Yelp Data Analysis

Group 5: Yi-Yun Liao, Hongfei Chen, Hsin Yeh, Xiaotian He

1. Abstract

In this project, we analyzed the restaurant rating and review datasets on Yelp to provide insights into the success patterns of popular restaurants in the US. We utilized Hadoop for data cleaning and Hive for analyzing Yelp reviews of successful US restaurants. By integrating this with US demographic data, we can gain a comprehensive understanding of the factors that contribute to a restaurant's success.

2. Introduction

In the analysis, we used only the data from users with an average star rating higher than 3 as 'Good Users', as their reviews may be more reliable and informative, and businesses with 50 or more reviews, as this can help ensure a sufficient sample size for analysis and reduce the impact of outliers or biased reviews.

In the highly competitive restaurant industry, understanding the factors that contribute to a successful restaurant can provide a competitive advantage. By analyzing Yelp data, this project can provide insights into user behavior, preferences, and opinions, which can help restaurant owners and managers make data-driven decisions to improve their business. This research can be beneficial for a wide range of users, including restaurant owners and managers, investors, researchers, and analysts who are interested in understanding the factors that influence restaurant success. Restaurant owners and managers can benefit from this analysis by gaining insights into the factors that contribute to successful restaurants, such as user ratings and reviews, and can make informed decisions to improve their business. Investors can use this analysis to identify potential profitable restaurant investments. Researchers and analysts can use this data to study trends and patterns in the restaurant industry.

3. Pre-processing

3.1. Data Source

Our original datasets contain three business, review and user datasets from Yelp and one demographic information dataset from NYC OpenData.

a. Dataset: Yelp academic dataset business

• Description: Businesses listed on Yelp

• Size: 118.86 MB

b. Dataset: Yelp academic dataset user

Description: User information from Yelp

• Size: 3.36 GB

c. Dataset: Yelp_academic_dataset_review

• Description: Millions of reviews from Yelp

• Size: 5.34 GB

d. Dataset: Demographic Statistics By Zip Code

Description: Demographic information from NYC OpenData

• Size: 29KB

3.2. Data Cleaning

The Yelp dataset contains a large number of attributes, but not all of them may be relevant for the analysis. Therefore, it is essential to identify the relevant features and ensure that they are included in the analysis. The demographic dataset contains gender, race, nationality and public assistance information with both actual numbers and percentages, given that the total people participated in the survey are different in different jurisdictions, we decided to use only the percentages data to ensure the comparability between different regions.

In this analysis, we use Mapreduce to clean the data and select relevant attributes on NYU Dataproc Platform.

