Database SystemsProject Documentation

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https://github.com/johnxnguyen/ElectionDB

Chapter 1

Project Iteration 1

1.1 Introduction

1.1.1 Project Aim

The aim of this project to provide an interactive visualisation of tweet data collected prior to the american elections in 2016 from former presidential candidates Hillary Clinton and Donald J. Trump. It focuses primarily on the usage *hashtags* and aims to facilitate a user-oriented analysis on these hashtags. This includes providing visualizations of relationships between hashtags regarding the frequency of their single and pairwise occurrences; the development of their usages over time; and their inherent importance¹. The final form of the project will be a web application that enables the user to interactively query the underlying database.

1.1.2 Requirements

The following list describes the type of requests that the web application must be able to respond to. Although this list is not exhaustive, the user should at least be able to retrieve the following information:

- which hashtags were used, by whom and in which tweets
- how often a single hashtag, or a pair of hashtags, was used
- who used the most hashtags
- which hashtags were used the most
- in which period hashtags (generally) used the most
- how did the usage of a hashtag (or pair) develop over time
- how the usage of a hashtag relates to the amount of times a tweet was retweet or favorited

1.1.3 The Team

This project is being developed by John Nguyen, Gavithishan Ravechandran and Khanh Bui Trong. We are all currently studying a Bachelor of Computer Science at the Freie Universität in Berlin, Germany.

¹The importance of a hashtag is interpreted as the accumulative impact of the tweets that it belongs to.

1.2 Data Analysis

The data used in this project is a collation of tweet data produced by former american presidential candidates Hillary Clinton and Donald J. Trump prior to the 2016 election. The tweets were created between January and October and consist of the following components:

- handle: indicates the user who posted the tweet
- text: the actual tweet content
- is retweet: indicates if the tweet is a retweet
- original author: indicates the original author of the tweet (applicable only to retweets)
- time: indicates the date and time the tweet was created
- in reply to screen name: if the tweet is a reply, this indicates who the reply is directed to
- is quote status: indicates if the tweet was quoted tweet
- retweet count: indicates the number of times the tweet was retweeted (by other users)
- favorite count: indicates how many times the tweet was favorited (by other users)
- source url: indicates from which platform (device) the tweet was published
- truncated: indicates if the tweet content was truncated

It is important note that not all components are necessarily relevant for the purposes of this project. For this reason the components in reply to screen name, is quote status, source url and truncated will be excluded from the data model (section 3).

For the remaining components we will briefly describe their importance: handle, text and time are core components that identify and describe particular tweet and the hash-tags contained within; favorite count and retweet count describes the impact of the tweet; and finally is retweet and original author may provide additional contextual information about the tweet.

1.3 Data Model

1.3.1 The Entity-Relationship Model

The Entity-Relationship model is depicted in Figure 1.1. It consists of a single tweetentity with a single value attribute for each component with the exception of ID and hashtags. Hashtags contained within the tweet text are represented as a multivalued attribute.

The key ID attribute is a composed from handle and time. The reason for this is because no single simple attribute forms a key since it can not be guaranteed that the value of this attribute will be distinct among multiple tweets objects. In other words, a pair of tweets may share a common value for each of the attributes. In order to uniquely identify a tweet, one must consider a combination of simple attributes. On the assumption that a single user may not publish two distinct tweets simultaneously, the paired value of handle and time forms a key².

A hashtag is not represented as an entity (nor their relationship to a tweet through a corresponding relationship type) because it increases unnecessary complexity to the model. A hashtag entity would have only a single attribute (to specify the hashtag name) and participate in a single relationship with the tweet entity. This is more simply expressed using the multivalued approach as in Figure 1.1. One concern with the current model, however, is its capacity to express pairwise relationships between distinct hashtags. As we shall see in the next section, our ER-model does not limit us to fulfill this requirement, since its translation to the relational model expresses a hashtag as distinct relation³.

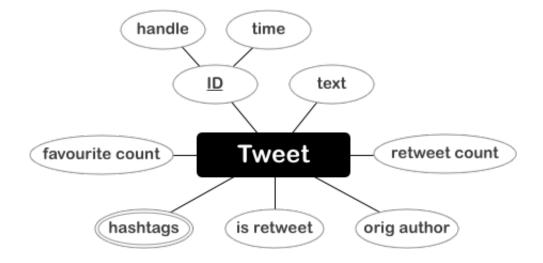


Figure 1.1: The Entity-Relationship model of the dataset

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,

²Although the time attribute is only precise to the second, the key remains valid in our (static) dataset.

