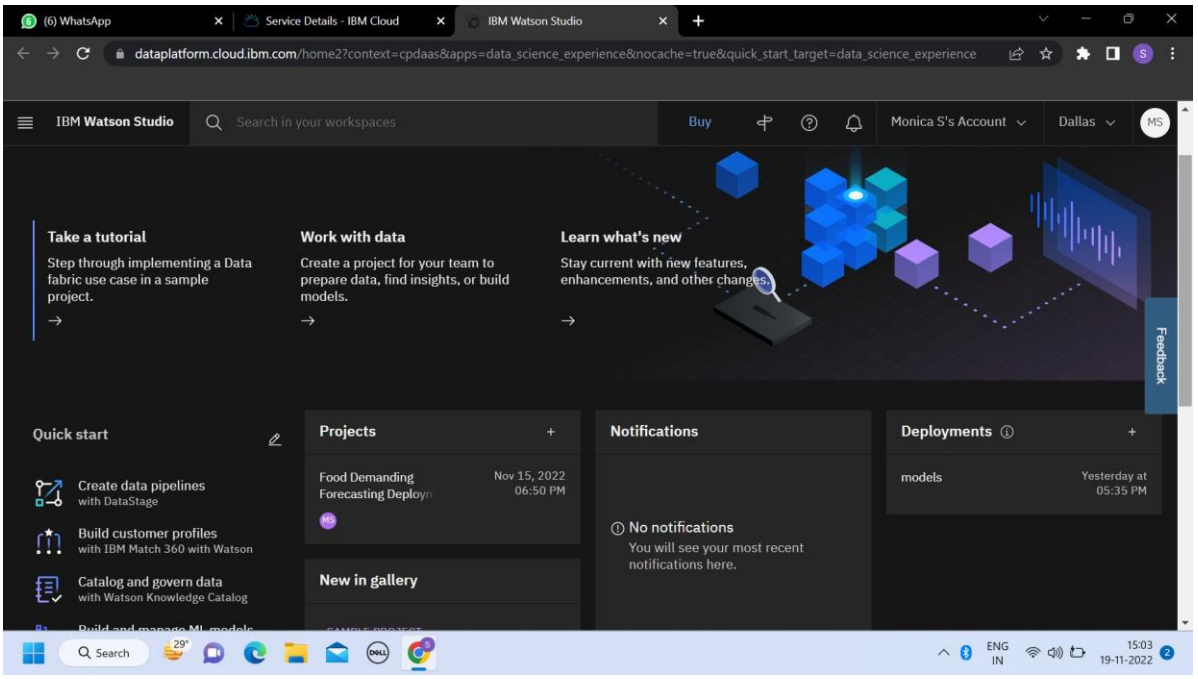
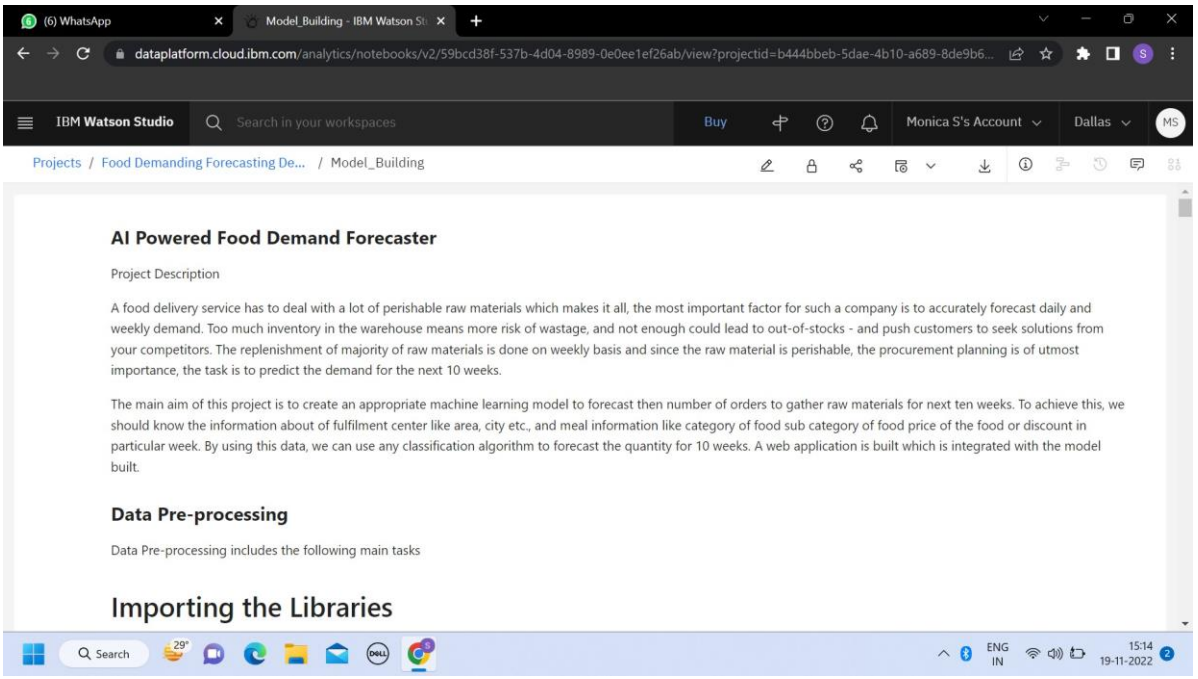
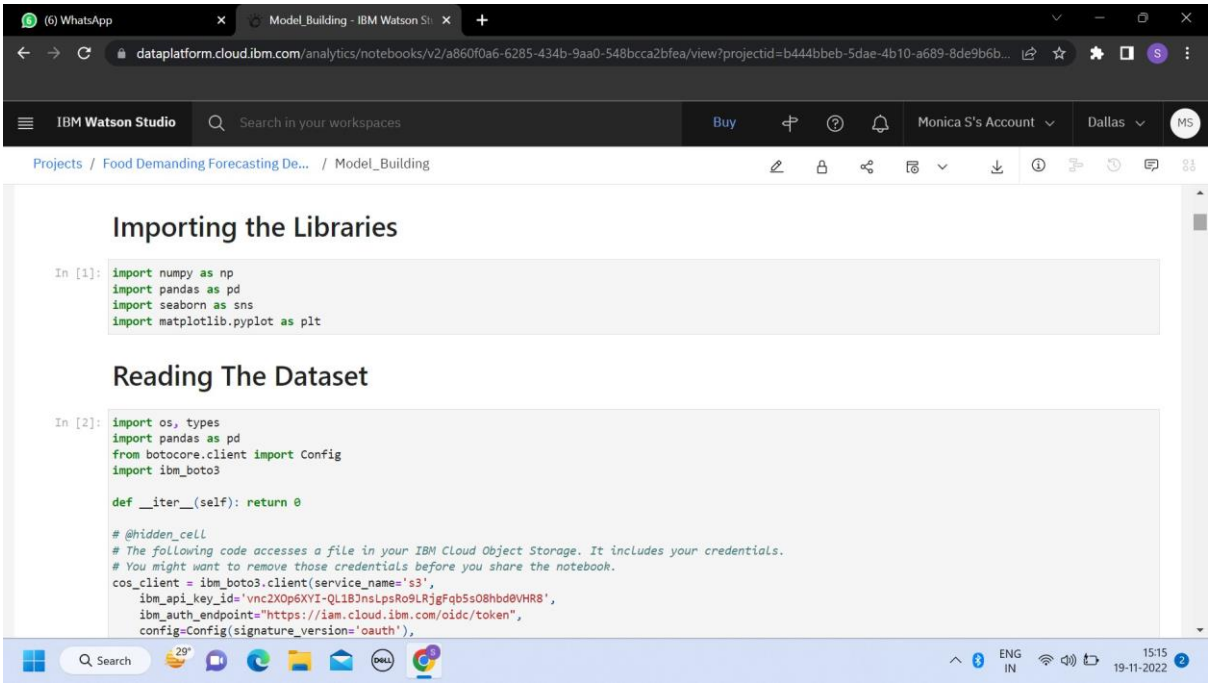
**TEAM ID: PNT2022TMID21777**

**PROJECT NAME: DemandEst - AI powered Food Demand Forecaster**

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Importing the Libraries
In [1]:
import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
Reading The Dataset
In [2]:
import os, types
import pandas as pd
from botocore.client import Config
import ibm_boto3
def __iter__(self): return 0
# @hidden_cell
# The following code accesses a file in your IBM Cloud Object Storage. It includes your credentials.
# You might want to remove those credentials before you share the notebook.
cos_client = ibm_boto3.client(service_name='s3',
ibm_api_key_id='vnc2XOp6XYI-QL1BJnsLpsRo9LRjgFqb5sO8hbd0VHR8',
ibm_auth_endpoint="https://iam.cloud.ibm.com/oidc/token",
config=Config(signature_version='oauth'),
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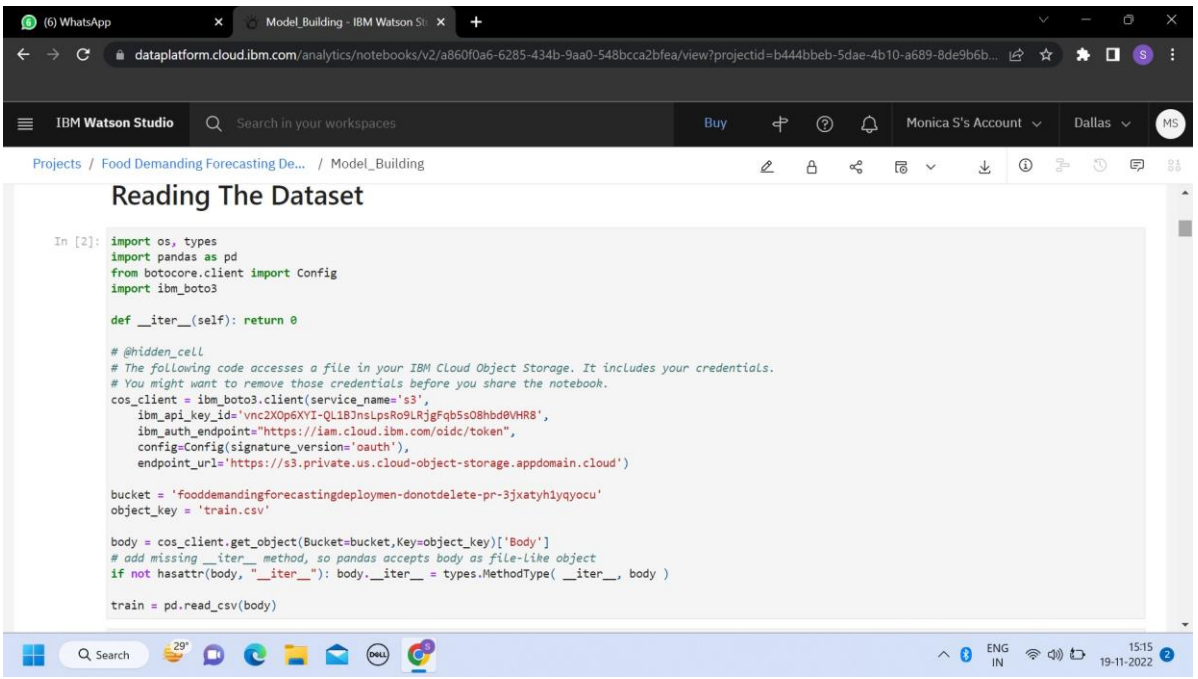
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Reading The Dataset
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import pandas as pd
from botocore.client import Config
import ibm_boto3
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ibm_api_key_id='vnc2XOp6XYI-QL1BJnsLpsRo9LRjgFqb5sO8hbd0VHR8',
ibm_auth_endpoint="https://iam.cloud.ibm.com/oidc/token",
config=Config(signature_version='oauth'),
endpoint_url='https://s3.private.us.cloud-object-storage.appdomain.cloud')
bucket = 'fooddemandingforecastingdeploymen-donotdelete-pr-3jxatyh1yqyocu'
object_key = 'train.csv'
body = cos_client.get_object(Bucket=bucket,Key=object_key)['Body']
# add missing __iter__ method, so pandas accepts body as file-like object
if not hasattr(body, "__iter__"): body.__iter__ = types.MethodType( __iter__, body )
train = pd.read_csv(body)
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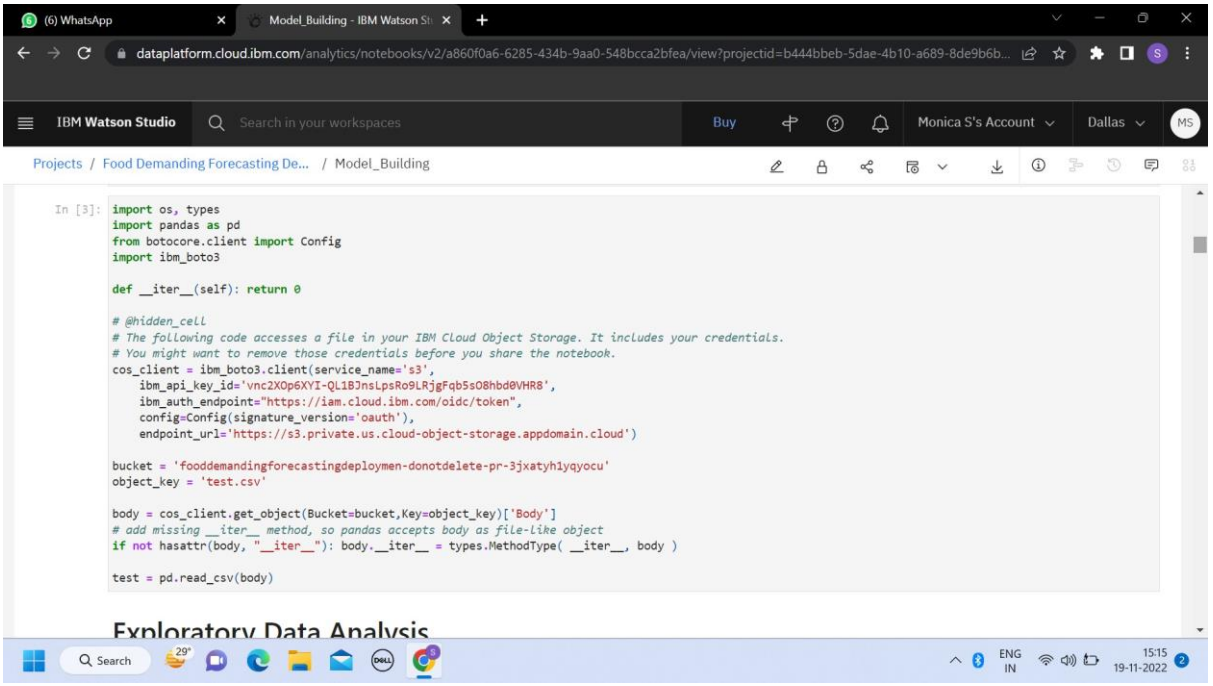

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In [3]:

```
import os, types
import pandas as pd
from botocore.client import Config
import ibm_boto3

def __iter__(self): return 0

# @hidden_cell
# The following code accesses a file in your IBM Cloud Object Storage. It includes your credentials.
# You might want to remove those credentials before you share the notebook.
cos_client = ibm_boto3.client(service_name='s3',
    ibm_api_key_id='vnc2XOp6XYI-QL1BJnsLpsRo9LRjgFqb5sO8hbd0VHR8',
    ibm_auth_endpoint="https://iam.cloud.ibm.com/oidc/token",
    config=Config(signature_version='oauth'),
    endpoint_url='https://s3.private.us.cloud-object-storage.appdomain.cloud')

bucket = 'fooddemandingforecastingdeploymen-donotdelete-pr-3jxatyh1yqyocu'
object_key = 'test.csv'

body = cos_client.get_object(Bucket=bucket,Key=object_key)['Body']
# add missing __iter__ method, so pandas accepts body as file-like object
if not hasattr(body, "__iter__"): body.__iter__ = types.MethodType( __iter__, body )

test = pd.read_csv(body)
```