3.3. Data Sample

a. Dataset: Yelp academic dataset business

1	business id	name	address	city	state	postal code	latitude	longitude	stars	review count	is open	attributes
2	_		1616 Chapala St, Ste 2	Santa Barba		93101				7		('ByAppointmentOnly': 'True')
3	mpf3x-BjTdTEA3yCZrAYPw		87 Grasso Plaza Shopping Center	Affton	МО	63123	38.55113	-90.3357	3	15		{'BusinessAcceptsCreditCards': 'True'
4	tUFrWirKiKi TAnsVWINQQ		5255 E Broadway Blvd	Tucson	ΑZ	85711	32.22324	-110.88	3.5	22		{'BikeParking': 'True', 'BusinessAccer
5	MTSW4McQd7CbVtyjgoe9mw	St Honore Pastries	935 Race St	Philadelphia	PA	19107	39.95551	-75.1556	4	80	1	{'RestaurantsDelivery': 'False', 'Outde
6	mWMc6 wTdE0EUBKIGXDVfA	Perkiomen Valley Bre	101 Walnut St	Green Lane	PA	18054	40.33818	-75.4717	4.5	13	1	{'BusinessAcceptsCreditCards': 'True'
7	CF33F8-E6oudUQ46HnavjQ	Sonic Drive-In	615 S Main St	Ashland City	TN	37015	36.26959	-87.0589	2	6	1	{'BusinessParking': 'None', 'BusinessA
8	n 0UpQx1hsNbnPUSlodU8w	Famous Footwear	8522 Eager Road, Dierbergs Brent	Brentwood	мо	63144	38.6277	-90.3405	2.5	13	1	{'BusinessAcceptsCreditCards': 'True'
9	gkRM 2X51Ygxk3btlwAQlg	Temple Beth-El	400 Pasadena Ave S	St. Petersbu	FL	33707	27.76659	-82.733	3.5	5	1	
10	k0hlBqXX-Bt0vf1op7Jr1w	Tsevi's Pub And Grill	8025 Mackenzie Rd	Affton	мо	63123	38.56516	-90.3211	3	19		('Caters': 'True', 'Alcohol': "u'full_bai
11	bBDDEgkFA1Otx9Lfe7BZUQ	Sonic Drive-In	2312 Dickerson Pike	Nashville	TN	37207	36.2081	-86.7682	1.5	10	1	{'RestaurantsAttire': "'casual'", 'Resta
12	UJsufbvfyfONHeWdvAHKjA	Marshalls	21705 Village Lakes Sc Dr	Land O' Lake	FL	34639	28.19046	-82.4574	3.5	6	1	{'RestaurantsPriceRange2': '2', 'Bikel
13	eEOYSgkmpB90uNA7IDOMRA	Vietnamese Food Truc	k	Tampa Bay	FL	33602	27.95527	-82.4563	4	10	1	{'Alcohol': "'none'", 'OutdoorSeating'
14	il_Ro8jwPlHresjw9EGmBg	Denny's	8901 US 31 S	Indianapolis	IN	46227	39.63713	-86.1272	2.5	28	1	{'RestaurantsReservations': 'False', 'F
15	jaxMSolnw8Poo3XeMJt8IQ	Adams Dental	15 N Missouri Ave	Clearwater	FL	33755	27.96624	-82.7874	5	10	1	{'ByAppointmentOnly': 'True'}
16	0bPLkL0QhhPO5kt1_EXmNQ	Zio's Italian Market	2575 E Bay Dr	Largo	FL	33771	27.91612	-82.7605	4.5	100	(('OutdoorSeating': 'False', 'Restaurar
17	MUTTqe8uqyMdBl186RmNeA	Tuna Bar	205 Race St	Philadelphia	PA	19106	39.95395	-75.1432	4	245	1	{'RestaurantsReservations': 'True', 'R
18	rBmpy_Y1UbBx8ggHlyb7hA	Arizona Truck Outfitte	625 N Stone Ave	Tucson	ΑZ	85705	32.22987	-110.972	4.5	10	1	{'DriveThru': 'False', 'BusinessAccept
19	M0XSSHqrASOnhgbWDJlpQA	Herb Import Co	712 Adams St	New Orlean	LA	70118	29.94147	-90.13	4	5	1	{'BusinessParking': "{'garage': False,
20	8wGISYjYkE2tSqn3cDMu8A	Nifty Car Rental	1241 Airline Dr	Kenner	LA	70062	29.98118	-90.254	3.5	14	1	

b. Dataset: Yelp_academic_dataset_user

1	user_id	name	review_count	yelping_since	average_stars	friends_count	useful	funny	cool
2	qVc8ODYU5SZjKXVBgXdI7w	Walker	585	2007	3.91	14995	7217	1259	5994
3	j14WgRoU2ZE1aw1dXrJg	Daniel	4333	2009	3.74	4646	43091	13066	27281
4	2WnXYQFK0hXEoTxPtV2zvg	Steph	665	2008	3.32	381	2086	1010	1003
5	SZDeASXq7o05mMNLshsdIA	Gwen	224	2005	4.27	131	512	330	299
6	hA5IMy-EnncsH4JoR-hFGQ	Karen	79	2007	3.54	27	29	15	. 7
7	q_QQ5kBBwlCcbL1s4NVK3g	Jane	1221	2005	3.85	5843	14953	9940	11211
8	cxuxXkcihfCbqt5Byrup8Q	Rob	12	2009	2.75	23	6	1	. 0
9	E9kcWJdJUHuTKfQurPljwA	Mike	358	2008	3.73	82	399	102	143
10	IO1iq-f75hnPNZkTy3Zerg	Rachelle	40	2008	4.04	488	109	40	46
11	AUi8MPWJ0mLkMfwbui27lg	John	109	2010	3.4	64	154	20	23
12	iYzhPPqnrjJkg1JHZyMhzA	Chris	4	2010	4	241	1	0	1
13	xoZvMJPDW6Q9pDAXI0e_Ww	Ryan	535	2009	3.89	356	1130	487	573
14	vVukUtqoLF5BvH_VtQFNoA	Charlene	37	2011	4.51	154	63	3	27
15	_crlokUeTCHVK_JVOy-0qQ	Kenny	11	2009	3.08	64	30	3	0
16	1McG5Rn_UDkmlkZOrsdptg	Teresa	7	2009	4.29	14	18	3	13
17	SgiBkhXeqlKl1PlFpZOycQ	Eugene	682	2006	3.75	187	1819	1138	1297
18	fJZO_skqpnhk1kvomI4dmA	Jennifer	25	2008	4.15	13	29	2	19
19	x7YtLnBW2dUnrrpwaofVQQ	Ronskee	37	2010	3.84	84	56	29	29
20	QF1Kuhs8iwLWANNZxebTow	Catherine	607	2009	4.11	487	4573	3714	4149

