

## Food Service Opportunities and Competition in the Kansas City Area: Data Sources and Python

Code

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### **Links to Data Sources**

The two data sources for this project are:

- [KCMO Food Permits](#)
- [Food Establishment Complaints](#)

### **Data Preparation and Python Code**

Data cleaning is at the heart of data preparation. The KCMO Food Permits data set is thoroughly cleaned to eliminate and correct irrelevant parts of the data, particularly in the “Establishment Name” column. In Python, irrelevant columns are dropped on both data sets. As shown in Fig. 1, there are many irrelevant parts to data defining Taco Bell establishments. The objective of data preparation is to have one stable format and name for all taco bell establishments. Failure to do so will result in inaccurate analysis and decision-making (data quality). Duplicate records, as shown in Fig. 2, are also removed for an accurate analysis.

The data set rows are also formatted to achieve consistency before analysis. Inconsistency in parts of data could eventuate into a contradictory analysis due to poor data quality. Failure to address inconsistency and irrelevant parts to the data could cost the entire project time, as the entire process ought to be repeated when subsequent issues transpire. Overall, the data preparation process, in particular data cleaning, consumes a sizeable chunk of time in this project. Gonzales (2004) notes “approximately 70 percent of data mining activity is focused on data preparation and cleaning” (p. 14).

Facility Permit #	Establishment Name	Facility Permit #	Establishment Name
106790	Taco Bell #29844	1	2600 Taco Bell
11703	TACO BELL #2874	1	4479 Taco Bell
31432	TACO BELL	1	11298 Taco Bell
4479	Taco Bell	1	11703 Taco Bell
101161	Taco Bell #022580	1	29981 Taco Bell
101566	Taco Bell @ Tuileries Plaza Center	1	31432 Taco Bell
29981	Taco Bell #2953	1	100216 Taco Bell
11298	Taco Bell #4136	1	101161 Taco Bell
2600	TACO BELL #3266	1	101566 Taco Bell
102362	Taco Bell #21436	1	102362 Taco Bell
103809	Taco Bell #420 - Linwood	1	103809 Taco Bell
106763	Taco Bell #29842	1	106762 Taco Bell
106763	Taco Bell #29842	1	106763 Taco Bell
106763	Taco Bell #29842	1	106790 Taco Bell
106763	Taco Bell #29842	1	106799 Taco Bell
106763	Taco Bell #29842	1	106973 Taco Bell

*Fig 1. Removal of irrelevant parts of data.*

e) Establishment Name	Facility Na Facility Address	Facility Permit #	Establishment Name	Facility Address
3 Aramark	Bartle Hall 301 W. 13th. St.	435	100583 Aramark	301 W. 13th. St.
0 Aramark	Bartle Hall 301 W. 13th. St.			
5 Aramark	Aramark a 601 E 12th St	549	101495 Aramark	601 E 12th St
9 Aramark	Royals Ara 1 Royal Way			
0 Aramark	Royals Ara 1 Royal Way			
0 Aramark	Royals Ara 1 Royal Way			
2 Aramark	Royals Ara 1 Royal Way			
4 Aramark	Royals Ara 1 Royal Way			
5 Aramark	Royals Ara 1 Royal Way			
6 Aramark	Royals Ara 1 Royal Way			
7 Aramark	Royals Ara 1 Royal Way			
8 Aramark	Royals Ara 1 Royal Way			
2 Aramark	Chiefs Ara 1 Arrowhead Dr	753	103969 Aramark	1 Royal Way
1 Aramark	Chiefs Ara 1 Arrowhead Drive			
4 Aramark	Chiefs Ara 1 Arrowhead Drive			
5 Aramark	Chiefs Ara 1 Arrowhead Drive			
6 Aramark	Chiefs Ara 1 Arrowhead Drive			
8 Aramark	Chiefs Ara 1 Arrowhead Drive			

Fig 2. Removal of inconsistent parts of data.

```
1 import pandas as pd
2 #from matplotlib import pyplot as plt
3 #from sklearn.cluster import KMeans
4
5 complaints = pd.read_csv('C:/Users/18166/Downloads/Food_Establishment_Complaints.csv')
6 food_permits = pd.read_csv('C:/Users/18166/Downloads/KCMO_Food_Permits.csv',encoding='latin1')
7
8
9 complaints.drop(['SOURCE', 'DEPARTMENT', 'WORK GROUP', 'CREATION MONTH', 'CREATION YEAR', 'STATUS', 'EXCEEDED EST TIMEFRAME', 'CLOSED DATE',
10                'CLOSED MONTH', 'CLOSED YEAR', 'DAYS TO CLOSE', 'ADDRESS WITH GEOCODE', 'NEIGHBORHOOD', 'COUNCIL DISTRICT', 'PARCEL ID NO',
11                'LATITUDE', 'LONGITUDE'], axis=1, inplace=True)
12 food_permits.drop(['Facility Name', 'Business Status', 'Facility Type', 'Permit Type', 'Web Site', 'Operational Status', 'Location 1'], axis=1, inplace=True)
13
14
15 food_permits = food_permits.drop_duplicates(subset=['Establishment Name', 'Facility Address'])
16
17 food_permits.to_csv('food_permits.csv')
18 complaints.to_csv('complaints.csv')
19
```

*Fig 5. Code to remove redundant data and irrelevant columns.*

## References

- Andrienko, N., Lammarsch, T., Andrienko, G., Fuchs, G., Keim, D., Miksch, S., & Rind, A. (2018). Viewing visual analytics as model building. *Computer Graphics Forum*, 37(6), 275-299. doi:<http://dx.doi.org/10.1111/cgf.13324>
- Gonzales, M. L. (2004). The architecture of enterprise data quality. *Intelligent Enterprise*, 7(9), 14-17. Retrieved from <https://search.proquest.com/trade-journals/architecture-enterprise-data-quality/docview/200663640/se-2?accountid=28370>