Exploratory Data Analysis

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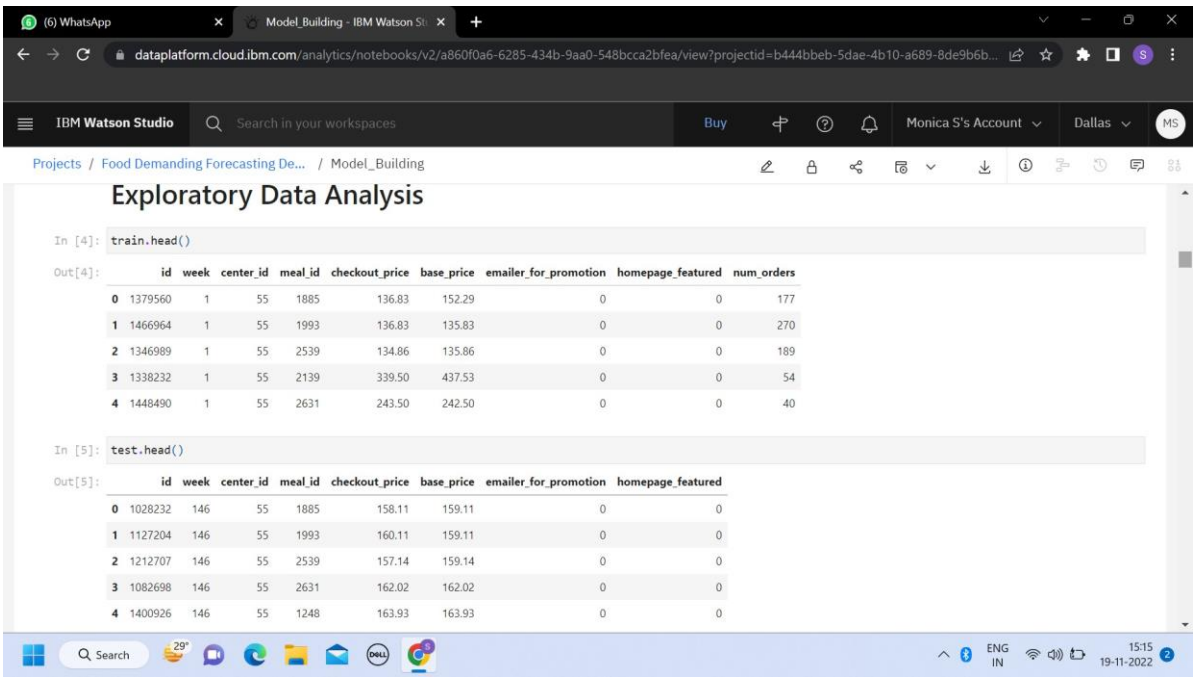

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# Exploratory Data Analysis

In [4]:

```
train.head()
```

Out[4]:

| | id | week | center_id | meal_id | checkout_price | base_price | emailer_for_promotion | homepage_featured | num_orders |
|---|---|---|---|---|---|---|---|---|---|
| 0 | 1379560 | 1 | 55 | 1885 | 136.83 | 152.29 | 0 | 0 | 177 |
| 1 | 1466964 | 1 | 55 | 1993 | 136.83 | 135.83 | 0 | 0 | 270 |
| 2 | 1346989 | 1 | 55 | 2539 | 134.86 | 135.86 | 0 | 0 | 189 |
| 3 | 1338232 | 1 | 55 | 2139 | 339.50 | 437.53 | 0 | 0 | 54 |
| 4 | 1448490 | 1 | 55 | 2631 | 243.50 | 242.50 | 0 | 0 | 40 |

In [5]:

```
test.head()
```

Out[5]:

| | id | week | center_id | meal_id | checkout_price | base_price | emailer_for_promotion | homepage_featured |
|---|---|---|---|---|---|---|---|---|
| 0 | 1028232 | 146 | 55 | 1885 | 158.11 | 159.11 | 0 | 0 |
| 1 | 1127204 | 146 | 55 | 1993 | 160.11 | 159.11 | 0 | 0 |
| 2 | 1212707 | 146 | 55 | 2539 | 157.14 | 159.14 | 0 | 0 |
| 3 | 1082698 | 146 | 55 | 2631 | 162.02 | 162.02 | 0 | 0 |
| 4 | 1400926 | 146 | 55 | 1248 | 163.93 | 163.93 | 0 | 0 |