c. Dataset: Yelp_academic_dataset_review

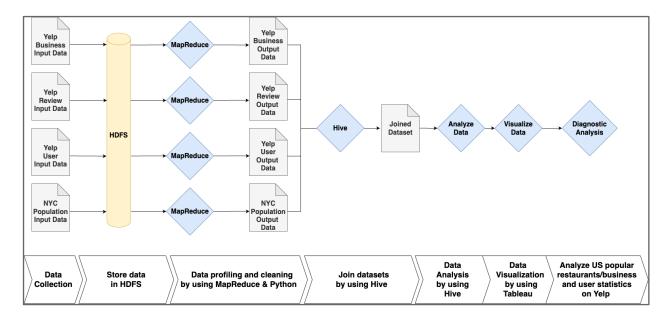
		• •	,	
1	review_id	user_id	business_id	stars
2	KU_O5udG6zpxOg-VcAEodg	mheMZ6K5RLWhZyISBhwA	XQfwVwDr-v0ZS3_CbbE5Xw	3
3	BiTunyQ73aT9WBnpR9DZGw	OyoGAe7OKpv6SyGZT5g77Q	7ATYjTlgM3jUlt4UM3lypQ	5
4	saUsX_uimxRICVr67Z4Jig	8g_iMtfSiwikVnbP2etR0A	YjUWPpI6HXG530lwP-fb2A	3
5	AqPFMleE6RsU23_auESxiA	_7bHUi9Uuf5HHc_Q8guQ	kxX2SOes4o-D3ZQBkiMRfA	5
6	Sx8TMOWLNuJBWer-0pcmoA	bcjbaE6dDog4jkNY91ncLQ	e4Vwtrqf-wpJfwesgvdgxQ	4
7	JrlxlS1TzJ-iCu79ul40cQ	eUta8W_HdHMXPzLBBZhL1A	04UD14gamNjLY0IDYVhHJg	1
8	6AxgBCNX_PNTOxmbRSwcKQ	r3zeYsv1XFBRA4dJpL78cw	gmjsEdUsKpj9Xxu6pdjH0g	5
9	_ZeMknuYdlQcUqng_Im3yg	yfFzsLmaWF2d4Sr0UNbBgg	LHSTtnW3YHCeUkRDGyJOyw	5
10	ZKvDG2sBvHVdF5oBNUOpAQ	wSTuiTk-sKNdcFyprzZAjg	B5XSoSG3SfvQGtKEGQ1tSQ	3
11	pUycOfUwM8vqX7KjRRhUEA	59MxRhNVhU9MYndMkz0wtw	gebiRewfieSdtt17PTW6Zg	3
12	rGQRf8UafX7OTIMNN19I8A	1WHRWwQmZOZDAhp2Qyny4g	uMvVYRgGNXf5boolA9HXTw	5
13	I3Wk_mvAog6XANIuGQ9C7Q	ZbqSHbgCjzVAqaa7NKWn5A	EQ-TZ2eeD_E0BHuvoaeG5Q	4
14	XW_LfMv0fV21l9c6xQd_lw	9OAtfnWag-ajVxRbUTGlyg	lj-E32x9_FA7GmUrBGBEWg	4
15	8JFGBuHMoiNDyfcxuWNtrA	smOvOajNG0lS4Pq7d8g4JQ	RZtGWDLCAtuipwaZ-UfjmQ	4
16	UBp0zWyH60Hmw6Fsasei7w	4Uh27DgGzsp6PqrH913giQ	otQS34_MymijPTdNBoBdCw	4
17	OAhBYw8IQ6wlfw1owXWRWw	1C2lxzUo1Hyye4RFIXly3g	BVndHaLihEYbr76Z0CMEGw	5
18	oyaMhzBSwfGgemSGuZCdwQ	Dd1jQj7S-BFGqRbApFzCFw	YtSqYv1Q_pOltsVPSx54SA	5
19	LnGZB0fjfgeVDVz5IHuEVA	j2wlzrntrbKwyOcOiB3l3w	rBdG_23USc7DletfZ11xGA	4
20	u2vzZaOqJ2feRshaaF1doQ	NDZvyYHTUWWu-kqgQzzDGQ	CLEWowfkj-wKYJIQDqT1aw	5

