

# grocery-sales-forecasting

September 20, 2023

## 1 Machine Learning Techniques for Sales Forecasting

### 1.1 Importing Libraries

```
[ ]: %pip install xgboost
      %pip install statsmodels
      %pip install pandas numpy statsmodels
```

Requirement already satisfied: xgboost in c:\users\srumel\anaconda3\lib\site-packages (2.0.0)

Requirement already satisfied: numpy in

c:\users\srumel\appdata\roaming\python\python311\site-packages (from xgboost) (1.25.1)

Requirement already satisfied: scipy in

c:\users\srumel\appdata\roaming\python\python311\site-packages (from xgboost) (1.11.1)

Note: you may need to restart the kernel to use updated packages.

Requirement already satisfied: statsmodels in c:\users\srumel\anaconda3\lib\site-packages (0.14.0)

Requirement already satisfied: numpy>=1.18 in

c:\users\srumel\appdata\roaming\python\python311\site-packages (from statsmodels) (1.25.1)

Requirement already satisfied: scipy!=1.9.2,>=1.4 in

c:\users\srumel\appdata\roaming\python\python311\site-packages (from statsmodels) (1.11.1)

Requirement already satisfied: pandas>=1.0 in

c:\users\srumel\appdata\roaming\python\python311\site-packages (from statsmodels) (2.0.3)

Requirement already satisfied: patsy>=0.5.2 in

c:\users\srumel\anaconda3\lib\site-packages (from statsmodels) (0.5.3)

Requirement already satisfied: packaging>=21.3 in

c:\users\srumel\appdata\roaming\python\python311\site-packages (from statsmodels) (23.1)

Requirement already satisfied: python-dateutil>=2.8.2 in

c:\users\srumel\appdata\roaming\python\python311\site-packages (from pandas>=1.0->statsmodels) (2.8.2)

Requirement already satisfied: pytz>=2020.1 in

c:\users\srumel\appdata\roaming\python\python311\site-packages (from

pandas>=1.0->statsmodels) (2023.3)  
Requirement already satisfied: tzdata>=2022.1 in  
c:\users\srum\appdata\roaming\python\python311\site-packages (from  
pandas>=1.0->statsmodels) (2023.3)  
Requirement already satisfied: six in  
c:\users\srum\appdata\roaming\python\python311\site-packages (from  
patsy>=0.5.2->statsmodels) (1.16.0)  
Note: you may need to restart the kernel to use updated packages.  
Requirement already satisfied: pandas in  
c:\users\srum\appdata\roaming\python\python311\site-packages (2.0.3)  
Requirement already satisfied: numpy in  
c:\users\srum\appdata\roaming\python\python311\site-packages (1.25.1)  
Requirement already satisfied: statsmodels in c:\users\srum\anaconda3\lib\site-  
packages (0.14.0)  
Requirement already satisfied: python-dateutil>=2.8.2 in  
c:\users\srum\appdata\roaming\python\python311\site-packages (from pandas)  
(2.8.2)  
Requirement already satisfied: pytz>=2020.1 in  
c:\users\srum\appdata\roaming\python\python311\site-packages (from pandas)  
(2023.3)  
Requirement already satisfied: tzdata>=2022.1 in  
c:\users\srum\appdata\roaming\python\python311\site-packages (from pandas)  
(2023.3)  
Requirement already satisfied: scipy!=1.9.2,>=1.4 in  
c:\users\srum\appdata\roaming\python\python311\site-packages (from statsmodels)  
(1.11.1)  
Requirement already satisfied: patsy>=0.5.2 in  
c:\users\srum\anaconda3\lib\site-packages (from statsmodels) (0.5.3)  
Requirement already satisfied: packaging>=21.3 in  
c:\users\srum\appdata\roaming\python\python311\site-packages (from statsmodels)  
(23.1)  
Requirement already satisfied: six in  
c:\users\srum\appdata\roaming\python\python311\site-packages (from  
patsy>=0.5.2->statsmodels) (1.16.0)  
Note: you may need to restart the kernel to use updated packages.

```
[ ]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import statsmodels.api as sm
import scipy.stats as stats
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
from sklearn.linear_model import LogisticRegression
from sklearn.svm import SVC
from sklearn.tree import DecisionTreeClassifier
```

```

from sklearn.ensemble import RandomForestRegressor
from sklearn.linear_model import SGDRegressor
from sklearn.ensemble import ExtraTreesRegressor
from sklearn.linear_model import Ridge
from xgboost import XGBRegressor
from sklearn.linear_model import Lasso
from statsmodels.tsa.arima.model import ARIMA
from sklearn.linear_model import BayesianRidge
from sklearn.neighbors import KNeighborsRegressor
from sklearn.ensemble import AdaBoostRegressor
from sklearn.metrics import mean_squared_error
from sklearn.metrics import mean_absolute_error
from sklearn.metrics import mean_squared_error
import panel as pn
pn.extension()
import hvplot.pandas
from statsmodels.tsa.stattools import adfuller

```

## 1.2 Importing Datasets & Read all csv files

files available at: <https://www.kaggle.com/datasets/ndarshan2797/english-converted-datasets>

1. item\_categories.csv - item\_category\_name, item\_category\_id
2. items.csv - item\_name, item\_id, category\_id
3. sales\_train.csv - date, date\_block\_num, shop\_id, item\_id, item\_price, item\_cnt\_day
4. shops.csv - shop\_name, shop\_id
5. test.csv - ID, shop\_id, item\_id

```

[ ]: #importing data
item_categories = pd.read_csv('./data-set/item_categories.csv')
items = pd.read_csv('./data-set/items.csv')
sales_train = pd.read_csv('./data-set/sales_train.csv')
shops = pd.read_csv('./data-set/shops.csv')
test = pd.read_csv('./data-set/test.csv')

```

```

[ ]: #checking the shape of the data
print("Shape of item_categories:", item_categories.shape)
print("Shape of items:", items.shape)
print("Shape of sales_train:", sales_train.shape)
print("Shape of shops:", shops.shape)
print("Shape of test:", test.shape)

```

```

Shape of item_categories: (84, 2)
Shape of items: (22170, 3)
Shape of sales_train: (2935849, 6)
Shape of shops: (60, 2)
Shape of test: (214200, 3)

```

```
[ ]: #checking the columns of the data
print("\n\nColumns of item_categories:\n")
print(item_categories.info())

print("-----")

print("\n\nColumns of items:\n")
print(items.info())

print("-----")

print("\n\nColumns of sales_train:\n")
print(sales_train.info())

print("-----")

print("\n\nColumns of shops:\n")
print(shops.info())

print("-----")

print("\n\nColumns of test:\n")
print(test.info())
```

Columns of item\_categories:

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 84 entries, 0 to 83
Data columns (total 2 columns):
#   Column                Non-Null Count  Dtype
---  ---
0   item_category_name     84 non-null    object
1   item_category_id       84 non-null    int64
dtypes: int64(1), object(1)
memory usage: 1.4+ KB
None
-----
```

Columns of items:

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 22170 entries, 0 to 22169
Data columns (total 3 columns):
#   Column                Non-Null Count  Dtype
---  ---
-----
```

```

0    item_name      22170 non-null  object
1    item_id        22170 non-null  int64
2    category_id    22170 non-null  int64
dtypes: int64(2), object(1)
memory usage: 519.7+ KB
None

```

---

Columns of sales\_train:

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2935849 entries, 0 to 2935848
Data columns (total 6 columns):
#   Column          Dtype
---  ----
0    date            object
1    date_block_num  int64
2    shop_id         int64
3    item_id         int64
4    item_price      float64
5    item_cnt_day    float64
dtypes: float64(2), int64(3), object(1)
memory usage: 134.4+ MB
None

```

---

Columns of shops:

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 60 entries, 0 to 59
Data columns (total 2 columns):
#   Column          Non-Null Count  Dtype
---  ----
0    shop_name       60 non-null     object
1    shop_id         60 non-null     int64
dtypes: int64(1), object(1)
memory usage: 1.1+ KB
None

```

---

Columns of test:

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 214200 entries, 0 to 214199
Data columns (total 3 columns):

```

#	Column	Non-Null Count	Dtype
0	ID	214200 non-null	int64
1	shop_id	214200 non-null	int64
2	item_id	214200 non-null	int64

dtypes: int64(3)  
memory usage: 4.9 MB  
None

```
[ ]: #checking the head and tail of the data

print("\n\nHead of item_categories:\n")
print(item_categories.head())

print("\n\nTail of item_categories:\n")
print(item_categories.tail())

print("-----")

print("\n\nHead of items:\n")
print(items.head())

print("\n\nTail of items:\n")
print(items.tail())

print("-----")

print("\n\nHead of sales_train:\n")
print(sales_train.head())

print("\n\nTail of sales_train:\n")
print(sales_train.tail())

print("-----")

print("\n\nHead of shops:\n")
print(shops.head())

print("\n\nTail of shops:\n")
print(shops.tail())

print("-----")

print("\n\nHead of test:\n")
print(test.head())

print("\n\nTail of test:\n")
```

```
print(test.tail())
```

Head of item\_categories:

	item_category_name	item_category_id
0	PC - Headsets / Headphones	0
1	Accessories - PS2	1
2	Accessories - PS3	2
3	Accessories - PS4	3
4	Accessories - PSP	4

Tail of item\_categories:

	item_category_name	item_category_id
79	Service	79
80	Service - Tickets	80
81	Blank media (spire)	81
82	Blank media (piece)	82
83	Batteries	83

-----

Head of items:

	item_name	item_id	category_id
0	!! IN THE POWER OF HAPPINESS (PLAST) D	0	40
1	! ABBYY FineReader 12 Professional Edition Ful...	1	76
2	*** IN THE GLORY OF THE GLORY (UNV) D	2	40
3	*** BLUE WAVE (Univ) D	3	40
4	*** BOX (GLASS) D	4	40

Tail of items:

	item_name	item_id	category_id
22165	Nuclear Titbit 2 [PC, Digital Version]	22165	31
22166	Query language 1C: Enterprise [Digital version]	22166	54
22167	The query language is 1C: Enterprise 8 (+ CD)...	22167	49
22168	Egg for Little Inu	22168	62
22169	Egg of the Dragon (Game of Thrones)	22169	69

-----

Head of sales\_train:

	date	date_block_num	shop_id	item_id	item_price	item_cnt_day
0	02.01.2013	0	59	22154	999.00	1.0
1	03.01.2013	0	25	2552	899.00	1.0
2	05.01.2013	0	25	2552	899.00	-1.0
3	06.01.2013	0	25	2554	1709.05	1.0
4	15.01.2013	0	25	2555	1099.00	1.0

Tail of sales\_train:

	date	date_block_num	shop_id	item_id	item_price	\
2935844	10.10.2015	33	25	7409	299.0	
2935845	09.10.2015	33	25	7460	299.0	
2935846	14.10.2015	33	25	7459	349.0	
2935847	22.10.2015	33	25	7440	299.0	
2935848	03.10.2015	33	25	7460	299.0	

	item_cnt_day
2935844	1.0
2935845	1.0
2935846	1.0
2935847	1.0
2935848	1.0

Head of shops:

	shop_name	shop_id
0	! Yakutsk Ordzhonikidze, 56 francs	0
1	! Yakutsk TC "Central" fran	1
2	Adygea TC "Mega"	2
3	Balashikha TC "Oktyabr-Kinomir"	3
4	Volga TC "Volga Mall"	4

Tail of shops:

	shop_name	shop_id
55	Digital warehouse 1C-Online	55
56	Chekhov TC" Karnaval "	56
57	Yakutsk Ordzhonikidze, 56	57
58	Yakutsk TC" Central "	58
59	Yaroslavl TC" Altair "	59

Head of test:



	ID	shop_id	item_id
0	0	5	5037
1	1	5	5320
2	2	5	5233
3	3	5	5232
4	4	5	5268

Tail of test:

	ID	shop_id	item_id
214195	214195	45	18454
214196	214196	45	16188
214197	214197	45	15757
214198	214198	45	19648
214199	214199	45	969

### 1.3 Data Preprocessing & Feature Engineering

```
[ ]: #merging the data for better understand the data
```

```
[ ]: #Merge sales_train.csv with items.csv on the "item_id" column
sales_with_items = sales_train.merge(items, on='item_id', how='left')
print("\n\nHead of sales_with_items:\n")
print(sales_with_items.head(20))
print(sales_with_items.shape)
```

Head of sales\_with\_items:

	date	date_block_num	shop_id	item_id	item_price	item_cnt_day	\
0	02.01.2013	0	59	22154	999.00	1.0	
1	03.01.2013	0	25	2552	899.00	1.0	
2	05.01.2013	0	25	2552	899.00	-1.0	
3	06.01.2013	0	25	2554	1709.05	1.0	
4	15.01.2013	0	25	2555	1099.00	1.0	
5	10.01.2013	0	25	2564	349.00	1.0	
6	02.01.2013	0	25	2565	549.00	1.0	
7	04.01.2013	0	25	2572	239.00	1.0	
8	11.01.2013	0	25	2572	299.00	1.0	
9	03.01.2013	0	25	2573	299.00	3.0	
10	03.01.2013	0	25	2574	399.00	2.0	
11	05.01.2013	0	25	2574	399.00	1.0	
12	07.01.2013	0	25	2574	399.00	1.0	
13	08.01.2013	0	25	2574	399.00	2.0	
14	10.01.2013	0	25	2574	399.00	1.0	

15	11.01.2013	0	25	2574	399.00	2.0
16	13.01.2013	0	25	2574	399.00	1.0
17	16.01.2013	0	25	2574	399.00	1.0
18	26.01.2013	0	25	2574	399.00	1.0
19	27.01.2013	0	25	2574	399.00	1.0

	item_name	category_id
0	SCENE 2012 (BD)	37
1	DEEP PURPLE The House Of Blue Light LP	58
2	DEEP PURPLE The House Of Blue Light LP	58
3	DEEP PURPLE Who Do You Think We Are LP	58
4	DEEP PURPLE 30 Very Best Of 2CD (Businesses).	56
5	DEEP PURPLE Perihelion: Live In Concert DVD (C...	59
6	DEEP PURPLE Stormbringer (firms).	56
7	DEFTONES Koi No Yokan	55
8	DEFTONES Koi No Yokan	55
9	DEL REY LANA Born To Die	55
10	DEL REY LANA Born To Die The Paradise Editio...	55
11	DEL REY LANA Born To Die The Paradise Editio...	55
12	DEL REY LANA Born To Die The Paradise Editio...	55
13	DEL REY LANA Born To Die The Paradise Editio...	55
14	DEL REY LANA Born To Die The Paradise Editio...	55
15	DEL REY LANA Born To Die The Paradise Editio...	55
16	DEL REY LANA Born To Die The Paradise Editio...	55
17	DEL REY LANA Born To Die The Paradise Editio...	55
18	DEL REY LANA Born To Die The Paradise Editio...	55
19	DEL REY LANA Born To Die The Paradise Editio...	55

(2935849, 8)

```
[ ]: #Merge the result with item_categories.csv on the "category_id"
sales_with_items_and_categories = sales_with_items.merge(item_categories,
    right_on='item_category_id', left_on='category_id', how='left')
print("\n\nHead of sales_with_items_and_categories:\n")
print(sales_with_items_and_categories.head(20))
print(sales_with_items_and_categories.shape)
```

Head of sales\_with\_items\_and\_categories:

	date	date_block_num	shop_id	item_id	item_price	item_cnt_day	\
0	02.01.2013	0	59	22154	999.00	1.0	
1	03.01.2013	0	25	2552	899.00	1.0	
2	05.01.2013	0	25	2552	899.00	-1.0	
3	06.01.2013	0	25	2554	1709.05	1.0	
4	15.01.2013	0	25	2555	1099.00	1.0	
5	10.01.2013	0	25	2564	349.00	1.0	
6	02.01.2013	0	25	2565	549.00	1.0	

7	04.01.2013	0	25	2572	239.00	1.0
8	11.01.2013	0	25	2572	299.00	1.0
9	03.01.2013	0	25	2573	299.00	3.0
10	03.01.2013	0	25	2574	399.00	2.0
11	05.01.2013	0	25	2574	399.00	1.0
12	07.01.2013	0	25	2574	399.00	1.0
13	08.01.2013	0	25	2574	399.00	2.0
14	10.01.2013	0	25	2574	399.00	1.0
15	11.01.2013	0	25	2574	399.00	2.0
16	13.01.2013	0	25	2574	399.00	1.0
17	16.01.2013	0	25	2574	399.00	1.0
18	26.01.2013	0	25	2574	399.00	1.0
19	27.01.2013	0	25	2574	399.00	1.0

	item_name	category_id	\
0	SCENE 2012 (BD)	37	
1	DEEP PURPLE The House Of Blue Light LP	58	
2	DEEP PURPLE The House Of Blue Light LP	58	
3	DEEP PURPLE Who Do You Think We Are LP	58	
4	DEEP PURPLE 30 Very Best Of 2CD (Businesses).	56	
5	DEEP PURPLE Perihelion: Live In Concert DVD (C...	59	
6	DEEP PURPLE Stormbringer (firms).	56	
7	DEFTONES Koi No Yokan	55	
8	DEFTONES Koi No Yokan	55	
9	DEL REY LANA Born To Die	55	
10	DEL REY LANA Born To Die The Paradise Editio...	55	
11	DEL REY LANA Born To Die The Paradise Editio...	55	
12	DEL REY LANA Born To Die The Paradise Editio...	55	
13	DEL REY LANA Born To Die The Paradise Editio...	55	
14	DEL REY LANA Born To Die The Paradise Editio...	55	
15	DEL REY LANA Born To Die The Paradise Editio...	55	
16	DEL REY LANA Born To Die The Paradise Editio...	55	
17	DEL REY LANA Born To Die The Paradise Editio...	55	
18	DEL REY LANA Born To Die The Paradise Editio...	55	
19	DEL REY LANA Born To Die The Paradise Editio...	55	

	item_category_name	item_category_id
0	Cinema - Blu-Ray	37
1	Music - Vinyl	58
2	Music - Vinyl	58
3	Music - Vinyl	58
4	Music - CD of corporate production	56
5	Music - Music video	59
6	Music - CD of corporate production	56
7	Music - CD of local production	55
8	Music - CD of local production	55
9	Music - CD of local production	55
10	Music - CD of local production	55

```

11      Music - CD of local production      55
12      Music - CD of local production      55
13      Music - CD of local production      55
14      Music - CD of local production      55
15      Music - CD of local production      55
16      Music - CD of local production      55
17      Music - CD of local production      55
18      Music - CD of local production      55
19      Music - CD of local production      55
(2935849, 10)

```

```

[ ]: # Check if the two columns are the same
if sales_with_items_and_categories['item_category_id'].
    equals(sales_with_items_and_categories['category_id']):
    # If they are the same, drop one of the columns
    sales_with_items_and_categories.drop(columns=['item_category_id'],
    inplace=True)

```

```

[ ]: print("\n\nHead of sales_with_items_and_categories:\n")
print(sales_with_items_and_categories.head(20))
print(sales_with_items_and_categories.shape)

```

Head of sales\_with\_items\_and\_categories:

	date	date_block_num	shop_id	item_id	item_price	item_cnt_day	\
0	02.01.2013	0	59	22154	999.00	1.0	
1	03.01.2013	0	25	2552	899.00	1.0	
2	05.01.2013	0	25	2552	899.00	-1.0	
3	06.01.2013	0	25	2554	1709.05	1.0	
4	15.01.2013	0	25	2555	1099.00	1.0	
5	10.01.2013	0	25	2564	349.00	1.0	
6	02.01.2013	0	25	2565	549.00	1.0	
7	04.01.2013	0	25	2572	239.00	1.0	
8	11.01.2013	0	25	2572	299.00	1.0	
9	03.01.2013	0	25	2573	299.00	3.0	
10	03.01.2013	0	25	2574	399.00	2.0	
11	05.01.2013	0	25	2574	399.00	1.0	
12	07.01.2013	0	25	2574	399.00	1.0	
13	08.01.2013	0	25	2574	399.00	2.0	
14	10.01.2013	0	25	2574	399.00	1.0	
15	11.01.2013	0	25	2574	399.00	2.0	
16	13.01.2013	0	25	2574	399.00	1.0	
17	16.01.2013	0	25	2574	399.00	1.0	
18	26.01.2013	0	25	2574	399.00	1.0	
19	27.01.2013	0	25	2574	399.00	1.0	

	item_name	category_id	\
0	SCENE 2012 (BD)	37	
1	DEEP PURPLE The House Of Blue Light LP	58	
2	DEEP PURPLE The House Of Blue Light LP	58	
3	DEEP PURPLE Who Do You Think We Are LP	58	
4	DEEP PURPLE 30 Very Best Of 2CD (Businesses).	56	
5	DEEP PURPLE Perihelion: Live In Concert DVD (C...	59	
6	DEEP PURPLE Stormbringer (firms).	56	
7	DEFTONES Koi No Yokan	55	
8	DEFTONES Koi No Yokan	55	
9	DEL REY LANA Born To Die	55	
10	DEL REY LANA Born To Die The Paradise Editio...	55	
11	DEL REY LANA Born To Die The Paradise Editio...	55	
12	DEL REY LANA Born To Die The Paradise Editio...	55	
13	DEL REY LANA Born To Die The Paradise Editio...	55	
14	DEL REY LANA Born To Die The Paradise Editio...	55	
15	DEL REY LANA Born To Die The Paradise Editio...	55	
16	DEL REY LANA Born To Die The Paradise Editio...	55	
17	DEL REY LANA Born To Die The Paradise Editio...	55	
18	DEL REY LANA Born To Die The Paradise Editio...	55	
19	DEL REY LANA Born To Die The Paradise Editio...	55	

	item_category_name
0	Cinema - Blu-Ray
1	Music - Vinyl
2	Music - Vinyl
3	Music - Vinyl
4	Music - CD of corporate production
5	Music - Music video
6	Music - CD of corporate production
7	Music - CD of local production
8	Music - CD of local production
9	Music - CD of local production
10	Music - CD of local production
11	Music - CD of local production
12	Music - CD of local production
13	Music - CD of local production
14	Music - CD of local production
15	Music - CD of local production
16	Music - CD of local production
17	Music - CD of local production
18	Music - CD of local production
19	Music - CD of local production

(2935849, 9)

```
[ ]: #Merge the result with shops.csv on the "shop_id"
```

```

final_dataset = sales_with_items_and_categories.merge(shops, on='shop_id',
↳how='left')
print("\n\nHead of final_dataset:\n")
print(final_dataset.head(20))
print(final_dataset.shape)

```

Head of final\_dataset:

	date	date_block_num	shop_id	item_id	item_price	item_cnt_day	\
0	02.01.2013	0	59	22154	999.00	1.0	
1	03.01.2013	0	25	2552	899.00	1.0	
2	05.01.2013	0	25	2552	899.00	-1.0	
3	06.01.2013	0	25	2554	1709.05	1.0	
4	15.01.2013	0	25	2555	1099.00	1.0	
5	10.01.2013	0	25	2564	349.00	1.0	
6	02.01.2013	0	25	2565	549.00	1.0	
7	04.01.2013	0	25	2572	239.00	1.0	
8	11.01.2013	0	25	2572	299.00	1.0	
9	03.01.2013	0	25	2573	299.00	3.0	
10	03.01.2013	0	25	2574	399.00	2.0	
11	05.01.2013	0	25	2574	399.00	1.0	
12	07.01.2013	0	25	2574	399.00	1.0	
13	08.01.2013	0	25	2574	399.00	2.0	
14	10.01.2013	0	25	2574	399.00	1.0	
15	11.01.2013	0	25	2574	399.00	2.0	
16	13.01.2013	0	25	2574	399.00	1.0	
17	16.01.2013	0	25	2574	399.00	1.0	
18	26.01.2013	0	25	2574	399.00	1.0	
19	27.01.2013	0	25	2574	399.00	1.0	

	item_name	category_id	\
0	SCENE 2012 (BD)	37	
1	DEEP PURPLE The House Of Blue Light LP	58	
2	DEEP PURPLE The House Of Blue Light LP	58	
3	DEEP PURPLE Who Do You Think We Are LP	58	
4	DEEP PURPLE 30 Very Best Of 2CD (Businesses).	56	
5	DEEP PURPLE Perihelion: Live In Concert DVD (C...	59	
6	DEEP PURPLE Stormbringer (firms).	56	
7	DEFTONES Koi No Yokan	55	
8	DEFTONES Koi No Yokan	55	
9	DEL REY LANA Born To Die	55	
10	DEL REY LANA Born To Die The Paradise Editio...	55	
11	DEL REY LANA Born To Die The Paradise Editio...	55	
12	DEL REY LANA Born To Die The Paradise Editio...	55	
13	DEL REY LANA Born To Die The Paradise Editio...	55	
14	DEL REY LANA Born To Die The Paradise Editio...	55	

15	DEL REY LANA	Born To Die	The Paradise Editio...	55
16	DEL REY LANA	Born To Die	The Paradise Editio...	55
17	DEL REY LANA	Born To Die	The Paradise Editio...	55
18	DEL REY LANA	Born To Die	The Paradise Editio...	55
19	DEL REY LANA	Born To Die	The Paradise Editio...	55

	item_category_name	shop_name
0	Cinema - Blu-Ray	Yaroslavl TC" Altair "
1	Music - Vinyl	Moscow TEC" Atrium "
2	Music - Vinyl	Moscow TEC" Atrium "
3	Music - Vinyl	Moscow TEC" Atrium "
4	Music - CD of corporate production	Moscow TEC" Atrium "
5	Music - Music video	Moscow TEC" Atrium "
6	Music - CD of corporate production	Moscow TEC" Atrium "
7	Music - CD of local production	Moscow TEC" Atrium "
8	Music - CD of local production	Moscow TEC" Atrium "
9	Music - CD of local production	Moscow TEC" Atrium "
10	Music - CD of local production	Moscow TEC" Atrium "
11	Music - CD of local production	Moscow TEC" Atrium "
12	Music - CD of local production	Moscow TEC" Atrium "
13	Music - CD of local production	Moscow TEC" Atrium "
14	Music - CD of local production	Moscow TEC" Atrium "
15	Music - CD of local production	Moscow TEC" Atrium "
16	Music - CD of local production	Moscow TEC" Atrium "
17	Music - CD of local production	Moscow TEC" Atrium "
18	Music - CD of local production	Moscow TEC" Atrium "
19	Music - CD of local production	Moscow TEC" Atrium "

(2935849, 10)