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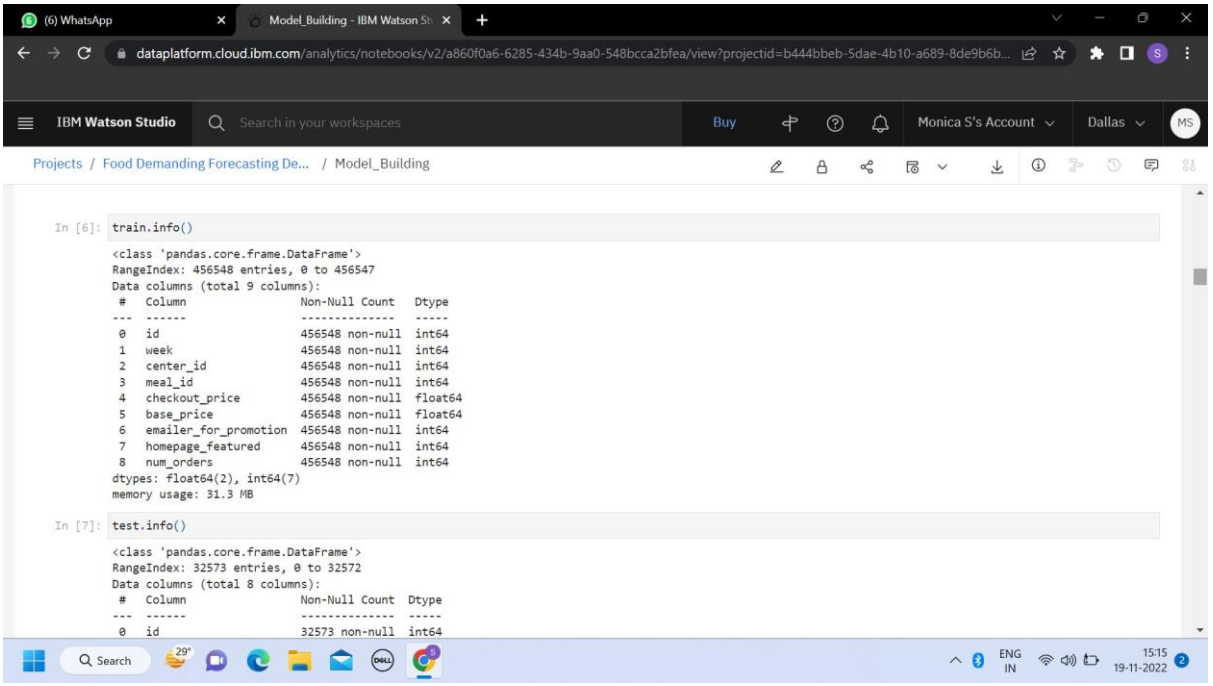
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In [6]: train.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 456548 entries, 0 to 456547
Data columns (total 9 columns):
# Column Non-Null Count Dtype
--- ------ -------------- -----
0 id 456548 non-null int64
1 week 456548 non-null int64
2 center_id 456548 non-null int64
3 meal_id 456548 non-null int64
4 checkout_price 456548 non-null float64
5 base_price 456548 non-null float64
6 emailer_for_promotion 456548 non-null int64
7 homepage_featured 456548 non-null int64
8 num_orders 456548 non-null int64
dtypes: float64(2), int64(7)
memory usage: 31.3 MB
In [7]: test.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 32573 entries, 0 to 32572
Data columns (total 8 columns):
# Column Non-Null Count Dtype
--- ------ -------------- -----
0 id 32573 non-null int64
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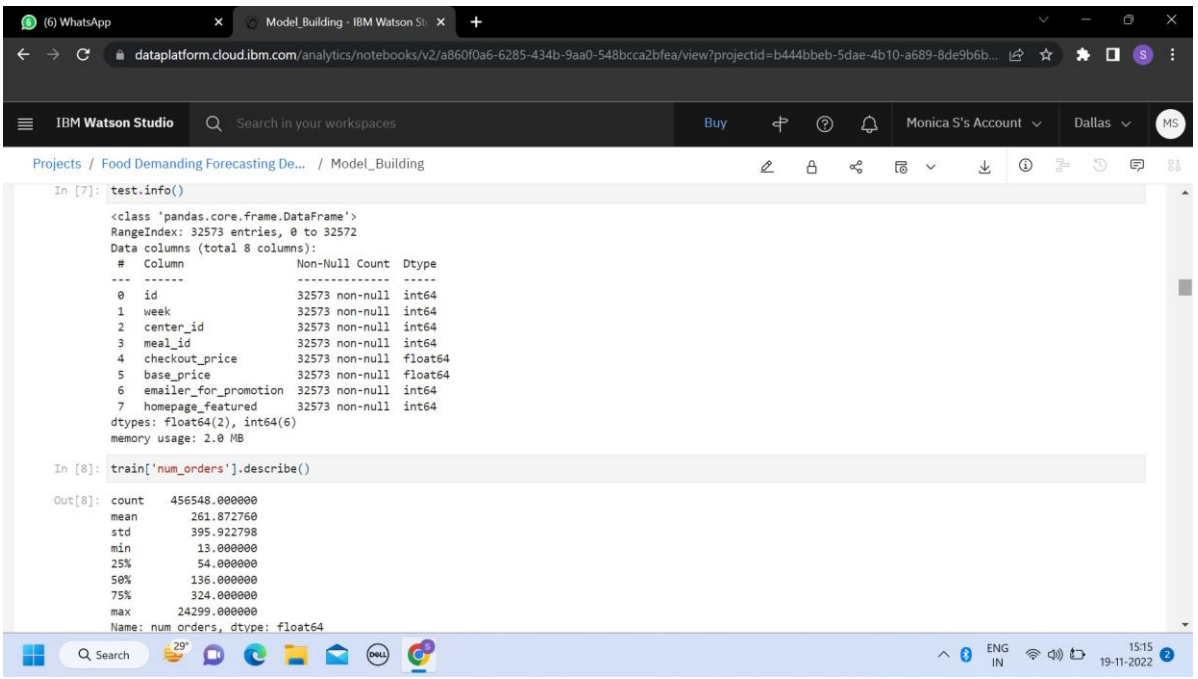
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In [7]: test.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 32573 entries, 0 to 32572
Data columns (total 8 columns):
# Column Non-Null Count Dtype
--- ------ -------------- -----
0 id 32573 non-null int64
1 week 32573 non-null int64
2 center_id 32573 non-null int64
3 meal_id 32573 non-null int64
4 checkout_price 32573 non-null float64
5 base_price 32573 non-null float64
6 emailer_for_promotion 32573 non-null int64
7 homepage_featured 32573 non-null int64
dtypes: float64(2), int64(6)
memory usage: 2.0 MB
In [8]: train['num_orders'].describe()
Out[8]: count 456548.000000
mean 261.872760
std 395.922798
min 13.000000
25% 54.000000
50% 136.000000
75% 324.000000
max 24299.000000
Name: num_orders, dtype: float64
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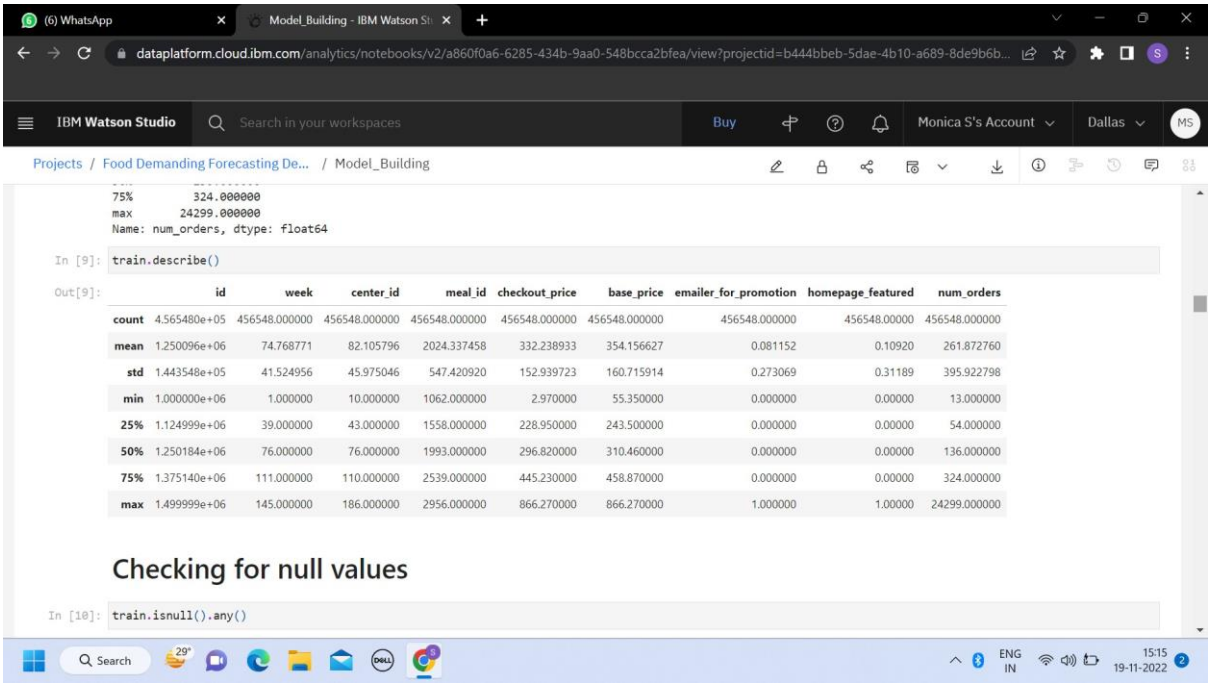
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75% 324.000000
max 24299.000000
Name: num_orders, dtype: float64
In [9]: train.describe()
Out[9]:
id week center_id meal_id checkout_price base_price emailer_for_promotion homepage_featured num_orders
count 4.565480e+05 456548.000000 456548.000000 456548.000000 456548.000000 456548.000000 456548.000000 456548.00000 456548.000000
mean 1.250096e+06 74.768771 82.105796 2024.337458 332.238933 354.156627 0.081152 0.10920 261.872760
std 1.443548e+05 41.524956 45.975046 547.420920 152.939723 160.715914 0.273069 0.31189 395.922798
min 1.000000e+06 1.000000 10.000000 1062.000000 2.970000 55.350000 0.000000 0.00000 13.000000
25% 1.124999e+06 39.000000 43.000000 1558.000000 228.950000 243.500000 0.000000 0.00000 54.000000
50% 1.250184e+06 76.000000 76.000000 1993.000000 296.820000 310.460000 0.000000 0.00000 136.000000
75% 1.375140e+06 111.000000 110.000000 2539.000000 445.230000 458.870000 0.000000 0.00000 324.000000
max 1.499999e+06 145.000000 186.000000 2956.000000 866.270000 866.270000 1.000000 1.00000 24299.000000
Checking for null values
In [10]: train.isnull().any()
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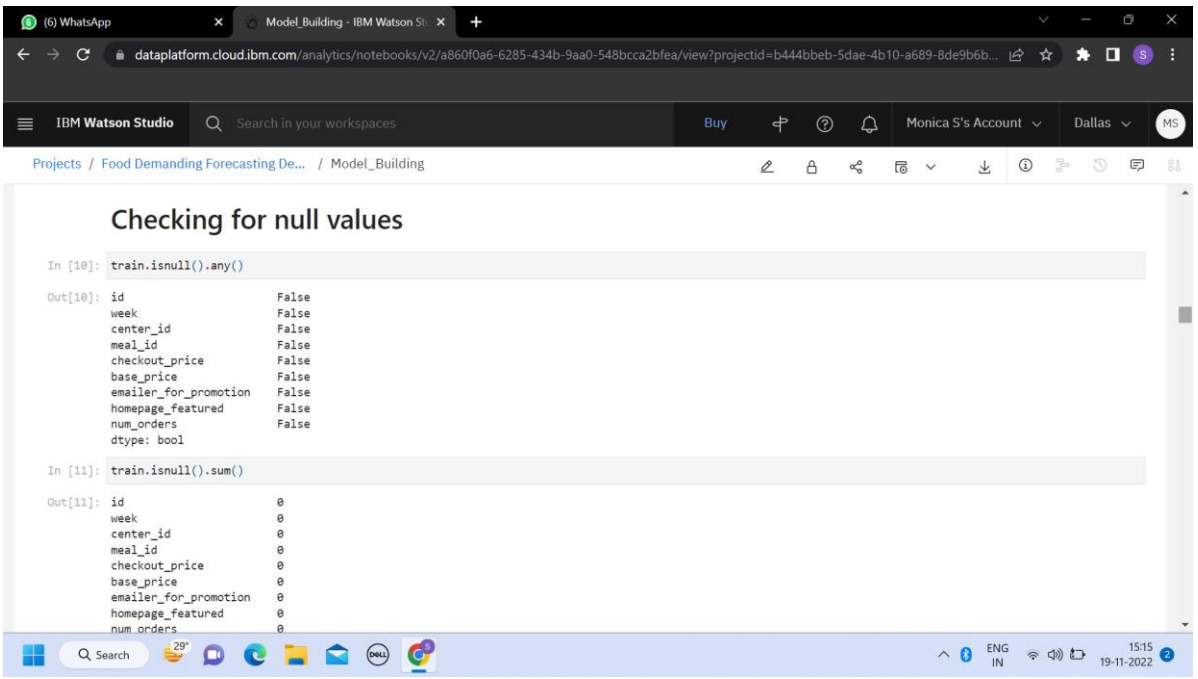
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Checking for null values
In [10]: train.isnull().any()
Out[10]:
id False
week False
center_id False
meal_id False
checkout_price False
base_price False
emailer_for_promotion False
homepage_featured False
num_orders False
dtype: bool
In [11]: train.isnull().sum()
Out[11]:
id 0
week 0
center_id 0
meal_id 0
checkout_price 0
base_price 0
emailer_for_promotion 0
homepage_featured 0
num_orders 0
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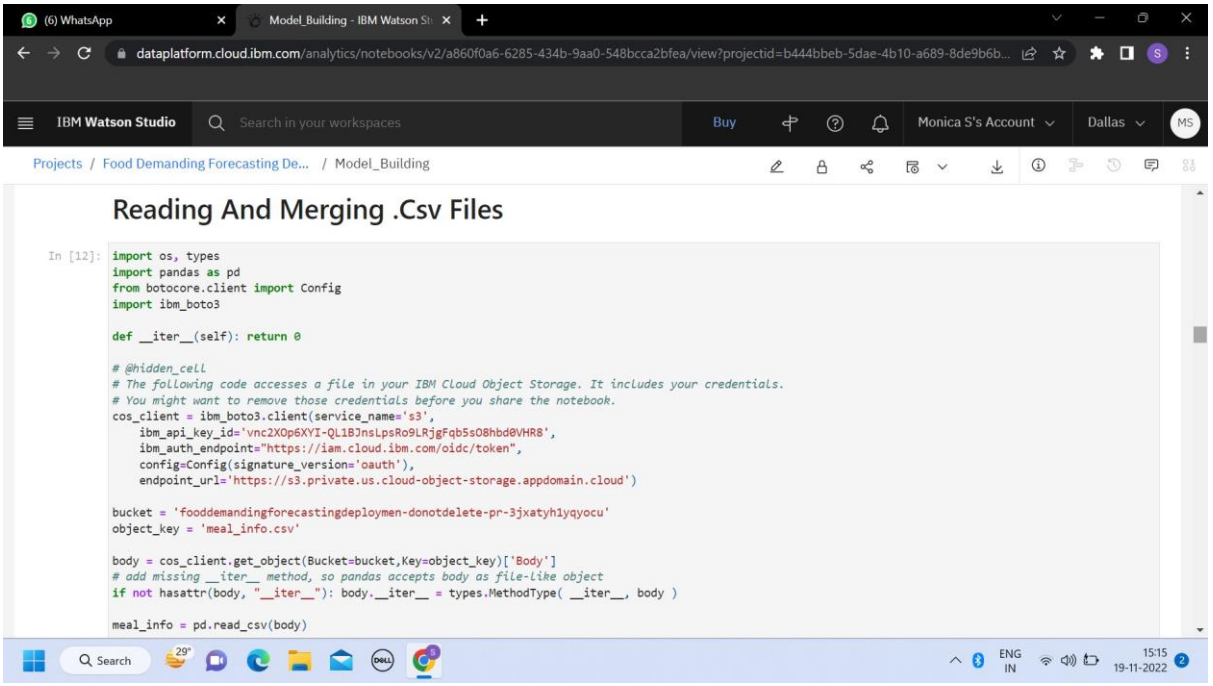
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Reading And Merging .Csv Files
In [12]:
import os, types
import pandas as pd
from botocore.client import Config
import ibm_boto3
def __iter__(self): return 0
# @hidden_cell
# The following code accesses a file in your IBM Cloud Object Storage. It includes your credentials.
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ibm_auth_endpoint="https://iam.cloud.ibm.com/oidc/token",
config=Config(signature_version='oauth'),
endpoint_url='https://s3.private.us.cloud-object-storage.appdomain.cloud')
bucket = 'fooddemandingforecastingdeploymen-donotdelete-pr-3jxatyh1yqyocu'
object_key = 'meal_info.csv'
body = cos_client.get_object(Bucket=bucket,Key=object_key)['Body']
# add missing __iter__ method, so pandas accepts body as file-like object
if not hasattr(body, "__iter__"): body.__iter__ = types.MethodType( __iter__, body )
meal_info = pd.read_csv(body)
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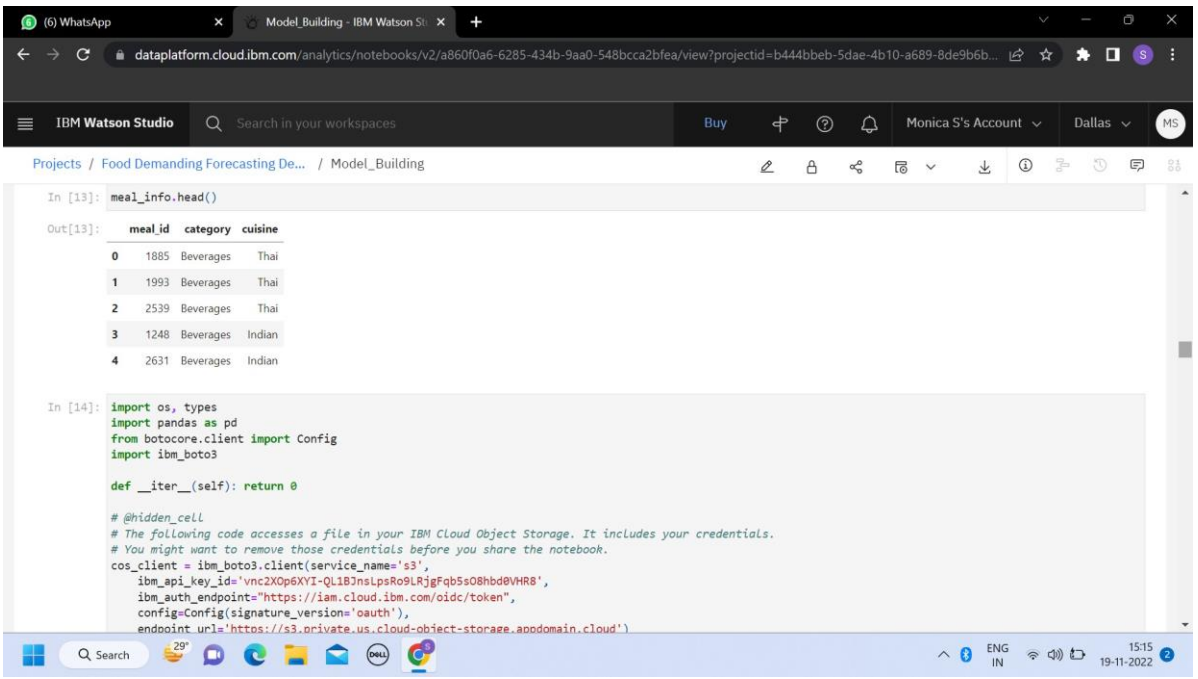
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In [13]:
meal_info.head()
Out[13]:
meal_id category cuisine
0 1885 Beverages Thai
1 1993 Beverages Thai
2 2539 Beverages Thai
3 1248 Beverages Indian
4 2631 Beverages Indian
In [14]:
import os, types
import pandas as pd
from botocore.client import Config
import ibm_boto3
def __iter__(self): return 0
# @hidden_cell
# The following code accesses a file in your IBM Cloud Object Storage. It includes your credentials.
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ibm_auth_endpoint="https://iam.cloud.ibm.com/oidc/token",
config=Config(signature_version='oauth'),
endpoint_url='https://s3.private.us.cloud-object-storage.appdomain.cloud')
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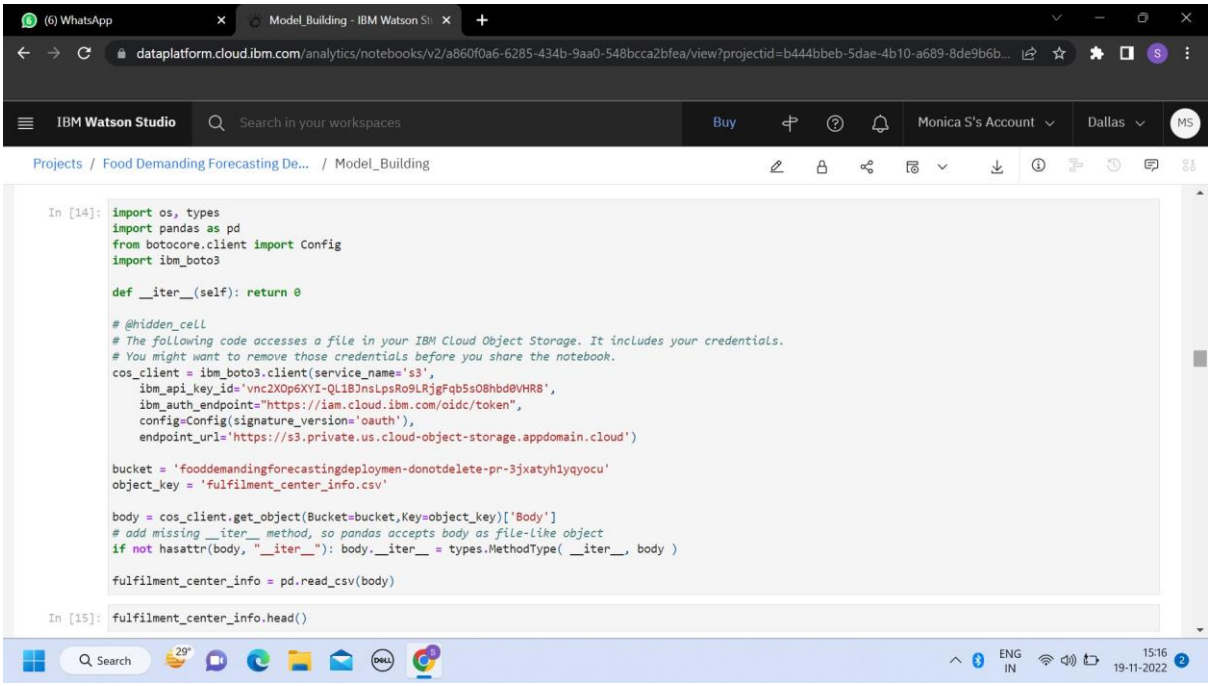
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import pandas as pd
from botocore.client import Config
import ibm_boto3
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# @hidden_cell
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ibm_auth_endpoint="https://iam.cloud.ibm.com/oidc/token",
config=Config(signature_version='oauth'),
endpoint_url='https://s3.private.us.cloud-object-storage.appdomain.cloud')
bucket = 'fooddemandingforecastingdeploymen-donotdelete-pr-3jxatyh1yqyocu'
object_key = 'fulfilment_center_info.csv'
body = cos_client.get_object(Bucket=bucket,Key=object_key)['Body']
# add missing __iter__ method, so pandas accepts body as file-like object
if not hasattr(body, "__iter__"): body.__iter__ = types.MethodType( __iter__, body )