³When discussing relational models, we will use the terms *relation* and *table* interchangeably.

1.3.2 The Relational Model

The ER-model translates⁴ into the relational model as illustrated in Figure 1.2. It consists of two relations, one corresponding to the a tweet and the other to a hashtag. The primary key of the Tweet-relation is the tuple (handle, time), as defined in section 1.3.1.

The Hashtag-relation contains three attributes, the first of which, tag, stores the actual hashtag. The remaining two attributes tweetHandle and tweetTime form a foreign key referencing the primary key of the Tweet-relation. All three attributes then form the primary key of Hashtag-relation.

The following list specifies the domains and constraints of each attribute:

handle: The domain is all alphanumeric (including underscores '-') strings with maximum 15 characters⁵. This attribute must have a value, i.e, it is constrained be NOT NULL, because it is part of the primary key. These constraints are defined by Twitter⁶.

text: The domain is all strings containing up to 140 characters. This attribute is constrained to be NOT NULL. These constrainst are also defined by Twitter⁷.

time: The domain is all timestamps in the format YYYY-MM-DD HH:MM:SS. This attribute is constrained to be NOT NULL because it is part of the primary key.

handle text time isRetweet origAuth retweetCount favCount Hashtag tag tweetHandle tweetTime

Figure 1.2: The relational model of the dataset

Tweet

⁴The algorithm used is from: Fundamentals of Database Systems (7ED), Elmasri & Navathe, 2016, ch. 9

⁵More specifically, this is any string matching the regular expression [a-zA-Z0-9_]{1,15}.

⁶See https://support.twitter.com/articles/101299#unavailable

⁷See https://support.twitter.com/articles/15367

isRetweet: The domain is the set of boolean values $\{true, false\}$. This attribute is naturally constrained to be NOT NULL.

origAuthor: The domain is the same as the handle attribute since it refers to another twitter user. The attribute will contain a value only if isRetweet set to true.

retweetCount & favCount: The domain is the set of positive whole numbers and the attribute is constrainted to be NOT NULL, since a tweet must be retweeted 0 or more times.

tag: The domain is the set of all valid hashtags; strings containing alphanumeric characters (including underscores) but beginning with the # symbol and otherwise containing at least one alphabetic letter. Because the entire tag is part of the tweet text, it may not exceed 140 characters in length⁸. Due to this attribute being part of the primary key for the Hashtag-relation, it is constrained to be NOT NULL.

1.3.3 Important Considerations

The Hashtag-relation does not represent single hashtag objects, but rather the occurrences of hashtags. Thus each tuple in the Hashtag-relation pairs together the hashtag used, and the tweet it was used in.

Since the primary key of the Hashtag-relation is the set of the tag, tweetHandle and tweetTime attributes, each tuple within the relation must have a unique combination of values for these attributes. As a consequence, no single hashtag may occur multiple times in a single tweet because an attempt to insert two occurrences of the same hashtag from the same tweet leads to two tuples with equal attribute values. This is a violation of the primary key constraint. To avoid this violation it is required that the occurrence of a hashtag within a tweet is to be counted at most once. This requirement does not adversely affect the goals of the projects since the functionality of a hashtag depends solely on its existence rather than its quantity.

⁸This is any string matching the regular expression $\#\w*[a-zA-Z]\w*$.

1.4 Constructing the Database

The database management system used in this project is PostgreSQL (version 9.6). It can be installed by the $Homebrew^9$ package manager via the command line.

The PostgreSQL installation includes the command line application psql, with which we can execute the SQL statement required to create the database. Entering the command psql starts the application. Once inside psql, we can execute the following SQL statement:

CREATE DATABASE election;

This command creates a empty database called **election** based on the default settings. At this point in the project this is sufficient for experimentation purposes, however it may be necessary later to adjust the database settings to specify user access privileges and server information. We can confirm the database was created by entering the command:

 $\backslash 1$.

This lists all the exisiting databases. To quit psql simply enter the command:

 $\backslash q$

⁹for macOS see https://brew.sh, or alternatively for linux see http://linuxbrew.sh. For instructions on installing PostgreSQL see https://gist.github.com/sgnl/609557ebacd3378f3b72.