```
[ ]: #checks the columns of the final dataset
print("\n\nColumns of final_dataset:\n")
print(final_dataset.info())
```

Columns of final\_dataset:

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2935849 entries, 0 to 2935848
Data columns (total 10 columns):
#   Column              Dtype
---  -
0   date                 object
1   date_block_num       int64
2   shop_id              int64
3   item_id              int64
4   item_price           float64
5   item_cnt_day         float64
```

```

6   item_name          object
7   category_id        int64
8   item_category_name object
9   shop_name          object
dtypes: float64(2), int64(4), object(4)
memory usage: 224.0+ MB
None

```

```

[ ]: #prints the date and date_block_num column to check whether they are related
columns_to_print = ['date', 'date_block_num']
print(final_dataset[columns_to_print])

```

```

      date  date_block_num
0    02.01.2013           0
1    03.01.2013           0
2    05.01.2013           0
3    06.01.2013           0
4    15.01.2013           0
...
2935844  10.10.2015        33
2935845  09.10.2015        33
2935846  14.10.2015        33
2935847  22.10.2015        33
2935848  03.10.2015        33

```

[2935849 rows x 2 columns]

```

[ ]: # Rename the column
final_dataset.rename(columns={'date_block_num': 'month_num'}, inplace=True)

```

```

[ ]: #Rename the column
final_dataset.rename(columns={'item_cnt_day': 'item_cnt_month'}, inplace=True)

```

```

[ ]: print("\n\nHead of final_dataset:\n")
print(final_dataset.head(20))
print(final_dataset.shape)

```

Head of final\_dataset:

```

      date  month_num  shop_id  item_id  item_price  item_cnt_month  \
0  02.01.2013         0      59   22154      999.00           1.0
1  03.01.2013         0      25   2552      899.00           1.0
2  05.01.2013         0      25   2552      899.00          -1.0
3  06.01.2013         0      25   2554     1709.05           1.0
4  15.01.2013         0      25   2555     1099.00           1.0
5  10.01.2013         0      25   2564      349.00           1.0
6  02.01.2013         0      25   2565      549.00           1.0

```



7	04.01.2013	0	25	2572	239.00	1.0
8	11.01.2013	0	25	2572	299.00	1.0
9	03.01.2013	0	25	2573	299.00	3.0
10	03.01.2013	0	25	2574	399.00	2.0
11	05.01.2013	0	25	2574	399.00	1.0
12	07.01.2013	0	25	2574	399.00	1.0
13	08.01.2013	0	25	2574	399.00	2.0
14	10.01.2013	0	25	2574	399.00	1.0
15	11.01.2013	0	25	2574	399.00	2.0
16	13.01.2013	0	25	2574	399.00	1.0
17	16.01.2013	0	25	2574	399.00	1.0
18	26.01.2013	0	25	2574	399.00	1.0
19	27.01.2013	0	25	2574	399.00	1.0

	item_name	category_id	\
0	SCENE 2012 (BD)	37	
1	DEEP PURPLE The House Of Blue Light LP	58	
2	DEEP PURPLE The House Of Blue Light LP	58	
3	DEEP PURPLE Who Do You Think We Are LP	58	
4	DEEP PURPLE 30 Very Best Of 2CD (Businesses).	56	
5	DEEP PURPLE Perihelion: Live In Concert DVD (C...	59	
6	DEEP PURPLE Stormbringer (firms).	56	
7	DEFTONES Koi No Yokan	55	
8	DEFTONES Koi No Yokan	55	
9	DEL REY LANA Born To Die	55	
10	DEL REY LANA Born To Die The Paradise Editio...	55	
11	DEL REY LANA Born To Die The Paradise Editio...	55	
12	DEL REY LANA Born To Die The Paradise Editio...	55	
13	DEL REY LANA Born To Die The Paradise Editio...	55	
14	DEL REY LANA Born To Die The Paradise Editio...	55	
15	DEL REY LANA Born To Die The Paradise Editio...	55	
16	DEL REY LANA Born To Die The Paradise Editio...	55	
17	DEL REY LANA Born To Die The Paradise Editio...	55	
18	DEL REY LANA Born To Die The Paradise Editio...	55	
19	DEL REY LANA Born To Die The Paradise Editio...	55	

	item_category_name	shop_name
0	Cinema - Blu-Ray	Yaroslavl TC" Altair "
1	Music - Vinyl	Moscow TEC" Atrium "
2	Music - Vinyl	Moscow TEC" Atrium "
3	Music - Vinyl	Moscow TEC" Atrium "
4	Music - CD of corporate production	Moscow TEC" Atrium "
5	Music - Music video	Moscow TEC" Atrium "
6	Music - CD of corporate production	Moscow TEC" Atrium "
7	Music - CD of local production	Moscow TEC" Atrium "
8	Music - CD of local production	Moscow TEC" Atrium "
9	Music - CD of local production	Moscow TEC" Atrium "
10	Music - CD of local production	Moscow TEC" Atrium "

```

11      Music - CD of local production      Moscow TEC" Atrium "
12      Music - CD of local production      Moscow TEC" Atrium "
13      Music - CD of local production      Moscow TEC" Atrium "
14      Music - CD of local production      Moscow TEC" Atrium "
15      Music - CD of local production      Moscow TEC" Atrium "
16      Music - CD of local production      Moscow TEC" Atrium "
17      Music - CD of local production      Moscow TEC" Atrium "
18      Music - CD of local production      Moscow TEC" Atrium "
19      Music - CD of local production      Moscow TEC" Atrium "
(2935849, 10)

```

```

[ ]: #checks the columns of the final dataset
print("\n\nColumns of final_dataset:\n")
print(final_dataset.info())

```

Columns of final\_dataset:

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2935849 entries, 0 to 2935848
Data columns (total 10 columns):
#   Column                Dtype
---  -
0   date                  object
1   month_num             int64
2   shop_id               int64
3   item_id               int64
4   item_price            float64
5   item_cnt_month        float64
6   item_name             object
7   category_id           int64
8   item_category_name    object
9   shop_name             object
dtypes: float64(2), int64(4), object(4)
memory usage: 224.0+ MB
None

```

```

[ ]: #export the final dataset to csv file
final_dataset.to_csv('./data-set/output/final_dataset_without_cleaning.csv',
    ↪index=False)

```

```

[ ]: #Data Cleaning

#checking for missing values
print("\n\nMissing values in final_dataset:\n")
print(final_dataset.isnull().sum())

```

Missing values in final\_dataset:

```
date            0
month_num       0
shop_id         0
item_id         0
item_price      0
item_cnt_month  0
item_name       0
category_id     0
item_category_name 0
shop_name       0
dtype: int64
```

```
[ ]: #checking for null values
print("\n\nNull values in final_dataset:\n")
print(final_dataset.isnull().sum())
```

Null values in final\_dataset:

```
date            0
month_num       0
shop_id         0
item_id         0
item_price      0
item_cnt_month  0
item_name       0
category_id     0
item_category_name 0
shop_name       0
dtype: int64
```

```
[ ]: print(final_dataset.shape)
```

```
(2935849, 10)
```

```
[ ]: #handles the missing values in final_dataset
final_dataset['item_name'].fillna('Unknown', inplace=True)
final_dataset['item_category_name'].fillna('Unknown', inplace=True)
```

```
[ ]: print(final_dataset.shape)
```

```
(2935849, 10)
```

```
[ ]: #removes duplicates rows in final_dataset
final_dataset.drop_duplicates(inplace=True)
```

```
[ ]: print(final_dataset.shape)
```

```
(2935843, 10)
```

```
[ ]: #checks and solves the data type of the columns
print("\n\nData types of final_dataset:\n")
print(final_dataset.dtypes)
```

Data types of final\_dataset:

```
date                object
month_num           int64
shop_id            int64
item_id            int64
item_price          float64
item_cnt_month      float64
item_name           object
category_id         int64
item_category_name  object
shop_name           object
dtype: object
```

```
[ ]: # #seems like item_cnt_month should be int64
final_dataset['item_cnt_month'] = final_dataset['item_cnt_month'].
    .astype('int64')
```

```
[ ]: print(final_dataset.dtypes)
```

```
date                object
month_num           int64
shop_id            int64
item_id            int64
item_price          float64
item_cnt_month      int64
item_name           object
category_id         int64
item_category_name  object
shop_name           object
dtype: object
```

```
[ ]: #prints item_cnt_month column to check whether it is int64
print(final_dataset['item_cnt_month'].head(30))
```

```
0    1
1    1
2   -1
3    1
```

```

4      1
5      1
6      1
7      1
8      1
9      3
10     2
11     1
12     1
13     2
14     1
15     2
16     1
17     1
18     1
19     1
20     1
21     1
22     1
23     1
24     1
25     1
26     1
27     1
28     1
29     1
Name: item_cnt_month, dtype: int64

```

```
[ ]: print(final_dataset.shape)
```

```
(2935843, 10)
```

```
[ ]: #removes -1 and 307980 from item_cnt_month column because it is an outlier
#it is not possible to sell -1 and 307980 items in a day because 307980 is the
↳total number of items sold in a day
#which means that the data is incorrect
#and -1 is not possible

final_dataset = final_dataset[(final_dataset['item_cnt_month'] > 0) &
    ↳(final_dataset['item_cnt_month'] < 307980)]

print(final_dataset.shape)
```

```
(2928487, 10)
```

```
[ ]: #outlier treatment

#checks for outliers in the item_cnt_month column
```

```

print("\n\nOutliers in item_cnt_month column:\n")
print(final_dataset[final_dataset['item_cnt_month'] > 1000])

#removes the outliers in the item_cnt_month column
final_dataset = final_dataset[final_dataset['item_cnt_month'] < 1000]

print("\n\nHead of final_dataset:\n")
print(final_dataset.head(20))
print(final_dataset.shape)

```

Outliers in item\_cnt\_month column:

	date	month_num	shop_id	item_id	item_price	item_cnt_month	\
2909818	28.10.2015	33	12	11373	0.908714	2169	

	item_name	category_id	\
2909818	Delivery to the point of issue (Boxberry)	9	

	item_category_name	shop_name
2909818	Delivery of goods	Internet-shop of emergency situations

Head of final\_dataset:

	date	month_num	shop_id	item_id	item_price	item_cnt_month	\
0	02.01.2013	0	59	22154	999.00	1	
1	03.01.2013	0	25	2552	899.00	1	
3	06.01.2013	0	25	2554	1709.05	1	
4	15.01.2013	0	25	2555	1099.00	1	
5	10.01.2013	0	25	2564	349.00	1	
6	02.01.2013	0	25	2565	549.00	1	
7	04.01.2013	0	25	2572	239.00	1	
8	11.01.2013	0	25	2572	299.00	1	
9	03.01.2013	0	25	2573	299.00	3	
10	03.01.2013	0	25	2574	399.00	2	
11	05.01.2013	0	25	2574	399.00	1	
12	07.01.2013	0	25	2574	399.00	1	
13	08.01.2013	0	25	2574	399.00	2	
14	10.01.2013	0	25	2574	399.00	1	
15	11.01.2013	0	25	2574	399.00	2	
16	13.01.2013	0	25	2574	399.00	1	
17	16.01.2013	0	25	2574	399.00	1	
18	26.01.2013	0	25	2574	399.00	1	
19	27.01.2013	0	25	2574	399.00	1	
20	09.01.2013	0	25	2593	279.00	1	

	item_name	category_id	\
0	SCENE 2012 (BD)	37	
1	DEEP PURPLE The House Of Blue Light LP	58	
3	DEEP PURPLE Who Do You Think We Are LP	58	
4	DEEP PURPLE 30 Very Best Of 2CD (Businesses).	56	
5	DEEP PURPLE Perihelion: Live In Concert DVD (C...	59	
6	DEEP PURPLE Stormbringer (firms).	56	
7	DEFTONES Koi No Yokan	55	
8	DEFTONES Koi No Yokan	55	
9	DEL REY LANA Born To Die	55	
10	DEL REY LANA Born To Die The Paradise Editio...	55	
11	DEL REY LANA Born To Die The Paradise Editio...	55	
12	DEL REY LANA Born To Die The Paradise Editio...	55	
13	DEL REY LANA Born To Die The Paradise Editio...	55	
14	DEL REY LANA Born To Die The Paradise Editio...	55	
15	DEL REY LANA Born To Die The Paradise Editio...	55	
16	DEL REY LANA Born To Die The Paradise Editio...	55	
17	DEL REY LANA Born To Die The Paradise Editio...	55	
18	DEL REY LANA Born To Die The Paradise Editio...	55	
19	DEL REY LANA Born To Die The Paradise Editio...	55	
20	DEPECHE MODE Music For The Masses	55	

	item_category_name	shop_name
0	Cinema - Blu-Ray	Yaroslavl TC" Altair "
1	Music - Vinyl	Moscow TEC" Atrium "
3	Music - Vinyl	Moscow TEC" Atrium "
4	Music - CD of corporate production	Moscow TEC" Atrium "
5	Music - Music video	Moscow TEC" Atrium "
6	Music - CD of corporate production	Moscow TEC" Atrium "
7	Music - CD of local production	Moscow TEC" Atrium "
8	Music - CD of local production	Moscow TEC" Atrium "
9	Music - CD of local production	Moscow TEC" Atrium "
10	Music - CD of local production	Moscow TEC" Atrium "
11	Music - CD of local production	Moscow TEC" Atrium "
12	Music - CD of local production	Moscow TEC" Atrium "
13	Music - CD of local production	Moscow TEC" Atrium "
14	Music - CD of local production	Moscow TEC" Atrium "
15	Music - CD of local production	Moscow TEC" Atrium "
16	Music - CD of local production	Moscow TEC" Atrium "
17	Music - CD of local production	Moscow TEC" Atrium "
18	Music - CD of local production	Moscow TEC" Atrium "
19	Music - CD of local production	Moscow TEC" Atrium "
20	Music - CD of local production	Moscow TEC" Atrium "

(2928485, 10)

```
[ ]: #deal with the incorrect data in the item_price column
      #the item_price should not be negative
```

```

#the item_price should not be zero
#the item_price should not be greater than 100000

final_dataset = final_dataset[(final_dataset['item_price'] > 0) &
    ↪(final_dataset['item_price'] < 100000)]

```

```
[ ]: print(final_dataset.shape)
```

```
(2928483, 10)
```

```
[ ]: #handles special characters and formatting in the data set
final_dataset['item_name'] = final_dataset['item_name'].str.
    ↪replace('[^A-Za-z0-9 - - ]+', ' ')

```

```
[ ]: print(final_dataset.shape)
```

```
(2928483, 10)
```

```
[ ]: #removes the noise in the item_name column
final_dataset['item_name'] = final_dataset['item_name'].str.replace(' ', ' ')

```

```
[ ]: print(final_dataset.head())
```

	date	month_num	shop_id	item_id	item_price	item_cnt_month \
0	02.01.2013	0	59	22154	999.00	1
1	03.01.2013	0	25	2552	899.00	1
3	06.01.2013	0	25	2554	1709.05	1
4	15.01.2013	0	25	2555	1099.00	1
5	10.01.2013	0	25	2564	349.00	1

	item_name	category_id \
0	SCENE 2012 (BD)	37
1	DEEP PURPLE The House Of Blue Light LP	58
3	DEEP PURPLE Who Do You Think We Are LP	58
4	DEEP PURPLE 30 Very Best Of 2CD (Businesses).	56
5	DEEP PURPLE Perihelion: Live In Concert DVD (C...	59

	item_category_name	shop_name
0	Cinema - Blu-Ray	Yaroslavl TC" Altair "
1	Music - Vinyl	Moscow TEC" Atrium "
3	Music - Vinyl	Moscow TEC" Atrium "
4	Music - CD of corporate production	Moscow TEC" Atrium "
5	Music - Music video	Moscow TEC" Atrium "

```
[ ]: #creates a new column called revenue
final_dataset['revenue'] = final_dataset['item_cnt_month'] *
    ↪final_dataset['item_price']

```



```
[ ]: print("\n\nHead of final_dataset:\n")
      print(final_dataset.head(20))
      print(final_dataset.shape)
```

Head of final\_dataset:

	date	month_num	shop_id	item_id	item_price	item_cnt_month	\
0	02.01.2013	0	59	22154	999.00	1	
1	03.01.2013	0	25	2552	899.00	1	
3	06.01.2013	0	25	2554	1709.05	1	
4	15.01.2013	0	25	2555	1099.00	1	
5	10.01.2013	0	25	2564	349.00	1	
6	02.01.2013	0	25	2565	549.00	1	
7	04.01.2013	0	25	2572	239.00	1	
8	11.01.2013	0	25	2572	299.00	1	
9	03.01.2013	0	25	2573	299.00	3	
10	03.01.2013	0	25	2574	399.00	2	
11	05.01.2013	0	25	2574	399.00	1	
12	07.01.2013	0	25	2574	399.00	1	
13	08.01.2013	0	25	2574	399.00	2	
14	10.01.2013	0	25	2574	399.00	1	
15	11.01.2013	0	25	2574	399.00	2	
16	13.01.2013	0	25	2574	399.00	1	
17	16.01.2013	0	25	2574	399.00	1	
18	26.01.2013	0	25	2574	399.00	1	
19	27.01.2013	0	25	2574	399.00	1	
20	09.01.2013	0	25	2593	279.00	1	

	item_name	category_id	\
0	SCENE 2012 (BD)	37	
1	DEEP PURPLE The House Of Blue Light LP	58	
3	DEEP PURPLE Who Do You Think We Are LP	58	
4	DEEP PURPLE 30 Very Best Of 2CD (Businesses).	56	
5	DEEP PURPLE Perihelion: Live In Concert DVD (C...	59	
6	DEEP PURPLE Stormbringer (firms).	56	
7	DEFTONES Koi No Yokan	55	
8	DEFTONES Koi No Yokan	55	
9	DEL REY LANA Born To Die	55	
10	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
11	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
12	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
13	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
14	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
15	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
16	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
17	DEL REY LANA Born To Die The Paradise Edition 2CD	55	

18	DEL REY LANA Born To Die The Paradise Edition 2CD	55
19	DEL REY LANA Born To Die The Paradise Edition 2CD	55
20	DEPECHE MODE Music For The Masses	55

	item_category_name	shop_name	revenue
0	Cinema - Blu-Ray	Yaroslavl TC" Altair "	999.00
1	Music - Vinyl	Moscow TEC" Atrium "	899.00
3	Music - Vinyl	Moscow TEC" Atrium "	1709.05
4	Music - CD of corporate production	Moscow TEC" Atrium "	1099.00
5	Music - Music video	Moscow TEC" Atrium "	349.00
6	Music - CD of corporate production	Moscow TEC" Atrium "	549.00
7	Music - CD of local production	Moscow TEC" Atrium "	239.00
8	Music - CD of local production	Moscow TEC" Atrium "	299.00
9	Music - CD of local production	Moscow TEC" Atrium "	897.00
10	Music - CD of local production	Moscow TEC" Atrium "	798.00
11	Music - CD of local production	Moscow TEC" Atrium "	399.00
12	Music - CD of local production	Moscow TEC" Atrium "	399.00
13	Music - CD of local production	Moscow TEC" Atrium "	798.00
14	Music - CD of local production	Moscow TEC" Atrium "	399.00
15	Music - CD of local production	Moscow TEC" Atrium "	798.00
16	Music - CD of local production	Moscow TEC" Atrium "	399.00
17	Music - CD of local production	Moscow TEC" Atrium "	399.00
18	Music - CD of local production	Moscow TEC" Atrium "	399.00
19	Music - CD of local production	Moscow TEC" Atrium "	399.00
20	Music - CD of local production	Moscow TEC" Atrium "	279.00

(2928483, 11)

```
[ ]: #creates a new column called revenue_per_item
final_dataset['revenue_per_item'] = final_dataset['revenue'] / \
    final_dataset['item_cnt_month']

print("\n\nHead of final_dataset:\n")
print(final_dataset.head(20))
print(final_dataset.shape)
```

Head of final\_dataset:

	date	month_num	shop_id	item_id	item_price	item_cnt_month	\
0	02.01.2013	0	59	22154	999.00	1	
1	03.01.2013	0	25	2552	899.00	1	
3	06.01.2013	0	25	2554	1709.05	1	
4	15.01.2013	0	25	2555	1099.00	1	
5	10.01.2013	0	25	2564	349.00	1	
6	02.01.2013	0	25	2565	549.00	1	
7	04.01.2013	0	25	2572	239.00	1	

8	11.01.2013	0	25	2572	299.00	1
9	03.01.2013	0	25	2573	299.00	3
10	03.01.2013	0	25	2574	399.00	2
11	05.01.2013	0	25	2574	399.00	1
12	07.01.2013	0	25	2574	399.00	1
13	08.01.2013	0	25	2574	399.00	2
14	10.01.2013	0	25	2574	399.00	1
15	11.01.2013	0	25	2574	399.00	2
16	13.01.2013	0	25	2574	399.00	1
17	16.01.2013	0	25	2574	399.00	1
18	26.01.2013	0	25	2574	399.00	1
19	27.01.2013	0	25	2574	399.00	1
20	09.01.2013	0	25	2593	279.00	1

	item_name	category_id	\
0	SCENE 2012 (BD)	37	
1	DEEP PURPLE The House Of Blue Light LP	58	
3	DEEP PURPLE Who Do You Think We Are LP	58	
4	DEEP PURPLE 30 Very Best Of 2CD (Businesses).	56	
5	DEEP PURPLE Perihelion: Live In Concert DVD (C...	59	
6	DEEP PURPLE Stormbringer (firms).	56	
7	DEFTONES Koi No Yokan	55	
8	DEFTONES Koi No Yokan	55	
9	DEL REY LANA Born To Die	55	
10	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
11	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
12	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
13	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
14	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
15	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
16	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
17	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
18	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
19	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
20	DEPECHE MODE Music For The Masses	55	

	item_category_name	shop_name	revenue	\
0	Cinema - Blu-Ray	Yaroslavl TC" Altair "	999.00	
1	Music - Vinyl	Moscow TEC" Atrium "	899.00	
3	Music - Vinyl	Moscow TEC" Atrium "	1709.05	
4	Music - CD of corporate production	Moscow TEC" Atrium "	1099.00	
5	Music - Music video	Moscow TEC" Atrium "	349.00	
6	Music - CD of corporate production	Moscow TEC" Atrium "	549.00	
7	Music - CD of local production	Moscow TEC" Atrium "	239.00	
8	Music - CD of local production	Moscow TEC" Atrium "	299.00	
9	Music - CD of local production	Moscow TEC" Atrium "	897.00	
10	Music - CD of local production	Moscow TEC" Atrium "	798.00	
11	Music - CD of local production	Moscow TEC" Atrium "	399.00	

12	Music - CD of local production	Moscow TEC" Atrium "	399.00
13	Music - CD of local production	Moscow TEC" Atrium "	798.00
14	Music - CD of local production	Moscow TEC" Atrium "	399.00
15	Music - CD of local production	Moscow TEC" Atrium "	798.00
16	Music - CD of local production	Moscow TEC" Atrium "	399.00
17	Music - CD of local production	Moscow TEC" Atrium "	399.00
18	Music - CD of local production	Moscow TEC" Atrium "	399.00
19	Music - CD of local production	Moscow TEC" Atrium "	399.00
20	Music - CD of local production	Moscow TEC" Atrium "	279.00