fulfilment_center_info = pd.read_csv(body)
In [15]:
fulfilment_center_info.head()
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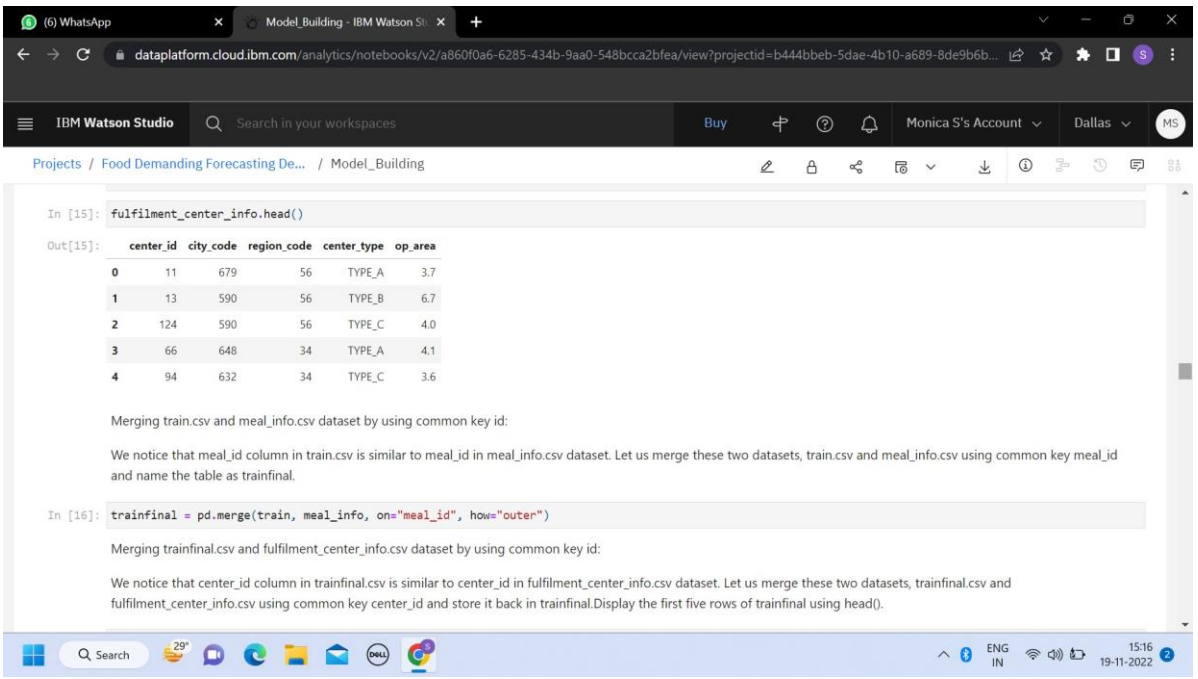
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In [15]:
fulfilment_center_info.head()
Out[15]:
center_id city_code region_code center_type op_area
0 11 679 56 TYPE_A 3.7
1 13 590 56 TYPE_B 6.7
2 124 590 56 TYPE_C 4.0
3 66 648 34 TYPE_A 4.1
4 94 632 34 TYPE_C 3.6
Merging train.csv and meal_info.csv dataset by using common key id:
We notice that meal_id column in train.csv is similar to meal_id in meal_info.csv dataset. Let us merge these two datasets, train.csv and meal_info.csv using common key meal_id and name the table as trainfinal.
In [16]:
trainfinal = pd.merge(train, meal_info, on="meal_id", how="outer")
Merging trainfinal.csv and fulfilment_center_info.csv dataset by using common key id:
We notice that center_id column in trainfinal.csv is similar to center_id in fulfilment_center_info.csv dataset. Let us merge these two datasets, trainfinal.csv and fulfilment_center_info.csv using common key center_id and store it back in trainfinal.Display the first five rows of trainfinal using head().
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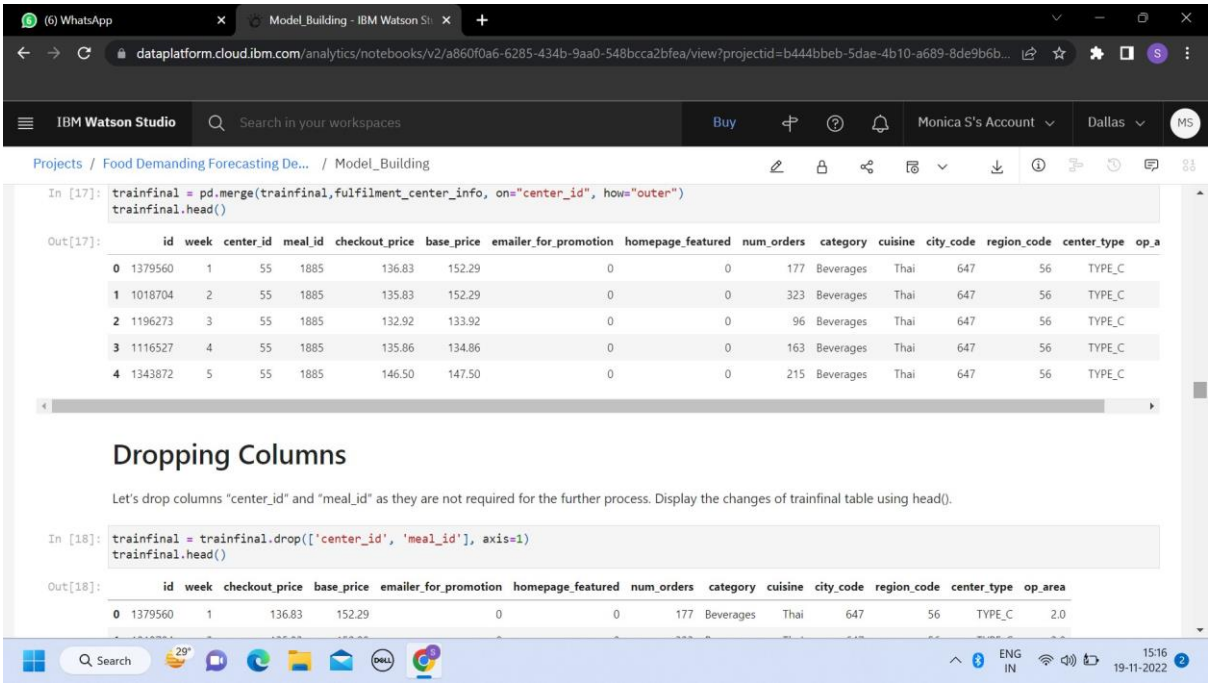
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In [17]:
trainfinal = pd.merge(trainfinal,fulfilment_center_info, on="center_id", how="outer")
trainfinal.head()
Out[17]:
id week center_id meal_id checkout_price base_price emailer_for_promotion homepage_featured num_orders category cuisine city_code region_code center_type op_a
0 1379560 1 55 1885 136.83 152.29 0 0 177 Beverages Thai 647 56 TYPE_C
1 1018704 2 55 1885 135.83 152.29 0 0 323 Beverages Thai 647 56 TYPE_C
2 1196273 3 55 1885 132.92 133.92 0 0 96 Beverages Thai 647 56 TYPE_C
3 1116527 4 55 1885 135.86 134.86 0 0 163 Beverages Thai 647 56 TYPE_C
4 1343872 5 55 1885 146.50 147.50 0 0 215 Beverages Thai 647 56 TYPE_C
Dropping Columns
Let's drop columns "center_id" and "meal_id" as they are not required for the further process. Display the changes of trainfinal table using head().
In [18]:
trainfinal = trainfinal.drop(['center_id', 'meal_id'], axis=1)
trainfinal.head()
Out[18]:
id week checkout_price base_price emailer_for_promotion homepage_featured num_orders category cuisine city_code region_code center_type op_area
0 1379560 1 136.83 152.29 0 0 177 Beverages Thai 647 56 TYPE_C 2.0
Search
ENG
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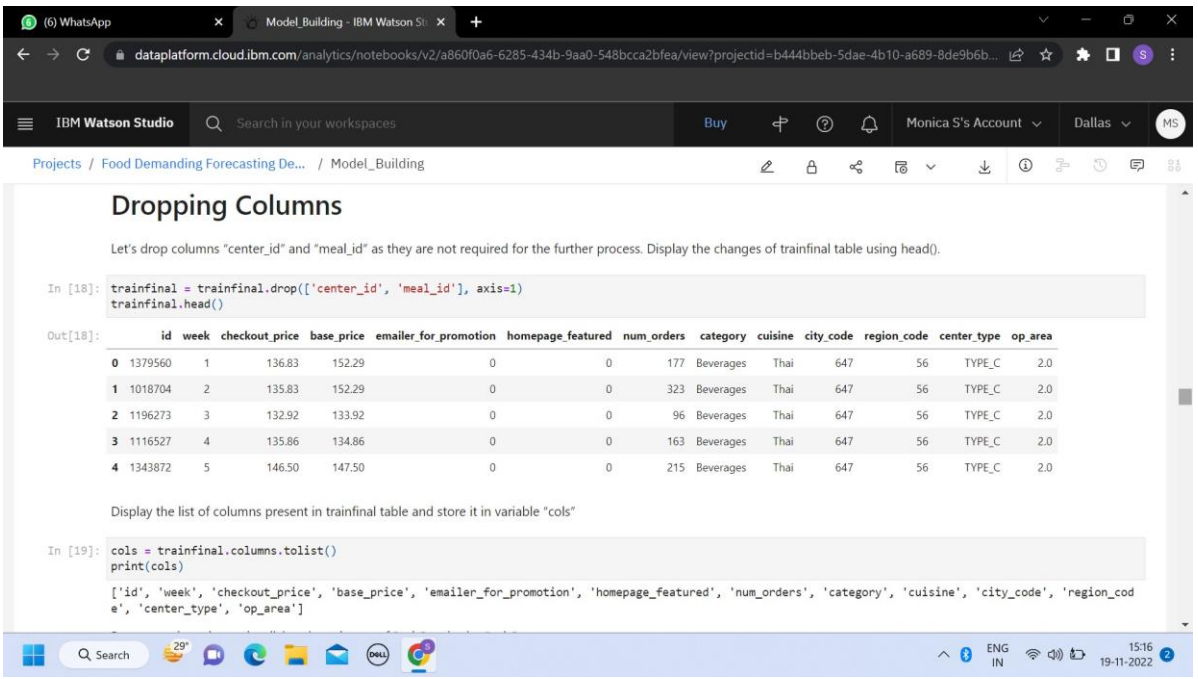
15:16
19-11-2022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Dropping Columns
Let's drop columns "center_id" and "meal_id" as they are not required for the further process. Display the changes of trainfinal table using head().
In [18]:
trainfinal = trainfinal.drop(['center_id', 'meal_id'], axis=1)
trainfinal.head()
Out[18]:
id week checkout_price base_price emailer_for_promotion homepage_featured num_orders category cuisine city_code region_code center_type op_area
0 1379560 1 136.83 152.29 0 0 177 Beverages Thai 647 56 TYPE_C 2.0
1 1018704 2 135.83 152.29 0 0 323 Beverages Thai 647 56 TYPE_C 2.0
2 1196273 3 132.92 133.92 0 0 96 Beverages Thai 647 56 TYPE_C 2.0
3 1116527 4 135.86 134.86 0 0 163 Beverages Thai 647 56 TYPE_C 2.0
4 1343872 5 146.50 147.50 0 0 215 Beverages Thai 647 56 TYPE_C 2.0
Display the list of columns present in trainfinal table and store it in variable "cols"
In [19]:
cols = trainfinal.columns.tolist()
print(cols)
['id', 'week', 'checkout_price', 'base_price', 'emailer_for_promotion', 'homepage_featured', 'num_orders', 'category', 'cuisine', 'city_code', 'region_cod
e', 'center_type', 'op_area']
Search
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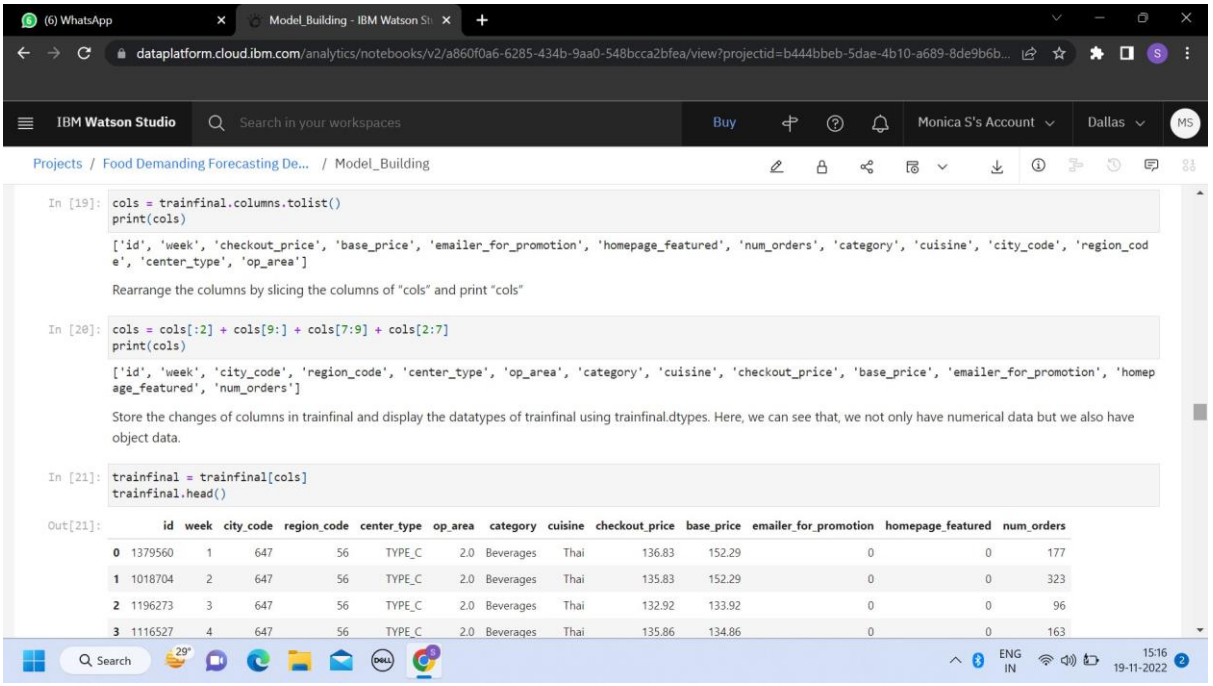
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In [19]:
cols = trainfinal.columns.tolist()
print(cols)
['id', 'week', 'checkout_price', 'base_price', 'emailer_for_promotion', 'homepage_featured', 'num_orders', 'category', 'cuisine', 'city_code', 'region_code', 'center_type', 'op_area']
Rearrange the columns by slicing the columns of "cols" and print "cols"
In [20]:
cols = cols[:2] + cols[9:] + cols[7:9] + cols[2:7]
print(cols)
['id', 'week', 'city_code', 'region_code', 'center_type', 'op_area', 'category', 'cuisine', 'checkout_price', 'base_price', 'emailer_for_promotion', 'homepage_featured', 'num_orders']
Store the changes of columns in trainfinal and display the datatypes of trainfinal using trainfinal.dtypes. Here, we can see that, we not only have numerical data but we also have object data.
In [21]:
trainfinal = trainfinal[cols]
trainfinal.head()
Out[21]:
id week city_code region_code center_type op_area category cuisine checkout_price base_price emailer_for_promotion homepage_featured num_orders
0 1379560 1 647 56 TYPE_C 2.0 Beverages Thai 136.83 152.29 0 0 177
1 1018704 2 647 56 TYPE_C 2.0 Beverages Thai 135.83 152.29 0 0 323
2 1196273 3 647 56 TYPE_C 2.0 Beverages Thai 132.92 133.92 0 0 96
3 1116527 4 647 56 TYPE_C 2.0 Beverages Thai 135.86 134.86 0 0 163
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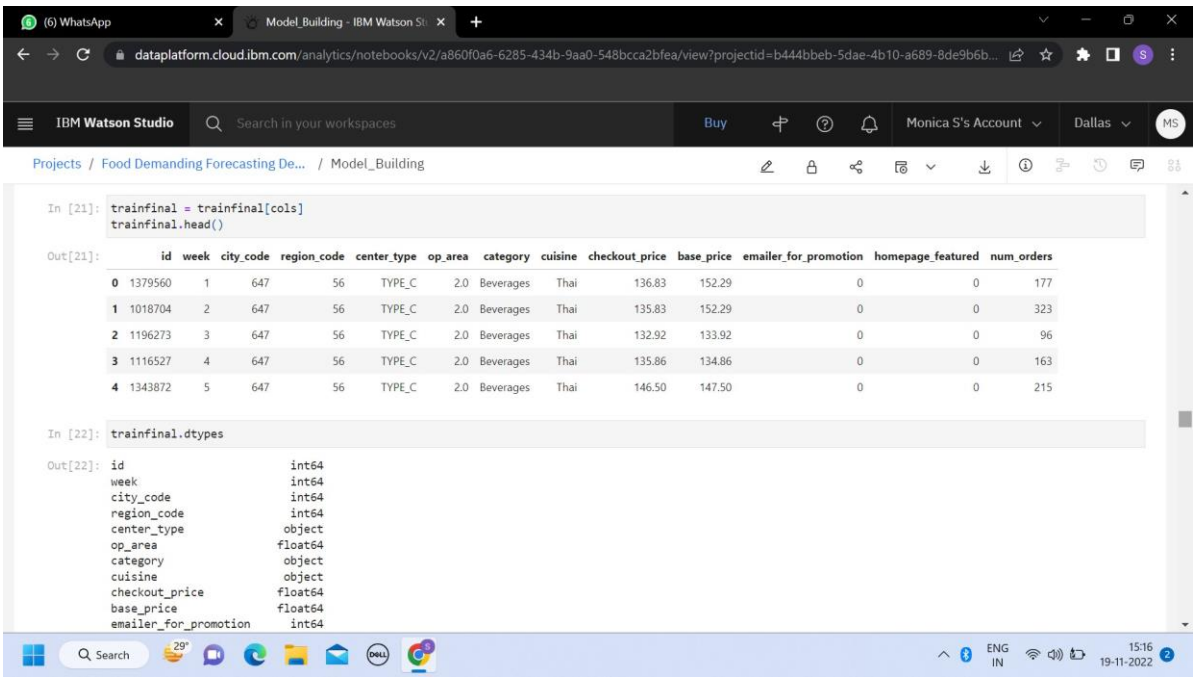
19-11-2022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In [21]:
trainfinal = trainfinal[cols]
trainfinal.head()
Out[21]:
id week city_code region_code center_type op_area category cuisine checkout_price base_price emailer_for_promotion homepage_featured num_orders
0 1379560 1 647 56 TYPE_C 2.0 Beverages Thai 136.83 152.29 0 0 177
1 1018704 2 647 56 TYPE_C 2.0 Beverages Thai 135.83 152.29 0 0 323
2 1196273 3 647 56 TYPE_C 2.0 Beverages Thai 132.92 133.92 0 0 96
3 1116527 4 647 56 TYPE_C 2.0 Beverages Thai 135.86 134.86 0 0 163
4 1343872 5 647 56 TYPE_C 2.0 Beverages Thai 146.50 147.50 0 0 215
In [22]:
trainfinal.dtypes
Out[22]:
id int64
week int64
city_code int64
region_code int64
center_type object
op_area float64
category object
cuisine object
checkout_price float64
base_price float64
emailer_for_promotion int64
Search
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## Label Encoding

