

Restaurant Recommendations

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Notes:

- This project is an exploratory work for my own learning and is not suitable for use in a production environment.
- The dataset is too large to include in an upload of this project. You can find and download the Yelp Dataset at <https://www.yelp.com/dataset>. Place all dataset json files into the folder 'yelp_dataset'

In [2]:

```
import pandas as pd
import json
import numpy as np
from sklearn.linear_model import LinearRegression
```

In [3]:

```
df_b = pd.read_json("yelp_dataset/yelp_academic_dataset_business.json", lines
```

In [187...]

```
display(df_b.head(3))
```

		business_id	name	address	city	state	postal_code	latitude
0	f9NumwFMBDn751xgFiRbNA		The Range At Lake Norman	10913 Bailey Rd	Cornelius	NC	28031	35.46272
1	Yzvvg0SayhoZgCljUJRF9Q		Carlos Santo, NMD	8880 E Via Linda, Ste 107	Scottsdale	AZ	85258	33.56940
2	XNoUzKckATkOD1hP6vghZg		Felinus	3554 Rue Notre-Dame O	Montreal	QC	H4C 1P4	45.47998

In [5]:

```
df_u = pd.read_json("yelp_dataset/yelp_academic_dataset_user.json", lines=True
```

In [6]:

```
TRAINING_REVIEW_COUNT_THRESHOLD_ = 250
df_u_abbrev = df_u[df_u["review_count"]>TRAINING_REVIEW_COUNT_THRESHOLD_]
```

In [188...]

```
display(df_u_abbrev.head(5))
```

		user_id	name	review_count	yelping_since	useful	funny	cool
0	ntlvfPzc8eglqvk92iDIAw		Rafael	553	2007-07-06 03:27:11	628	225	22

		user_id	name	review_count	yelping_since	useful	funny	cool
1	FOBRPIBHa3WPHFB5qYDIVg		Michelle	564	2008-04-28 01:29:25	790	316	40
4	xvu8G900tezTzbbfqmTKvA		Anne	485	2008-08-09 00:30:27	1265	400	51
7		f4_MRNVHvN- yRn7EA8YWRxg	Jennifer	822	2011-01-17 00:18:23	4127	2446	287
11	l_6wY8_RsewziNnKhGZg4g		Jeff	405	2010-08-05 18:42:29	799	244	31

5 rows × 22 columns

◀		▶
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```
In [9]: df_r = pd.read_json("yelp_dataset/yelp_academic_dataset_review.json", lines=True)
```

```
In [10]: df_r = df_r.drop(labels=["useful", "funny", "cool", "text", "date"], axis=1)
```

```
In [189... display(df_r.head(5))
```

	review_id	user_id	business_id	stars
0	xQY8N_XvtGbearJ5X4QryQ	OwjRMXRC0KyPrllcjaXeFQ	-MhfebM0QIsKt87iDN-FNw	
1	UmFMZ8PyXZTY2QcwzsfQYA	nIJD_7ZXHq-FX8byPMOkMQ	lbrU8StCq3yDfr-QMnGrmQ	
2	LG2ZaYiOgpr2DK_90pYjNw	V34qejxNsCbcgD8C0HVk-Q	HQI28KMwrEKHqhFrrDqVnQ	
3	i6g_oA9Yf9Y31qt0wibXpw	ofKDKjKXSKZXu5xJNGiIBQ	5JxlZaqCnk1MnbgRirs40Q	
4	6TdNDKywdbjoTkizeMce8A	UgMW8bLE0QMJDCKQ1Ax5Mg	IS4cv902ykd8wj1TR0N3-A	

◀		▶
---	--	---

Create business feature vectors, place into a dictionary

```
In [14]: #df_b[df_b["attributes"].map(lambda x: "Alcohol" in x)]
#set(df_b["attributes"].map(lambda x: x["RestaurantsAttire"] if isinstance(x,
```

```
In [15]: featureAttributesBool = [
    "RestaurantsPriceRange2",
    "RestaurantsReservations",
    "BikeParking",
    "BusinessAcceptsCreditCards",
    "ByAppointmentOnly",
    "DriveThru",
    "GoodForKids",
    'HappyHour',
    "HasTV",
    "OutdoorSeating",
    "RestaurantsDelivery",
    "RestaurantsGoodForGroups",
    "RestaurantsTakeOut",
    "WheelchairAccessible"
]

featureAttributesOneHot = [
    ['Alcohol', "None", "'none'", "u'none'"],
    ['Alcohol', "'beer_and_wine'", "u'beer_and_wine'"],
```