```

revenue_per_item
0      999.00
1      899.00
3     1709.05
4     1099.00
5      349.00
6      549.00
7      239.00
8      299.00
9      299.00
10     399.00
11     399.00
12     399.00
13     399.00
14     399.00
15     399.00
16     399.00
17     399.00
18     399.00
19     399.00
20     279.00
(2928483, 12)

```

```

[ ]: #checks whether the revenue_per_item column and revenue column are the same

if final_dataset['revenue_per_item'].equals(final_dataset['revenue']):
    # If they are the same, drop one of the columns
    final_dataset.drop(columns=['revenue_per_item'], inplace=True)

print("\n\nHead of final_dataset:\n")
print(final_dataset.head(20))
print(final_dataset.shape)

```

Head of final\_dataset:

```

date month_num shop_id item_id item_price item_cnt_month \

```

0	02.01.2013	0	59	22154	999.00	1
1	03.01.2013	0	25	2552	899.00	1
3	06.01.2013	0	25	2554	1709.05	1
4	15.01.2013	0	25	2555	1099.00	1
5	10.01.2013	0	25	2564	349.00	1
6	02.01.2013	0	25	2565	549.00	1
7	04.01.2013	0	25	2572	239.00	1
8	11.01.2013	0	25	2572	299.00	1
9	03.01.2013	0	25	2573	299.00	3
10	03.01.2013	0	25	2574	399.00	2
11	05.01.2013	0	25	2574	399.00	1
12	07.01.2013	0	25	2574	399.00	1
13	08.01.2013	0	25	2574	399.00	2
14	10.01.2013	0	25	2574	399.00	1
15	11.01.2013	0	25	2574	399.00	2
16	13.01.2013	0	25	2574	399.00	1
17	16.01.2013	0	25	2574	399.00	1
18	26.01.2013	0	25	2574	399.00	1
19	27.01.2013	0	25	2574	399.00	1
20	09.01.2013	0	25	2593	279.00	1

	item_name	category_id	\
0	SCENE 2012 (BD)	37	
1	DEEP PURPLE The House Of Blue Light LP	58	
3	DEEP PURPLE Who Do You Think We Are LP	58	
4	DEEP PURPLE 30 Very Best Of 2CD (Businesses).	56	
5	DEEP PURPLE Perihelion: Live In Concert DVD (C...	59	
6	DEEP PURPLE Stormbringer (firms).	56	
7	DEFTONES Koi No Yokan	55	
8	DEFTONES Koi No Yokan	55	
9	DEL REY LANA Born To Die	55	
10	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
11	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
12	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
13	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
14	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
15	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
16	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
17	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
18	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
19	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
20	DEPECHE MODE Music For The Masses	55	

	item_category_name	shop_name	revenue	\
0	Cinema - Blu-Ray	Yaroslavl TC" Altair "	999.00	
1	Music - Vinyl	Moscow TEC" Atrium "	899.00	
3	Music - Vinyl	Moscow TEC" Atrium "	1709.05	
4	Music - CD of corporate production	Moscow TEC" Atrium "	1099.00	

5	Music - Music video	Moscow TEC" Atrium "	349.00
6	Music - CD of corporate production	Moscow TEC" Atrium "	549.00
7	Music - CD of local production	Moscow TEC" Atrium "	239.00
8	Music - CD of local production	Moscow TEC" Atrium "	299.00
9	Music - CD of local production	Moscow TEC" Atrium "	897.00
10	Music - CD of local production	Moscow TEC" Atrium "	798.00
11	Music - CD of local production	Moscow TEC" Atrium "	399.00
12	Music - CD of local production	Moscow TEC" Atrium "	399.00
13	Music - CD of local production	Moscow TEC" Atrium "	798.00
14	Music - CD of local production	Moscow TEC" Atrium "	399.00
15	Music - CD of local production	Moscow TEC" Atrium "	798.00
16	Music - CD of local production	Moscow TEC" Atrium "	399.00
17	Music - CD of local production	Moscow TEC" Atrium "	399.00
18	Music - CD of local production	Moscow TEC" Atrium "	399.00
19	Music - CD of local production	Moscow TEC" Atrium "	399.00
20	Music - CD of local production	Moscow TEC" Atrium "	279.00

```

revenue_per_item
0      999.00
1      899.00
3     1709.05
4     1099.00
5      349.00
6      549.00
7      239.00
8      299.00
9      299.00
10     399.00
11     399.00
12     399.00
13     399.00
14     399.00
15     399.00
16     399.00
17     399.00
18     399.00
19     399.00
20     279.00
(2928483, 12)

```

```
[ ]: #creates a new column called date_num
final_dataset['date_num'] = final_dataset['date'].str[:2]
```

```
[ ]: print("\n\nHead of final_dataset:\n")
print(final_dataset.head(20))
print(final_dataset.shape)
```

Head of final\_dataset:

	date	month_num	shop_id	item_id	item_price	item_cnt_month	\
0	02.01.2013	0	59	22154	999.00	1	
1	03.01.2013	0	25	2552	899.00	1	
3	06.01.2013	0	25	2554	1709.05	1	
4	15.01.2013	0	25	2555	1099.00	1	
5	10.01.2013	0	25	2564	349.00	1	
6	02.01.2013	0	25	2565	549.00	1	
7	04.01.2013	0	25	2572	239.00	1	
8	11.01.2013	0	25	2572	299.00	1	
9	03.01.2013	0	25	2573	299.00	3	
10	03.01.2013	0	25	2574	399.00	2	
11	05.01.2013	0	25	2574	399.00	1	
12	07.01.2013	0	25	2574	399.00	1	
13	08.01.2013	0	25	2574	399.00	2	
14	10.01.2013	0	25	2574	399.00	1	
15	11.01.2013	0	25	2574	399.00	2	
16	13.01.2013	0	25	2574	399.00	1	
17	16.01.2013	0	25	2574	399.00	1	
18	26.01.2013	0	25	2574	399.00	1	
19	27.01.2013	0	25	2574	399.00	1	
20	09.01.2013	0	25	2593	279.00	1	

	item_name	category_id	\
0	SCENE 2012 (BD)	37	
1	DEEP PURPLE The House Of Blue Light LP	58	
3	DEEP PURPLE Who Do You Think We Are LP	58	
4	DEEP PURPLE 30 Very Best Of 2CD (Businesses).	56	
5	DEEP PURPLE Perihelion: Live In Concert DVD (C...	59	
6	DEEP PURPLE Stormbringer (firms).	56	
7	DEFTONES Koi No Yokan	55	
8	DEFTONES Koi No Yokan	55	
9	DEL REY LANA Born To Die	55	
10	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
11	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
12	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
13	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
14	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
15	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
16	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
17	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
18	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
19	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
20	DEPECHE MODE Music For The Masses	55	

item_category_name	shop_name	revenue	\
--------------------	-----------	---------	---

0	Cinema - Blu-Ray	Yaroslavl TC" Altair "	999.00
1	Music - Vinyl	Moscow TEC" Atrium "	899.00
3	Music - Vinyl	Moscow TEC" Atrium "	1709.05
4	Music - CD of corporate production	Moscow TEC" Atrium "	1099.00
5	Music - Music video	Moscow TEC" Atrium "	349.00
6	Music - CD of corporate production	Moscow TEC" Atrium "	549.00
7	Music - CD of local production	Moscow TEC" Atrium "	239.00
8	Music - CD of local production	Moscow TEC" Atrium "	299.00
9	Music - CD of local production	Moscow TEC" Atrium "	897.00
10	Music - CD of local production	Moscow TEC" Atrium "	798.00
11	Music - CD of local production	Moscow TEC" Atrium "	399.00
12	Music - CD of local production	Moscow TEC" Atrium "	399.00
13	Music - CD of local production	Moscow TEC" Atrium "	798.00
14	Music - CD of local production	Moscow TEC" Atrium "	399.00
15	Music - CD of local production	Moscow TEC" Atrium "	798.00
16	Music - CD of local production	Moscow TEC" Atrium "	399.00
17	Music - CD of local production	Moscow TEC" Atrium "	399.00
18	Music - CD of local production	Moscow TEC" Atrium "	399.00
19	Music - CD of local production	Moscow TEC" Atrium "	399.00
20	Music - CD of local production	Moscow TEC" Atrium "	279.00

	revenue_per_item	date_num
0	999.00	02
1	899.00	03
3	1709.05	06
4	1099.00	15
5	349.00	10
6	549.00	02
7	239.00	04
8	299.00	11
9	299.00	03
10	399.00	03
11	399.00	05
12	399.00	07
13	399.00	08
14	399.00	10
15	399.00	11
16	399.00	13
17	399.00	16
18	399.00	26
19	399.00	27
20	279.00	09

(2928483, 13)

```
[ ]: #creates a new column called year_num
final_dataset['year_num'] = final_dataset['date'].str[6:]
```



```
[ ]: print("\n\nHead of final_dataset:\n")
      print(final_dataset.head(20))
      print(final_dataset.shape)
```

Head of final\_dataset:

	date	month_num	shop_id	item_id	item_price	item_cnt_month	\
0	02.01.2013	0	59	22154	999.00	1	
1	03.01.2013	0	25	2552	899.00	1	
3	06.01.2013	0	25	2554	1709.05	1	
4	15.01.2013	0	25	2555	1099.00	1	
5	10.01.2013	0	25	2564	349.00	1	
6	02.01.2013	0	25	2565	549.00	1	
7	04.01.2013	0	25	2572	239.00	1	
8	11.01.2013	0	25	2572	299.00	1	
9	03.01.2013	0	25	2573	299.00	3	
10	03.01.2013	0	25	2574	399.00	2	
11	05.01.2013	0	25	2574	399.00	1	
12	07.01.2013	0	25	2574	399.00	1	
13	08.01.2013	0	25	2574	399.00	2	
14	10.01.2013	0	25	2574	399.00	1	
15	11.01.2013	0	25	2574	399.00	2	
16	13.01.2013	0	25	2574	399.00	1	
17	16.01.2013	0	25	2574	399.00	1	
18	26.01.2013	0	25	2574	399.00	1	
19	27.01.2013	0	25	2574	399.00	1	
20	09.01.2013	0	25	2593	279.00	1	

	item_name	category_id	\
0	SCENE 2012 (BD)	37	
1	DEEP PURPLE The House Of Blue Light LP	58	
3	DEEP PURPLE Who Do You Think We Are LP	58	
4	DEEP PURPLE 30 Very Best Of 2CD (Businesses).	56	
5	DEEP PURPLE Perihelion: Live In Concert DVD (C...	59	
6	DEEP PURPLE Stormbringer (firms).	56	
7	DEFTONES Koi No Yokan	55	
8	DEFTONES Koi No Yokan	55	
9	DEL REY LANA Born To Die	55	
10	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
11	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
12	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
13	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
14	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
15	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
16	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
17	DEL REY LANA Born To Die The Paradise Edition 2CD	55	

18	DEL REY LANA Born To Die The Paradise Edition 2CD	55
19	DEL REY LANA Born To Die The Paradise Edition 2CD	55
20	DEPECHE MODE Music For The Masses	55

	item_category_name	shop_name	revenue \
0	Cinema - Blu-Ray	Yaroslavl TC" Altair "	999.00
1	Music - Vinyl	Moscow TEC" Atrium "	899.00
3	Music - Vinyl	Moscow TEC" Atrium "	1709.05
4	Music - CD of corporate production	Moscow TEC" Atrium "	1099.00
5	Music - Music video	Moscow TEC" Atrium "	349.00
6	Music - CD of corporate production	Moscow TEC" Atrium "	549.00
7	Music - CD of local production	Moscow TEC" Atrium "	239.00
8	Music - CD of local production	Moscow TEC" Atrium "	299.00
9	Music - CD of local production	Moscow TEC" Atrium "	897.00
10	Music - CD of local production	Moscow TEC" Atrium "	798.00
11	Music - CD of local production	Moscow TEC" Atrium "	399.00
12	Music - CD of local production	Moscow TEC" Atrium "	399.00
13	Music - CD of local production	Moscow TEC" Atrium "	798.00
14	Music - CD of local production	Moscow TEC" Atrium "	399.00
15	Music - CD of local production	Moscow TEC" Atrium "	798.00
16	Music - CD of local production	Moscow TEC" Atrium "	399.00
17	Music - CD of local production	Moscow TEC" Atrium "	399.00
18	Music - CD of local production	Moscow TEC" Atrium "	399.00
19	Music - CD of local production	Moscow TEC" Atrium "	399.00
20	Music - CD of local production	Moscow TEC" Atrium "	279.00

	revenue_per_item	date_num	year_num
0	999.00	02	2013
1	899.00	03	2013
3	1709.05	06	2013
4	1099.00	15	2013
5	349.00	10	2013
6	549.00	02	2013
7	239.00	04	2013
8	299.00	11	2013
9	299.00	03	2013
10	399.00	03	2013
11	399.00	05	2013
12	399.00	07	2013
13	399.00	08	2013
14	399.00	10	2013
15	399.00	11	2013
16	399.00	13	2013
17	399.00	16	2013
18	399.00	26	2013
19	399.00	27	2013
20	279.00	09	2013

(2928483, 14)

```
[ ]: print(final_dataset.shape)
      print(final_dataset.info())
```

```
(2928483, 14)
<class 'pandas.core.frame.DataFrame'>
Index: 2928483 entries, 0 to 2935848
Data columns (total 14 columns):
#   Column                Dtype
---  -
0   date                  object
1   month_num             int64
2   shop_id               int64
3   item_id               int64
4   item_price            float64
5   item_cnt_month        int64
6   item_name             object
7   category_id           int64
8   item_category_name     object
9   shop_name             object
10  revenue               float64
11  revenue_per_item       float64
12  date_num              object
13  year_num              object
dtypes: float64(3), int64(5), object(6)
memory usage: 335.1+ MB
None
```

```
[ ]: # rearrange the columns
final_dataset = final_dataset[['date', 'date_num', 'year_num', 'month_num',
    ↪ 'shop_id', 'shop_name', 'item_id', 'item_name', 'category_id',
    ↪ 'item_category_name', 'item_price', 'item_cnt_month', 'revenue']]

print(final_dataset.shape)
print(final_dataset.info())
```

```
(2928483, 13)
<class 'pandas.core.frame.DataFrame'>
Index: 2928483 entries, 0 to 2935848
Data columns (total 13 columns):
#   Column                Dtype
---  -
0   date                  object
1   date_num              object
2   year_num              object
3   month_num             int64
4   shop_id               int64
5   shop_name             object
6   item_id               int64
```

```

7   item_name          object
8   category_id        int64
9   item_category_name object
10  item_price          float64
11  item_cnt_month      int64
12  revenue             float64
dtypes: float64(2), int64(5), object(6)
memory usage: 312.8+ MB
None

```

```

[ ]: #data profiling

#descriptive statistics
print("\n\nDescriptive statistics of final_dataset:\n")
print(final_dataset.describe())

```

Descriptive statistics of final\_dataset:

	month_num	shop_id	item_id	category_id	item_price \
count	2.928483e+06	2.928483e+06	2.928483e+06	2.928483e+06	2.928483e+06
mean	1.456976e+01	3.300296e+01	1.020028e+04	4.001637e+01	8.893627e+02
std	9.422952e+00	1.622543e+01	6.324391e+03	1.709809e+01	1.718155e+03
min	0.000000e+00	0.000000e+00	0.000000e+00	0.000000e+00	7.000000e-02
25%	7.000000e+00	2.200000e+01	4.477000e+03	2.800000e+01	2.490000e+02
50%	1.400000e+01	3.100000e+01	9.355000e+03	4.000000e+01	3.990000e+02
75%	2.300000e+01	4.700000e+01	1.569100e+04	5.500000e+01	9.990000e+02
max	3.300000e+01	5.900000e+01	2.216900e+04	8.300000e+01	5.920000e+04

	item_cnt_month	revenue
count	2.928483e+06	2.928483e+06
mean	1.247257e+00	1.164267e+03
std	2.217429e+00	5.684853e+03
min	1.000000e+00	7.000000e-02
25%	1.000000e+00	2.490000e+02
50%	1.000000e+00	4.490000e+02
75%	1.000000e+00	1.090000e+03
max	6.690000e+02	1.829990e+06

```

[ ]: #data enrichment

#creates a new column called month name

```

```

final_dataset['month_name'] = final_dataset['month_num'].replace({0: 'January', 1: 'February', 2: 'March', 3: 'April', 4: 'May', 5: 'June', 6: 'July', 7: 'August', 8: 'September', 9: 'October', 10: 'November', 11: 'December', 12: 'January', 13: 'February', 14: 'March', 15: 'April', 16: 'May', 17: 'June', 18: 'July', 19: 'August', 20: 'September', 21: 'October', 22: 'November', 23: 'December', 24: 'January', 25: 'February', 26: 'March', 27: 'April', 28: 'May', 29: 'June', 30: 'July', 31: 'August', 32: 'September', 33: 'October'})

print("\n\nHead of final_dataset:\n")
print(final_dataset.head(20))
print(final_dataset.shape)

```

Head of final\_dataset:

	date	date_num	year_num	month_num	shop_id	shop_name	\
0	02.01.2013	02	2013	0	59	Yaroslavl TC	" Altair "
1	03.01.2013	03	2013	0	25	Moscow TEC	" Atrium "
3	06.01.2013	06	2013	0	25	Moscow TEC	" Atrium "
4	15.01.2013	15	2013	0	25	Moscow TEC	" Atrium "
5	10.01.2013	10	2013	0	25	Moscow TEC	" Atrium "
6	02.01.2013	02	2013	0	25	Moscow TEC	" Atrium "
7	04.01.2013	04	2013	0	25	Moscow TEC	" Atrium "
8	11.01.2013	11	2013	0	25	Moscow TEC	" Atrium "
9	03.01.2013	03	2013	0	25	Moscow TEC	" Atrium "
10	03.01.2013	03	2013	0	25	Moscow TEC	" Atrium "
11	05.01.2013	05	2013	0	25	Moscow TEC	" Atrium "
12	07.01.2013	07	2013	0	25	Moscow TEC	" Atrium "
13	08.01.2013	08	2013	0	25	Moscow TEC	" Atrium "
14	10.01.2013	10	2013	0	25	Moscow TEC	" Atrium "
15	11.01.2013	11	2013	0	25	Moscow TEC	" Atrium "
16	13.01.2013	13	2013	0	25	Moscow TEC	" Atrium "
17	16.01.2013	16	2013	0	25	Moscow TEC	" Atrium "
18	26.01.2013	26	2013	0	25	Moscow TEC	" Atrium "
19	27.01.2013	27	2013	0	25	Moscow TEC	" Atrium "
20	09.01.2013	09	2013	0	25	Moscow TEC	" Atrium "

	item_id	item_name	category_id	\
0	22154	SCENE 2012 (BD)	37	
1	2552	DEEP PURPLE The House Of Blue Light LP	58	
3	2554	DEEP PURPLE Who Do You Think We Are LP	58	
4	2555	DEEP PURPLE 30 Very Best Of 2CD (Businesses).	56	
5	2564	DEEP PURPLE Perihelion: Live In Concert DVD (C...	59	
6	2565	DEEP PURPLE Stormbringer (firms).	56	
7	2572	DEFTONES Koi No Yokan	55	
8	2572	DEFTONES Koi No Yokan	55	
9	2573	DEL REY LANA Born To Die	55	

10	2574	DEL REY LANA Born To Die The Paradise Edition 2CD	55
11	2574	DEL REY LANA Born To Die The Paradise Edition 2CD	55
12	2574	DEL REY LANA Born To Die The Paradise Edition 2CD	55
13	2574	DEL REY LANA Born To Die The Paradise Edition 2CD	55
14	2574	DEL REY LANA Born To Die The Paradise Edition 2CD	55
15	2574	DEL REY LANA Born To Die The Paradise Edition 2CD	55
16	2574	DEL REY LANA Born To Die The Paradise Edition 2CD	55
17	2574	DEL REY LANA Born To Die The Paradise Edition 2CD	55
18	2574	DEL REY LANA Born To Die The Paradise Edition 2CD	55
19	2574	DEL REY LANA Born To Die The Paradise Edition 2CD	55
20	2593	DEPECHE MODE Music For The Masses	55

	item_category_name	item_price	item_cnt_month	revenue \
0	Cinema - Blu-Ray	999.00	1	999.00
1	Music - Vinyl	899.00	1	899.00
3	Music - Vinyl	1709.05	1	1709.05
4	Music - CD of corporate production	1099.00	1	1099.00
5	Music - Music video	349.00	1	349.00
6	Music - CD of corporate production	549.00	1	549.00
7	Music - CD of local production	239.00	1	239.00
8	Music - CD of local production	299.00	1	299.00
9	Music - CD of local production	299.00	3	897.00
10	Music - CD of local production	399.00	2	798.00
11	Music - CD of local production	399.00	1	399.00
12	Music - CD of local production	399.00	1	399.00
13	Music - CD of local production	399.00	2	798.00
14	Music - CD of local production	399.00	1	399.00
15	Music - CD of local production	399.00	2	798.00
16	Music - CD of local production	399.00	1	399.00
17	Music - CD of local production	399.00	1	399.00
18	Music - CD of local production	399.00	1	399.00
19	Music - CD of local production	399.00	1	399.00
20	Music - CD of local production	279.00	1	279.00

	month_name
0	January
1	January
3	January
4	January
5	January
6	January
7	January
8	January
9	January
10	January
11	January
12	January
13	January

```

14    January
15    January
16    January
17    January
18    January
19    January
20    January
(2928483, 14)

```

```

[ ]: #removes month_num column

final_dataset.drop(columns=['month_num'], inplace=True)

print("\n\nHead of final_dataset:\n")
print(final_dataset.head(20))
print(final_dataset.shape)

```

Head of final\_dataset:

	date	date_num	year_num	shop_id	shop_name	item_id	\
0	02.01.2013	02	2013	59	Yaroslavl TC" Altair "	22154	
1	03.01.2013	03	2013	25	Moscow TEC" Atrium "	2552	
3	06.01.2013	06	2013	25	Moscow TEC" Atrium "	2554	
4	15.01.2013	15	2013	25	Moscow TEC" Atrium "	2555	
5	10.01.2013	10	2013	25	Moscow TEC" Atrium "	2564	
6	02.01.2013	02	2013	25	Moscow TEC" Atrium "	2565	
7	04.01.2013	04	2013	25	Moscow TEC" Atrium "	2572	
8	11.01.2013	11	2013	25	Moscow TEC" Atrium "	2572	
9	03.01.2013	03	2013	25	Moscow TEC" Atrium "	2573	
10	03.01.2013	03	2013	25	Moscow TEC" Atrium "	2574	
11	05.01.2013	05	2013	25	Moscow TEC" Atrium "	2574	
12	07.01.2013	07	2013	25	Moscow TEC" Atrium "	2574	
13	08.01.2013	08	2013	25	Moscow TEC" Atrium "	2574	
14	10.01.2013	10	2013	25	Moscow TEC" Atrium "	2574	
15	11.01.2013	11	2013	25	Moscow TEC" Atrium "	2574	
16	13.01.2013	13	2013	25	Moscow TEC" Atrium "	2574	
17	16.01.2013	16	2013	25	Moscow TEC" Atrium "	2574	
18	26.01.2013	26	2013	25	Moscow TEC" Atrium "	2574	
19	27.01.2013	27	2013	25	Moscow TEC" Atrium "	2574	
20	09.01.2013	09	2013	25	Moscow TEC" Atrium "	2593	

	item_name	category_id	\
0	SCENE 2012 (BD)	37	
1	DEEP PURPLE The House Of Blue Light LP	58	
3	DEEP PURPLE Who Do You Think We Are LP	58	
4	DEEP PURPLE 30 Very Best Of 2CD (Businesses).	56	

5	DEEP PURPLE Perihelion: Live In Concert DVD (C...	59
6	DEEP PURPLE Stormbringer (firms).	56
7	DEFTONES Koi No Yokan	55
8	DEFTONES Koi No Yokan	55
9	DEL REY LANA Born To Die	55
10	DEL REY LANA Born To Die The Paradise Edition 2CD	55
11	DEL REY LANA Born To Die The Paradise Edition 2CD	55
12	DEL REY LANA Born To Die The Paradise Edition 2CD	55
13	DEL REY LANA Born To Die The Paradise Edition 2CD	55
14	DEL REY LANA Born To Die The Paradise Edition 2CD	55
15	DEL REY LANA Born To Die The Paradise Edition 2CD	55
16	DEL REY LANA Born To Die The Paradise Edition 2CD	55
17	DEL REY LANA Born To Die The Paradise Edition 2CD	55
18	DEL REY LANA Born To Die The Paradise Edition 2CD	55
19	DEL REY LANA Born To Die The Paradise Edition 2CD	55
20	DEPECHE MODE Music For The Masses	55

	item_category_name	item_price	item_cnt_month	revenue	\
0	Cinema - Blu-Ray	999.00	1	999.00	
1	Music - Vinyl	899.00	1	899.00	
3	Music - Vinyl	1709.05	1	1709.05	
4	Music - CD of corporate production	1099.00	1	1099.00	
5	Music - Music video	349.00	1	349.00	
6	Music - CD of corporate production	549.00	1	549.00	
7	Music - CD of local production	239.00	1	239.00	
8	Music - CD of local production	299.00	1	299.00	
9	Music - CD of local production	299.00	3	897.00	
10	Music - CD of local production	399.00	2	798.00	
11	Music - CD of local production	399.00	1	399.00	
12	Music - CD of local production	399.00	1	399.00	
13	Music - CD of local production	399.00	2	798.00	
14	Music - CD of local production	399.00	1	399.00	
15	Music - CD of local production	399.00	2	798.00	
16	Music - CD of local production	399.00	1	399.00	
17	Music - CD of local production	399.00	1	399.00	
18	Music - CD of local production	399.00	1	399.00	
19	Music - CD of local production	399.00	1	399.00	
20	Music - CD of local production	279.00	1	279.00	

	month_name
0	January
1	January
3	January
4	January
5	January
6	January
7	January
8	January