Typically, any structured dataset includes multiple columns with combination of numerical as well as categorical variables. A machine can only understand the numbers. It cannot understand the text.That's essentially the case with Machine Learning algorithms too.

We need to convert each text category to numbers in order for the machine to process those using mathematical equations. Label Encoding is a popular encoding technique for handling categorical variables implemented using the scikit-learn library in python.In this technique, each label is assigned a unique integer based on alphabetical ordering.

In [23]:
```
from sklearn.preprocessing import LabelEncoder

lb1 = LabelEncoder()
trainfinal['center_type'] = lb1.fit_transform(trainfinal['center_type'])
lb2 = LabelEncoder()
trainfinal['category'] = lb1.fit_transform(trainfinal['category'])
lb3 = LabelEncoder()
trainfinal['cuisine'] = lb1.fit_transform(trainfinal['cuisine'])
```

In the above code we have selected text class categorical columns for performing label encoding.

In [24]:
```
trainfinal.head()
```

Out[24]:

| | id | week | city_code | region_code | center_type | op_area | category | cuisine | checkout_price | base_price | emailer_for_promotion | homepage_featured | num_orders |
|---|---|---|---|---|---|---|---|---|---|---|---|---|---|
| 0 | 1379560 | 1 | 647 | 56 | 2 | 2.0 | 0 | 3 | 136.83 | 152.29 | 0 | 0 | 177 |
| 1 | 1018704 | 2 | 647 | 56 | 2 | 2.0 | 0 | 3 | 135.83 | 152.29 | 0 | 0 | 323 |





After performing label encoding, alphabetical classes - center type, category and city code are converted to numeric values.

Finally display number of rows and columns of trainfinal using shape()

In [25]:
```
trainfinal.shape
```

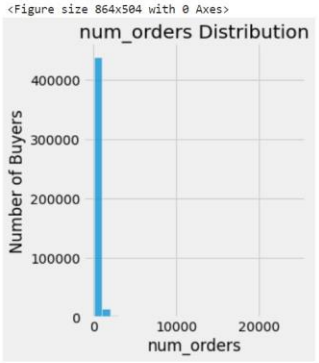
Out[25]:
```
(456548, 13)
```

## Data Visualization

In [26]:
```
plt.style.use('fivethirtyeight')
plt.figure(figsize=(12,7))
sns.displot(trainfinal.num_orders, bins = 25)
plt.xlabel("num_orders")
plt.ylabel("Number of Buyers")
plt.title("num_orders Distribution")
```

Out[26]:
```
Text(0.5, 1.0, 'num_orders Distribution')
<Figure size 864x504 with 0 Axes>
```

<Figure size 864x504 with 0 Axes>




Drop the column "id" and find the correlation between the columns.

In [27]:
```
trainfinal2 = trainfinal.drop(['id'],axis=1)
```



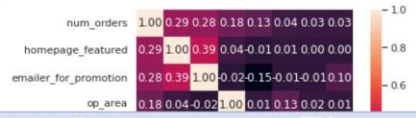
Drop the column "id" and find the correlation between the columns.

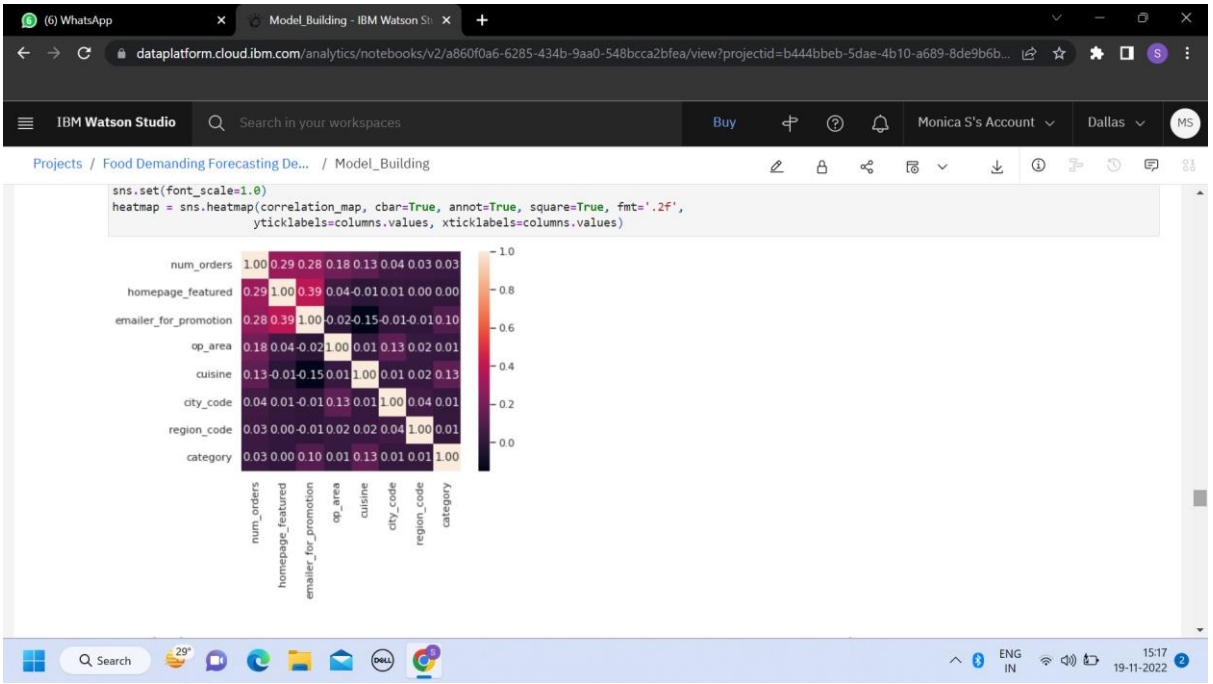
In [27]:
```
trainfinal2 = trainfinal.drop(['id'],axis=1)
correlation = trainfinal2.corr(method='pearson')
columns = correlation.nlargest(8, 'num_orders').index
columns
```

Out[27]:
```
Index(['num_orders', 'homepage_featured', 'emailer_for_promotion', 'op_area',
       'cuisine', 'city_code', 'region_code', 'category'],
      dtype='object')
```

Correlation is a statistical relationship between two variables and it could be positive, meaning both variables move in the same direction, or negative, meaning that when one variable's value increases, the other variables' values decrease. With the help of seaborn heatmap we will be plotting the heatmap and for finding the correlation between variable we have corr() available

In [28]:
```
correlation_map = np.corrcoef(trainfinal2[columns].values.T)
sns.set(font_scale=1.0)
heatmap = sns.heatmap(correlation_map, cbar=True, annot=True, square=True, fmt='.2f',
                      yticklabels=columns.values, xticklabels=columns.values)
```

