS', mydb)
k = r.sqlCMDToPD('KEYS', mydb)
pap = r.sqlCMDToPD('PAPER', mydb)
af = r.sqlCMDToPD('AFFILIATIONS', mydb)
pk = r.sqlCMDToPD('PAPERKEY', mydb)
ap = r.sqlCMDToPD('AFFILIATIONPAPER', mydb)
pa = r.sqlCMDToPD('PAPERAUTHOR', mydb)
pk.tail(10)

```

```

(1141, 7)
(3, 2)
(27, 3)
(1951, 2)
(5560, 2)
(1103, 8)
(831, 2)
(17440, 2)
(1103, 2)
(3434, 2)

```

	paperID	abstract	title	terms	authors	affiliation	pubYear	confName
1101	1102	Without rigorous software development and main...	A case study in incremental architecture based...	['Computer programming languages', 'Computer a...	[Abi Antoun, Marwan', 'C', 'Coeiro, Wesley]	Institute for Software Research Int ISRI, C...	2005	WCSA
1102	1103	My IQ is one of the highest – and you all know...	this is the USA	'nuclear', 'Software', 'Research', 'four'	'The Donald', 'Drumh'	NBC and the apocalypse	2010	QoSA

#### Future Improvements:

The script for creation of the sql database has not been refactored optimally. The abstract text files and the difficulty to parse them correctly was hard to overcome. Hardcoding was removed in most instances, but can be improved. The chance to change the input (ie using JSON files) seems to render this point moot. This is the best way to enhance the speed, cleanness, and ease of the database. With the ability to now insert individual and multiple pseudo json abstract files directly into the database (ie using tags etc) from the running website, the expansion can be done much quicker and cleaner. With the successful implementation of the sqlite foreign keys, deletion will also be improved. Currently, deletion is done via paperID and a long cascade from that (deletionbyPaperID.py).

```
import os, sys

def iterativeJsoninsert(db = mydb, directory = 'static/data'):
    entries = []
    for sub_dir, dirs, files in os.walk(directory):
        for f in files:
            filepath = sub_dir + os.sep + f
            if filepath.endswith('.json'):
                print ("Entering : %s" % filepath)
                df = r.jsonDF(filepath)
                entries.append(filepath)
    return entries

x = iterativeJsoninsert()
pd.DataFrame(x).rename(columns = {0:'File'})