```

9      January
10     January
11     January
12     January
13     January
14     January
15     January
16     January
17     January
18     January
19     January
20     January
(2928483, 13)

```

```

[ ]: #rearrange the columns

final_dataset = final_dataset[['date', 'date_num', 'month_name', 'year_num',
    ↪ 'shop_id', 'shop_name', 'item_id', 'item_name', 'category_id',
    ↪ 'item_category_name', 'item_price', 'item_cnt_month', 'revenue']]

print("\n\nHead of final_dataset:\n")
print(final_dataset.head(20))
print(final_dataset.shape)

```

Head of final\_dataset:

	date	date_num	month_name	year_num	shop_id	shop_name	\
0	02.01.2013	02	January	2013	59	Yaroslavl TC	Altair "
1	03.01.2013	03	January	2013	25	Moscow TEC	Atrium "
3	06.01.2013	06	January	2013	25	Moscow TEC	Atrium "
4	15.01.2013	15	January	2013	25	Moscow TEC	Atrium "
5	10.01.2013	10	January	2013	25	Moscow TEC	Atrium "
6	02.01.2013	02	January	2013	25	Moscow TEC	Atrium "
7	04.01.2013	04	January	2013	25	Moscow TEC	Atrium "
8	11.01.2013	11	January	2013	25	Moscow TEC	Atrium "
9	03.01.2013	03	January	2013	25	Moscow TEC	Atrium "
10	03.01.2013	03	January	2013	25	Moscow TEC	Atrium "
11	05.01.2013	05	January	2013	25	Moscow TEC	Atrium "
12	07.01.2013	07	January	2013	25	Moscow TEC	Atrium "
13	08.01.2013	08	January	2013	25	Moscow TEC	Atrium "
14	10.01.2013	10	January	2013	25	Moscow TEC	Atrium "
15	11.01.2013	11	January	2013	25	Moscow TEC	Atrium "
16	13.01.2013	13	January	2013	25	Moscow TEC	Atrium "
17	16.01.2013	16	January	2013	25	Moscow TEC	Atrium "
18	26.01.2013	26	January	2013	25	Moscow TEC	Atrium "
19	27.01.2013	27	January	2013	25	Moscow TEC	Atrium "

20 09.01.2013 09 January 2013 25 Moscow TEC" Atrium "

	item_id	item_name	category_id \
0	22154	SCENE 2012 (BD)	37
1	2552	DEEP PURPLE The House Of Blue Light LP	58
3	2554	DEEP PURPLE Who Do You Think We Are LP	58
4	2555	DEEP PURPLE 30 Very Best Of 2CD (Businesses).	56
5	2564	DEEP PURPLE Perihelion: Live In Concert DVD (C...	59
6	2565	DEEP PURPLE Stormbringer (firms).	56
7	2572	DEFTONES Koi No Yokan	55
8	2572	DEFTONES Koi No Yokan	55
9	2573	DEL REY LANA Born To Die	55
10	2574	DEL REY LANA Born To Die The Paradise Edition 2CD	55
11	2574	DEL REY LANA Born To Die The Paradise Edition 2CD	55
12	2574	DEL REY LANA Born To Die The Paradise Edition 2CD	55
13	2574	DEL REY LANA Born To Die The Paradise Edition 2CD	55
14	2574	DEL REY LANA Born To Die The Paradise Edition 2CD	55
15	2574	DEL REY LANA Born To Die The Paradise Edition 2CD	55
16	2574	DEL REY LANA Born To Die The Paradise Edition 2CD	55
17	2574	DEL REY LANA Born To Die The Paradise Edition 2CD	55
18	2574	DEL REY LANA Born To Die The Paradise Edition 2CD	55
19	2574	DEL REY LANA Born To Die The Paradise Edition 2CD	55
20	2593	DEPECHE MODE Music For The Masses	55

	item_category_name	item_price	item_cnt_month	revenue
0	Cinema - Blu-Ray	999.00	1	999.00
1	Music - Vinyl	899.00	1	899.00
3	Music - Vinyl	1709.05	1	1709.05
4	Music - CD of corporate production	1099.00	1	1099.00
5	Music - Music video	349.00	1	349.00
6	Music - CD of corporate production	549.00	1	549.00
7	Music - CD of local production	239.00	1	239.00
8	Music - CD of local production	299.00	1	299.00
9	Music - CD of local production	299.00	3	897.00
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11	Music - CD of local production	399.00	1	399.00
12	Music - CD of local production	399.00	1	399.00
13	Music - CD of local production	399.00	2	798.00
14	Music - CD of local production	399.00	1	399.00
15	Music - CD of local production	399.00	2	798.00
16	Music - CD of local production	399.00	1	399.00
17	Music - CD of local production	399.00	1	399.00
18	Music - CD of local production	399.00	1	399.00
19	Music - CD of local production	399.00	1	399.00
20	Music - CD of local production	279.00	1	279.00

(2928483, 13)

```
[ ]: #data binning

#found the bins using the following code
print(final_dataset['item_price'].max())
print(final_dataset['item_price'].min())

#creates a new column called price range
final_dataset['price_range'] = pd.cut(final_dataset['item_price'], bins=[-1,
↪100, 200, 300, 400, 500, 600, 700, 800, 900, 100000], labels=['0-100',
↪'100-200', '200-300', '300-400', '400-500', '500-600', '600-700', '700-800',
↪'800-900', '900-100000'])
```

59200.0

0.07

```
[ ]: print("\n\nHead of final_dataset:\n")
print(final_dataset.head(20))
print(final_dataset.shape)
```

Head of final\_dataset:

	date	date_num	month_name	year_num	shop_id	shop_name	\
0	02.01.2013	02	January	2013	59	Yaroslavl TC	" Altair "
1	03.01.2013	03	January	2013	25	Moscow TEC	" Atrium "
3	06.01.2013	06	January	2013	25	Moscow TEC	" Atrium "
4	15.01.2013	15	January	2013	25	Moscow TEC	" Atrium "
5	10.01.2013	10	January	2013	25	Moscow TEC	" Atrium "
6	02.01.2013	02	January	2013	25	Moscow TEC	" Atrium "
7	04.01.2013	04	January	2013	25	Moscow TEC	" Atrium "
8	11.01.2013	11	January	2013	25	Moscow TEC	" Atrium "
9	03.01.2013	03	January	2013	25	Moscow TEC	" Atrium "
10	03.01.2013	03	January	2013	25	Moscow TEC	" Atrium "
11	05.01.2013	05	January	2013	25	Moscow TEC	" Atrium "
12	07.01.2013	07	January	2013	25	Moscow TEC	" Atrium "
13	08.01.2013	08	January	2013	25	Moscow TEC	" Atrium "
14	10.01.2013	10	January	2013	25	Moscow TEC	" Atrium "
15	11.01.2013	11	January	2013	25	Moscow TEC	" Atrium "
16	13.01.2013	13	January	2013	25	Moscow TEC	" Atrium "
17	16.01.2013	16	January	2013	25	Moscow TEC	" Atrium "
18	26.01.2013	26	January	2013	25	Moscow TEC	" Atrium "
19	27.01.2013	27	January	2013	25	Moscow TEC	" Atrium "
20	09.01.2013	09	January	2013	25	Moscow TEC	" Atrium "

	item_id	item_name	category_id	\
0	22154	SCENE 2012 (BD)	37	
1	2552	DEEP PURPLE The House Of Blue Light LP	58	

3	2554	DEEP PURPLE Who Do You Think We Are LP	58
4	2555	DEEP PURPLE 30 Very Best Of 2CD (Businesses).	56
5	2564	DEEP PURPLE Perihelion: Live In Concert DVD (C...	59
6	2565	DEEP PURPLE Stormbringer (firms).	56
7	2572	DEFTONES Koi No Yokan	55
8	2572	DEFTONES Koi No Yokan	55
9	2573	DEL REY LANA Born To Die	55
10	2574	DEL REY LANA Born To Die The Paradise Edition 2CD	55
11	2574	DEL REY LANA Born To Die The Paradise Edition 2CD	55
12	2574	DEL REY LANA Born To Die The Paradise Edition 2CD	55
13	2574	DEL REY LANA Born To Die The Paradise Edition 2CD	55
14	2574	DEL REY LANA Born To Die The Paradise Edition 2CD	55
15	2574	DEL REY LANA Born To Die The Paradise Edition 2CD	55
16	2574	DEL REY LANA Born To Die The Paradise Edition 2CD	55
17	2574	DEL REY LANA Born To Die The Paradise Edition 2CD	55
18	2574	DEL REY LANA Born To Die The Paradise Edition 2CD	55
19	2574	DEL REY LANA Born To Die The Paradise Edition 2CD	55
20	2593	DEPECHE MODE Music For The Masses	55

	item_category_name	item_price	item_cnt_month	revenue \
0	Cinema - Blu-Ray	999.00	1	999.00
1	Music - Vinyl	899.00	1	899.00
3	Music - Vinyl	1709.05	1	1709.05
4	Music - CD of corporate production	1099.00	1	1099.00
5	Music - Music video	349.00	1	349.00
6	Music - CD of corporate production	549.00	1	549.00
7	Music - CD of local production	239.00	1	239.00
8	Music - CD of local production	299.00	1	299.00
9	Music - CD of local production	299.00	3	897.00
10	Music - CD of local production	399.00	2	798.00
11	Music - CD of local production	399.00	1	399.00
12	Music - CD of local production	399.00	1	399.00
13	Music - CD of local production	399.00	2	798.00
14	Music - CD of local production	399.00	1	399.00
15	Music - CD of local production	399.00	2	798.00
16	Music - CD of local production	399.00	1	399.00
17	Music - CD of local production	399.00	1	399.00
18	Music - CD of local production	399.00	1	399.00
19	Music - CD of local production	399.00	1	399.00
20	Music - CD of local production	279.00	1	279.00

	price_range
0	900-100000
1	800-900
3	900-100000
4	900-100000
5	300-400
6	500-600

```

7      200-300
8      200-300
9      200-300
10     300-400
11     300-400
12     300-400
13     300-400
14     300-400
15     300-400
16     300-400
17     300-400
18     300-400
19     300-400
20     200-300
(2928483, 14)

```

```

[ ]: #log transformation

#creates a new column called log_revenue
final_dataset['log_revenue'] = np.log(final_dataset['revenue'])

```

```

[ ]: print("\n\nHead of final_dataset:\n")
print(final_dataset.head(20))
print(final_dataset.shape)

```

Head of final\_dataset:

	date	date_num	month_name	year_num	shop_id	shop_name	\
0	02.01.2013	02	January	2013	59	Yaroslavl TC"	Altair "
1	03.01.2013	03	January	2013	25	Moscow TEC"	Atrium "
3	06.01.2013	06	January	2013	25	Moscow TEC"	Atrium "
4	15.01.2013	15	January	2013	25	Moscow TEC"	Atrium "
5	10.01.2013	10	January	2013	25	Moscow TEC"	Atrium "
6	02.01.2013	02	January	2013	25	Moscow TEC"	Atrium "
7	04.01.2013	04	January	2013	25	Moscow TEC"	Atrium "
8	11.01.2013	11	January	2013	25	Moscow TEC"	Atrium "
9	03.01.2013	03	January	2013	25	Moscow TEC"	Atrium "
10	03.01.2013	03	January	2013	25	Moscow TEC"	Atrium "
11	05.01.2013	05	January	2013	25	Moscow TEC"	Atrium "
12	07.01.2013	07	January	2013	25	Moscow TEC"	Atrium "
13	08.01.2013	08	January	2013	25	Moscow TEC"	Atrium "
14	10.01.2013	10	January	2013	25	Moscow TEC"	Atrium "
15	11.01.2013	11	January	2013	25	Moscow TEC"	Atrium "
16	13.01.2013	13	January	2013	25	Moscow TEC"	Atrium "
17	16.01.2013	16	January	2013	25	Moscow TEC"	Atrium "
18	26.01.2013	26	January	2013	25	Moscow TEC"	Atrium "

19	27.01.2013	27	January	2013	25	Moscow TEC" Atrium "
20	09.01.2013	09	January	2013	25	Moscow TEC" Atrium "

	item_id	item_name	category_id	\
0	22154	SCENE 2012 (BD)	37	
1	2552	DEEP PURPLE The House Of Blue Light LP	58	
3	2554	DEEP PURPLE Who Do You Think We Are LP	58	
4	2555	DEEP PURPLE 30 Very Best Of 2CD (Businesses).	56	
5	2564	DEEP PURPLE Perihelion: Live In Concert DVD (C...	59	
6	2565	DEEP PURPLE Stormbringer (firms).	56	
7	2572	DEFTONES Koi No Yokan	55	
8	2572	DEFTONES Koi No Yokan	55	
9	2573	DEL REY LANA Born To Die	55	
10	2574	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
11	2574	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
12	2574	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
13	2574	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
14	2574	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
15	2574	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
16	2574	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
17	2574	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
18	2574	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
19	2574	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
20	2593	DEPECHE MODE Music For The Masses	55	

	item_category_name	item_price	item_cnt_month	revenue	\
0	Cinema - Blu-Ray	999.00	1	999.00	
1	Music - Vinyl	899.00	1	899.00	
3	Music - Vinyl	1709.05	1	1709.05	
4	Music - CD of corporate production	1099.00	1	1099.00	
5	Music - Music video	349.00	1	349.00	
6	Music - CD of corporate production	549.00	1	549.00	
7	Music - CD of local production	239.00	1	239.00	
8	Music - CD of local production	299.00	1	299.00	
9	Music - CD of local production	299.00	3	897.00	
10	Music - CD of local production	399.00	2	798.00	
11	Music - CD of local production	399.00	1	399.00	
12	Music - CD of local production	399.00	1	399.00	
13	Music - CD of local production	399.00	2	798.00	
14	Music - CD of local production	399.00	1	399.00	
15	Music - CD of local production	399.00	2	798.00	
16	Music - CD of local production	399.00	1	399.00	
17	Music - CD of local production	399.00	1	399.00	
18	Music - CD of local production	399.00	1	399.00	
19	Music - CD of local production	399.00	1	399.00	
20	Music - CD of local production	279.00	1	279.00	

price\_range log\_revenue

0	900-100000	6.906755
1	800-900	6.801283
3	900-100000	7.443693
4	900-100000	7.002156
5	300-400	5.855072
6	500-600	6.308098
7	200-300	5.476464
8	200-300	5.700444
9	200-300	6.799056
10	300-400	6.682109
11	300-400	5.988961
12	300-400	5.988961
13	300-400	6.682109
14	300-400	5.988961
15	300-400	6.682109
16	300-400	5.988961
17	300-400	5.988961
18	300-400	5.988961
19	300-400	5.988961
20	200-300	5.631212

(2928483, 15)

```
[ ]: #encoding

#encodes the year_num column to 0, 1, 2

final_dataset['year_num'] = final_dataset['year_num'].replace({'2013': 0,
↪ '2014': 1, '2015': 2})

print("\n\nHead of final_dataset:\n")
print(final_dataset.head(20))
print(final_dataset.shape)
```

Head of final\_dataset:

	date	date_num	month_name	year_num	shop_id	\
0	02.01.2013	02	January	0	59	
1	03.01.2013	03	January	0	25	
3	06.01.2013	06	January	0	25	
4	15.01.2013	15	January	0	25	
5	10.01.2013	10	January	0	25	
6	02.01.2013	02	January	0	25	
7	04.01.2013	04	January	0	25	
8	11.01.2013	11	January	0	25	
9	03.01.2013	03	January	0	25	

10	03.01.2013	03	January	0	25
11	05.01.2013	05	January	0	25
12	07.01.2013	07	January	0	25
13	08.01.2013	08	January	0	25
14	10.01.2013	10	January	0	25
15	11.01.2013	11	January	0	25
16	13.01.2013	13	January	0	25
17	16.01.2013	16	January	0	25
18	26.01.2013	26	January	0	25
19	27.01.2013	27	January	0	25
20	09.01.2013	09	January	0	25

	shop_name	item_id	\
0	Yaroslavl TC" Altair "	22154	
1	Moscow TEC" Atrium "	2552	
3	Moscow TEC" Atrium "	2554	
4	Moscow TEC" Atrium "	2555	
5	Moscow TEC" Atrium "	2564	
6	Moscow TEC" Atrium "	2565	
7	Moscow TEC" Atrium "	2572	
8	Moscow TEC" Atrium "	2572	
9	Moscow TEC" Atrium "	2573	
10	Moscow TEC" Atrium "	2574	
11	Moscow TEC" Atrium "	2574	
12	Moscow TEC" Atrium "	2574	
13	Moscow TEC" Atrium "	2574	
14	Moscow TEC" Atrium "	2574	
15	Moscow TEC" Atrium "	2574	
16	Moscow TEC" Atrium "	2574	
17	Moscow TEC" Atrium "	2574	
18	Moscow TEC" Atrium "	2574	
19	Moscow TEC" Atrium "	2574	
20	Moscow TEC" Atrium "	2593	

	item_name	category_id	\
0	SCENE 2012 (BD)	37	
1	DEEP PURPLE The House Of Blue Light LP	58	
3	DEEP PURPLE Who Do You Think We Are LP	58	
4	DEEP PURPLE 30 Very Best Of 2CD (Businesses).	56	
5	DEEP PURPLE Perihelion: Live In Concert DVD (C...	59	
6	DEEP PURPLE Stormbringer (firms).	56	
7	DEFTONES Koi No Yokan	55	
8	DEFTONES Koi No Yokan	55	
9	DEL REY LANA Born To Die	55	
10	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
11	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
12	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
13	DEL REY LANA Born To Die The Paradise Edition 2CD	55	



14	DEL REY LANA Born To Die The Paradise Edition 2CD	55
15	DEL REY LANA Born To Die The Paradise Edition 2CD	55
16	DEL REY LANA Born To Die The Paradise Edition 2CD	55
17	DEL REY LANA Born To Die The Paradise Edition 2CD	55
18	DEL REY LANA Born To Die The Paradise Edition 2CD	55
19	DEL REY LANA Born To Die The Paradise Edition 2CD	55
20	DEPECHE MODE Music For The Masses	55

	item_category_name	item_price	item_cnt_month	revenue	\
0	Cinema - Blu-Ray	999.00	1	999.00	
1	Music - Vinyl	899.00	1	899.00	
3	Music - Vinyl	1709.05	1	1709.05	
4	Music - CD of corporate production	1099.00	1	1099.00	
5	Music - Music video	349.00	1	349.00	
6	Music - CD of corporate production	549.00	1	549.00	
7	Music - CD of local production	239.00	1	239.00	
8	Music - CD of local production	299.00	1	299.00	
9	Music - CD of local production	299.00	3	897.00	
10	Music - CD of local production	399.00	2	798.00	
11	Music - CD of local production	399.00	1	399.00	
12	Music - CD of local production	399.00	1	399.00	
13	Music - CD of local production	399.00	2	798.00	
14	Music - CD of local production	399.00	1	399.00	
15	Music - CD of local production	399.00	2	798.00	
16	Music - CD of local production	399.00	1	399.00	
17	Music - CD of local production	399.00	1	399.00	
18	Music - CD of local production	399.00	1	399.00	
19	Music - CD of local production	399.00	1	399.00	
20	Music - CD of local production	279.00	1	279.00	

	price_range	log_revenue
0	900-100000	6.906755
1	800-900	6.801283
3	900-100000	7.443693
4	900-100000	7.002156
5	300-400	5.855072
6	500-600	6.308098
7	200-300	5.476464
8	200-300	5.700444
9	200-300	6.799056
10	300-400	6.682109
11	300-400	5.988961
12	300-400	5.988961
13	300-400	6.682109
14	300-400	5.988961
15	300-400	6.682109
16	300-400	5.988961
17	300-400	5.988961

```

18      300-400      5.988961
19      300-400      5.988961
20      200-300      5.631212
(2928483, 15)

```

```

[ ]: #grouping and aggregation

#grouping the data set by shop_id and year_num and aggregating the
    ↪ item_cnt_month column using sum

grouped_by_shop_id_and_year_num = final_dataset.groupby(['shop_id',
    ↪ 'year_num']).agg({'item_cnt_month': 'sum'})

print("\n\nHead of grouped_by_shop_id_and_year_num:\n")
print(grouped_by_shop_id_and_year_num.head(60))
print(grouped_by_shop_id_and_year_num.shape)

```

Head of grouped\_by\_shop\_id\_and\_year\_num:

shop_id	year_num	item_cnt_month
0	0	11705
1	0	6311
2	0	9989
	1	12247
	2	8470
3	0	10242
	1	11039
	2	7194
4	0	19054
	1	15909
	2	9106
5	0	14717
	1	17041
	2	11089
6	0	46707
	1	35496
	2	18554
7	0	28218
	1	24530
	2	14411
8	0	3602
9	0	6531
	1	6155
	2	3188
10	0	11132

	1	8859
	2	4532
11	2	572
12	0	19494
	1	26379
	2	24947
13	0	13529
	1	6234
14	0	20114
	1	16003
	2	10397
15	0	32339
	1	24854
	2	14138
16	0	30250
	1	19569
	2	11953
17	0	9913
	1	15308
	2	697
18	0	33632
	1	18815
	2	13174
19	0	32860
	1	25353
	2	15455
20	1	3262
	2	2611
21	0	25258
	1	25167
	2	18259
22	0	26943
	1	20341
	2	13050
23	0	7722

(159, 1)