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sns.set(font_scale=1.0)
heatmap = sns.heatmap(correlation_map, cbar=True, annot=True, square=True, fmt='.2f',
yticklabels=columns.values, xticklabels=columns.values)
num_orders 1.00 0.29 0.28 0.18 0.13 0.04 0.03 0.03
homepage_featured 0.29 1.00 0.39 0.04 -0.01 0.01 0.00 0.00
emailer_for_promotion 0.28 0.39 1.00 -0.02 -0.15 -0.01 -0.01 0.10
op_area 0.18 0.04 -0.02 1.00 0.01 0.13 0.02 0.01
cuisine 0.13 -0.01 -0.15 0.01 1.00 0.01 0.02 0.13
city_code 0.04 0.01 -0.01 0.13 0.01 1.00 0.04 0.01
region_code 0.03 0.00 -0.01 0.02 0.02 0.04 1.00 0.01
category 0.03 0.00 0.10 0.01 0.13 0.01 0.01 1.00
num_orders
homepage_featured
emailer_for_promotion
op_area
cuisine
city_code
region_code
category
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0.2
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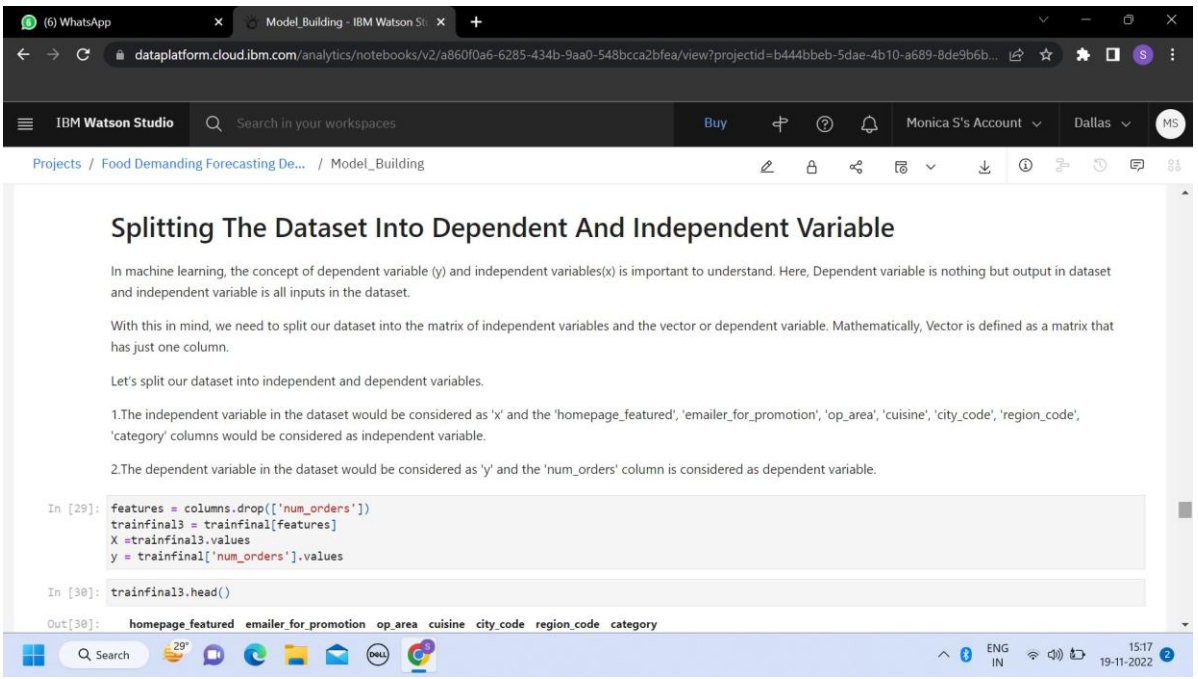
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Splitting The Dataset Into Dependent And Independent Variable
In machine learning, the concept of dependent variable (y) and independent variables(x) is important to understand. Here, Dependent variable is nothing but output in dataset and independent variable is all inputs in the dataset.
With this in mind, we need to split our dataset into the matrix of independent variables and the vector or dependent variable. Mathematically, Vector is defined as a matrix that has just one column.
Let's split our dataset into independent and dependent variables.
1.The independent variable in the dataset would be considered as 'x' and the 'homepage_featured', 'emailer_for_promotion', 'op_area', 'cuisine', 'city_code', 'region_code', 'category' columns would be considered as independent variable.
2.The dependent variable in the dataset would be considered as 'y' and the 'num_orders' column is considered as dependent variable.
In [29]:
features = columns.drop(['num_orders'])
trainfinal3 = trainfinal[features]
X =trainfinal3.values
y = trainfinal['num_orders'].values
In [30]:
trainfinal3.head()
Out[30]:
homepage_featured emailer_for_promotion op_area cuisine city_code region_code category
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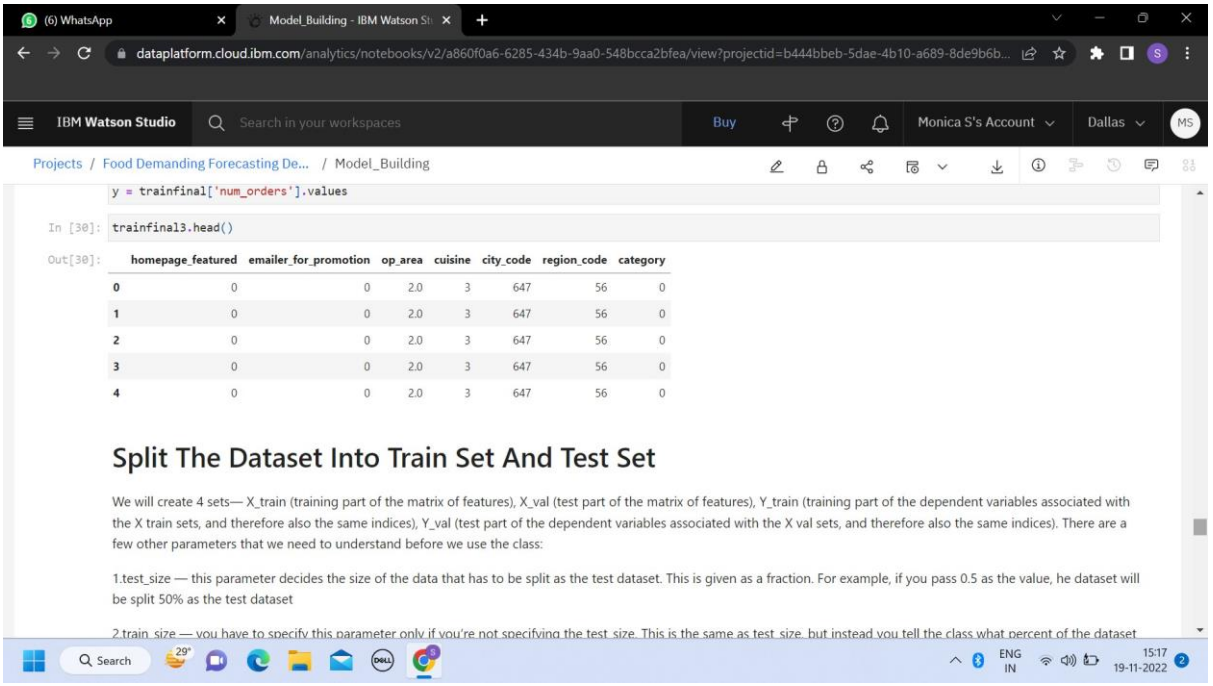
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y = trainfinal['num_orders'].values
In [30]:
trainfinal3.head()
Out[30]:
homepage_featured emailer_for_promotion op_area cuisine city_code region_code category
0 0 0 2.0 3 647 56 0
1 0 0 2.0 3 647 56 0
2 0 0 2.0 3 647 56 0
3 0 0 2.0 3 647 56 0
4 0 0 2.0 3 647 56 0
Split The Dataset Into Train Set And Test Set
We will create 4 sets— X_train (training part of the matrix of features), X_val (test part of the matrix of features), Y_train (training part of the dependent variables associated with the X train sets, and therefore also the same indices), Y_val (test part of the dependent variables associated with the X val sets, and therefore also the same indices). There are a few other parameters that we need to understand before we use the class:
1.test_size — this parameter decides the size of the data that has to be split as the test dataset. This is given as a fraction. For example, if you pass 0.5 as the value, he dataset will be split 50% as the test dataset
2.train_size — you have to specify this parameter only if you're not specifying the test_size. This is the same as test_size, but instead you tell the class what percent of the dataset
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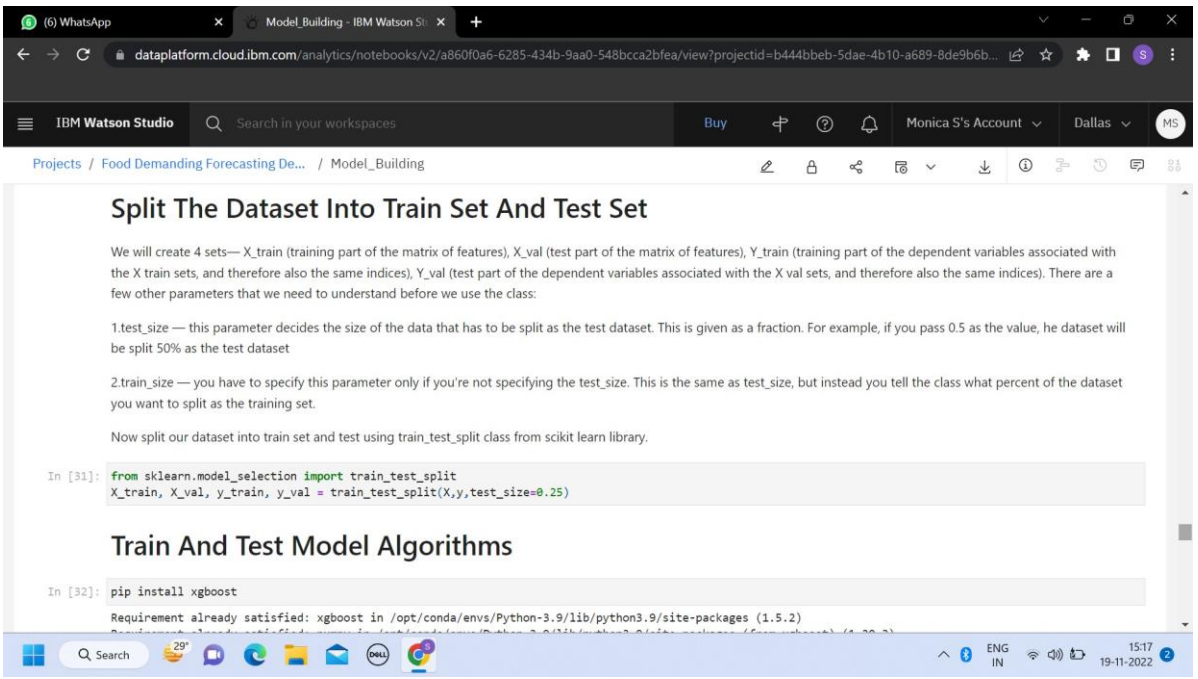
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Split The Dataset Into Train Set And Test Set
We will create 4 sets— X_train (training part of the matrix of features), X_val (test part of the matrix of features), Y_train (training part of the dependent variables associated with the X train sets, and therefore also the same indices), Y_val (test part of the dependent variables associated with the X val sets, and therefore also the same indices). There are a few other parameters that we need to understand before we use the class:
1.test_size — this parameter decides the size of the data that has to be split as the test dataset. This is given as a fraction. For example, if you pass 0.5 as the value, he dataset will be split 50% as the test dataset
2.train_size — you have to specify this parameter only if you're not specifying the test_size. This is the same as test_size, but instead you tell the class what percent of the dataset you want to split as the training set.
Now split our dataset into train set and test using train_test_split class from scikit learn library.
In [31]:
from sklearn.model_selection import train_test_split
X_train, X_val, y_train, y_val = train_test_split(X,y,test_size=0.25)
Train And Test Model Algorithms
In [32]:
pip install xgboost
Requirement already satisfied: xgboost in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (1.5.2)
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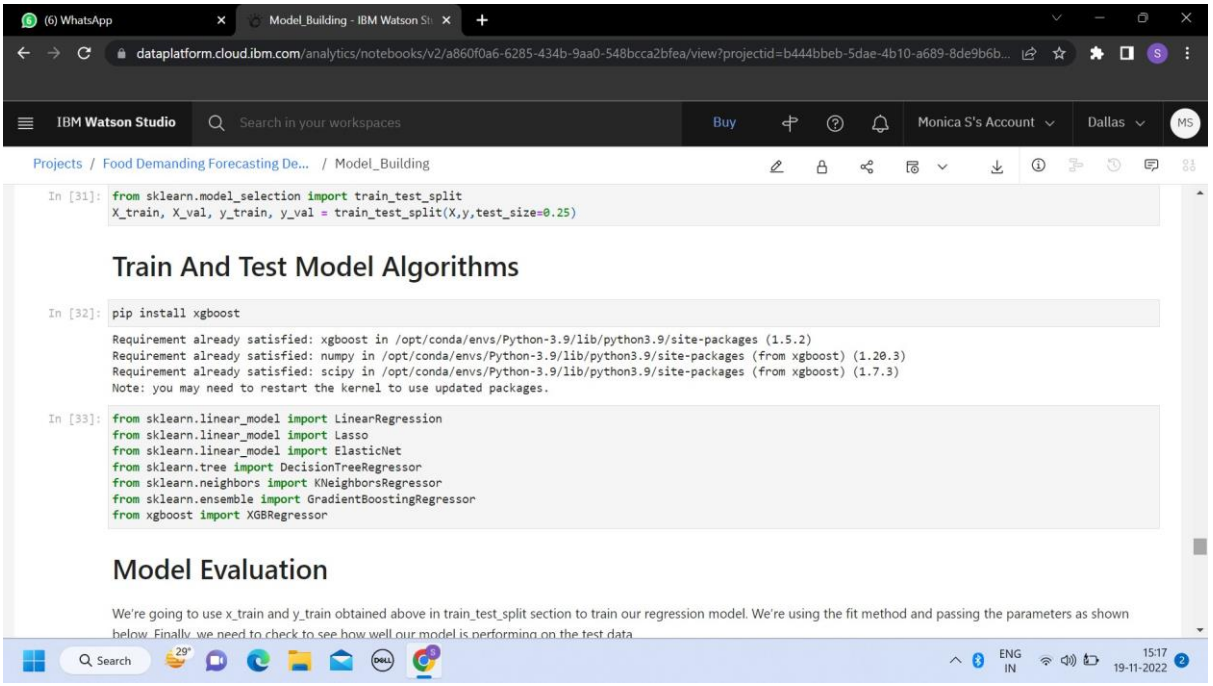
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In [31]:
from sklearn.model_selection import train_test_split
X_train, X_val, y_train, y_val = train_test_split(X,y,test_size=0.25)
Train And Test Model Algorithms
In [32]:
pip install xgboost
Requirement already satisfied: xgboost in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (1.5.2)
Requirement already satisfied: numpy in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from xgboost) (1.20.3)
Requirement already satisfied: scipy in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from xgboost) (1.7.3)
Note: you may need to restart the kernel to use updated packages.
In [33]:
from sklearn.linear_model import LinearRegression
from sklearn.linear_model import Lasso
from sklearn.linear_model import ElasticNet
from sklearn.tree import DecisionTreeRegressor
from sklearn.neighbors import KNeighborsRegressor
from sklearn.ensemble import GradientBoostingRegressor
from xgboost import XGBRegressor
Model Evaluation
We're going to use x_train and y_train obtained above in train_test_split section to train our regression model. We're using the fit method and passing the parameters as shown below. Finally, we need to check to see how well our model is performing on the test data