d. Dataset: Demographic Statistics By Zip Code

1	JURISDIC'	COUNT PA	PERCENT F	ERCENT													
2	10001	44	0.5	0.5	0	0.36	0	0.07	0.02	0.48	0.07	0	0.05	0.95	0	0.45	0.55
3	10002	35	0.54	0.46	0	0.03	0	0.8	0.17	0	0	0	0.06	0.94	0	0.06	0.94
4	10003	1	1	0	0	0	0	1	0	0	0	0	0	1	0	0	1
5	10004	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
6	10005	2	1	0	0	0	0	0.5	0	0.5	0	0	0.5	0.5	0	0	1
7	10006	6	0.33	0.67	0	0.33	0	0	0.17	0.5	0	0	0	1	0	0	1
8	10007	1	0	1	0	0	0	1	0	0	0	0	0	1	0	1	0
9	10009	2	0	1	0	0	0	1	0	0	0	0	0	1	0	0	1
10	10010	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
11	10011	3	0.67	0.33	0	0.33	0	0	0	0.33	0.33	0	0	1	0	0	1
12	10012	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
13	10013	8	0.13	0.88	0	0	0	1	0	0	0	0	0	1	0	0.13	0.88
14	10014	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
15	10016	17	0.71	0.29	0	0.53	0	0	0	0.47	0	0	0	1	0	0.53	0.47
16	10017	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
17	10018	3	0.67	0.33	0	0	0	1	0	0	0	0	0.33	0.67	0	0	1
18	10019	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
19	10020	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
20	10021	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
21	10022	1	1	0	0	0	0	1	0	0	0	0	0	1	0	0	1
22	10023	7	0.71	0.29	0	0.43	0	0.14	0	0.43	0	0	0	1	0	0.71	0.29
23	10024	4	1	0	0	0	0	0	0.75	0	0	0.25	0	1	0	0.25	0.75
24	10025	27	0.63	0.37	0	0.56	0	0	0	0.41	0.04	0	0.11	0.89	0	0.3	0.7
25	10026	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
26	10027	7	0.57	0.43	0	0.14	0	0.14	0	0.57	0.14	0	0.14	0.86	0	0.14	0.86
27	10028	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
28	10029	20	0.65	0.35	0	0.2	0	0	0	0.75	0.05	0	0	1	0	0.4	0.6

4. Data Analysis

4.1. Analysis Diagram



4.2. Code Challenge

a. Extract/Calculate/Store useful values in Yelp User.json

The yelp_since in the user dataset contains complete timestamp yyyy-mm-dd hh:mm:ss, which has redundant information we don't need for further data analysis in Hive or Tableau. Besides, the friends in the user dataset contains user ids for all friends which are

useful for data analysis or data visualization.

Our solution is to substring yelp_since and only extract year value. Create a new field friends_count to store the number of friends for each user instead of the user ids list for their friends.

```
// retrieve year
String join_year = "";
if (yelping_since != null) {
    join_year = yelping_since.substring(0, 4);
}

// count the number of friends and identify None value and store it as 0 friends_count
String[] friendsList = friends.split(",");
long friends_count = 0;
if (friendsList.length == 1 && friendsList[0].equals("None")) {
    friends_count = 0;
} else {
    friends_count = friendsList.length;
}
```

b. Data Separation in Yelp Business.json and Yelp User.json

When separating data using commas as separators, issues can arise. For example, when dealing with names that include a comma in between the first and last name, the comma may be mistakenly treated as a separator, leading to incorrect separation of the data. Our solution is that using spaces to replace commas for values in specific fields can clean up the data, help alleviate this issue and ensure accurate separation of the data when creating tables in Hive.

```
String user_id = (String) json.get("user_id");
String name = ((String) json.get("name")).replace(",", ""); // replace commas with space
String business_id = (String) json.get("business_id");
String name = ((String) json.get("name")).replace(",", ""); // replace commas with space
String city = ((String) json.get("city")).replace(",", ""); // replace commas with space
```