Chapter 2

Project Iteration 2

2.1 Database Schema

Section 1.3.2 outlines in detail the database schemas needed to be constructed. We must simply express the domains and constraints with SQL statements. Firstly we define two domains, handle_type and natural_num, which correspond to the set of all valid twitter handles and the set of all positive whole numbers. Note that PostgreSQL allows the use of regular expressions within CHECK constraints, which we will utilize to validate the twitter handles. The following SQL statement defines the domains:

```
CREATE DOMAIN HANDLE_TYPE AS VARCHAR(15) CHECK (VALUE ~ '^\w{1,15}$');
CREATE DOMAIN NATURAL_NUM AS INTEGER NOT NULL CHECK (VALUE >= 0);
```

Next we define the schemas corresponding to the *Tweets* and *Hashtags* relations with the following SQL:

```
CREATE TABLE Tweets (
      handle
                       HANDLE_TYPE
                                        NOT NULL,
      content
                       VARCHAR(140)
                                        NOT NULL,
                       TIMESTAMP
                                        NOT NULL,
      t_stamp
                                        NOT NULL,
      is_retweet
                       BOOLEAN
      orig_author
                       HANDLE_TYPE,
      retweet_count
                       NATURAL_NUM,
      fav_count
                       NATURAL_NUM,
      PRIMARY KEY (handle, t_stamp)
  );
  CREATE TABLE Hashtags (
                       VARCHAR(140)
                                     NOT NULL
        CHECK (tag ^{\sim} , ^{+}\w*[a-zA-Z]\w*$,),
        tweet_handle HANDLE_TYPE
                                      NOT NULL,
        tweet_t_stamp TIMESTAMP
                                      NOT NULL,
        PRIMARY KEY (tag, tweet_handle, tweet_t_stamp),
        FOREIGN KEY (tweet_handle, tweet_t_stamp)
        REFERENCES Tweets(handle, t_stamp)
        ON UPDATE CASCADE ON DELETE CASCADE
    );
```

We note that if a tweet is updated or deleted from the Tweets relation, then any hashtags from the Hashtag relation that references this tweet will also be updated or deleted.

The Python script in listing 2.1 creates the relation schemas by first connecting to the election databse and then sequentially executing each of the SQL statements above. Note that before the domains and tables are created, the script first checks to see if they exist, and if so, they are deleted. This allows the program to be run even if the database is not empty and is included as a convenience. The *psycopg2* Python package is used to communicate with PostgreSQL.

Listing 2.1: table_creator.py

```
1 #!/usr/bin/python
  import psycopg2
  from config import db_config
4
5
   table_commands = (
        11 11 11
6
7
       DROP TABLE IF EXISTS Tweets CASCADE;
8
       DROP TABLE IF EXISTS Hashtags;
9
       DROP DOMAIN IF EXISTS HANDLE_TYPE;
10
       DROP DOMAIN IF EXISTS NATURAL_NUM;
        " " ,
11
        11 11 11
12
13
        CREATE DOMAIN HANDLE_TYPE AS VARCHAR(15) CHECK (VALUE ~ '^\w-
           {1,15}$');
        CREATE DOMAIN NATURAL_NUM AS INTEGER NOT NULL CHECK (VALUE \hookleftarrow
14
        " " ,
15
        11 11 11
16
17
        CREATE TABLE Tweets (
18
                              HANDLE_TYPE
                                                NOT NULL,
            handle
19
            content
                              VARCHAR (140)
                                                NOT NULL,
20
            time
                              TIMESTAMP
                                                NOT NULL,
21
            is_retweet
                              BOOLEAN
                                                NOT NULL,
22
            orig_author
                              HANDLE_TYPE,
            retweet_count
                              NATURAL_NUM,
23
24
            fav_count
                              NATURAL_NUM,
            PRIMARY KEY (handle, time)
25
       );
26
27
        " " ,
        11 11 11
28
29
        CREATE TABLE Hashtags (
```

```
30
                           VARCHAR(140) NOT NULL CHECK (tag ~ ', ** ', w←
              *[a-zA-Z]\setminus w*\$'),
            tweet_handle HANDLE_TYPE
31
                                          NOT NULL,
32
            tweet_time
                           TIMESTAMP
                                          NOT NULL,
33
            PRIMARY KEY (tag, tweet_handle, tweet_time),
34
            FOREIGN KEY (tweet_handle, tweet_time)
35
            REFERENCES Tweets(handle, time)
36
            ON UPDATE CASCADE ON DELETE CASCADE
37
       );
       11 11 11
38
39 )
40
41
  def create_tables():
42
43
       conn = None
44
45
       try:
            # connect to the PostgreSQL server
46
            print "\nconnecting to DB.."
47
            # get DB info
48
            params = db_config()
49
50
            conn = psycopg2.connect(**params)
51
            cur = conn.cursor()
            # execute commmands
52
            print "creating tables..."
53
54
            for command in table_commands:
                cur.execute(command)
55
56
57
            # commit the changes
58
            conn.commit()
            cur.close()
59
            print "DONE"
60
61
62
       except (Exception, psycopg2.DatabaseError) as error:
            print "ERROR:", error
63
64
            exit()
65
66
       finally:
67
            if conn is not None:
68
                conn.close()
```

