

Research and Innovation

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Abstract

TODO: modify the abstract s.t. it fits to the content The aim of this project is to find and visualize the events posted on Meetup.com in different cities in order to analyze the density of planned activities and to make a social study regarding typical meeting times and most usual activities and interests. One aspect may concern different meeting times in Spain and Germany or group size. We will focus our research on cultural activities like languages tandems and sports. The data shall be obtained via the Meetup-API and OpenStreetMap. The most important data source for this project is the Meetup-API where we have to request the events with different Python packages like “requests” or “meetup-api”. Our second task is to visualize the locations of the events on a given city map obtained by data resources like OpenStreetMap. To plot the locations we could use “matplotlib” or similar Python plotting libraries. One further aspect could be to look for interest of twitter users to make recommends to activities nearby. The geotag of the tweets provides the central information for the search.

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1 Summary and Goals

The human being is social. Every day there are millions of events. Thousands of them are posted at the webpage [Meetup.com](#). It is a website where people can seek and find other people with the same interest. It helps to organize events and makes them public.

This is the reason why we chose [Meetup.com](#) to analyze the behavior and the preferences of cities and its inhabitants all over the world. Almost all [Meetup.com](#) activities contain information about their location, expressed in terms of latitude and longitude. Thus, by using the [Meetup.com](#) API, we can request and extract a dataset "latitude-longitude" for all the available [Meetup.com](#) events. Then, we can use this dataset to plot all these located activities on a [GoogleMaps.com](#) map. Our aim, here, is to create and visualize a density map, a so-called heatmap, where one can see all the activities for a selected city.

Furthermore, [Meetup.com](#) API allows the user to get activities arranged by category. We want to use this tool to consider separately different types of events like "Food & Drink" or "Sports" and study their relative predominance in different cities. By combining this information with the geographical and/or political structure of a city, one may also identify trending districts and neighborhoods in a city. To do so we make use of GeoJSON-files which can contain the polygonal shape of a district and its name. There are several websites where one can get GeoJSON data or create his own data. By using the information from these websites, a GeoJSON dataset is created with all the coordinates of the districts for all the studied cities. This constitutes the third data source that is used in this work.

A fourth data source is [Wikipedia.org](#) where a web scraping script is applied on to extract some useful population data such as the population number, the population density or the land area of each district. With all this information, one can use the following factor to evaluate the social activity per district. This factor is given as

$$\text{Social activity} = \frac{\# \text{ events}}{\# \text{ person}}.$$

Once downloaded, we can access to our data off-line. This is due to the fact that we store it locally by using csv (comma-separated files).

This report is structured as follows in 6 sections. Firstly, in section 2 we discuss the data life-cycle of our data. In section 3, we present both, the third-party tools that we used and the custom methods that we developed in our own, to achieve this project. They mainly, are developed by using Python. Afterwards, we present our results in section 4 and, in section 5, we make a legal and ethical disquisition. Finally, in section 6 we discuss the limitations of our work and make proposals for future work.

2 Data Life-cycle

In the following section we have a closer look at the terms "data life-cycle" and "data life-cycle management".

"Data life-cycle" is a term which tries to describe the life of data and information. Since there is no unique definition of a data life-cycle, we define the following six stages: creation, storage, use, sharing, archiving and destruction [7], [9].

1. **Creation:** data is created. It can be created as a structured or unstructured set of data, can be obtained by collection, measurements, generated or gathered. We create our data by using the [Meetup.com API](#). By performing a web scraping at [Wikipedia.org](#) we gather the number of population, the population density and the land area for different cities and their districts. In this way, we request and create the data which is going to be needed for this project.
2. **Storage:** once the data is created, it has to be stored. Depending on where we want to apply this data for, there are different and optimal kinds of storage solutions. However, this is not subject of this report. In our case, both, the data gathered from [Meetup.com](#) and [Wikipedia.org](#), are saved as csv-files on the hard drive.
3. **Use:** data can be viewed, modified, analyzed, corrected, interpreted, visualized, joined,... We use our data to visualize the social activity of cities. We join our data from different sources and visualize it on a given map which can be seen as data as well. This data visualization and its further treatment can be used to obtain new data such as the measurement of the social activity per district introduced in section 1.
4. **Sharing:** data can be shared. There are several ways of sharing by using different file formats, applications or operating environments. In this work, we used csv-files and GeoJSON-files to share data from one

method to another. For instance, the district coordinates are read from a GeoJSON-file, the Meetup events and the population data from a csv-file. Then, these are the files created by the scripts to share data along different methods.