c. Handling Long and Messy String Data in Yelp Business.json

The Yelp Business.json contains many columns with very long and messy strings, such as the "categories" field. This can make the data difficult to work with and analyze. Our solution is to modify values in the "categories" field by replacing commas with the pipe symbol "|" to avoid values being separated incorrectly when creating tables in Hive. And also, it is easy to read subcategories when analyzing data in Hive and Tableau.

```
// replace commas with pipe in categories field
String categoriesNew = "";
if (categories != null) {
    categoriesNew = categories.replace(",", "|");
} else {
    categoriesNew = "";
}
```

d. Improve efficiency with Python to clean sample data on local first

Though we could use the Data Ingest Console to transfer large files to HDFS, and then run MapReduce codes on Hadoop to verify whether the MapReduce jobs work successfully, it is still not time-efficiency.

Our solution is that using Python for data cleaning with sample data on a local is a useful and efficient way to verify that the cleaning process produces correct and valid data for further usage in Hive and Tableau. Python libraries such as JSON and Pandas make tests on codes easier and more efficient. Once the data cleaning process has been validated, MapReduce codes can be written in Java and run on a larger dataset using a Hadoop cluster.

4.3. Data Analysis

After data cleaning and profiling, we joined the datasets and analyzed the joined table by using Hive on HDFS. Firstly, we joined the Business, Review and User datasets by using business_id and user_id as foreign keys, then performed analysis to provide insights into the success patterns of popular restaurants in the US and user behavior on Yelp.

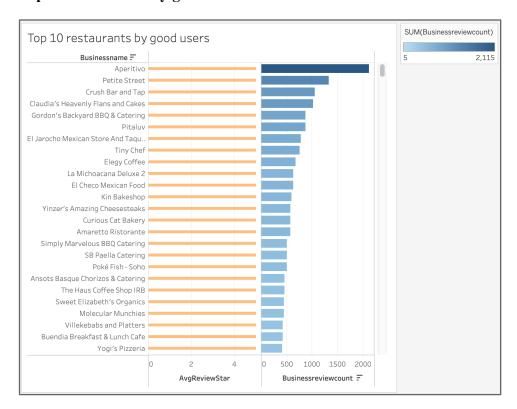
In the analysis of Businesses, we select users with an average star rating higher than 3 as 'good_users', and use their ratings to evaluate a business to ensure the quality of reviews. We first select the top 10 restaurants with highest average ranking and review count by good users, and then select top 10 Businesses with Highest Review Variability by 'Good' Users to show the businesses of different categories that have highly polarized reviews; We also selected top 10 businesses with the first category with highest review count and highest average rating to see the most popular and best rating businesses in different business types.

In the analysis of user behavior, we have computed user average review count, average user rating, and Number of Yelp users added per year from the joined table.

4.4. Data Visualization

After analyzing data on Hadoop, we visualized the result by using Tableau to transform the results into compelling and engaging visualizations.

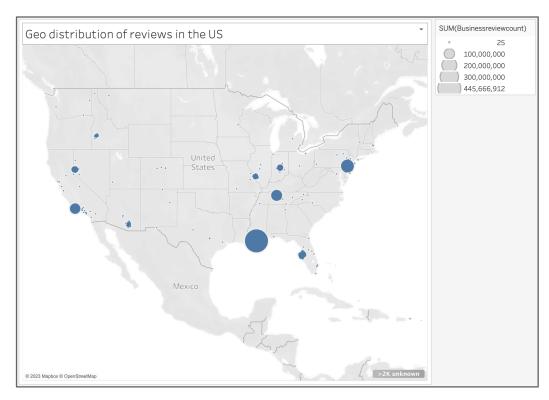
a. Top 10 restaurants by good users



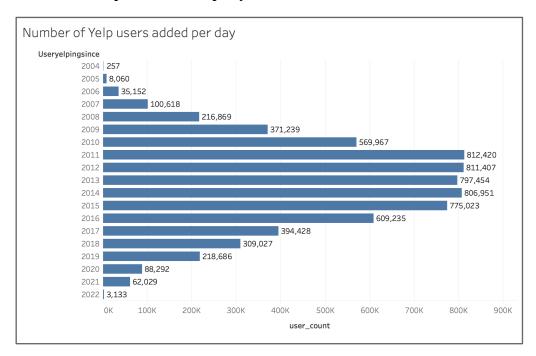
b. Top 10 restaurants by good users with business categories and state information