```

['Alcohol', "full_bar", "u'full_bar'"],
['RestaurantsAttire', "'None'"],
['RestaurantsAttire', "'casual'", "u'casual'"],
['RestaurantsAttire', "'dressy'", "u'dressy'"],
['RestaurantsAttire', "'formal'", "u'formal'"],
]

def extractAttributes(businessAttributes):
    features = list()
    for attr in featureAttributesBool:
        if isinstance(attr, str):
            if businessAttributes is not None and attr in businessAttributes:
                features.append(int(businessAttributes[attr] == 'True'))
            else:
                features.append(0)

    for attr in featureAttributesOneHot:
        if businessAttributes is not None and attr[0] in businessAttributes:
            found = False
            for attrTag in attr[1:]:
                if businessAttributes[attr[0]] == attrTag:
                    found = True
                    features.append(1)
                    break
            if not found:
                features.append(0)
        else:
            features.append(0)
    return features

```

```

In [16]: businessFeatureDict = dict()

def addBusiness(business):
    features = [business["stars"]] + [business["review_count"]] + extractAttr
    businessFeatureDict[business["business_id"]] = features

```

```

In [17]: df_b.apply(addBusiness, axis=1)

```

```

Out[17]: 0      None
1      None
2      None
3      None
4      None
...
209388  None
209389  None
209390  None
209391  None
209392  None
Length: 209393, dtype: object

```

```

In [19]: #addBusiness(df_b.iloc[209388])
         #print(businessFeatureDict["SYa2jlboLF8DcGV0YfHPcA"])

```

Testing on a random business

```

In [20]: a = extractAttributes(df_b[df_b["business_id"]=="SYa2jlboLF8DcGV0YfHPcA"]["at

```

```

In [21]: a

```

```
Out[21]: [0, 0, 1, 1, 0, 0, 1, 0, 0, 1, 0, 1, 1, 0, 1, 0, 0, 0, 1, 0, 0]
```

Weight Generation

```
In [22]: def generateWeights(user_id):
          user_reviews = df_r[df_r["user_id"]==user_id]
          user_review_scores = user_reviews["stars"].values
          #business_features = user_reviews.apply(lambda x: businessFeatureDict[x], u
          business_features = np.array(list(map(lambda x: businessFeatureDict[x], u
          #display(business_features)
          model = LinearRegression()
          model.fit(business_features, user_review_scores)
          return model.coef_
```

```
In [24]: weights = df_u_abbrev.apply(lambda x: generateWeights(x["user_id"]), axis=1)
```

```
In [192]... display(weights.head(5))
           print("Number of points: " + str(len(df_u_abbrev["user_id"].values)))
```

```
0      [1.0741416610495274, 0.0001243782230512169, 2....
1      [0.36931201740440167, 0.00036807714185722526, ...
4      [0.11302586666171928, 0.0002609166948490038, 0...
7      [0.5066197242566907, 0.0002036361778916583, -6...
11     [-1.9369369369368967, -0.00450450450450395, -9...
dtype: object
Number of points: 25763
```

Format weights to be acceptable input for Milvus

```
In [50]: weightsnp = np.stack(weights.values)
```

```
In [52]: weightsnp.shape
```

```
Out[52]: (25763, 23)
```

Locally hosted Milvus instance, partially taken from

<https://raw.githubusercontent.com/milvus-io/pymilvus/0.2.15/examples/example.py>

```
In [101]... from milvus import Milvus, IndexType, MetricType, Status

          # Milvus server IP address and port.
          # You may need to change _HOST and _PORT accordingly.
          _HOST = '127.0.0.1'
          _PORT = '19530' # default value
          # _PORT = '19121' # default http value

          # Vector parameters
          _DIM = 23 # dimension of vector

          _INDEX_FILE_SIZE = 32 # max file size of stored index
```

```
In [102]... milvus = Milvus(_HOST, _PORT)

          # Create collection demo_collection if it doesn't exist.
          collection_name = 'recomendation_collection_'
```

```

status, ok = milvus.has_collection(collection_name)
if not ok:
    param = {
        'collection_name': collection_name,
        'dimension': _DIM,
        'index_file_size': _INDEX_FILE_SIZE, # optional
        'metric_type': MetricType.L2 # optional
    }

    milvus.create_collection(param)

# Show collections in Milvus server
_, collections = milvus.list_collections()

# Describe demo_collection
_, collection = milvus.get_collection_info(collection_name)
print(collection)

```

CollectionSchema(collection_name='recomendation_collection_', dimension=23, index_file_size=32, metric_type=<MetricType: L2>)

```
In [104... #milvus.drop_collection(collection_name)
```

Retype ids into native python int types. Milvus does not seem to accept ids from numpy types, such as numpy.int64 or numpy.int32.