```
[ ]: #creates a new column called scaled_revenue
final_dataset['scaled_revenue'] = (final_dataset['revenue'] -
    ↪final_dataset['revenue'].min()) / (final_dataset['revenue'].max() -
    ↪final_dataset['revenue'].min())

print("\n\nHead of final_dataset:\n")
print(final_dataset.head(20))
print(final_dataset.shape)
```

Head of final\_dataset:

	date	date_num	month_name	year_num	shop_id \
0	02.01.2013	02	January	0	59
1	03.01.2013	03	January	0	25
3	06.01.2013	06	January	0	25
4	15.01.2013	15	January	0	25
5	10.01.2013	10	January	0	25
6	02.01.2013	02	January	0	25
7	04.01.2013	04	January	0	25
8	11.01.2013	11	January	0	25
9	03.01.2013	03	January	0	25
10	03.01.2013	03	January	0	25
11	05.01.2013	05	January	0	25
12	07.01.2013	07	January	0	25
13	08.01.2013	08	January	0	25
14	10.01.2013	10	January	0	25
15	11.01.2013	11	January	0	25
16	13.01.2013	13	January	0	25
17	16.01.2013	16	January	0	25
18	26.01.2013	26	January	0	25
19	27.01.2013	27	January	0	25
20	09.01.2013	09	January	0	25

	shop_name	item_id \
0	Yaroslavl TC" Altair "	22154
1	Moscow TEC" Atrium "	2552
3	Moscow TEC" Atrium "	2554
4	Moscow TEC" Atrium "	2555
5	Moscow TEC" Atrium "	2564
6	Moscow TEC" Atrium "	2565
7	Moscow TEC" Atrium "	2572
8	Moscow TEC" Atrium "	2572
9	Moscow TEC" Atrium "	2573
10	Moscow TEC" Atrium "	2574
11	Moscow TEC" Atrium "	2574
12	Moscow TEC" Atrium "	2574
13	Moscow TEC" Atrium "	2574
14	Moscow TEC" Atrium "	2574
15	Moscow TEC" Atrium "	2574
16	Moscow TEC" Atrium "	2574
17	Moscow TEC" Atrium "	2574
18	Moscow TEC" Atrium "	2574
19	Moscow TEC" Atrium "	2574
20	Moscow TEC" Atrium "	2593

	item_name	category_id \
0	SCENE 2012 (BD)	37

1	DEEP PURPLE The House Of Blue Light LP	58
3	DEEP PURPLE Who Do You Think We Are LP	58
4	DEEP PURPLE 30 Very Best Of 2CD (Businesses).	56
5	DEEP PURPLE Perihelion: Live In Concert DVD (C...	59
6	DEEP PURPLE Stormbringer (firms).	56
7	DEFTONES Koi No Yokan	55
8	DEFTONES Koi No Yokan	55
9	DEL REY LANA Born To Die	55
10	DEL REY LANA Born To Die The Paradise Edition 2CD	55
11	DEL REY LANA Born To Die The Paradise Edition 2CD	55
12	DEL REY LANA Born To Die The Paradise Edition 2CD	55
13	DEL REY LANA Born To Die The Paradise Edition 2CD	55
14	DEL REY LANA Born To Die The Paradise Edition 2CD	55
15	DEL REY LANA Born To Die The Paradise Edition 2CD	55
16	DEL REY LANA Born To Die The Paradise Edition 2CD	55
17	DEL REY LANA Born To Die The Paradise Edition 2CD	55
18	DEL REY LANA Born To Die The Paradise Edition 2CD	55
19	DEL REY LANA Born To Die The Paradise Edition 2CD	55
20	DEPECHE MODE Music For The Masses	55

	item_category_name	item_price	item_cnt_month	revenue \
0	Cinema - Blu-Ray	999.00	1	999.00
1	Music - Vinyl	899.00	1	899.00
3	Music - Vinyl	1709.05	1	1709.05
4	Music - CD of corporate production	1099.00	1	1099.00
5	Music - Music video	349.00	1	349.00
6	Music - CD of corporate production	549.00	1	549.00
7	Music - CD of local production	239.00	1	239.00
8	Music - CD of local production	299.00	1	299.00
9	Music - CD of local production	299.00	3	897.00
10	Music - CD of local production	399.00	2	798.00
11	Music - CD of local production	399.00	1	399.00
12	Music - CD of local production	399.00	1	399.00
13	Music - CD of local production	399.00	2	798.00
14	Music - CD of local production	399.00	1	399.00
15	Music - CD of local production	399.00	2	798.00
16	Music - CD of local production	399.00	1	399.00
17	Music - CD of local production	399.00	1	399.00
18	Music - CD of local production	399.00	1	399.00
19	Music - CD of local production	399.00	1	399.00
20	Music - CD of local production	279.00	1	279.00

	price_range	log_revenue	scaled_revenue
0	900-100000	6.906755	0.000546
1	800-900	6.801283	0.000491
3	900-100000	7.443693	0.000934
4	900-100000	7.002156	0.000601
5	300-400	5.855072	0.000191

6	500-600	6.308098	0.000300
7	200-300	5.476464	0.000131
8	200-300	5.700444	0.000163
9	200-300	6.799056	0.000490
10	300-400	6.682109	0.000436
11	300-400	5.988961	0.000218
12	300-400	5.988961	0.000218
13	300-400	6.682109	0.000436
14	300-400	5.988961	0.000218
15	300-400	6.682109	0.000436
16	300-400	5.988961	0.000218
17	300-400	5.988961	0.000218
18	300-400	5.988961	0.000218
19	300-400	5.988961	0.000218
20	200-300	5.631212	0.000152

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```
[ ]: #change month_name column to numeric

final_dataset['month_name'] = final_dataset['month_name'].replace({'January': 1,
↪ 'February': 2, 'March': 3, 'April': 4, 'May': 5, 'June': 6, 'July': 7,
↪ 'August': 8, 'September': 9, 'October': 10, 'November': 11, 'December': 12})

print("\n\nHead of final_dataset:\n")
print(final_dataset.head(20))
print(final_dataset.shape)
```

Head of final\_dataset:

	date	date_num	month_name	year_num	shop_id	\
0	02.01.2013	02	1	0	59	
1	03.01.2013	03	1	0	25	
3	06.01.2013	06	1	0	25	
4	15.01.2013	15	1	0	25	
5	10.01.2013	10	1	0	25	
6	02.01.2013	02	1	0	25	
7	04.01.2013	04	1	0	25	
8	11.01.2013	11	1	0	25	
9	03.01.2013	03	1	0	25	
10	03.01.2013	03	1	0	25	
11	05.01.2013	05	1	0	25	
12	07.01.2013	07	1	0	25	
13	08.01.2013	08	1	0	25	
14	10.01.2013	10	1	0	25	
15	11.01.2013	11	1	0	25	
16	13.01.2013	13	1	0	25	

17	16.01.2013	16	1	0	25
18	26.01.2013	26	1	0	25
19	27.01.2013	27	1	0	25
20	09.01.2013	09	1	0	25

	shop_name	item_id	\
0	Yaroslavl TC" Altair "	22154	
1	Moscow TEC" Atrium "	2552	
3	Moscow TEC" Atrium "	2554	
4	Moscow TEC" Atrium "	2555	
5	Moscow TEC" Atrium "	2564	
6	Moscow TEC" Atrium "	2565	
7	Moscow TEC" Atrium "	2572	
8	Moscow TEC" Atrium "	2572	
9	Moscow TEC" Atrium "	2573	
10	Moscow TEC" Atrium "	2574	
11	Moscow TEC" Atrium "	2574	
12	Moscow TEC" Atrium "	2574	
13	Moscow TEC" Atrium "	2574	
14	Moscow TEC" Atrium "	2574	
15	Moscow TEC" Atrium "	2574	
16	Moscow TEC" Atrium "	2574	
17	Moscow TEC" Atrium "	2574	
18	Moscow TEC" Atrium "	2574	
19	Moscow TEC" Atrium "	2574	
20	Moscow TEC" Atrium "	2593	

	item_name	category_id	\
0	SCENE 2012 (BD)	37	
1	DEEP PURPLE The House Of Blue Light LP	58	
3	DEEP PURPLE Who Do You Think We Are LP	58	
4	DEEP PURPLE 30 Very Best Of 2CD (Businesses).	56	
5	DEEP PURPLE Perihelion: Live In Concert DVD (C...	59	
6	DEEP PURPLE Stormbringer (firms).	56	
7	DEFTONES Koi No Yokan	55	
8	DEFTONES Koi No Yokan	55	
9	DEL REY LANA Born To Die	55	
10	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
11	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
12	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
13	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
14	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
15	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
16	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
17	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
18	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
19	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
20	DEPECHE MODE Music For The Masses	55	

	item_category_name	item_price	item_cnt_month	revenue	\
0	Cinema - Blu-Ray	999.00	1	999.00	
1	Music - Vinyl	899.00	1	899.00	
3	Music - Vinyl	1709.05	1	1709.05	
4	Music - CD of corporate production	1099.00	1	1099.00	
5	Music - Music video	349.00	1	349.00	
6	Music - CD of corporate production	549.00	1	549.00	
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8	Music - CD of local production	299.00	1	299.00	
9	Music - CD of local production	299.00	3	897.00	
10	Music - CD of local production	399.00	2	798.00	
11	Music - CD of local production	399.00	1	399.00	
12	Music - CD of local production	399.00	1	399.00	
13	Music - CD of local production	399.00	2	798.00	
14	Music - CD of local production	399.00	1	399.00	
15	Music - CD of local production	399.00	2	798.00	
16	Music - CD of local production	399.00	1	399.00	
17	Music - CD of local production	399.00	1	399.00	
18	Music - CD of local production	399.00	1	399.00	
19	Music - CD of local production	399.00	1	399.00	
20	Music - CD of local production	279.00	1	279.00	

	price_range	log_revenue	scaled_revenue
0	900-100000	6.906755	0.000546
1	800-900	6.801283	0.000491
3	900-100000	7.443693	0.000934
4	900-100000	7.002156	0.000601
5	300-400	5.855072	0.000191
6	500-600	6.308098	0.000300
7	200-300	5.476464	0.000131
8	200-300	5.700444	0.000163
9	200-300	6.799056	0.000490
10	300-400	6.682109	0.000436
11	300-400	5.988961	0.000218
12	300-400	5.988961	0.000218
13	300-400	6.682109	0.000436
14	300-400	5.988961	0.000218
15	300-400	6.682109	0.000436
16	300-400	5.988961	0.000218
17	300-400	5.988961	0.000218
18	300-400	5.988961	0.000218
19	300-400	5.988961	0.000218
20	200-300	5.631212	0.000152

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## 1.4 Data Exploration & Analysis

```
[ ]: #correlation

numeric_columns = final_dataset.select_dtypes(include=['number'])
print("\n\nCorrelation of final_dataset:\n")
print(numeric_columns.corr())
```

Correlation of final\_dataset:

	month_name	year_num	shop_id	item_id	category_id	\
month_name	1.000000	-0.134131	0.020646	-0.002103	-0.001055	
year_num	-0.134131	1.000000	0.011835	0.010486	0.030406	
shop_id	0.020646	0.011835	1.000000	0.029344	0.019339	
item_id	-0.002103	0.010486	0.029344	1.000000	0.369384	
category_id	-0.001055	0.030406	0.019339	0.369384	1.000000	
item_price	0.044361	0.080574	-0.023966	-0.134759	-0.255033	
item_cnt_month	0.021007	0.002510	-0.005660	0.018896	0.016922	
revenue	0.030326	0.029541	-0.013540	-0.063423	-0.117860	
log_revenue	0.057248	0.096145	-0.039415	-0.310144	-0.342896	
scaled_revenue	0.030326	0.029541	-0.013540	-0.063423	-0.117860	

	item_price	item_cnt_month	revenue	log_revenue	\
month_name	0.044361	0.021007	0.030326	0.057248	
year_num	0.080574	0.002510	0.029541	0.096145	
shop_id	-0.023966	-0.005660	-0.013540	-0.039415	
item_id	-0.134759	0.018896	-0.063423	-0.310144	
category_id	-0.255033	0.016922	-0.117860	-0.342896	
item_price	1.000000	0.014437	0.436314	0.601371	
item_cnt_month	0.014437	1.000000	0.481568	0.115253	
revenue	0.436314	0.481568	1.000000	0.331392	
log_revenue	0.601371	0.115253	0.331392	1.000000	
scaled_revenue	0.436314	0.481568	1.000000	0.331392	

	scaled_revenue
month_name	0.030326
year_num	0.029541
shop_id	-0.013540
item_id	-0.063423
category_id	-0.117860
item_price	0.436314
item_cnt_month	0.481568
revenue	1.000000
log_revenue	0.331392
scaled_revenue	1.000000

```
[ ]: #checks for missing values
print("\n\nMissing values in final_dataset:\n")
print(final_dataset.isnull().sum())

#checks for null values
print("\n\nNull values in final_dataset:\n")
print(final_dataset.isnull().sum())
```

Missing values in final\_dataset:

date	0
date_num	0
month_name	0
year_num	0
shop_id	0
shop_name	0
item_id	0
item_name	0
category_id	0
item_category_name	0
item_price	0
item_cnt_month	0
revenue	0
price_range	0
log_revenue	0
scaled_revenue	0
dtype: int64	

Null values in final\_dataset:

date	0
date_num	0
month_name	0
year_num	0
shop_id	0
shop_name	0
item_id	0
item_name	0
category_id	0
item_category_name	0
item_price	0
item_cnt_month	0
revenue	0
price_range	0
log_revenue	0

```
scaled_revenue      0
dtype: int64
```

```
[ ]: #Descriptive analytics
```

```
# Summary Statistics
print("\nDescriptive statistics of final_dataset:")
print(final_dataset.describe())
```

Descriptive statistics of final\_dataset:

	month_name	year_num	shop_id	item_id	category_id \
count	2.928483e+06	2.928483e+06	2.928483e+06	2.928483e+06	2.928483e+06
mean	6.248408e+00	7.767790e-01	3.300296e+01	1.020028e+04	4.001637e+01
std	3.535921e+00	7.684598e-01	1.622543e+01	6.324391e+03	1.709809e+01
min	1.000000e+00	0.000000e+00	0.000000e+00	0.000000e+00	0.000000e+00
25%	3.000000e+00	0.000000e+00	2.200000e+01	4.477000e+03	2.800000e+01
50%	6.000000e+00	1.000000e+00	3.100000e+01	9.355000e+03	4.000000e+01
75%	9.000000e+00	1.000000e+00	4.700000e+01	1.569100e+04	5.500000e+01
max	1.200000e+01	2.000000e+00	5.900000e+01	2.216900e+04	8.300000e+01

	item_price	item_cnt_month	revenue	log_revenue \
count	2.928483e+06	2.928483e+06	2.928483e+06	2.928483e+06
mean	8.893627e+02	1.247257e+00	1.164267e+03	6.254676e+00
std	1.718155e+03	2.217429e+00	5.684853e+03	1.171779e+00
min	7.000000e-02	1.000000e+00	7.000000e-02	-2.659260e+00
25%	2.490000e+02	1.000000e+00	2.490000e+02	5.517453e+00
50%	3.990000e+02	1.000000e+00	4.490000e+02	6.107023e+00
75%	9.990000e+02	1.000000e+00	1.090000e+03	6.993933e+00
max	5.920000e+04	6.690000e+02	1.829990e+06	1.441982e+01

	scaled_revenue
count	2.928483e+06
mean	6.361770e-04
std	3.106494e-03
min	0.000000e+00
25%	1.360281e-04
50%	2.453183e-04
75%	5.955934e-04
max	1.000000e+00

```
[ ]: #seasonality analysis
```

```
grouped_by_month_name = final_dataset.groupby(['month_name']).
    ↪agg({'item_cnt_month': 'sum'})

print("\n\nHead of grouped_by_month_name:\n")
print(grouped_by_month_name)
```

```
print(grouped_by_month_name.shape)
```

Head of grouped\_by\_month\_name:

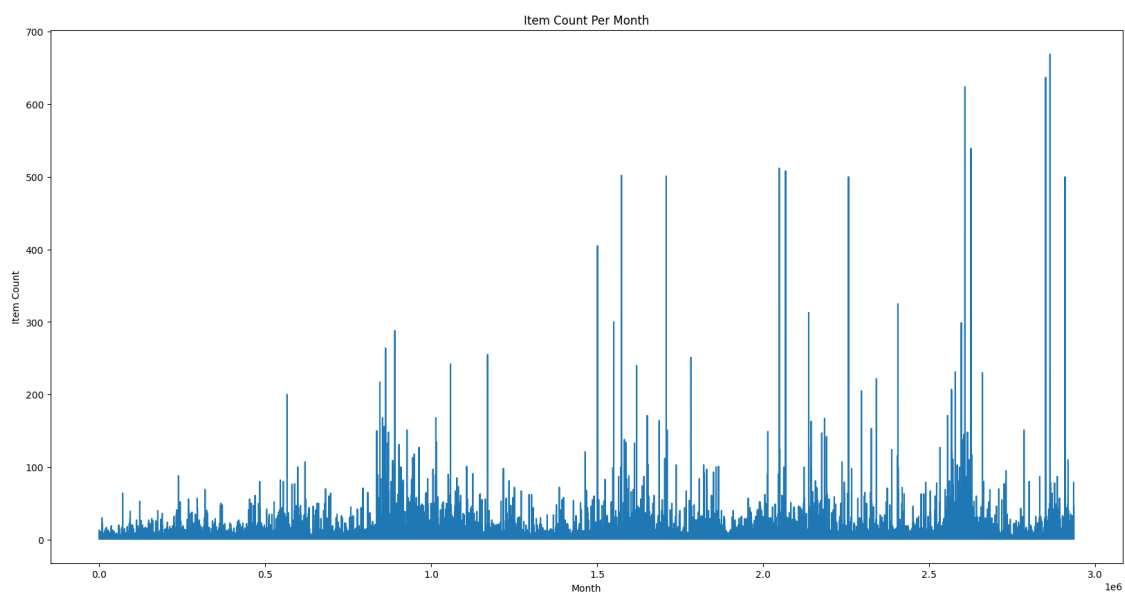
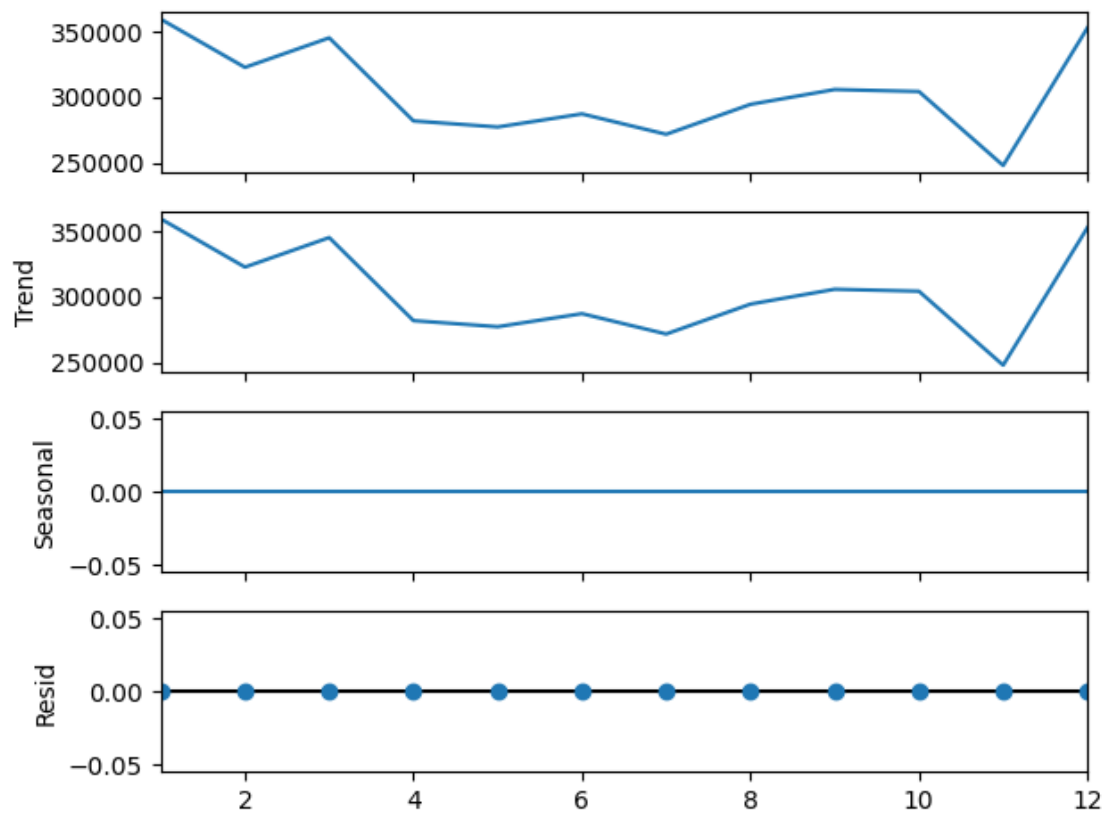
month_name	item_cnt_month
1	359391
2	322775
3	345171
4	282131
5	277519
6	287443
7	271985
8	294692
9	305967
10	304453
11	248320
12	352723

(12, 1)

```
[ ]: #performing seasonal decomposition
decomposition = sm.tsa.seasonal_decompose(grouped_by_month_name,
    ↪model='additive', period=1)

#plotting the seasonal decomposition
fig = decomposition.plot()
plt.show()

#plotting the item_cnt_month column
plt.figure(figsize=(20, 10))
plt.plot(final_dataset['item_cnt_month'])
plt.title('Item Count Per Month')
plt.xlabel('Month')
plt.ylabel('Item Count')
plt.show()
```



```
[ ]: #regulatory analytics

grouped_by_shop_id_and_year_num = final_dataset.groupby(['shop_id',
↳ 'year_num']).agg({'item_cnt_month': 'sum'})

print("\n\nHead of grouped_by_shop_id_and_year_num:\n")
print(grouped_by_shop_id_and_year_num.head(60))
```

Head of grouped\_by\_shop\_id\_and\_year\_num:

shop_id	year_num	item_cnt_month
0	0	11705
1	0	6311
2	0	9989
	1	12247
	2	8470
3	0	10242
	1	11039
	2	7194
4	0	19054
	1	15909
	2	9106
5	0	14717
	1	17041
	2	11089
6	0	46707
	1	35496
	2	18554
7	0	28218
	1	24530
	2	14411
8	0	3602
9	0	6531
	1	6155
	2	3188
10	0	11132
	1	8859
	2	4532
11	2	572
12	0	19494
	1	26379
	2	24947
13	0	13529
	1	6234
14	0	20114

	1	16003
	2	10397
15	0	32339
	1	24854
	2	14138
16	0	30250
	1	19569
	2	11953
17	0	9913
	1	15308
	2	697
18	0	33632
	1	18815
	2	13174
19	0	32860
	1	25353
	2	15455
20	1	3262
	2	2611
21	0	25258
	1	25167
	2	18259
22	0	26943
	1	20341
	2	13050
23	0	7722

```
[ ]: #Variable Identification
```

```
# Identify numerical and categorical variables
numerical_vars = final_dataset.select_dtypes(include=['int64', 'float64']).
    ↪columns
categorical_vars = final_dataset.select_dtypes(include=['object', 'category']).
    ↪columns

# Print the list of numerical and categorical variables
print("Numerical Variables:")
print(numerical_vars)

print("\nCategorical Variables:")
print(categorical_vars)
```

Numerical Variables:

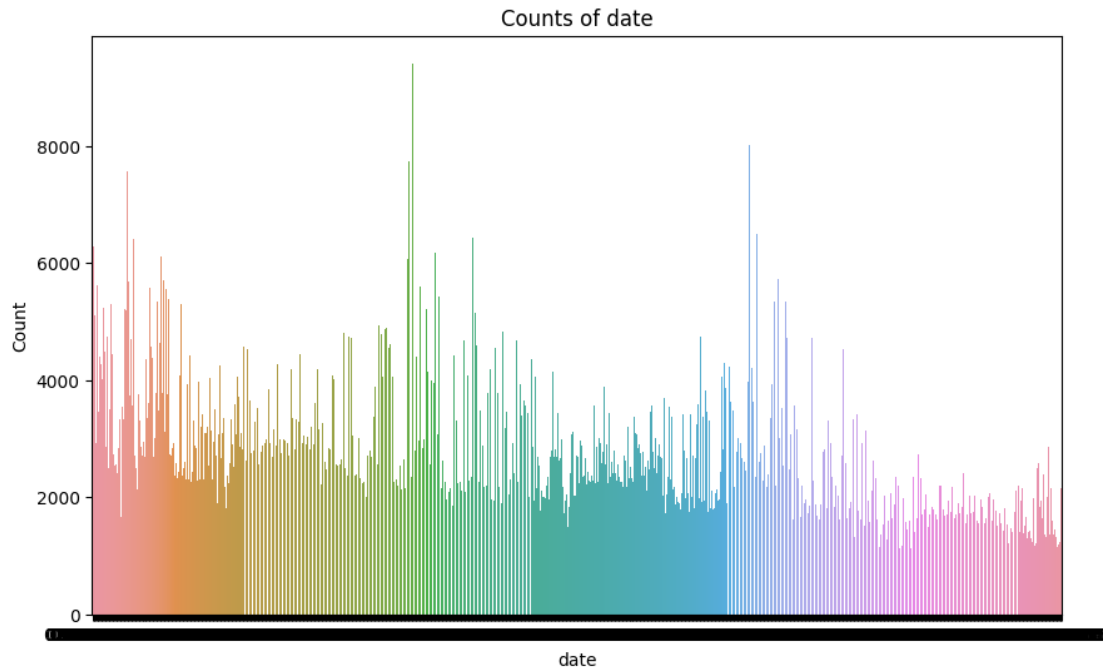
```
Index(['month_name', 'year_num', 'shop_id', 'item_id', 'category_id',
      'item_price', 'item_cnt_month', 'revenue', 'log_revenue',
      'scaled_revenue'],
      dtype='object')
```