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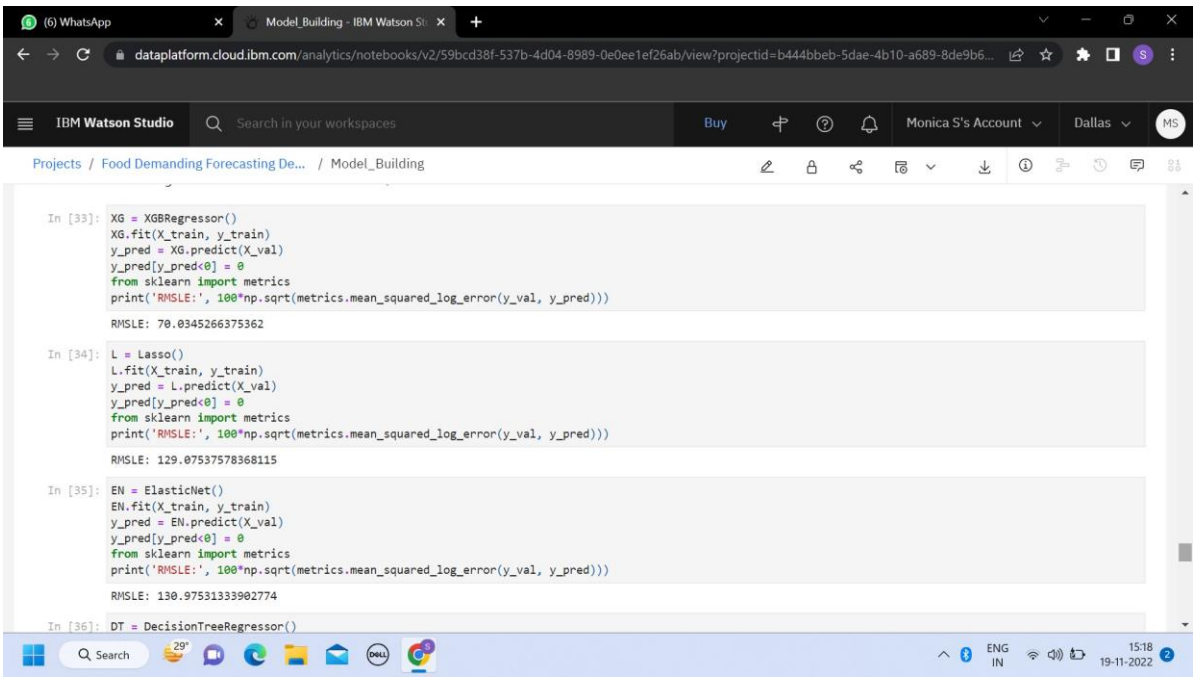
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In [33]:
XG = XGBRegressor()
XG.fit(X_train, y_train)
y_pred = XG.predict(X_val)
y_pred[y_pred<0] = 0
from sklearn import metrics
print('RMSLE:', 100*np.sqrt(metrics.mean_squared_log_error(y_val, y_pred)))
RMSLE: 70.0345266375362
In [34]:
L = Lasso()
L.fit(X_train, y_train)
y_pred = L.predict(X_val)
y_pred[y_pred<0] = 0
from sklearn import metrics
print('RMSLE:', 100*np.sqrt(metrics.mean_squared_log_error(y_val, y_pred)))
RMSLE: 129.07537578368115
In [35]:
EN = ElasticNet()
EN.fit(X_train, y_train)
y_pred = EN.predict(X_val)
y_pred[y_pred<0] = 0
from sklearn import metrics
print('RMSLE:', 100*np.sqrt(metrics.mean_squared_log_error(y_val, y_pred)))
RMSLE: 130.97531333902774
In [36]:
DT = DecisionTreeRegressor()
Search
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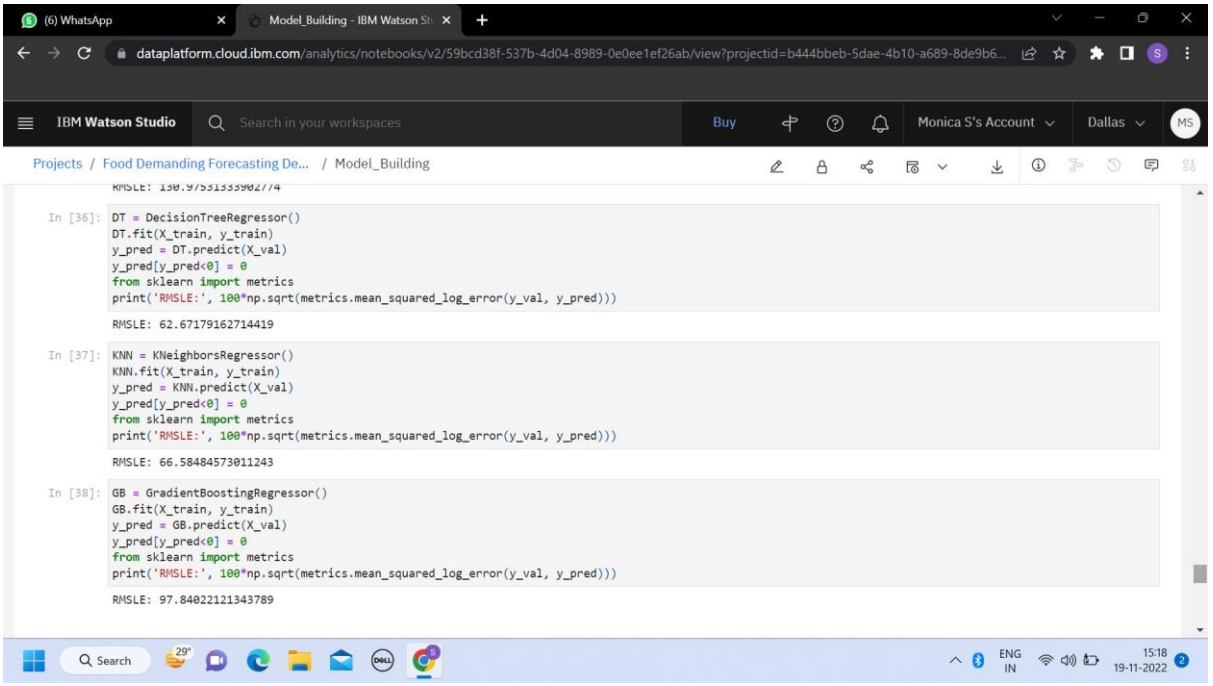
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Projects / Food Demanding Forecasting De... / Model_Building
RMSLE: 130.97513339027 74
In [36]:
DT = DecisionTreeRegressor()
DT.fit(X_train, y_train)
y_pred = DT.predict(X_val)
y_pred[y_pred<0] = 0
from sklearn import metrics
print('RMSLE:', 100*np.sqrt(metrics.mean_squared_log_error(y_val, y_pred)))
RMSLE: 62.67179162714419
In [37]:
KNN = KNeighborsRegressor()
KNN.fit(X_train, y_train)
y_pred = KNN.predict(X_val)
y_pred[y_pred<0] = 0
from sklearn import metrics
print('RMSLE:', 100*np.sqrt(metrics.mean_squared_log_error(y_val, y_pred)))
RMSLE: 66.58484573011243
In [38]:
GB = GradientBoostingRegressor()
GB.fit(X_train, y_train)
y_pred = GB.predict(X_val)
y_pred[y_pred<0] = 0
from sklearn import metrics
print('RMSLE:', 100*np.sqrt(metrics.mean_squared_log_error(y_val, y_pred)))
RMSLE: 97.84022121343789
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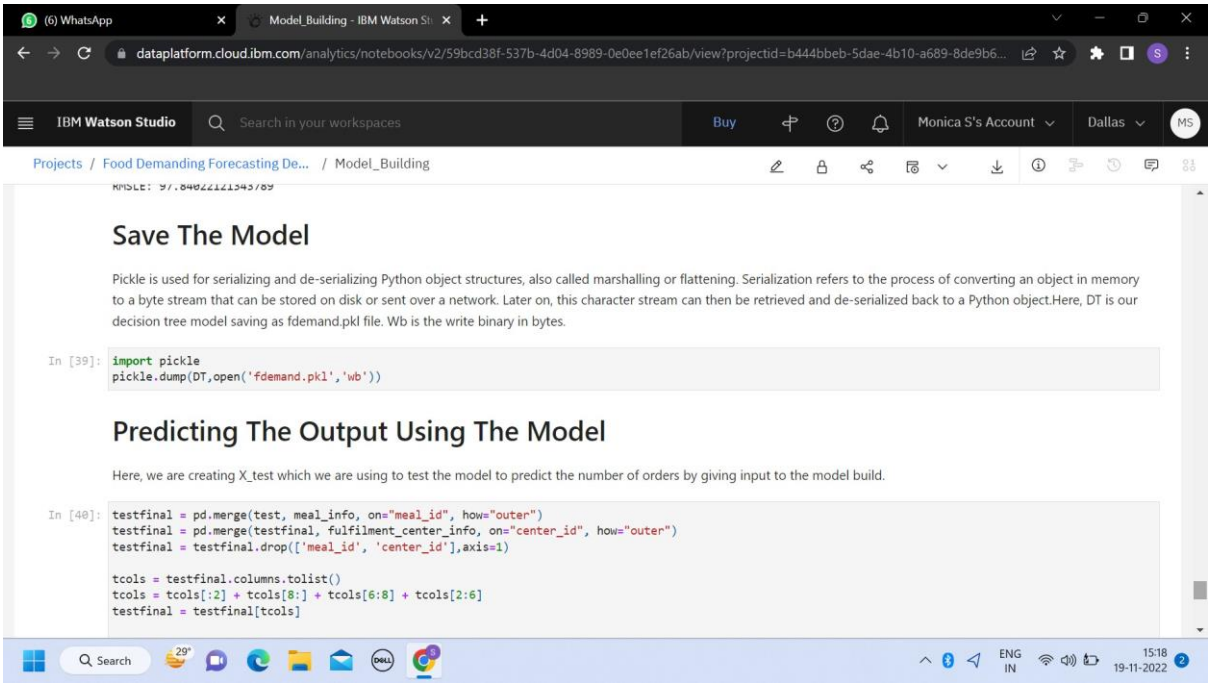
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RMSLE: 97.84022121345789
Save The Model
Pickle is used for serializing and de-serializing Python object structures, also called marshalling or flattening. Serialization refers to the process of converting an object in memory to a byte stream that can be stored on disk or sent over a network. Later on, this character stream can then be retrieved and de-serialized back to a Python object.Here, DT is our decision tree model saving as fdemand.pkl file. Wb is the write binary in bytes.
In [39]:
import pickle
pickle.dump(DT,open('fdemand.pkl','wb'))
Predicting The Output Using The Model
Here, we are creating X_test which we are using to test the model to predict the number of orders by giving input to the model build.
In [40]:
testfinal = pd.merge(test, meal_info, on="meal_id", how="outer")
testfinal = pd.merge(testfinal, fulfilment_center_info, on="center_id", how="outer")
testfinal = testfinal.drop(['meal_id', 'center_id'],axis=1)
tcols = testfinal.columns.tolist()
tcols = tcols[:2] + tcols[8:] + tcols[6:8] + tcols[2:6]
testfinal = testfinal[tcols]
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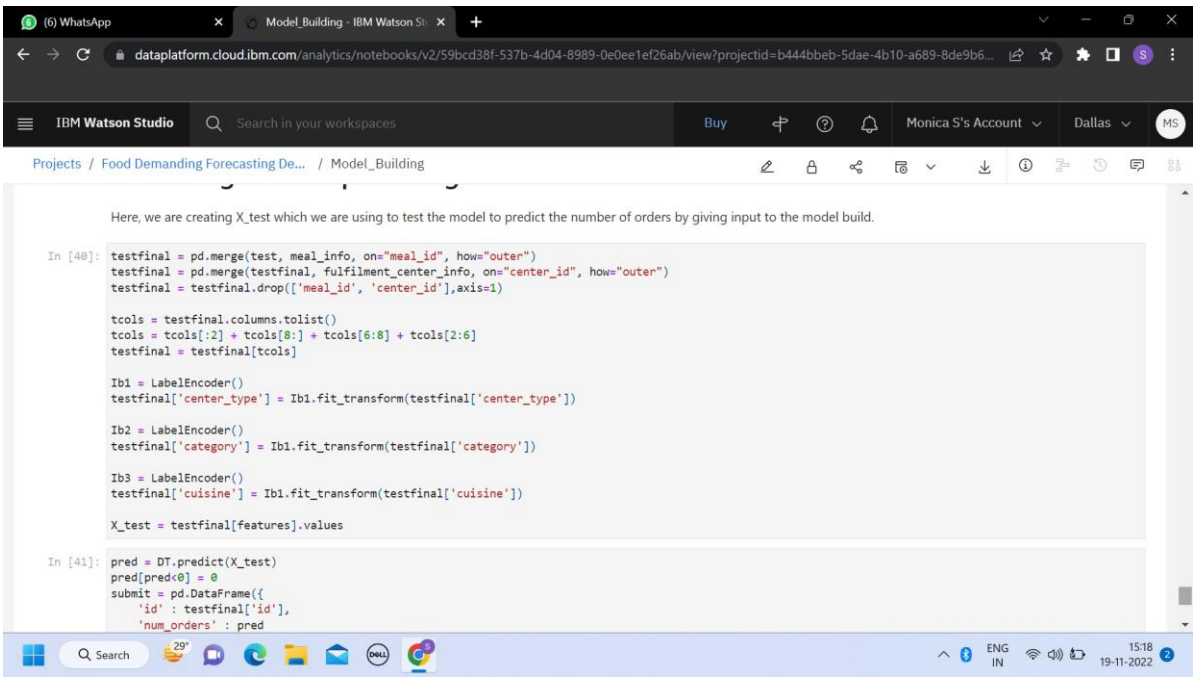
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Here, we are creating X_test which we are using to test the model to predict the number of orders by giving input to the model build.
In [40]:
testfinal = pd.merge(test, meal_info, on="meal_id", how="outer")
testfinal = pd.merge(testfinal, fulfilment_center_info, on="center_id", how="outer")
testfinal = testfinal.drop(['meal_id', 'center_id'],axis=1)
tcols = testfinal.columns.tolist()
tcols = tcols[:2] + tcols[8:] + tcols[6:8] + tcols[2:6]
testfinal = testfinal[tcols]
Ib1 = LabelEncoder()
testfinal['center_type'] = Ib1.fit_transform(testfinal['center_type'])
Ib2 = LabelEncoder()
testfinal['category'] = Ib1.fit_transform(testfinal['category'])
Ib3 = LabelEncoder()
testfinal['cuisine'] = Ib1.fit_transform(testfinal['cuisine'])
X_test = testfinal[features].values
In [41]:
pred = DT.predict(X_test)
pred[pred<0] = 0
submit = pd.DataFrame({
'id' : testfinal['id'],
'num_orders' : pred
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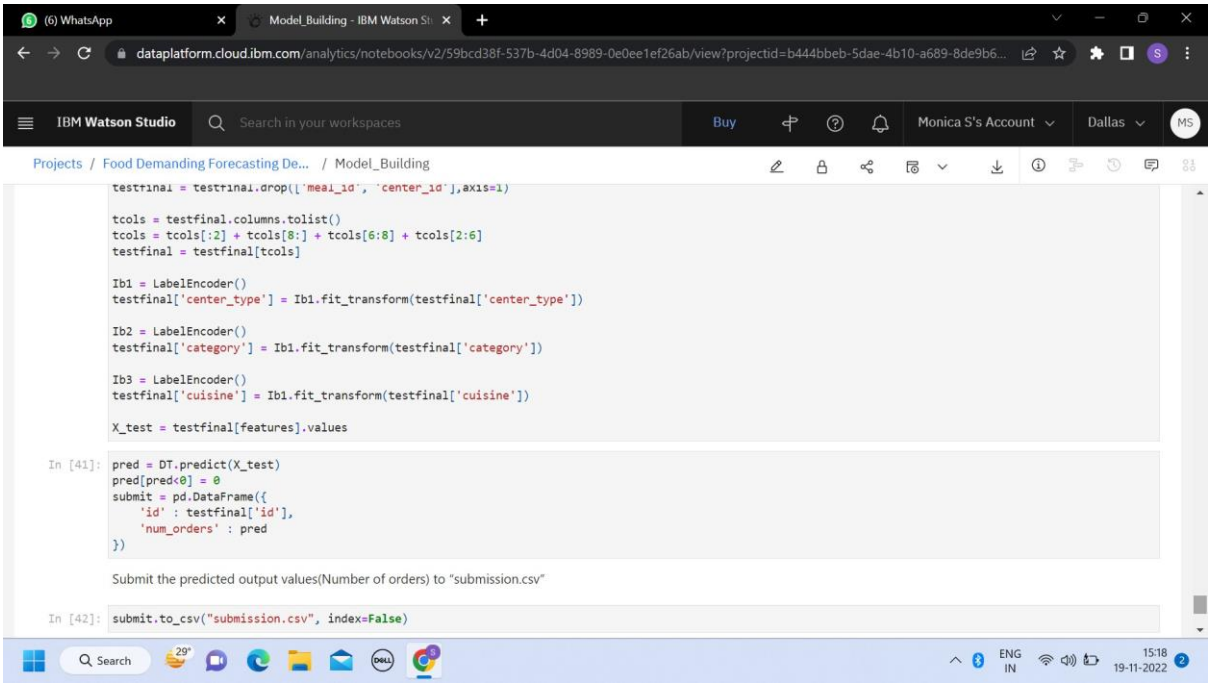
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testfinal = testfinal.drop(['meal_id', 'center_id'],axis=1)
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tcols = tcols[:2] + tcols[8:] + tcols[6:8] + tcols[2:6]
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testfinal['center_type'] = Ib1.fit_transform(testfinal['center_type'])
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In [41]:
pred = DT.predict(X_test)
pred[pred<0] = 0
submit = pd.DataFrame({
'id' : testfinal['id'],
'num_orders' : pred
})
Submit the predicted output values(Number of orders) to "submission.csv"
In [42]:
submit.to_csv("submission.csv", index=False)
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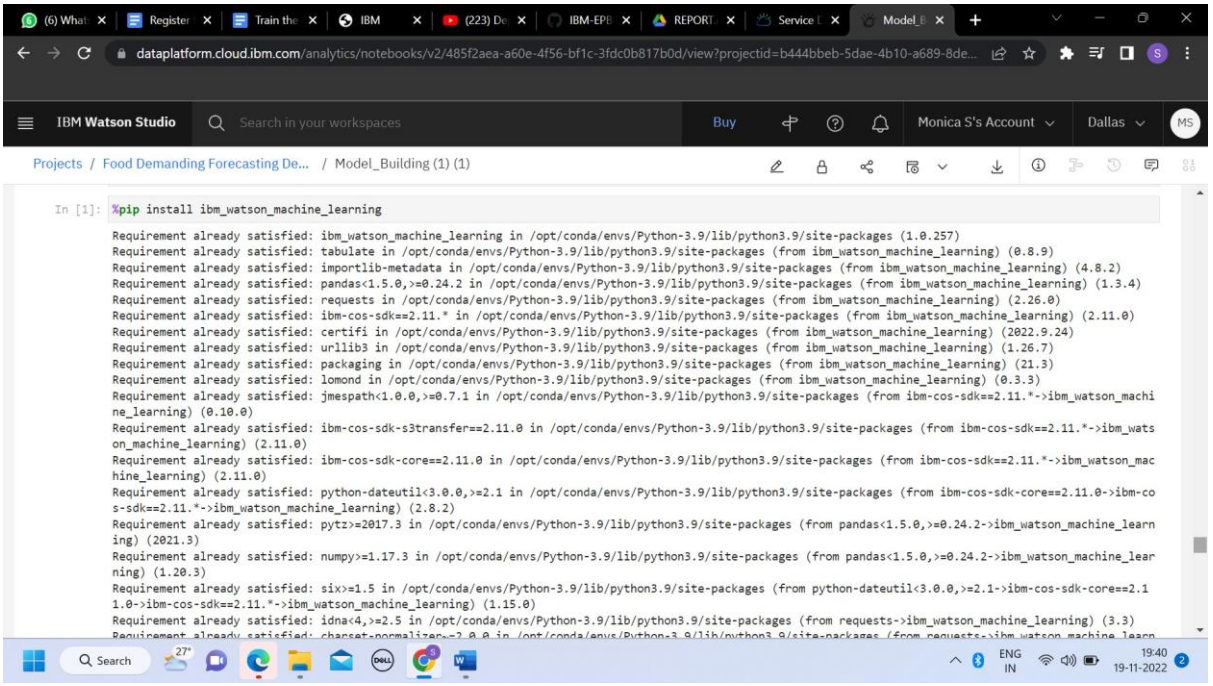
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In [1]:
%pip install ibm_watson_machine_learning
Requirement already satisfied: ibm_watson_machine_learning in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (1.0.257)
Requirement already satisfied: tabulate in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm_watson_machine_learning) (0.8.9)
Requirement already satisfied: importlib-metadata in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm_watson_machine_learning) (4.8.2)
Requirement already satisfied: pandas<1.5.0,>=0.24.2 in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm_watson_machine_learning) (1.3.4)
Requirement already satisfied: requests in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm_watson_machine_learning) (2.26.0)
Requirement already satisfied: ibm-cos-sdk==2.11.* in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm_watson_machine_learning) (2.11.0)
Requirement already satisfied: certifi in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm_watson_machine_learning) (2022.9.24)
Requirement already satisfied: urllib3 in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm_watson_machine_learning) (1.26.7)