```
In [103... weightindex = weights.index.values
weightids = [weightindex.item(i) for i in range(len(weightindex))]
```

```
In [105... # Insert vectors into demo_collection, return status and vectors id list
status, ids = milvus.insert(collection_name=collection_name, records=np.stack
if not status.OK():
    print("Insert failed: {}".format(status))

# Flush collection inserted data to disk.
milvus.flush([collection_name])
```

```
Out[105... Status(code=0, message='OK')
```

```
In [193... # Get collection row count
status, result = milvus.count_entities(collection_name)
print("Row count: " + str(result))
```

Row count: 25763

Build Index

We use the recommended value of nlist: $4 \times \sqrt{n}$, where n is the number of items in a segment

```
In [107... _, info = milvus.get_collection_stats(collection_name)
print(info)
```

```
{'partitions': [{'row_count': 25763, 'segments': [{'data_size': 2576300, 'index_name': 'IDMAP', 'name': '1613784371630425000', 'row_count': 25763}], 'tag': '_default'}], 'row_count': 25763}
```

```
In [109... 25763 ** (1/2) * 4
```

Out[109... 642.0342669982655

```
In [112... ivf_param = {'nlist': 642}
milvus.create_index(collection_name, IndexType.IVF_SQ8, ivf_param)
```

Out[112... Status(code=0, message='Build index successfully!')

nprobe is chosen semi-arbitrarily

```
In [194... search_param = {'nprobe': 16}
```

```
In [195... NUM_USERS_REC_ = 3
NUM_REC_PER_USER_ = 3

def retrieveRecs(uid, num_recs):
    user_id = df_u_abbrev.iloc[uid]["user_id"]
    toprecs = df_r[df_r["user_id"] == user_id].sort_values(by="stars", ascending=False)
    return df_b[df_b["business_id"].isin(toprecs.head(num_recs)["business_id"]]

def generateRecommendations(user_id):
    user_weights = np.stack(generateWeights(user_id)).reshape(1, 23)
    status, results = milvus.search(collection_name=collection_name, query_results=user_weights)
    print(status)
    rec_df = pd.DataFrame()
    for searchResult in results[0]:
        rec_df = rec_df.append(retrieveRecs(searchResult.id, NUM_REC_PER_USER_))
    return rec_df
```

Running

To generate a set of recommendations for a user, run

`generateRecommendations(user_id)`, where `user_id` is their id. This assumes that the user is present in the dataset. It will generate and output `NUM_USERS_REC_ * NUM_REC_PER_USER_` recommended businesses.

Example with a randomly selected user:

```
In [196... generateRecommendations("fL0jIsxSR2DSBeIRI80cTA")
```

Status(code=0, message='Search vectors successfully!')

```
Out[196...
```

		business_id	name	address	city	state	postal_code
148713	-Miw03v5yXJWjH9MN1aglw	The Nash	925 11th Street SE	Calgary	AB	T2G 0S	
172724	bcCfoAUpHY5SVrYd4alasA	Village Ice Cream	2406 34th Avenue SW	Calgary	AB	T2T 2C	
205504	TSzsZRpN09mqu54HWNn1PA	Mari Bakeshop	529 Riverfront Avenue SE	Calgary	AB	T2G 1K	

	business_id	name	address	city	state	postal_code
5257	UI6JwluSTm12PVDIqnNaTg	Kaya	2000 Smallman St	Pittsburgh	PA	1522
46641	TrCiLMGy_bxbeAQcuSSUeQ	Larry & Carols Pizza	410 Semple St	Pittsburgh	PA	1521
175195	dIsUtYng6IzaaLOqHlkOMA	Jerry's Records	2136 Murray Ave	Pittsburgh	PA	1521
8892	MN6HfA76VrdU4RjiGLwSug	A Elvis Chapel	727 S 9th St	Las Vegas	NV	8910
82036	Bkkwt8E9MHvgCHn4IUftow	Roy's Restaurant	5350 E Marriott Dr	Phoenix	AZ	8505
198269	dUffgo9Lh_Vk9TLuFR5ywg	Oregano's Pizza Bistro	1008 E Camelback Rd	Phoenix	AZ	8501