Categorical Variables:

```
Index(['date', 'date_num', 'shop_name', 'item_name', 'item_category_name',  
      'price_range'],  
      dtype='object')
```

```
[ ]: # univariate analysis  
  
for column in final_dataset.columns:  
    variable_type = final_dataset[column].dtype  
  
    summary_stats = final_dataset[column].describe()  
  
    plt.figure(figsize=(10, 6))  
  
    # For numerical variables, create a histogram  
    if variable_type in ['int64', 'float64']:  
        sns.histplot(data=final_dataset, x=column, kde=True)  
        plt.title(f'Distribution of {column}')  
        plt.xlabel(column)  
        plt.ylabel('Frequency')  
  
    # For categorical variables, create a bar plot  
    else:  
        sns.countplot(data=final_dataset, x=column)  
        plt.title(f'Counts of {column}')  
        plt.xlabel(column)  
        plt.ylabel('Count')  
  
    plt.show()  
  
    # Print summary statistics  
    print(f"Summary Statistics for {column}:")  
    print(summary_stats)
```

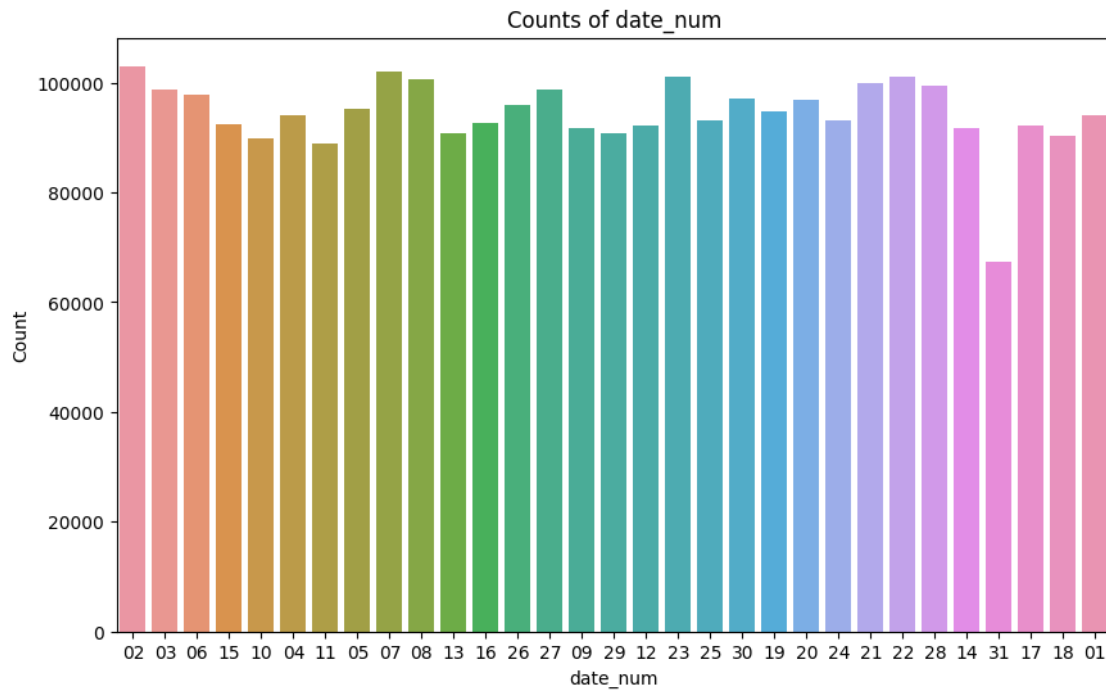




Summary Statistics for date:

count	2928483
unique	1034
top	28.12.2013
freq	9415

Name: date, dtype: object



Summary Statistics for date\_num:

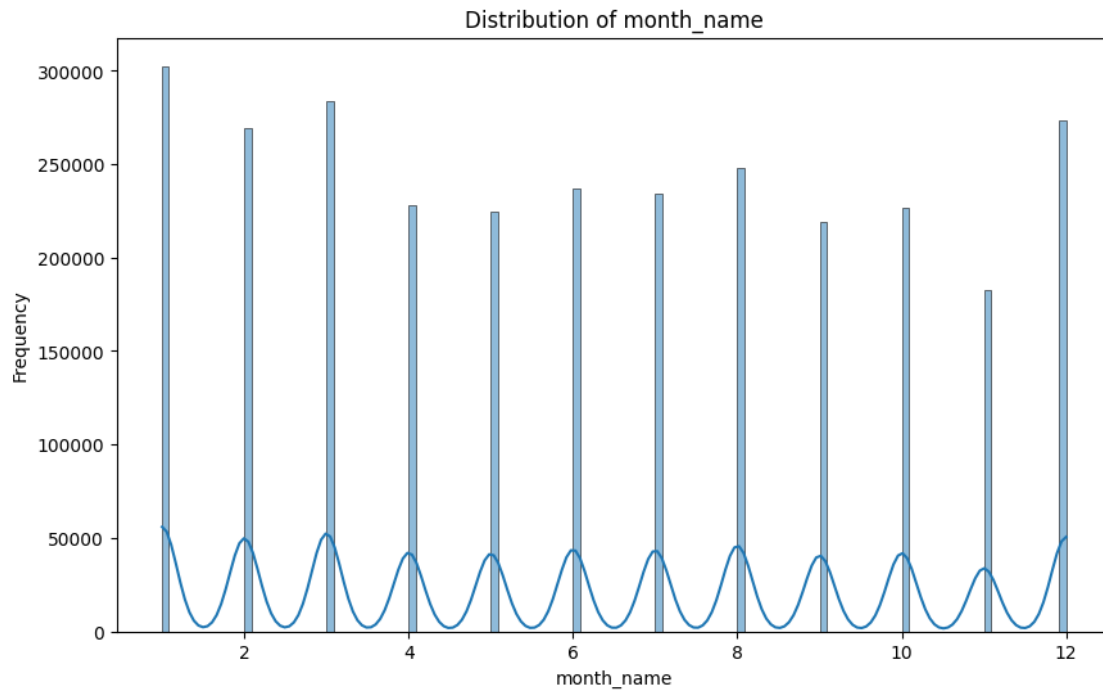
count 2928483

unique 31

top 02

freq 103081

Name: date\_num, dtype: object



Summary Statistics for month\_name:

count 2.928483e+06

mean 6.248408e+00

std 3.535921e+00

min 1.000000e+00

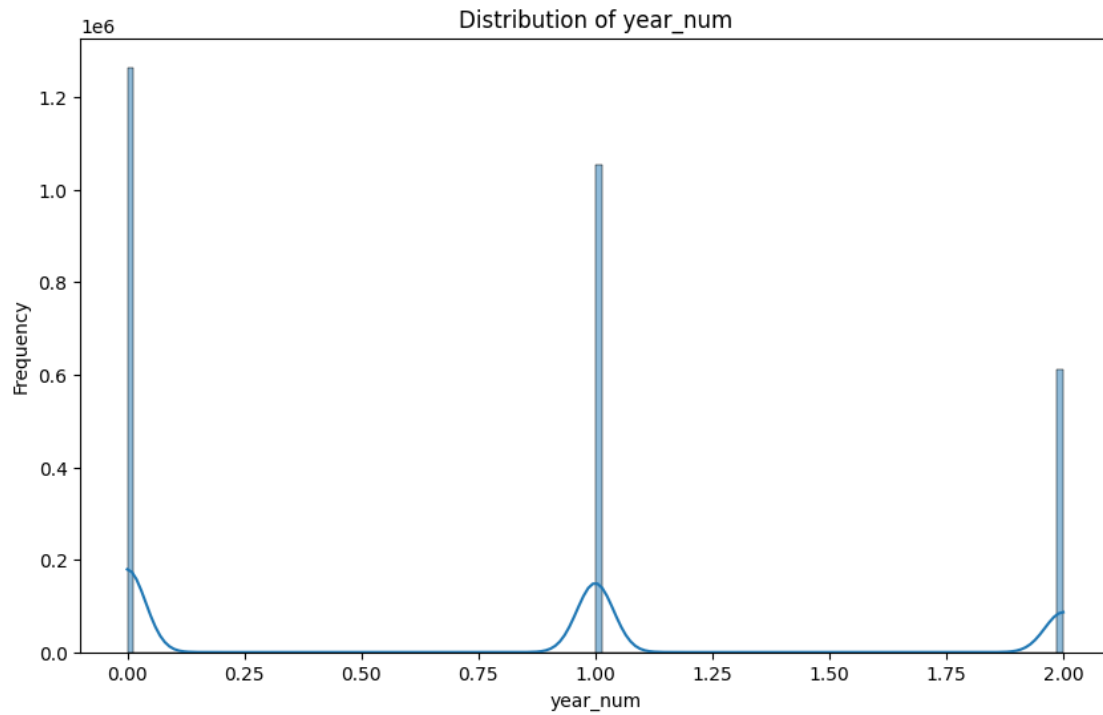
25% 3.000000e+00

50% 6.000000e+00

75% 9.000000e+00

max 1.200000e+01

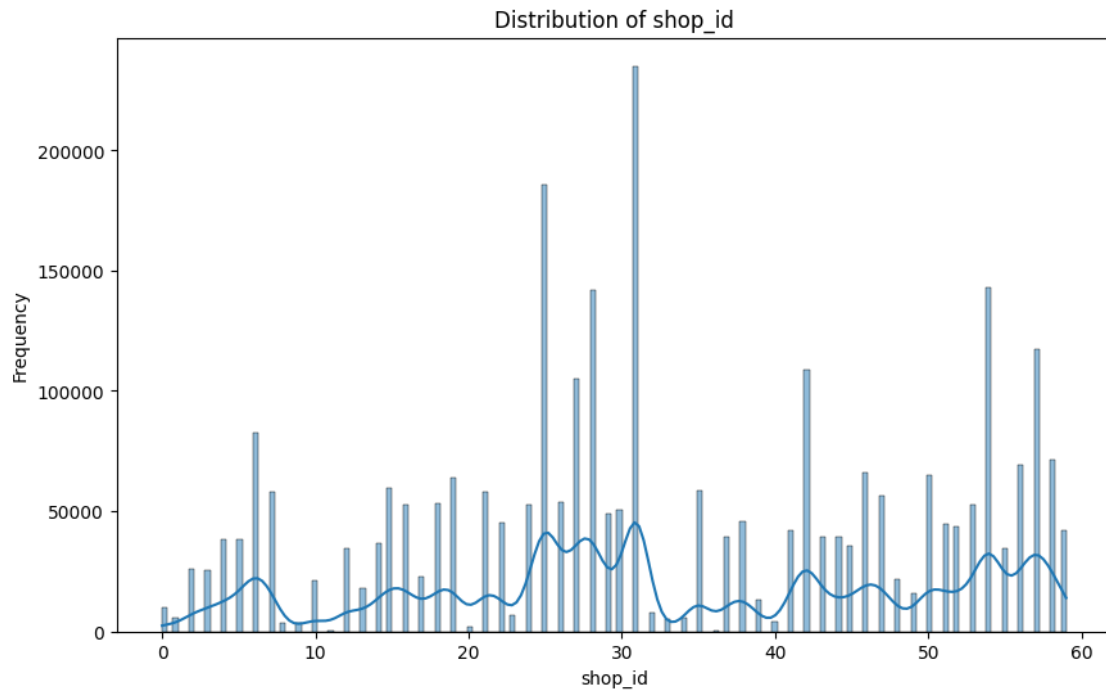
Name: month\_name, dtype: float64



Summary Statistics for year\_num:

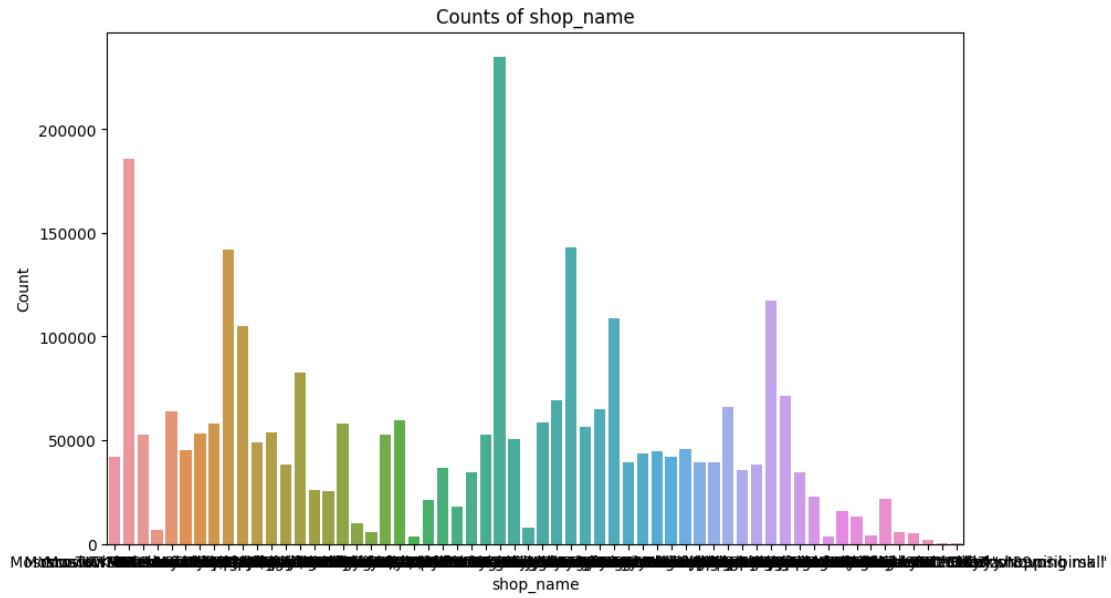
count	2.928483e+06
mean	7.767790e-01
std	7.684598e-01
min	0.000000e+00
25%	0.000000e+00
50%	1.000000e+00
75%	1.000000e+00
max	2.000000e+00

Name: year\_num, dtype: float64

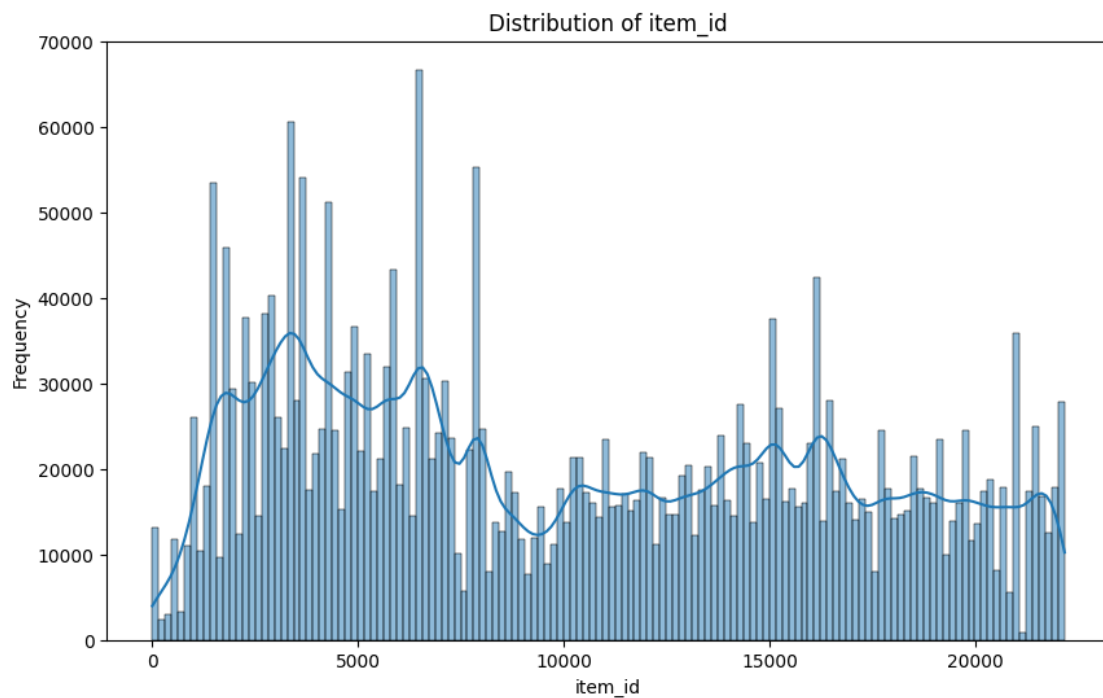


Summary Statistics for shop\_id:

```
count    2.928483e+06
mean      3.300296e+01
std       1.622543e+01
min       0.000000e+00
25%       2.200000e+01
50%       3.100000e+01
75%       4.700000e+01
max       5.900000e+01
Name: shop_id, dtype: float64
```



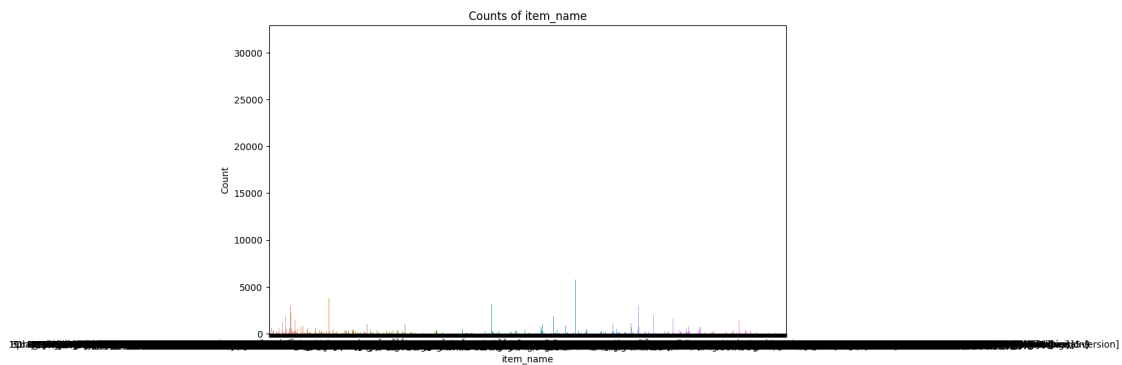
```
Summary Statistics for shop_name:
count          2928483
unique           60
top    Moscow TC" Semenovskiy "
freq          235185
Name: shop_name, dtype: object
```



Summary Statistics for item\_id:

count 2.928483e+06  
mean 1.020028e+04  
std 6.324391e+03  
min 0.000000e+00  
25% 4.477000e+03  
50% 9.355000e+03  
75% 1.569100e+04  
max 2.216900e+04

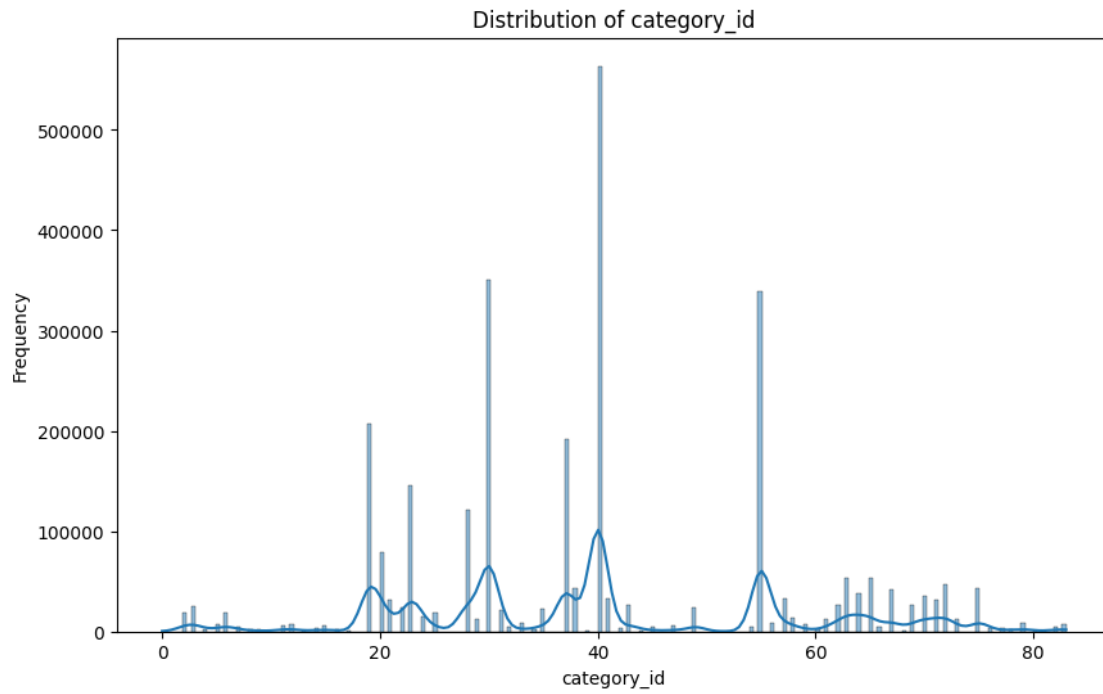
Name: item\_id, dtype: float64



Summary Statistics for item\_name:

count 2928483  
unique 21782  
top Corporate package T-shirt 1C Interest white (3..  
freq 31336

Name: item\_name, dtype: object

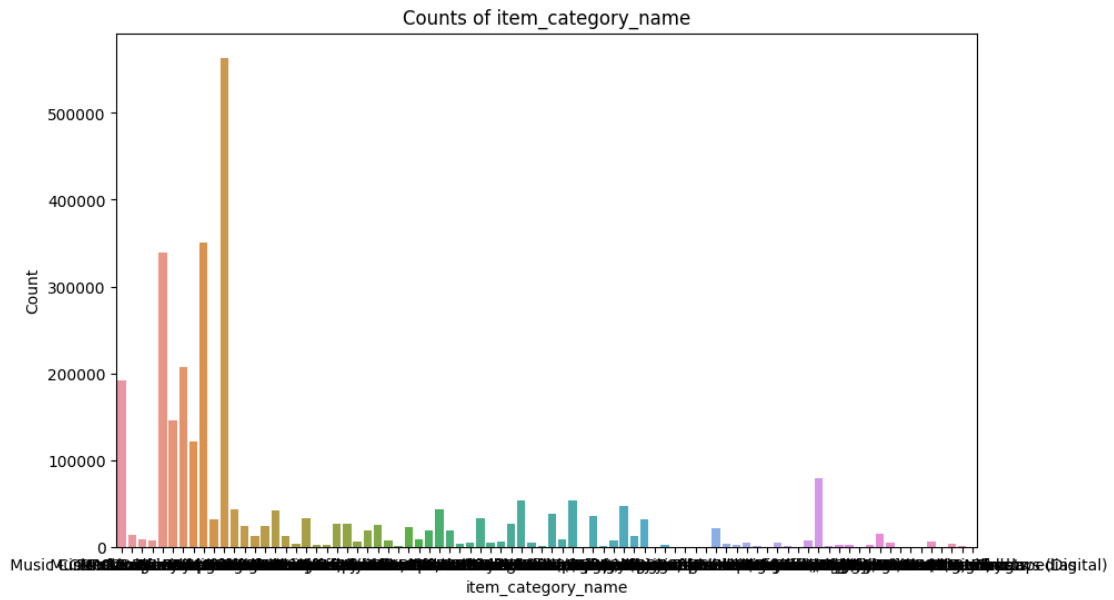


Summary Statistics for category\_id:

count	2.928483e+06
mean	4.001637e+01
std	1.709809e+01
min	0.000000e+00
25%	2.800000e+01
50%	4.000000e+01
75%	5.500000e+01
max	8.300000e+01

Name: category\_id, dtype: float64





Summary Statistics for item\_category\_name:

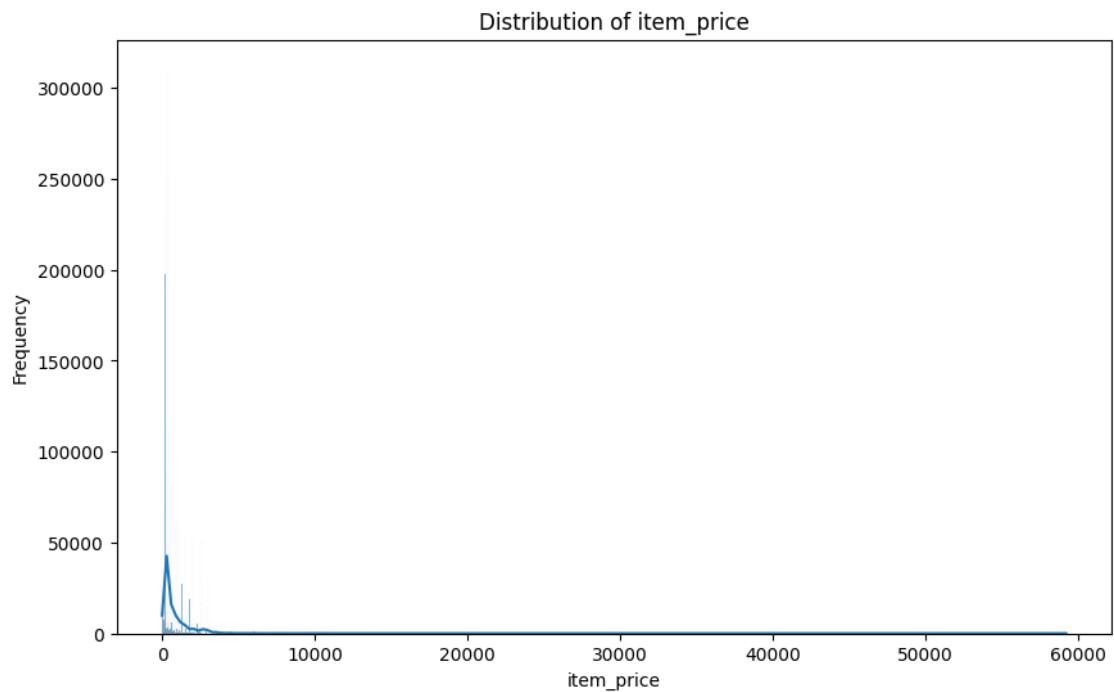
count 2928483

unique 84

top Cinema - DVD

freq 563937

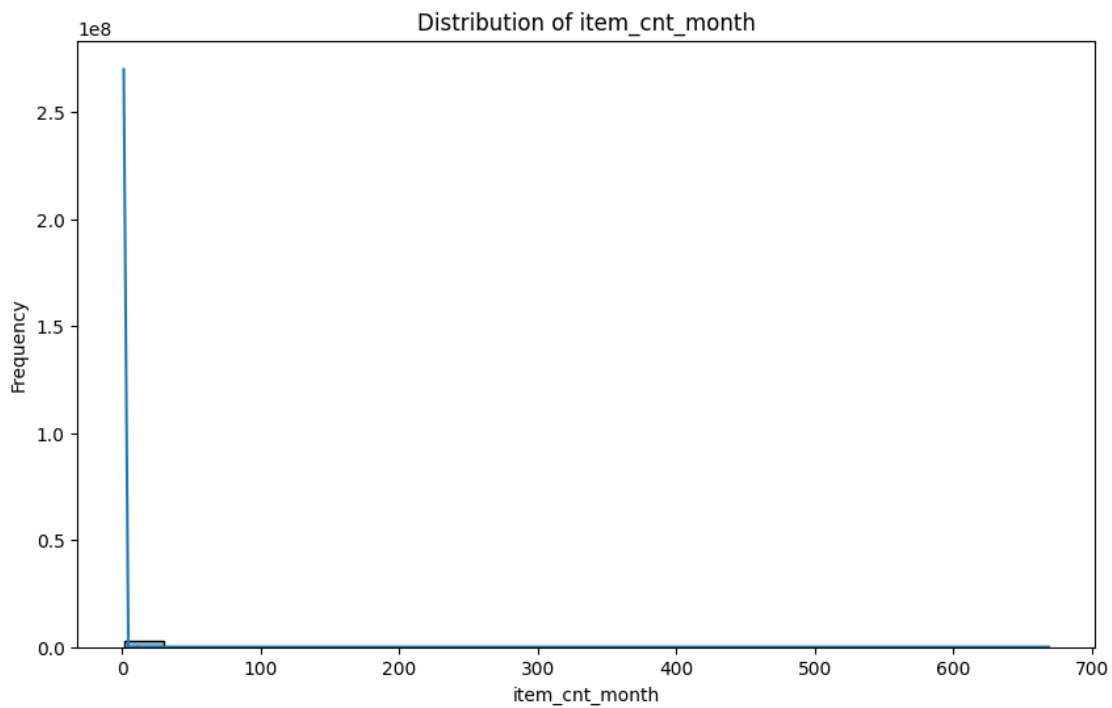
Name: item\_category\_name, dtype: object



Summary Statistics for item\_price:

count	2.928483e+06
mean	8.893627e+02
std	1.718155e+03
min	7.000000e-02
25%	2.490000e+02
50%	3.990000e+02
75%	9.990000e+02
max	5.920000e+04

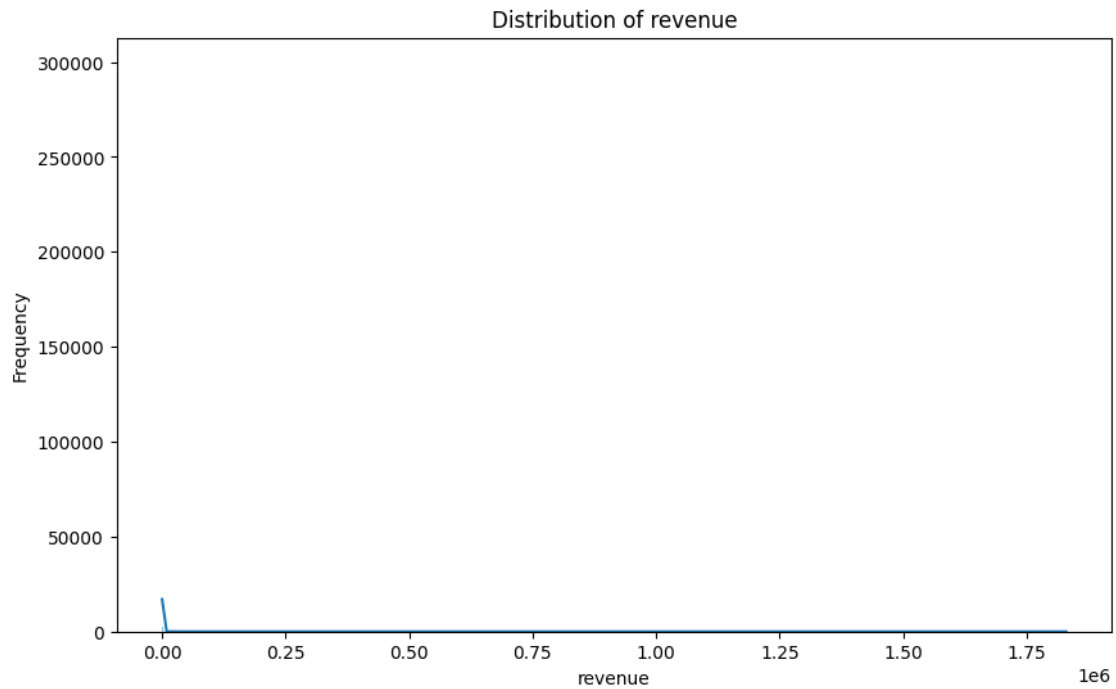
Name: item\_price, dtype: float64



Summary Statistics for item\_cnt\_month:

count	2.928483e+06
mean	1.247257e+00
std	2.217429e+00
min	1.000000e+00
25%	1.000000e+00
50%	1.000000e+00
75%	1.000000e+00
max	6.690000e+02

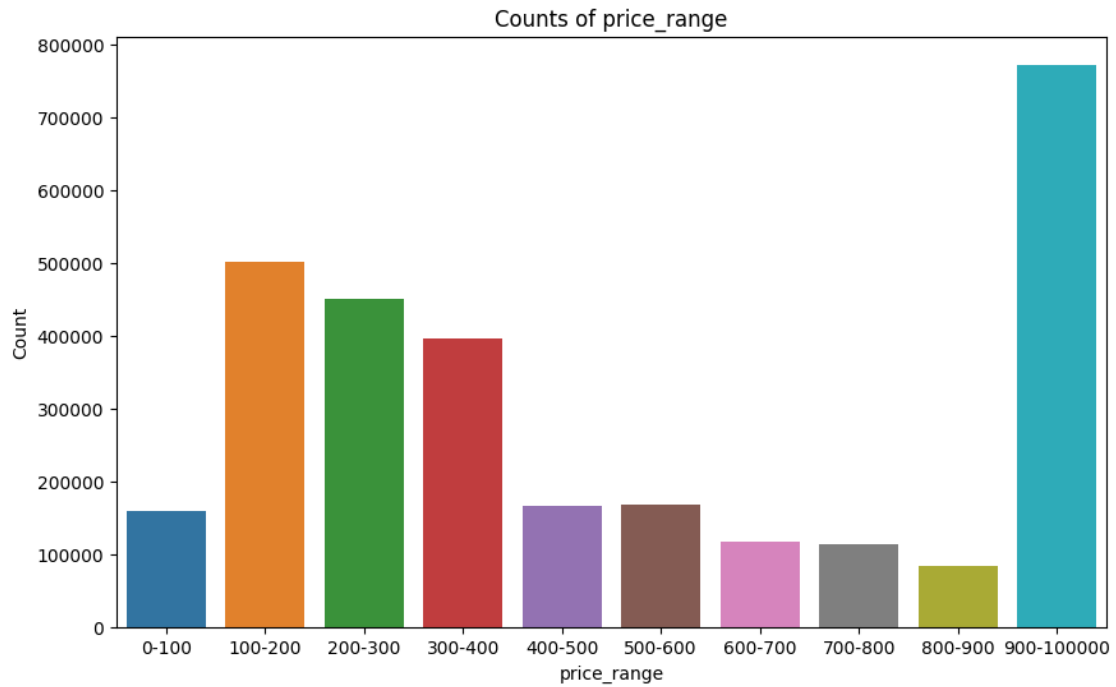
Name: item\_cnt\_month, dtype: float64



Summary Statistics for revenue:

count	2.928483e+06
mean	1.164267e+03
std	5.684853e+03
min	7.000000e-02
25%	2.490000e+02
50%	4.490000e+02
75%	1.090000e+03
max	1.829990e+06

Name: revenue, dtype: float64



Summary Statistics for price\_range:

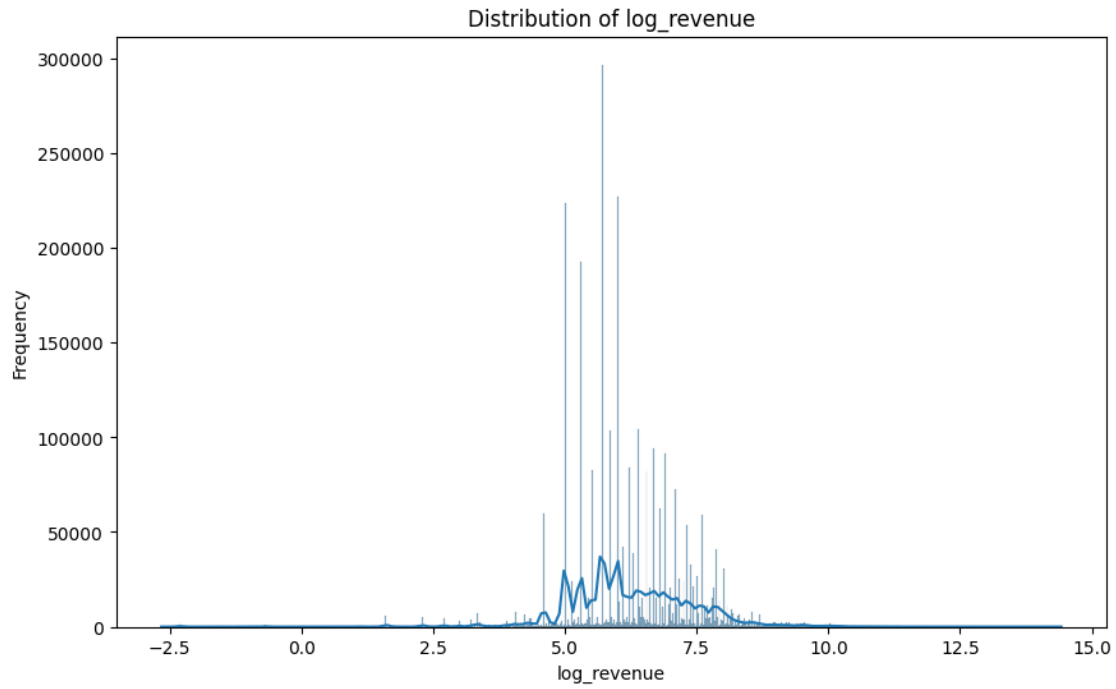
count 2928483

unique 10

top 900-100000

freq 772112

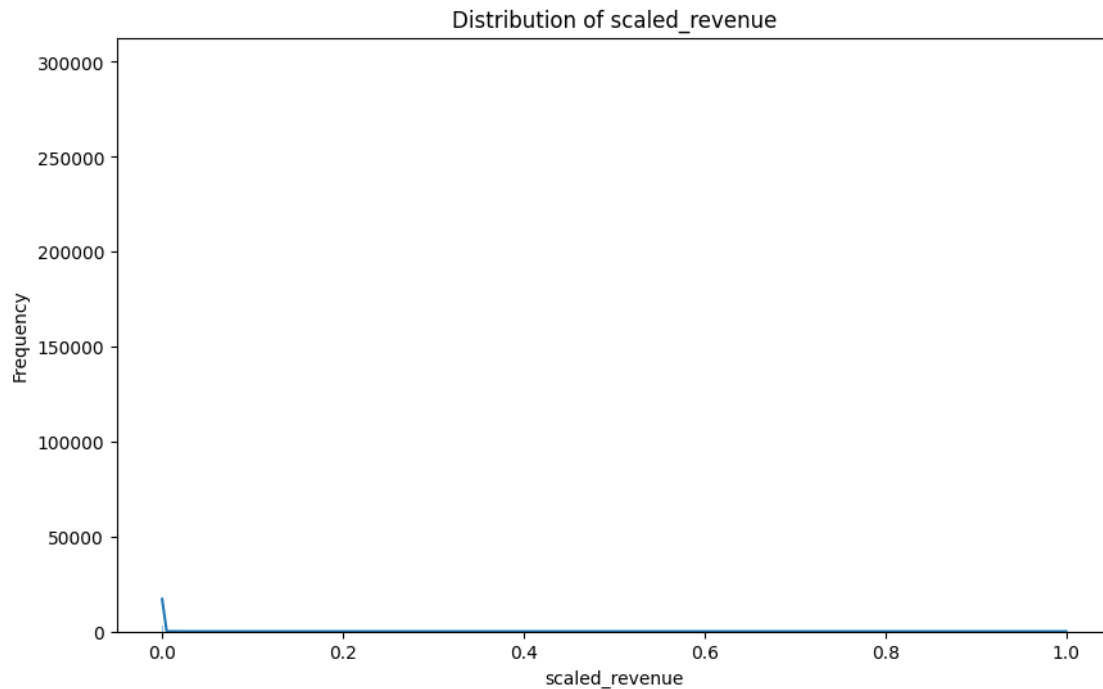
Name: price\_range, dtype: object



Summary Statistics for log\_revenue:

count	2.928483e+06
mean	6.254676e+00
std	1.171779e+00
min	-2.659260e+00
25%	5.517453e+00
50%	6.107023e+00
75%	6.993933e+00
max	1.441982e+01

Name: log\_revenue, dtype: float64



Summary Statistics for scaled\_revenue:

count	2.928483e+06
mean	6.361770e-04
std	3.106494e-03
min	0.000000e+00
25%	1.360281e-04
50%	2.453183e-04
75%	5.955934e-04
max	1.000000e+00

Name: scaled\_revenue, dtype: float64

```
[ ]: #bivariate analysis

#can analysis by changing var1 and var2
var1 = 'item_price'
var2 = 'item_cnt_month'

var1_type = final_dataset[var1].dtype
var2_type = final_dataset[var2].dtype

# Scatter Plot for Numerical vs. Numerical
if var1_type in ['int64', 'float64'] and var2_type in ['int64', 'float64']:
    plt.figure(figsize=(10, 6))
    sns.scatterplot(data=final_dataset, x=var1, y=var2)
    plt.title(f'Scatter Plot: {var1} vs. {var2}')
```

```

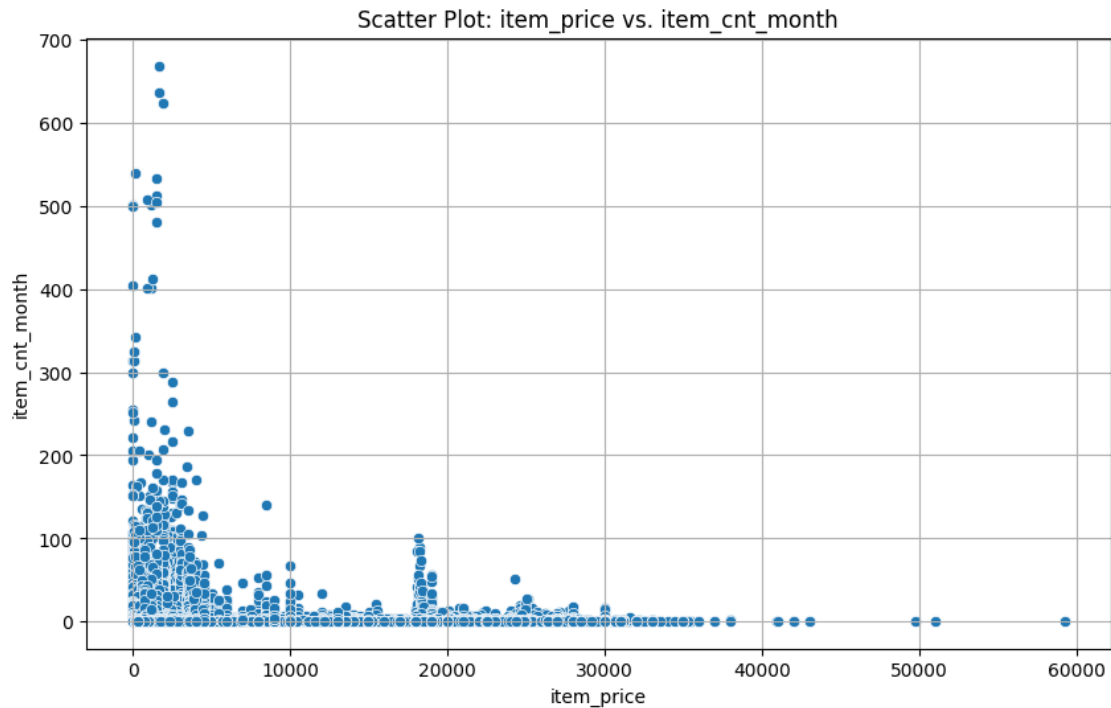
plt.xlabel(var1)
plt.ylabel(var2)
plt.grid(True)
plt.show()

# Box Plot for Categorical vs. Numerical
elif var1_type in ['object', 'category'] and var2_type in ['int64', 'float64']:
    plt.figure(figsize=(10, 6))
    sns.boxplot(data=final_dataset, x=var1, y=var2)
    plt.title(f'Box Plot: {var1} vs. {var2}')
    plt.xlabel(var1)
    plt.ylabel(var2)
    plt.grid(True)
    plt.show()

# Bar Plot for Categorical vs. Categorical
elif var1_type in ['object', 'category'] and var2_type in ['object', 'category']:
    crosstab = pd.crosstab(final_dataset[var1], final_dataset[var2])
    crosstab.plot(kind='bar', stacked=True, figsize=(10, 6))
    plt.title(f'Bar Plot: {var1} vs. {var2}')
    plt.xlabel(var1)
    plt.ylabel('Count')
    plt.grid(True)
    plt.show()

# Print correlation for Numerical vs. Numerical
if var1_type in ['int64', 'float64'] and var2_type in ['int64', 'float64']:
    correlation = final_dataset[[var1, var2]].corr().iloc[0, 1]
    print(f'Correlation between {var1} and {var2}: {correlation:.2f}')

```



Correlation between item\_price and item\_cnt\_month: 0.01

```
[ ]: #Exploratory Data Analysis (EDA)

print("Dataset Overview:")
print(final_dataset.info())

print("\nSummary Statistics for Numerical Variables:")
print(final_dataset.describe())

print("\nMissing Values:")
print(final_dataset.isnull().sum())

numerical_columns = ['month_name', 'year_num', 'shop_id', 'item_id',
                    ↪ 'category_id', 'item_price', 'item_cnt_month', 'revenue', 'log_revenue',
                    ↪ 'scaled_revenue']

for column in numerical_columns:
    plt.figure(figsize=(8, 4))
    sns.histplot(data=final_dataset, x=column, kde=True, bins=20)
    plt.title(f'Distribution of {column}')
    plt.xlabel(column)
    plt.ylabel('Frequency')
    plt.show()
```



```

# Visualize relationships between variables with a correlation matrix for
↳ numerical variables
correlation_matrix = final_dataset[numerical_columns].corr()
plt.figure(figsize=(10, 8))
sns.heatmap(correlation_matrix, annot=True, cmap='coolwarm', fmt=".2f")
plt.title("Correlation Heatmap for Numerical Variables")
plt.show()

# Explore categorical variables with bar plots
categorical_columns = ['shop_name', 'item_name', 'item_category_name',
↳ 'price_range']

for column in categorical_columns:
    plt.figure(figsize=(10, 6))
    sns.countplot(data=final_dataset, x=column)
    plt.title(f'Counts of {column}')
    plt.xlabel(column)
    plt.ylabel('Count')
    plt.xticks(rotation=90)
    plt.show()

```

Dataset Overview:

<class 'pandas.core.frame.DataFrame'>

Index: 2928483 entries, 0 to 2935848

Data columns (total 16 columns):

#	Column	Dtype
0	date	object
1	date_num	object
2	month_name	int64
3	year_num	int64
4	shop_id	int64
5	shop_name	object
6	item_id	int64
7	item_name	object
8	category_id	int64
9	item_category_name	object
10	item_price	float64
11	item_cnt_month	int64
12	revenue	float64
13	price_range	category
14	log_revenue	float64
15	scaled_revenue	float64

dtypes: category(1), float64(4), int64(6), object(5)

memory usage: 360.3+ MB

None

Summary Statistics for Numerical Variables:

	month_name	year_num	shop_id	item_id	category_id \
count	2.928483e+06	2.928483e+06	2.928483e+06	2.928483e+06	2.928483e+06
mean	6.248408e+00	7.767790e-01	3.300296e+01	1.020028e+04	4.001637e+01
std	3.535921e+00	7.684598e-01	1.622543e+01	6.324391e+03	1.709809e+01
min	1.000000e+00	0.000000e+00	0.000000e+00	0.000000e+00	0.000000e+00
25%	3.000000e+00	0.000000e+00	2.200000e+01	4.477000e+03	2.800000e+01
50%	6.000000e+00	1.000000e+00	3.100000e+01	9.355000e+03	4.000000e+01
75%	9.000000e+00	1.000000e+00	4.700000e+01	1.569100e+04	5.500000e+01
max	1.200000e+01	2.000000e+00	5.900000e+01	2.216900e+04	8.300000e+01