Entering : static/data/json6.json
Entering : static/data/json7.json
```

	File
0	static/data/json6.json
1	static/data/json7.json

Also, it should be explored whether REST api and sql commands are faster (both with regards to time and space) versus the creation and manipulation of Pandas DataFrames. Due to the limited size of the original database, this was not the focus of the project and thus not optimized for it.

***\*\*Regarding the AWS instance in particular for the Relational Database:***

Numerous packages are required to be installed. Due to the environment of Ubuntu/Unix pythonic packages can be difficult to import as is creating virtual environments to ensure proper packages and versions. Currently, the AWS deployment does not create and populate its own Relational database due to limits/instability of the Unix PyTables module. Thus it is currently created in the local and uploaded – ensuring to change permissions to the sudo Ubuntu user and allowing Read and Write. Perhaps, one should explore the use of AWS RDS. Another alternative would to utilize Docker instances to maintain environments. I did begin to explore this as well.

## *The Flask Web Application Queries and Visualizations*

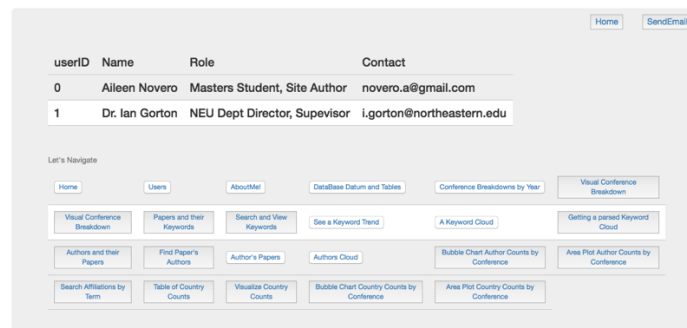
Why did I use Flask? Why not! It is a great tool and often used by companies to host lovely websites. Plus, you can launch a relational database and use python! It is compatible with many languages, has a great community, and can interface with many different packages and languages – including python and using json tools (jsonify!). And it is fun! “*Flask is a microframework for Python based on Werkzeug, Jinja 2...*”

<http://flask.pocoo.org/docs/0.11/>



Following the great tutorial available in the Flask documentations, utilizing Flask, jQuery, bootstrap, CSS, and the Python packages matplotlib and Seaborn the web app of the relational database was launched. The following queries and visualizations are currently available among others as they are constantly being added (I can't leave it be for more than a day if even!) – just send me an email! It is not the prettiest site, but please just explore!

Most of the tables support automatic searching and filtering, reordering the columns and rows interactively as well.



### **Extras:**

- About the site author: Aileen Novero
  - o Links to contact information, GitHub account, Linked In Profile, and Clickable email link
- User page, contact information of site author and academic supervisor for the project
- Welcome page
- Index/Home Page: with links to various tables and visualizations.

### **Authors:**

- Authors and their Papers
  - o View all authored papers
  - o Author Name, Paper Title, Conference, Year Published, Author's Total Publication Count,
- From Conference Links:



- View published authors given a conference and year (searchable)
  - Author Name, Paper ID number, Paper Title, Author's Count in total papers published per that given year, Link to view the paper, Link to delete the Paper
- Author's Papers
  - Search for and view author's metadata of queried author's name
  - Author Name, Paper ID, Paper Title, Conference, Year Published, Delete the Author (*not currently working*)
- Find a Paper's Authors
  - Find the Authors based on querying PaperID.
  - Paper ID number, Author Name, Paper Title, Conference, Year, Author's Count in total papers published per that given year, Link to view the paper, Link to delete the Paper
- Visualize the top overall 20 Authors broken down by count per conference per year:
  - Area Plot Author Counts by Conference
  - Bubble Chart Author Counts by Conference
- Authors Cloud
  - Visualize all Authors as a frequency based WordCloud over all years and conferences

### Conferences:

- Search Conferences
  - Search by pull downs, for and view metadata a queried conference and year
  - Paper ID, Paper Titles, Abstract, Link to view the paper in total, Link to Delete the Paper entry from the Database.
- Conference Break Down By Year:
  - View list of each conference and year with links to the published papers, authors, and top keywords
    - View papers published of given conference and year (searchable)
      - Paper ID, Paper Title, Abstract
    - View authors published of given conference and year (searchable)
    - View top keywords of given conference and year (searchable)
- Visualize Conference Break Down:
  - Visualize each conference broken down by year
  - Table of the publication count, Pie Chart breakdown, and Bar Plot breakdown

### Papers/Affiliations:

- Various Links allow to view metadata of Paper
  - Paper Title, Conference, Year Published, Abstract, Link to Paper's Authors, Link to Delete the Paper
- Search Affiliations by Term
  - View paper metadata from queried affiliation term
  - Paper Id, Affiliations, Link to paper information, Link to Delete the paper
- Table of Country Counts
  - View Country by Count table
  - *Caveat: string search is the way this is implemented.*
- Visualize Country Affiliation by Paper Counts:
  - Bar Chart
  - Area Plot
  - Bubble Chart

### Papers/Keywords:

- Papers and their keywords
  - View and search through papers and the list of their keywords
  - PaperID, Keywords list, Link to the Paper Information
- Search for a Keyword
  - Query for a Keyword and View keyword metadata
  - Paper ID, Paper Title, Conference, Year Published, Link to the Paper Total Entry, Delete the Paper
- See a Trend

- Search for and view metadata trend data queried keyword
  - Table of metadata (paperID, title, conference, and year), Table of Keyword count per conference/year, link to A Heat Map rendering of breakdown
- See Top 20 Trend
  - View table of Top 20 keywords with metadata and link to HeatMap rendering
  - Visualize overall Top 20 keywords broken down by year and conference in HeatMap
- A Keyword Cloud
  - Visualize keywords as a frequency based WordCloud
  - Over all years and conferences
- Getting Parsed Keyword Cloud
  - Cloud Broken down by year
  - Cloud Broken down by conference
    - Can be implemented for authors in future?