Requirement already satisfied: packaging in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm_watson_machine_learning) (21.3)
Requirement already satisfied: lomond in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm_watson_machine_learning) (0.3.3)
Requirement already satisfied: jmespath<1.0.0,>=0.7.1 in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm-cos-sdk==2.11.*->ibm_watson_machine_learning) (0.10.0)
Requirement already satisfied: ibm-cos-sdk-s3transfer==2.11.0 in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm-cos-sdk==2.11.*->ibm_watson_machine_learning) (2.11.0)
Requirement already satisfied: ibm-cos-sdk-core==2.11.0 in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm-cos-sdk==2.11.*->ibm_watson_machine_learning) (2.11.0)
Requirement already satisfied: python-dateutil<3.0.0,>=2.1 in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm-cos-sdk-core==2.11.0->ibm-cos-sdk==2.11.*->ibm_watson_machine_learning) (2.8.2)
Requirement already satisfied: pytz>=2017.3 in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from pandas<1.5.0,>=0.24.2->ibm_watson_machine_learning) (2021.3)
Requirement already satisfied: numpy>=1.17.3 in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from pandas<1.5.0,>=0.24.2->ibm_watson_machine_learning) (1.20.3)
Requirement already satisfied: six>=1.5 in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from python-dateutil<3.0.0,>=2.1->ibm-cos-sdk-core==2.11.0->ibm-cos-sdk==2.11.*->ibm_watson_machine_learning) (1.15.0)
Requirement already satisfied: idna<4,>=2.5 in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from requests->ibm_watson_machine_learning) (3.3)
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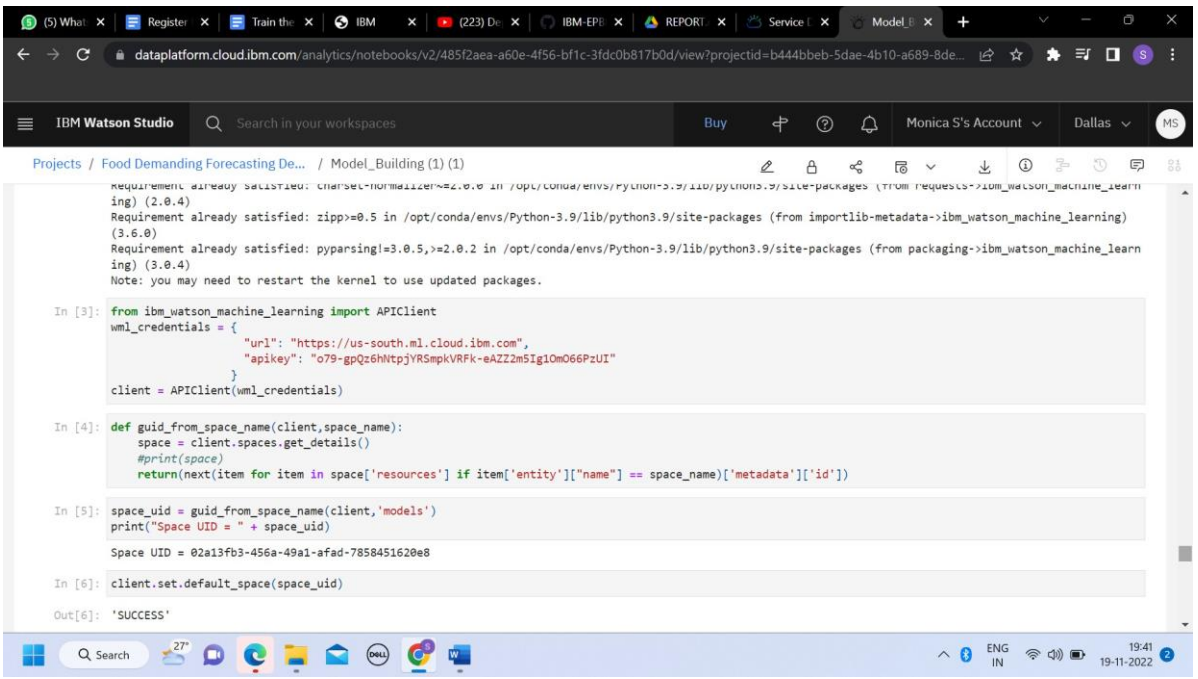
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ing) (2.0.4)
Requirement already satisfied: zipp>=0.5 in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from importlib-metadata->ibm_watson_machine_learning) (3.6.0)
Requirement already satisfied: pyparsing!=3.0.5,>=2.0.2 in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from packaging->ibm_watson_machine_learning) (3.0.4)
Note: you may need to restart the kernel to use updated packages.
In [3]:
from ibm_watson_machine_learning import APIClient
wml_credentials = {
"url": "https://us-south.ml.cloud.ibm.com",
"apikey": "o79-gpQz6hNtpjYRSmpkVRFk-eAZZ2m5Ig1OmO66PzUI"
}
client = APIClient(wml_credentials)
In [4]:
def guid_from_space_name(client,space_name):
space = client.spaces.get_details()
#print(space)
return(next(item for item in space['resources'] if item['entity']["name"] == space_name)['metadata']['id'])
In [5]:
space_uid = guid_from_space_name(client,'models')
print("Space UID = " + space_uid)
Space UID = 02a13fb3-456a-49a1-afad-7858451620e8
In [6]:
client.set.default_space(space_uid)
Out[6]:
'SUCCESS'
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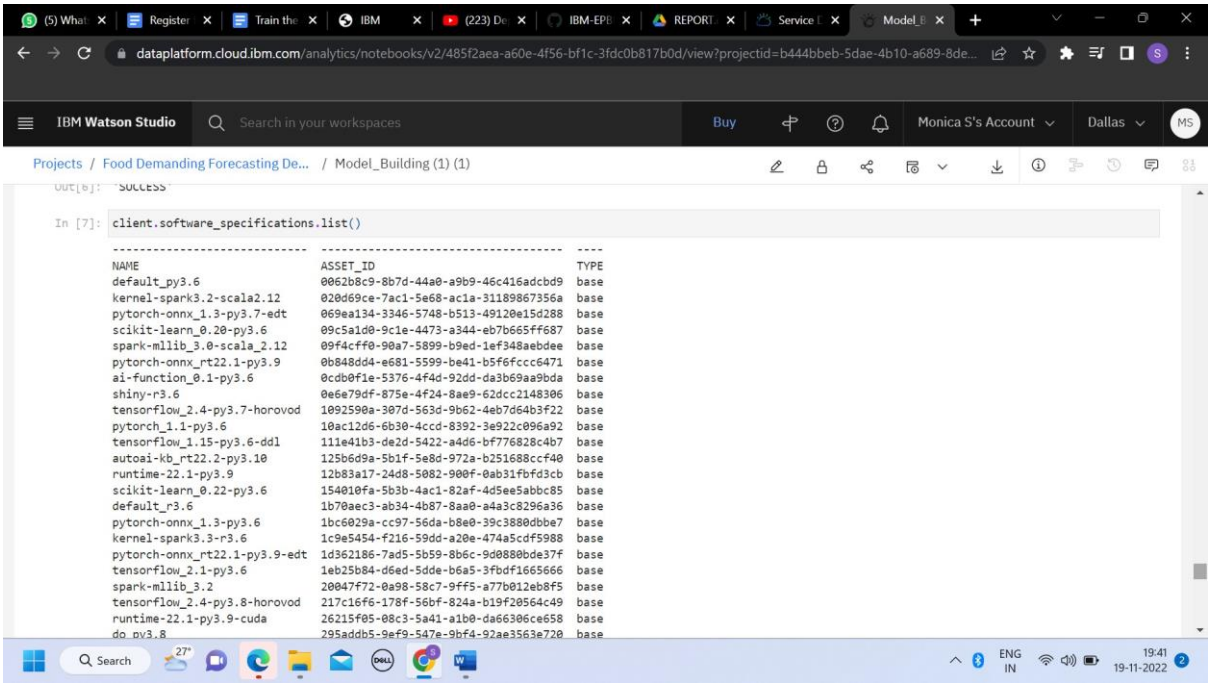
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Out[6]: 'SUCCESS'
In [7]: client.software_specifications.list()
NAME ASSET_ID TYPE
default_py3.6 0062b8c9-8b7d-44a0-a9b9-46c416adcbd9 base
kernel-spark3.2-scala2.12 020d69ce-7ac1-5e68-ac1a-31189867356a base
pytorch-onnx_1.3-py3.7-edt 069ea134-3346-5748-b513-49120e15d288 base
scikit-learn_0.20-py3.6 09c5a1d0-9c1e-4473-a344-eb7b665ff687 base
spark-mllib_3.0-scala_2.12 09f4cff0-90a7-5899-b9ed-1ef348aebdee base
pytorch-onnx_rt22.1-py3.9 0b848dd4-e681-5599-be41-b5f6fccc6471 base
ai-function_0.1-py3.6 0cdb0f1e-5376-4f4d-92dd-da3b69aa9bda base
shiny-r3.6 0e6e79df-875e-4f24-8ae9-62dcc2148306 base
tensorflow_2.4-py3.7-horovod 1092590a-307d-563d-9b62-4eb7d64b3f22 base
pytorch_1.1-py3.6 10ac12d6-6b30-4ccd-8392-3e922c096a92 base
tensorflow_1.15-py3.6-ddl 111e41b3-de2d-5422-a4d6-bf776828c4b7 base
autoai-kb_rt22.2-py3.10 125b6d9a-5b1f-5e8d-972a-b251688ccf40 base
runtime-22.1-py3.9 12b83a17-24d8-5082-900f-0ab31fbfd3cb base
scikit-learn_0.22-py3.6 154010fa-5b3b-4ac1-82af-4d5ee5abbc85 base
default_r3.6 1b70aec3-ab34-4b87-8aa0-a4a3c8296a36 base
pytorch-onnx_1.3-py3.6 1bc6029a-cc97-56da-b8e0-39c3880dbbe7 base
kernel-spark3.3-r3.6 1c9e5454-f216-59dd-a20e-474a5cdf5988 base
pytorch-onnx_rt22.1-py3.9-edt 1d362186-7ad5-5b59-8b6c-9d0880bde37f base
tensorflow_2.1-py3.6 1eb25b84-d6ed-5dde-b6a5-3fbdf1665666 base
spark-mllib_3.2 20047f72-0a98-58c7-9ff5-a77b012eb8f5 base
tensorflow_2.4-py3.8-horovod 217c16f6-178f-56bf-824a-b19f20564c49 base
runtime-22.1-py3.9-cuda 26215f05-08c3-5a41-a1b0-da66306ce658 base
do_py3.8 295addb5-9ef9-547e-9bf4-92ae3563e720 base
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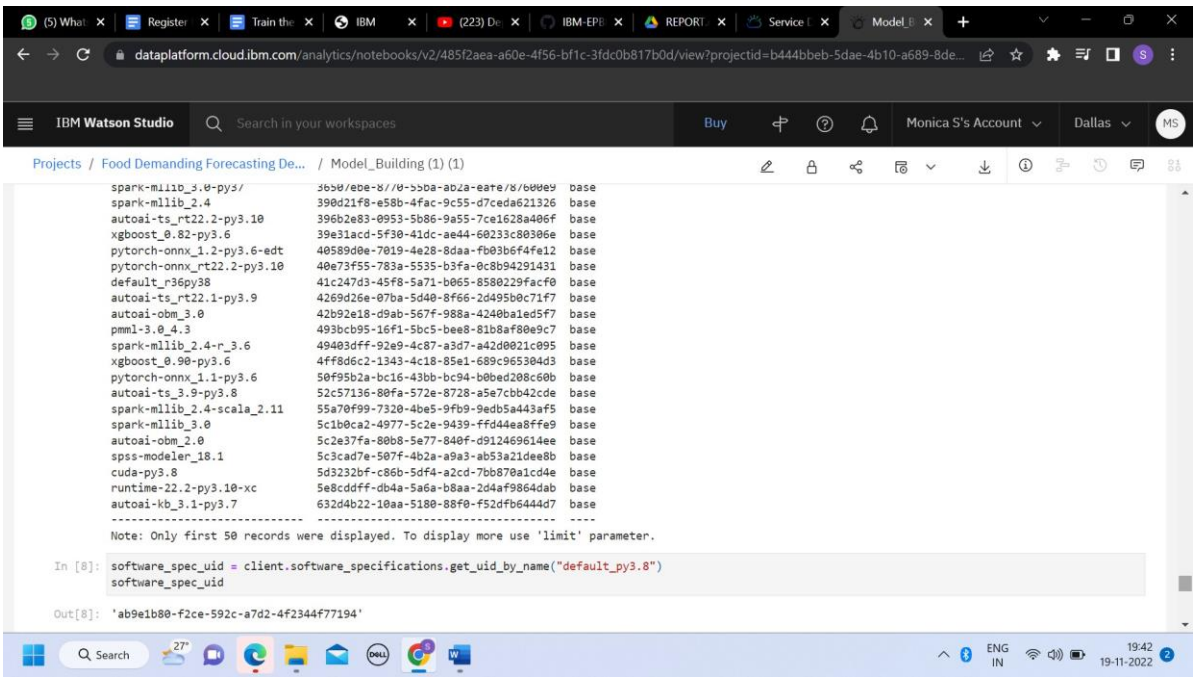
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spark-mllib_3.0-py37 3650/ebe-8770-55ba-ab2a-ea1e/8/b00e9 base
spark-mllib_2.4 390d21f8-e58b-4fac-9c55-d7ceda621326 base
autoai-ts_rt22.2-py3.10 396b2e83-0953-5b86-9a55-7ce1628a406f base
xgboost_0.82-py3.6 39e31acd-5f30-41dc-ae44-60233c80306e base
pytorch-onnx_1.2-py3.6-edt 40589d0e-7019-4e28-8daa-fb03b6f4fe12 base
pytorch-onnx_rt22.2-py3.10 40e73f55-783a-5535-b3fa-0c8b94291431 base
default_r36py38 41c247d3-45f8-5a71-b065-8580229facf0 base
autoai-ts_rt22.1-py3.9 4269d26e-07ba-5d40-8f66-2d495b0c71f7 base
autoai-obm_3.0 42b92e18-d9ab-567f-988a-4240ba1ed5f7 base
pmml-3.0_4.3 493bcb95-16f1-5bc5-bee8-81b8af80e9c7 base
spark-mllib_2.4-r_3.6 49403dff-92e9-4c87-a3d7-a42d0021c095 base
xgboost_0.90-py3.6 4ff8d6c2-1343-4c18-85e1-689c965304d3 base
pytorch-onnx_1.1-py3.6 50f95b2a-bc16-43bb-bc94-b0bed208c60b base
autoai-ts_3.9-py3.8 52c57136-80fa-572e-8728-a5e7cbb42cde base
spark-mllib_2.4-scala_2.11 55a70f99-7320-4be5-9fb9-9edb5a443af5 base
spark-mllib_3.0 5c1b0ca2-4977-5c2e-9439-ffd44ea8ffe9 base
autoai-obm_2.0 5c2e37fa-80b8-5e77-840f-d912469614ee base
spss-modeler_18.1 5c3cad7e-507f-4b2a-a9a3-ab53a21dee8b base
cuda-py3.8 5d3232bf-c86b-5df4-a2cd-7bb870a1cd4e base
runtime-22.2-py3.10-xc 5e8cddff-db4a-5a6a-b8aa-2d4af9864dab base
autoai-kb_3.1-py3.7 632d4b22-10aa-5180-88f0-f52dfb6444d7 base
Note: Only first 50 records were displayed. To display more use 'limit' parameter.
In [8]: software_spec_uid = client.software_specifications.get_uid_by_name("default_py3.8")
software_spec_uid
Out[8]: 'ab9e1b80-f2ce-592c-a7d2-4f2344f77194'
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