5. **Archiving:** when the data leaves the active use, one should archive it in a suitable way. Since archiving is very similar to storaging, one can use the same methods. One difference may be that saved data needs to be compressed to save storage or protected to prevent unwanted changes. Therefore, the access is slightly more difficult than for data in active use. The file formats used in this project does not satisfy this statement. This is due to the fact that the file formats are chosen to ease the access to their data rather than to save storage space. If we want to keep this data for a long time, we should think on developing a method to compress this data.
6. **Destruction:** there is a last stage in the data life-cycle. There are several reasons why one should destroy data. On one hand, the volume of archived data increases and one needs to save storage to store new data. On the other hand, data gets old and could be not needed anymore. Or, if it is not up to date, it can be replaced by newer versions. The second case is applicable for our work. During every single run of the program new csv-files with event data and new csv-files containing populations data are created. Especially the event data underlie a rapid change.

The term "data life-cycle management" describes the process that helps to organize the flow of data throughout its life-cycle.

3 Tools and Data

3.1 Tools

The main work for our project is done in the Python programming language in version 3.6 [8]. The most notable used Python libraries are requests [3], gmaps [6], Jupyter [4] and matplotlib [2]. The requests package is used to call and request data from the Meetup API [5].

With Jupyter Notebook and gmaps we are able to plot the obtained data from the Meetup API, Wikipedia and other resources in a browser. In detail Jupyter provides the general browser session functionality whereas gmaps extends Jupyter with an interactive map provided by Google Maps [1]. One

can visualize specific points or areas in various colors and densities to create meaningful maps. Matplotlib is used to transform the data from population density into specific color schemes.

3.2 Data

As already noted we used the following data sources:

- Meetup API: To get the events in the examined cities.
- Wikipedia: For obtaining data about the population density of the cities and their districts.
- Various resources for GeoJSON files: To visualize districts onto Google Maps. **TODO**
- Google Maps: As a general back-end for our visualization.

4 Results

To draw meaningful conclusions we created several types of visualization of the data. First and foremost we plotted the open events onto a map - either as a point map or as a occurrences distribution map for each city district.

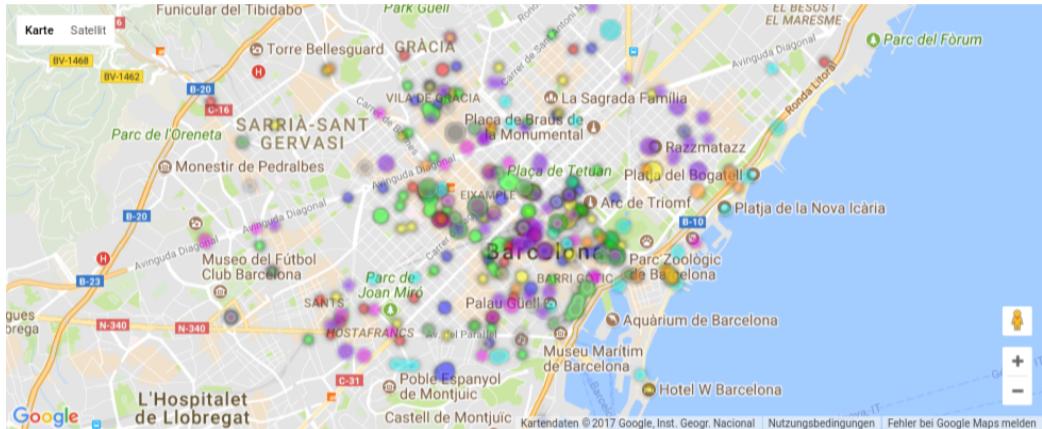
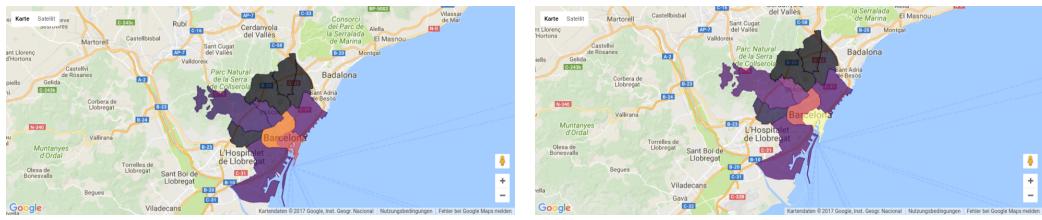


Figure 1: Snippet of Barcelona with Open Events

In Figure 1 one can see an example of Barcelona with open events plotted as points. The different colors represent different meetup categories.