Businessname	=	Businesscategories	Businessstat
Aperitivo		Desserts Food Bars Wine Bars C	CA
Petite Street		Bartenders Sandwiches Event Pla	NV
Crush Bar and Tap		Nightlife Gay Bars Wine Bars Bar	CA
Claudia's Heavenly Flans and Ca	akes	Bakeries Sandwiches Restaurant	FL
Gordon's Backyard BBQ & Cater	ing	Caterers Barbeque Food Trucks E	NV
Pitaluv		Salad Restaurants Greek Medite	FL
El Jarocho Mexican Store And Ta	aq	Restaurants International Grocery	MO
Tiny Chef		Korean Pizza Restaurants	MO
Elegy Coffee		Coffee & Tea Food Bakeries Rest	TN
La Michoacana Deluxe 2		Ice Cream & Frozen Yogurt Dessert	TN
El Checo Mexican Food		Mexican Food Trucks Food Food	AZ
Gyro Express		Mediterranean Restaurants Pakis	PA
Kin Bakeshop		Pop-Up Restaurants Restaurants	CA
Yinzer's Amazing Cheesesteaks		Restaurants Cheesesteaks Nightl	LA
Curious Cat Bakery		Bakeries Food Patisserie/Cake Sh	FL
Amaretto Ristorante		Italian Restaurants	FL
Simply Marvelous BBQ Catering)	Restaurants Event Planning & Ser	CA
SB Paella Catering		Caterers Restaurants Spanish Pe	CA
Poké Fish - Soho		Food Poke Hawaiian Restaurants	FL
Ansots Basque Chorizos & Cater	ring	Basque Caterers Event Planning	ID
The Haus Coffee Shop IRB		Wine Bars Bars Restaurants Foo	FL
Sweet Elizabeth's Organics		Bakeries Food Coffee & Tea Waff	FL
Molecular Munchies		Burgers Food Trucks Restaurants	AZ
Villekebabs and Platters		Restaurants Mediterranean Afgh	PA

c. Geographical distribution of reviews in the U.S.



d. Number of Yelp users added per year



5. Results

a. Top 10 restaurants by good users

Selected top 10 restaurants with highest average ranking and review count by good users whose average rating stars >3.

businessname	averagerating	totalgoodreviews
Aperitivo	-+ 5.0	2209
Petite Street	5.0	1369
Crush Bar and Tap	5.0	1156
Claudia's Heavenly Flans and Cakes	5.0	1088
Tiny Chef	5.0	812
El Checo Mexican Food	5.0	650
Kin Bakeshop	5.0	625
La Michoacana Deluxe 2	5.0	625
Yinzer's Amazing Cheesesteaks	5.0	600
Curious Cat Bakery	5.0	575

The highest ranking restaurant is Aperitivo, with 5.0 average rating star and 2209 Reviews.

b. Top 10 restaurants by good users where yelping since = 2022 or 2021 or 2020

Selected top 10 restaurants with highest average ranking and review count by good users whose average rating stars >3 and joined yelp in recent years.

+ businessname	firsttwocategories	++ businessstate +
3rd And Lindsley	Restaurants American (Traditional)	IN I
Acme Oyster House	Seafood Sandwiches	LA
1000 Degrees Neapolitan Pizzeria	Pizza Restaurants	FL
104St Grill	Restaurants Steakhouses	AB
1750 Bistro	Event Planning & Services Hotels & Travel	PA
13th Street Pub & Grill	Restaurants Bars	ID
3 Southern Girls	Restaurants Soul Food	LA
2 Pickles	Restaurants Sandwiches	TN
2-D Wok	Food Taiwanese	NV
51 Fifty First Kitchen & Bar	American (New) Food	TN
+	+	++

c. Top 10 restaurants reviewed by good users and opinion leaders (who has friends more than avg friends count)