2.2 Data Cleaning

In order to correctly import the data into the database, we must first prepare and clean the raw data. This involves extracting only the necessary data, formatting the data components into the expected data types as defined by the corresponding relation schema, as well as removing or correcting any other anomalies.

Of all the attributes listed in the schemas of the previous section, only content and t_stamp attributes need to be reformatted.

content: In the raw data file, some tweet entries contain symbols encoded for web usages. These include &, <, and >, and appear in the raw data as &, <, and > respectively. Thus we must replace each occurrence of these encoded symbols with the decoded symbols they represent.

t_stamp: In the raw data file, this element has the form yyyy-mm-ddThh:mm:ss; for example 2017-01-01T00:00:00 represents midnight on the 1st of January 2017. However, the TIMESTAMP data type in PostgreSQL is expected to be in the form yyyy-mm-dd hh:mm:ss; the only difference is absence of the letter T. Therefore we must replace the letter T with a space for every entry of this data element in the raw data.

All the other data elements corresponding to the remaining attributes are already in the appropriate format and thus no additional conversions are required.

The Python script shown in listing 2.2 automates this process by reading the raw data, performing the conversions, and writing only the needed (and cleaned) data into a new file. The raw data has been provided in two different file formats: xlsx (excel spreadsheet) and csv (comma separated values). We have chosen to work with the xlsx file because it is better structured compare to the csv file. Each line of csv file does not always represent a complete tweet entry; multiline tweets (those with newline characters) are spread across multiple lines within the csv file and thus pose a problem when trying to parse this file.

This problem does not arise when parsing a spreadsheet; all tweet elements are contained within individual cells, including multiline tweets. Thus the program can easily iterate through each row and then through each cell to extract the elements accordingly. Furthermore, the standard Python csv parsing module does not support unicode strings (and therefore also the emojis contained in the tweets), whereas the spreadsheet does. The program utilizes the openpyxl package for reading and writing the xlsx file format.

```
1 #!/usr/bin/python
2 from openpyxl import Workbook
3 from openpyxl import load_workbook
4 from config import data_config
5
6 # data is an array: [handle, content, isRetweet, origAuthor, \hookleftarrow
      time, inReplyTo, isQuoteStatus, retweetCount, favCount, \hookleftarrow
      sourceURLl
  def clean_row(data):
8
       # unpack columns needed
9
       handle, content, isRetweet, origAuthor, time = data[:5]
10
       retweetCount, favCount = data[7:9]
11
       # convert encoded symbols, if any
12
       content = content.replace("&", "&")
       content = content.replace("<", "<")</pre>
13
14
       content = content.replace(">", ">")
15
       # reformat time stamp
       time = time.replace("T"," ")
16
17
       # return only needed columns
18
       return [handle, content, time, isRetweet, origAuthor, \leftarrow
          retweetCount, favCount]
19
20
  def clean_data():
21
       # open raw data file & get the worksheet
22
       input_file = data_config['input_filename']
23
       print '\nopening file:', input_file, '...'
       # a workbook is an xlsx file
24
25
       input_wb = load_workbook(input_file, read_only = True)
       # a worksheet is a sheet containing rows
26
27
       input_ws = input_wb[data_config['sheet_name']]
28
29
       # create new workbook with a new worksheet
30
       output_wb = Workbook()
31
       output_ws = output_wb.active
32
       output_ws.title = data_config['sheet_name']
33
34
       print 'cleaning data...'
35
       # for each row in input sheet
       for row in list(input_ws.rows):
36
```

```
37
           # extract cell data
           row_data = []
38
           for cell in row:
39
               row_data.append(cell.value)
40
41
           # clean the data & add to new sheet
42
           output_ws.append(clean_row(row_data))
43
       # save wb
44
       output_wb.save(filename = data_config['clean_filename'])
45
       print 'DONE: saved file:', data_config['clean_filename']
46
```