Figure 2: Snippet of Barcelona with Language & Ethnic Identity Events



(a) Activity per District

(b) Activity per Capita per District

Figure 3: Barcelona



(a) Activity per District

(b) Activity per Capita per District

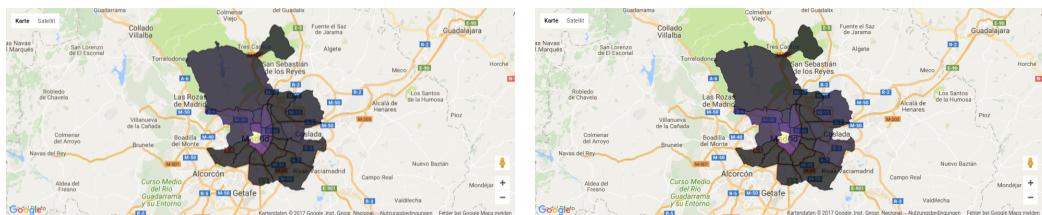
Figure 4: London



(a) Activity per District

(b) Activity per Capita per District

Figure 5: Berlin



(a) Activity per District

(b) Activity per Capita per District

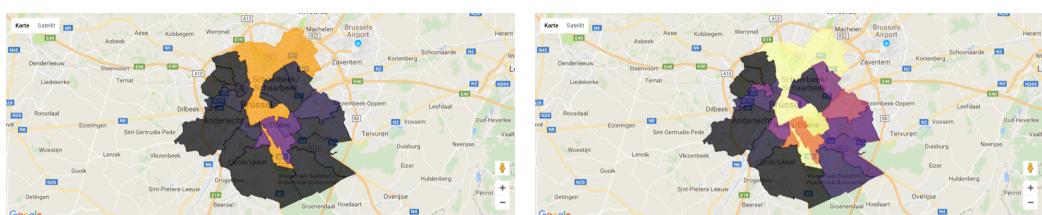
Figure 6: Madrid



(a) Activity per District

(b) Activity per Capita per District

Figure 7: Paris



(a) Activity per District

(b) Activity per Capita per District

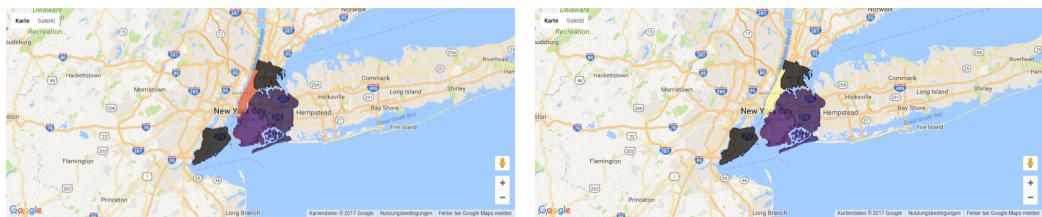
Figure 8: Brussels



(a) Activity per District

(b) Activity per Capita per District

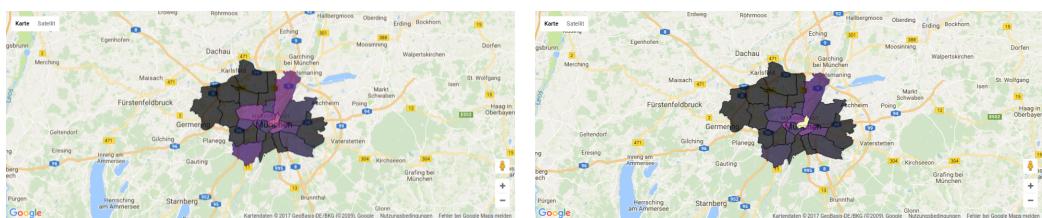
Figure 9: Hamburg



(a) Activity per District

(b) Activity per Capita per District

Figure 10: New York City



(a) Activity per District

(b) Activity per Capita per District

Figure 11: Munich



(a) Activity per District

(b) Activity per Capita per District

Figure 12: Hong Kong

5 Legal and Ethical Issues

6 Limitations and Future Work

One major issue is that this project is depending strongly on the quality of the data available at [Meetup.com](#). We have to rely on the users of [Meetup.com](#) so this issue can hardly be improved. One idea would be to integrate other social services like [Facebook.com](#) to extend our data source and to get more independent. In this way one could also fix the problem that [Meetup.com](#) might be less known in non-English speaking countries. Having a wide base as data source one gets more independent of local and regional preferences regarding the social networks.

Another problem is that this project has the potential to be extended in the following way: So far our program is restricted to predefined cities. It is not possible to select new cities or regions. This is due to the fact that we work with GeoJSON-files which are saved locally on the hard drive. **The second problem is that some Wikipedia pages for the cities are not coherent so that one needs to be careful.** One improvement could be to create a user interface and the possibility to select new cities and regions. By selecting single categories of events one could get special information, e.g. where to go running. Additional, in an application, one could automatically create suitable graphics, charts and tables.

Another limitation is as well related to [Meetup.com](#). Since we only have one data source for events, we only can give sufficient criterions for districts. A district having lots of events seems to be interesting. However, a district having less events is not automatically less interesting and popular. It just means that its inhabitants are not using [Meetup.com](#). Again, one should take more data sources into account.

Our tool could help social scientists to provide their hypothesis with proving data. Beside literature, studies and surveys our tool seems to be a good method for analyzing the behaviour and preferences of populations.

Appendix

Sources of GeoJSON Files

References

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