Selected top 10 restaurants with highest average ranking and review count by good users whose average rating stars >3 and who are also opinion leaders in terms of having more friends than average friends count.

```
businessname
                                                        firsttwocategories
#1 Mongolian BBQ - Best Stir Fried Noodles In Boise | Chinese| Restaurants
10th Street Diner
                                                    | Vegan| Vegetarian
12 South Bistro
                                                    | Restaurants| Pizza
1911 Grill
                                                    | Restaurants| Barbeque
3 Brothers Pub & Grub
&pizza - Willow Grove
                                                    | Pizza| Vegetarian
312 Pizza Company
                                                     Restaurants| Caterers
                                                     Nightlife| Bars
317 Burger
12th Street Diner
                                                     Diners| Restaurants
                                                      Restaurants | Food
'feine
```

d. Top 10 Businesses with Highest Review Variability by 'Good' Users

Selected top 10 Businesses with Highest Review Variability by good users whose average rating stars >3.

 businessname	-+
AVIANNA	Thai
Adele's	Cocktail Bars
Alma de Cuba	Latin American
Amuse	Food
Apricot Stone	Food
Batter & Dough	Waffles
Bell's Bike Shop	Active Life
Bistro at Cherry Hill	American (New)
Cafe Passe	Food
1st RND	Nightlife
+	-+

The highest Review Variability business is AVIANNA, and the business category is Thai.

e. Top 10 review count business with first categories

Selected top 10 businesses with the first category with highest review count to see the most popular businesses in different business types.



The most popular types of business are Local Services, Food Trucks and Laundry Services.

f. Top 10 rating business with first category

Selected top 10 businesses with the first category with highest average rating to see the best rating businesses in different business types.

 businessname	-+
\$155 Flat Rate Hauling Trash Removal 'Merica Food Truck \$225 Cleaners \$99 Pool Pumps & Pool Motors '81 Barbers \$5 Fresh Burger Stop &pizza - UPenn Xtreme Laser Tag Avon Grow Academy &pizza - Walnut	Local Services Food Trucks Laundry Services Pool Cleaners Barbers Restaurants Vegetarian Arts & Entertainment Preschools Pizza

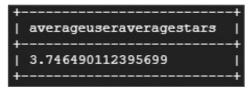
The most popular businesses with the highest reviews count and the best businesses with the highest average rating are highly identical. The highest rating types of business are also Local Services, Food Trucks and Laundry Services.

g. User average review count



The average user review count is 123.83.

h. Average user rating



The average user rating stars is 3.75.

i. Number of Yelp users added per day

+	-+
joinyear	numberofusers
+	
NULL	2
2004	257
2005	8060
2006	35152
2007	100618
2008	216869
2009	371239
2010	569967
2011	812420
2012	811407
2013	797454
2014	806951
2015	775023
2016	609235
2017	394428
2018	309027
2019	218686
2020	88292
2021	62029
2022	3133
+	-++

Yelp's highest number of new users was from 2011 to 2015, after which it showed a year-on-year decline

6. Conclusion - Insights

In the analysis of popular restaurants on Yelp, we can see that the most popular restaurants with the highest rating by good users are mostly in state CA, NV, MO, TN, PA and AZ, and different states show their preference on different kinds of cuisine, for example, we observe that in CA, popular restaurants are bars or catering while in PA, popular restaurants are Mediterranean food, and in FL, popular restaurants are foods like salad, bakery and poke. These preferences could have comprehensive analysis with population, lifestyle and weather dataset with corresponding states.

In the analysis of Businesses, we can see that the most popular businesses with the highest reviews count and the best businesses with highest average rating are highly identical. Which means that popularity and service quality are highly positively correlated, so a company can attract more customers by improving the quality of its food and service.

By analyzing the number of reviews and average ratings for all types of business, we found that the top three most popular types of business are Local Services, Food Trucks and Laundry Services, So if there are users who want to start a business, they can consider choosing these areas and will have a greater chance of success.

We also analyzed the variance of business rating. This analysis allowed us to know which companies' scores are highly polarized, and these companies can look deeper into the causes of

this phenomenon and make improvements.

In the analysis of user behavior, we found that Yelp's highest number of new users was from 2011 to 2015, after which it showed a year-on-year decline, suggesting that Yelp could address this decline in new users by making improvements, such as developing new features or new incentives to attract more users.

7. Acknowledgments

We would like to express our sincere gratitude to HPC for their invaluable support and assistance throughout this project. Their expertise and resources were instrumental in enabling us to effectively analyze and process the large Yelp dataset. We would also like to extend our thanks to Kaggle for providing us with the Yelp dataset that served as the foundation for our analysis. Without the support of HPC and the resources provided by Kaggle, this project would not have been possible. We are truly grateful for their contributions and assistance.

8. References

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