### **JSON/REST Implementations**

- JSON files stored in the static/data folder:
- View the JSON file as a pandas dataframe
- Insert the JSON file into the Database
- Delete any abstract via paperID (links in various tables above)
  - CASCADE delete is worked around (currently works for the Key tables and Affiliation primary tables, other than just the composites)
- Delete a table row by element (pk or value; manually employed not cascading)

### **Tables, Relational Database Returns** (all searchable)

- Total Table
  - Abstract, Author affiliation, Authors (nonparsed), Conference Name, Paper Title, Keywords (nonparsed), Publication Year
- Abstracts
  - Paper Title, Year Published, Conference Name, Abstract
- Affiliations
  - Affiliation ID, Affiliation Text
- Authors
  - Author ID, Author Name
- Conferences
  - Conference ID, Conference Name
- Keywords
  - Keyword ID, Keyword
- Papers
  - Paper ID, Paper Title, Publication Year, Conference Name, Authors, Affiliation, Abstract, Keywords
- Publications
  - Publication ID, Conference Name, Publication Year
- Composite Tables
  - AffiliationPaper
    - Paper ID, Affiliation ID
  - PaperKey
    - Paper ID, Keyword ID
  - PaperAuthor
    - Paper ID, Author ID
- Sqlite\_sequence
  - A table of the primary tables (non composite) and their counts.



Research		Search:			
paperID	Title	Conference	PublicationYear	See Total Entry	Delete Paper
54	An exploratory study of architectural practices and challenges in using agile software development approaches	ECSA	2009	<a href="#">View Paper</a>	<a href="#">Delete Paper!</a>
55	An exploratory study of architectural practices and challenges in using agile software development approaches	WICSA	2009	<a href="#">View Paper</a>	<a href="#">Delete Paper!</a>
56	An exploratory study of architectural practices and challenges in using agile software development approaches	WICSA	2011	<a href="#">View Paper</a>	<a href="#">Delete Paper!</a>
83	Stakeholder identification and quality attribute prioritization for a global vehicle control system	ECSA	2010	<a href="#">View Paper</a>	<a href="#">Delete Paper!</a>
94	Characterizing real time reflexion based architecture recovery An in vivo multi case study	QoSA	2012	<a href="#">View Paper</a>	<a href="#">Delete Paper!</a>
403	Evolving industrial software architectures into a software product line A case study	QoSA	2009	<a href="#">View Paper</a>	<a href="#">Delete Paper!</a>

For the future.

One will note the flaskapp is currently run via one python script (welcome\_flask3.py and flaskapp.py on the local and AWS instances respectively). Blueprints can be utilized for this purpose, perhaps. But that was not the focus of the project.

Succinctness of the code, refactoring can be improved as well. This was begun and has made for a bit cleaner code, but improvement is always helpful!

Extendibility will come through the Json and REST api most likely. Such scripts can be found in the folder scripts.

Images/Visualizations:

The reason python and DataFrames were originally employed for the database is the ease and breadth of the data visualization packages available.

Due to the way some images are rendered (statically saved versus dynamic), there is collision potential between multiple users. This was handled effectively in the local instance originally, however moving to AWS caused issues and thus the dynamic rendering was lost – especially with the word cloud rendering. This too causes load times to suffer. Too, to ensure, especially on the AWS instance of the correct image, one must reload the page due to caching bug.

File	Reason
<a href="#">welcome_flask3.py</a>	Run the Flask
<a href="#">f_images.py</a>	Image scripts (Bar Chart, HeatMap, Area Plot, Pie Chart, Bubble Chart ...)
<a href="#">f_wordcloudmaker.py</a>	WordCloud functions
<a href="#">f_RESTful.py</a>	Json inserts and deletes Querying the database
<a href="#">f_deletionbyPaperID.py</a>	Deleting papers from the database Cascade Delete by Paper ID
<a href="#">f_SqlScripts.py</a>	Pythonic SQL scripts
in Scripts/	
<a href="#">load_createSqlDB.py</a>	Run From Command line to build the Database (linked to .hdf5 file of the First Parsing of the text files)
<a href="#">sqlcommands.py</a>	Parses the initial DataFrame from the .hdf5 file and builds the sqlite relational database
<a href="#">terms.py</a>	Parsing of the terms.
Also in Scripts/ there are a few more sample scripts	

*Amazon Web Services  
Public Deployment with Ubuntu Server*

With Professor Gorton as the instigator, AWS via Ubuntu server, was utilized to publicly deploy the database. To explore in real time, one can point the browser to 54.201.113.183.

The basic approach for the initialization the instance followed:  
<http://www.datasciencebytes.com/bytes/2015/02/24/running-a-flask-app-on-aws-ec2/>

Focus on the mod\_wsgi and apache configs – here are the most likely source of errors!

Some key commands to remember:

Command	Reason
ssh -i <abstractsPK.pem> <a href="#">ubuntu@54.201.113.183</a>	SSH into the instance with the proper .pem/.ppk
sudo apachectl restart	If changes are made in the codes while on the AWS instance, to update the live site
<i>While on the remote instance: on remote:</i> ~/flaskapp/ ipython notebook --no-browser --port=8888 <i>Restart a new terminal:</i> ssh -L 8000:localhost:8888 -i Desktop/abstractsPK.pem ubuntu@54.201.113.183 <i>Point your browser to</i> localhost:8000	Launch a Jupyter notebook as an IDE of choice on the local while SSH'ed into Ubuntu.
/var/log/apache2/error.log	Error log! Oh bugs....
source /home/ubuntu/.bashrc	Updating the bash
~/etc/apache2/sites-enabled/000-default.conf	Location of the mod_wsgi configs.
curl http://169.254.169.254/latest/meta-data/public-ipv4	Get the public ip address
grep nameserver /etc/resolv.conf	Get the private ip address
dpkg --get-selections   grep -v deinstall	Get Ubuntu packages
sudo apt-get install <package>	Install Ubuntu packages

The most important idea to remember, and perhaps only due to my naiveté, for the AWS instance were permission and package issues.

If one is running into an error:

1. Check permissions!
2. Check the path!
3. Caching.