2.3 Importing the data

The Python script in listing 2.3 automates the insertion of the cleaned data into the database. It first makes a connection to the database, then iterates through each row of the data file (xlsx format) and inserts both the tweet data and the hashtag data contained in this row. The structure of the data file corresponds to the Tweets table and thus we can simply insert the whole row directly into this table. The hashtag data must be extracted from the current tweet content, which is achieved by using a regular expression to find any embedded hashtags. As noted in section 1.3.3, only distinct hashtags are extracted. Once a list of distinct hashtags is obtained, the program then inserts a new row into the Hashtags table for each hashtag with a reference to the current tweet.

Finally, the insertions are committed to the database and the number of insertions is reported back to the user.

Listing 2.3: table_populator.py

```
1 #!/usr/bin/python
2 import psycopg2
3 from config import db_config
4 from config import data_config
  from openpyxl import load_workbook
6
  import re
7
8
   def populate_tables():
9
10
       conn = None
11
       tweet_count = 0
12
       hashtag_count = 0
13
14
       # open the xlsx spreadsheet containing clean data
15
       filename = data_config['clean_filename']
       print "\nopening data file:", filename
16
       wb = load_workbook(filename, read_only = True)
17
       ws = wb[data_config['sheet_name']]
18
19
20
       # exclude first row (column names)
21
       rows = list(ws.rows)[1:]
22
       print "Number of tweets to insert:", len(rows)
23
```

```
24
       try:
25
            # connect to the PostgreSQL server
            print "connecting to DB.."
26
27
            params = db_config()
28
            conn = psycopg2.connect(**params)
            cur = conn.cursor()
29
30
            print "populating tables..."
31
32
            for row in rows:
                # extract cell values from row
33
                row = map(lambda cell: cell.value, row)
34
35
                # insert tweet
36
                sql = "INSERT INTO Tweets VALUES (%s, %s, %s, %s, %s, %s, %s ←
                   ,%s)"
37
                cur.execute(sql, row)
38
                tweet_count += 1
39
                # get hashtags from tweet
40
41
                hashtags = extract_hashtags(row)
                sql = "INSERT INTO Hashtags VALUES (%s, %s, %s)"
42
                # for each tag
43
44
                for tag in hashtags:
45
                    # insert tag, handle, timestamp
                    cur.execute(sql, (tag, row[0], row[2]))
46
                    hashtag_count += 1
47
48
            # commit the changes
49
50
            conn.commit()
51
            cur.close()
            print "DONE:", tweet_count, "tweets and", hashtag_count, ←
52
                "hashtags inserted.\n"
53
       except (Exception, psycopg2.DatabaseError) as error:
54
            print "ERROR:", error
55
            exit
56
57
58
       finally:
59
            if conn is not None:
60
                conn.close()
61
62
```

```
63 def extract_hashtags(tweet):
64  # extract hashtags
65  hashtags = re.findall("#\\w*[a-zA-Z]\\w*", tweet[1])
66  # remove duplicate tags
67  return list(set(hashtags))
```

Executing the code 2.4

For simplicity, the code presented in the previous three sections have been collated into a

single Python script called dbBuilder.py. Running this script within terminal from the

working directory will clean the raw data, save it to a new file, connect to the database,

create the schemas and finally import the cleaned data. It is assumed the database

election already exists and is configured, however it is not required that the database is

empty because the program will drop all schemas before recreating them.

The database name, username, and password are stored in the config.py file, which

is used to establish the connection to PostgreSQL. These parameters must match the

configuration of the local machine in order for the connection to be successful.

Web Server 2.5

We have chosen to use the open source server software Apache, which is included in the

macOS Sierra operating system. It is configured at both the system wide level and the

user level. The server root directory on the system level is located at:

/Library/WebServer/Documents

and can be accessed in the web browers at the url:

http://localhost

The server root directory on the user level is located at:

~/Sites

and can be accessed in the web browers at the url:

http://localhost/~username

where username is a placeholder for the username of on the local machine.