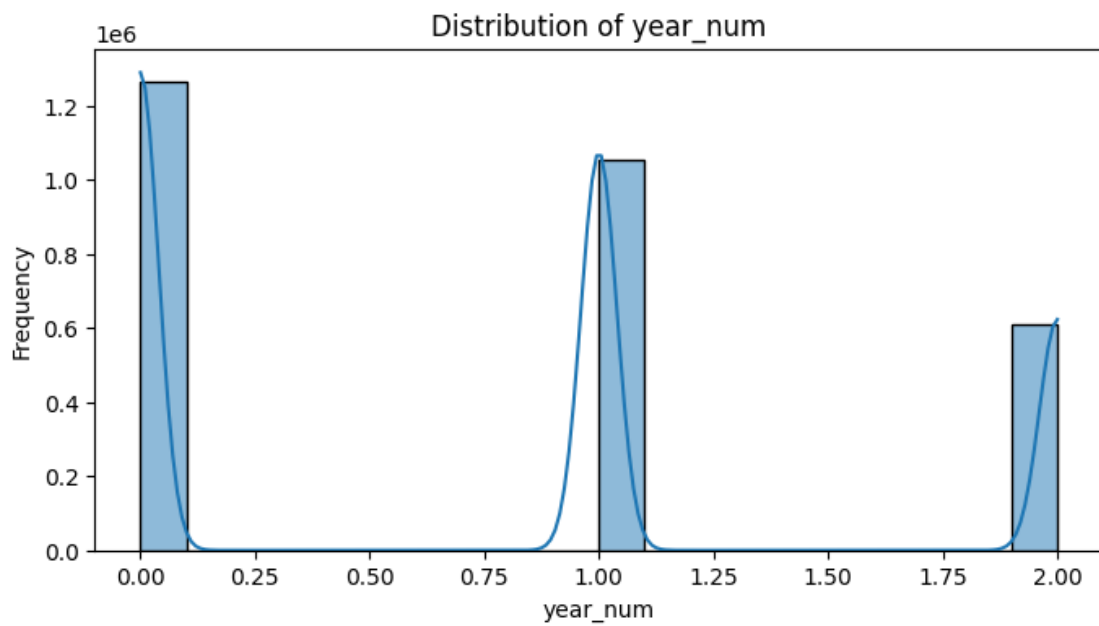
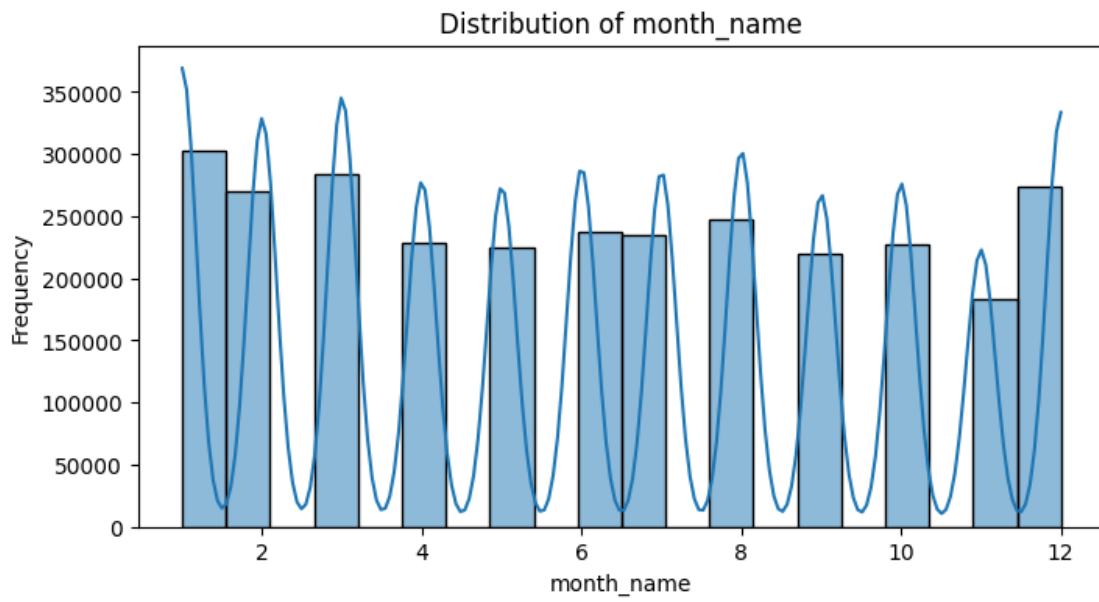
	item_price	item_cnt_month	revenue	log_revenue \
count	2.928483e+06	2.928483e+06	2.928483e+06	2.928483e+06
mean	8.893627e+02	1.247257e+00	1.164267e+03	6.254676e+00
std	1.718155e+03	2.217429e+00	5.684853e+03	1.171779e+00
min	7.000000e-02	1.000000e+00	7.000000e-02	-2.659260e+00
25%	2.490000e+02	1.000000e+00	2.490000e+02	5.517453e+00
50%	3.990000e+02	1.000000e+00	4.490000e+02	6.107023e+00
75%	9.990000e+02	1.000000e+00	1.090000e+03	6.993933e+00
max	5.920000e+04	6.690000e+02	1.829990e+06	1.441982e+01

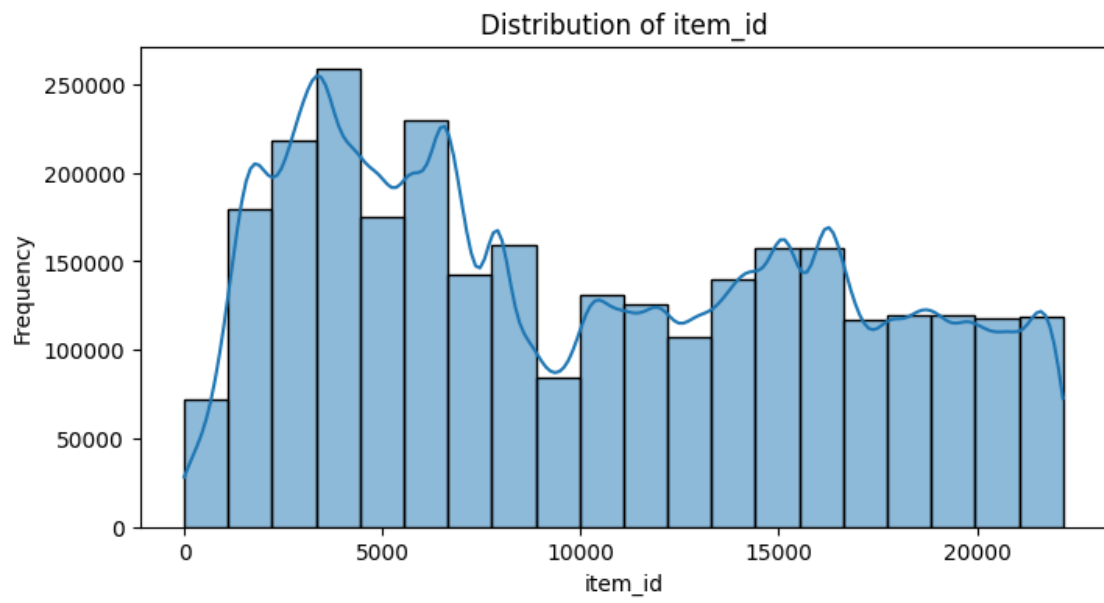
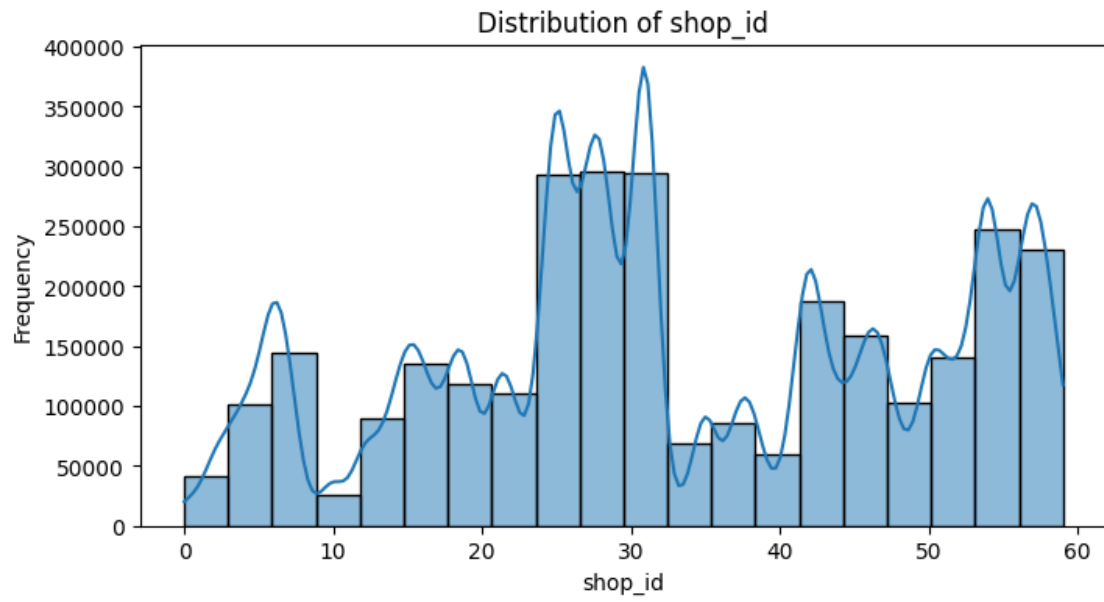
	scaled_revenue
count	2.928483e+06
mean	6.361770e-04
std	3.106494e-03
min	0.000000e+00
25%	1.360281e-04
50%	2.453183e-04
75%	5.955934e-04
max	1.000000e+00

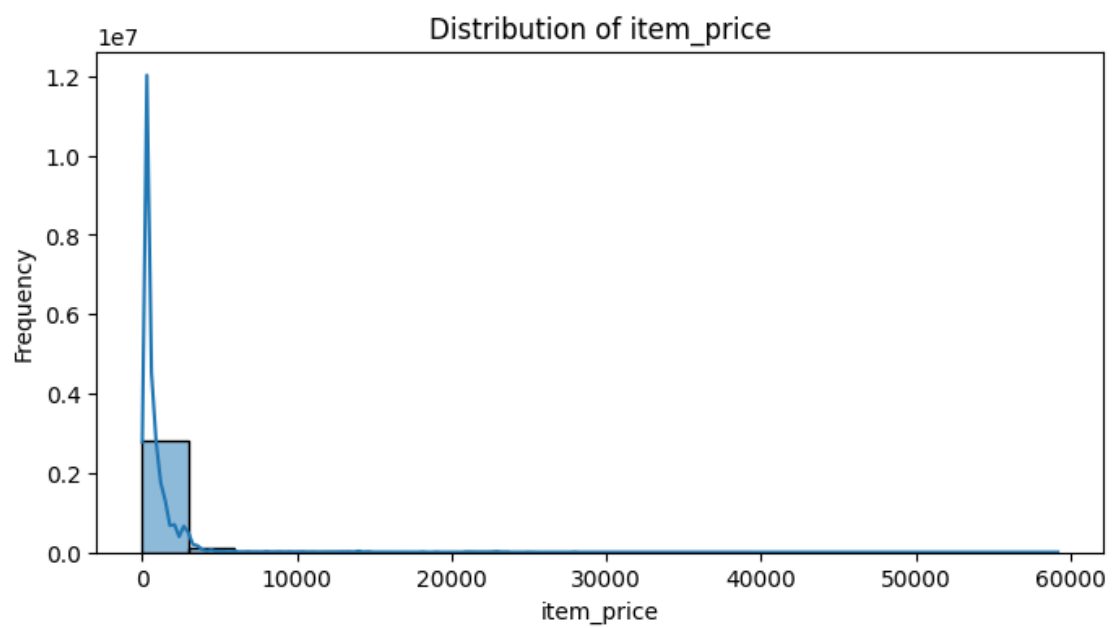
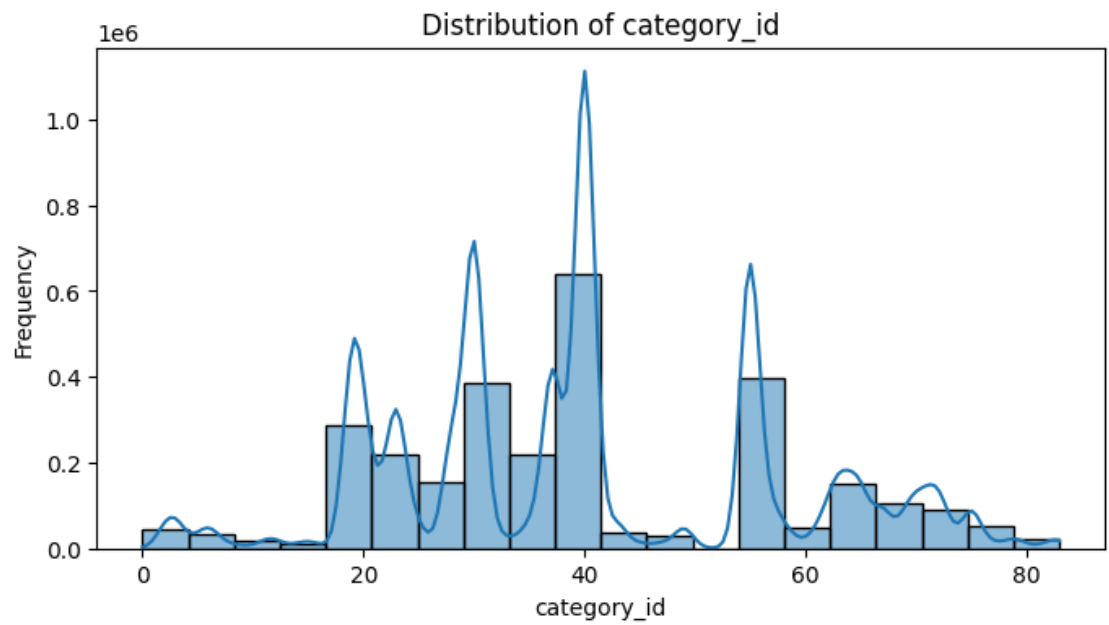
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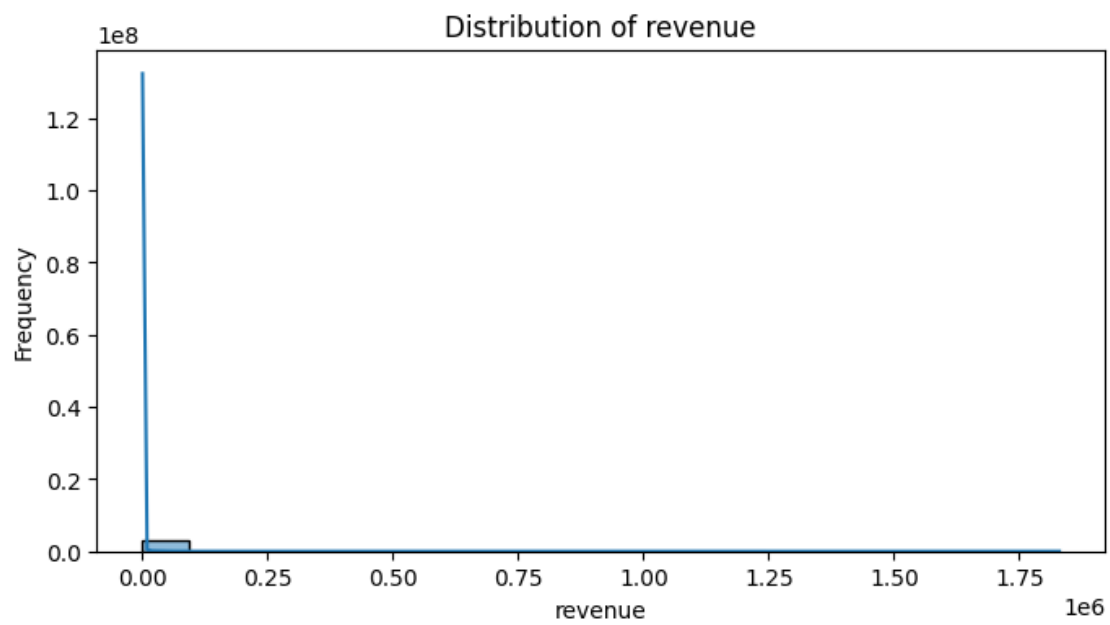
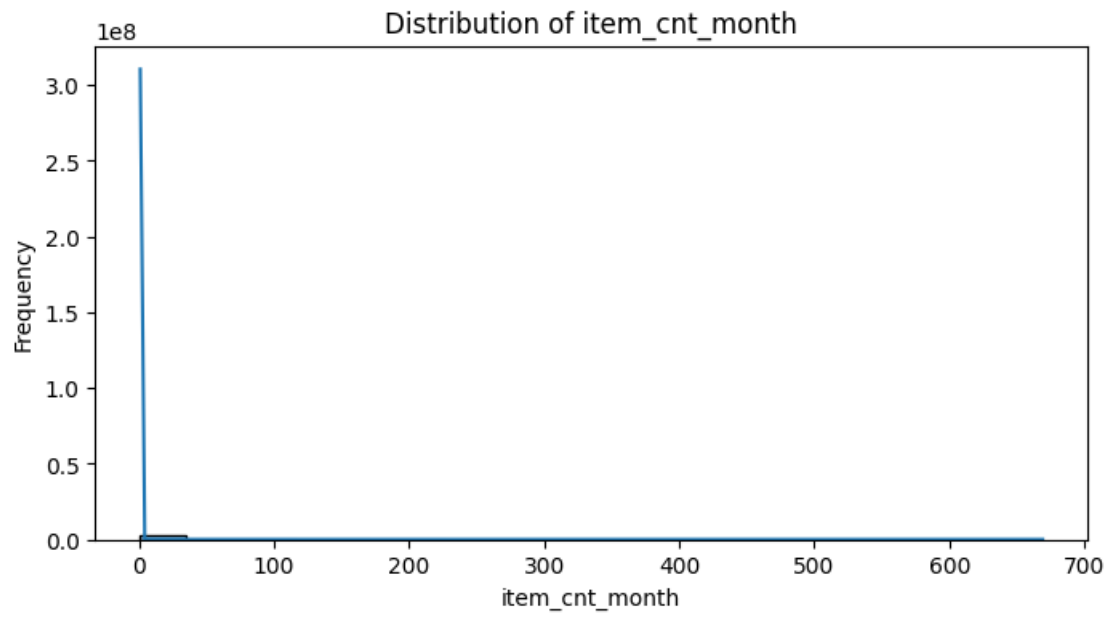
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date_num	0
month_name	0
year_num	0
shop_id	0
shop_name	0
item_id	0
item_name	0
category_id	0
item_category_name	0
item_price	0
item_cnt_month	0
revenue	0
price_range	0
log_revenue	0

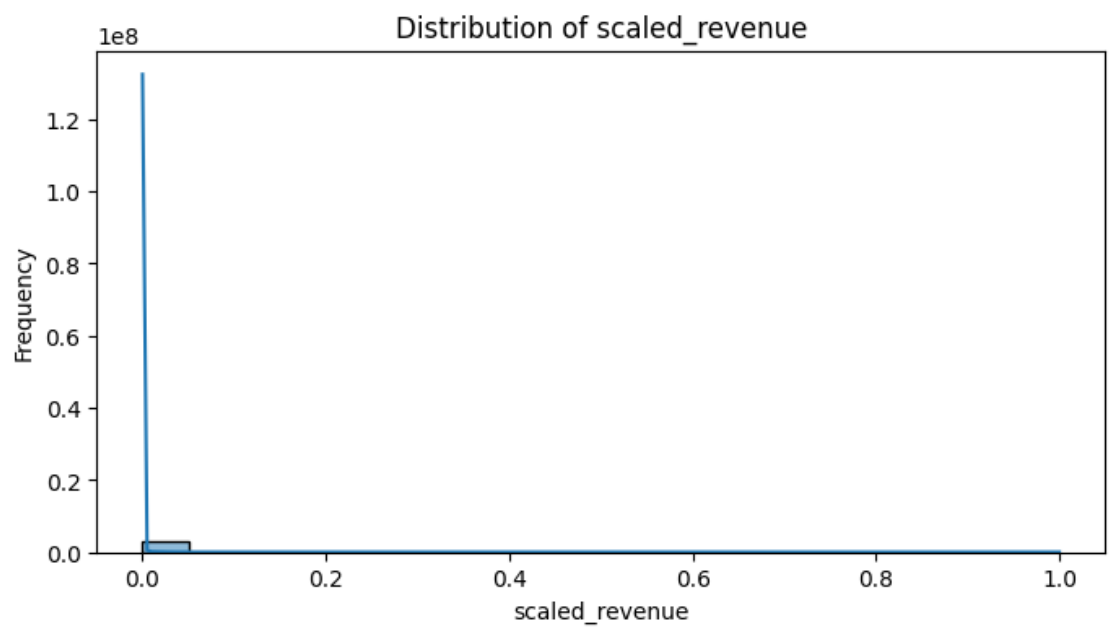
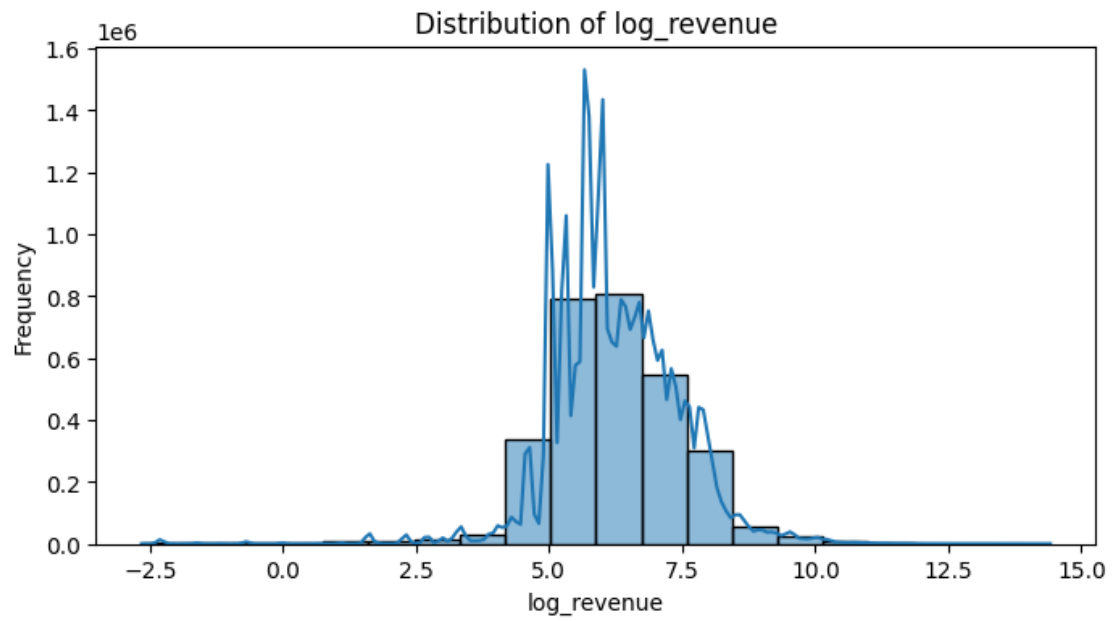
```
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dtype: int64
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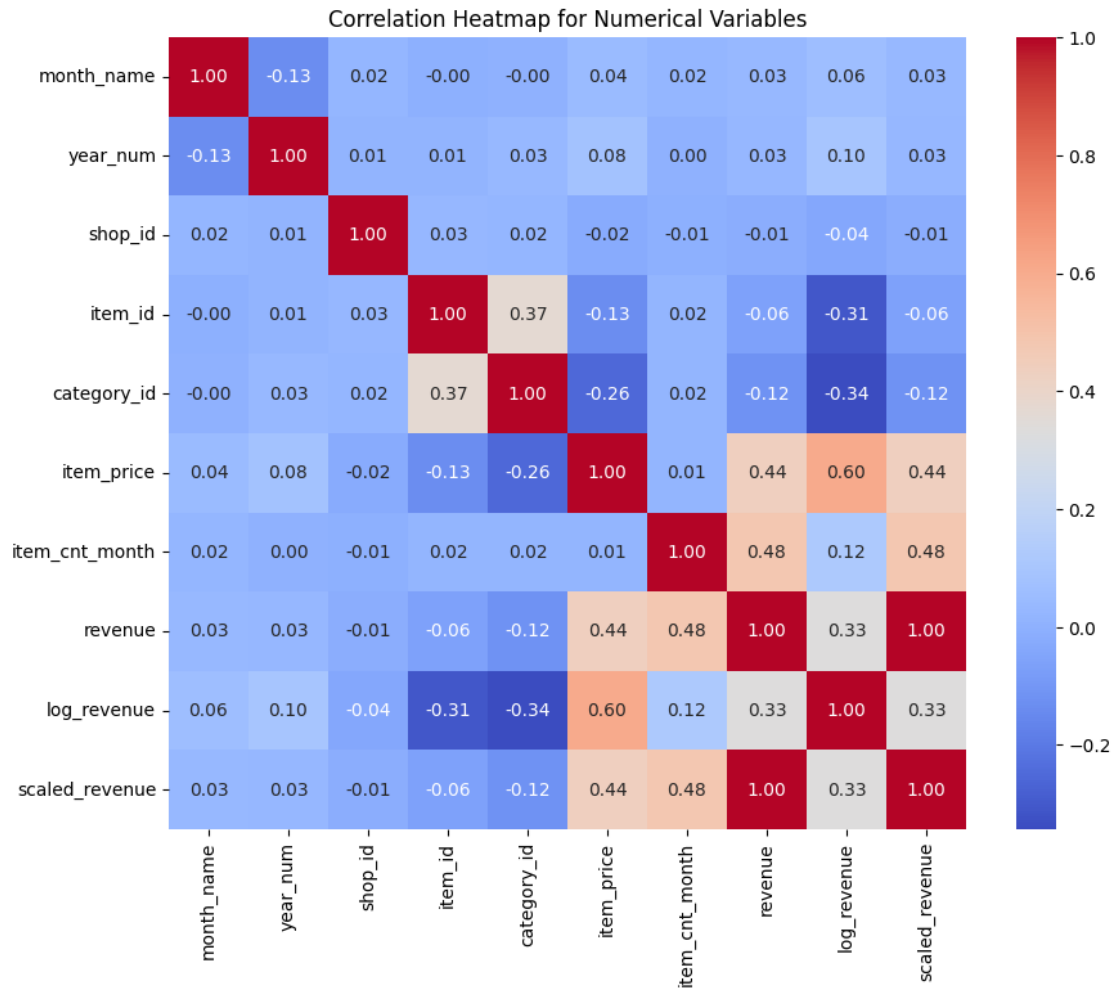




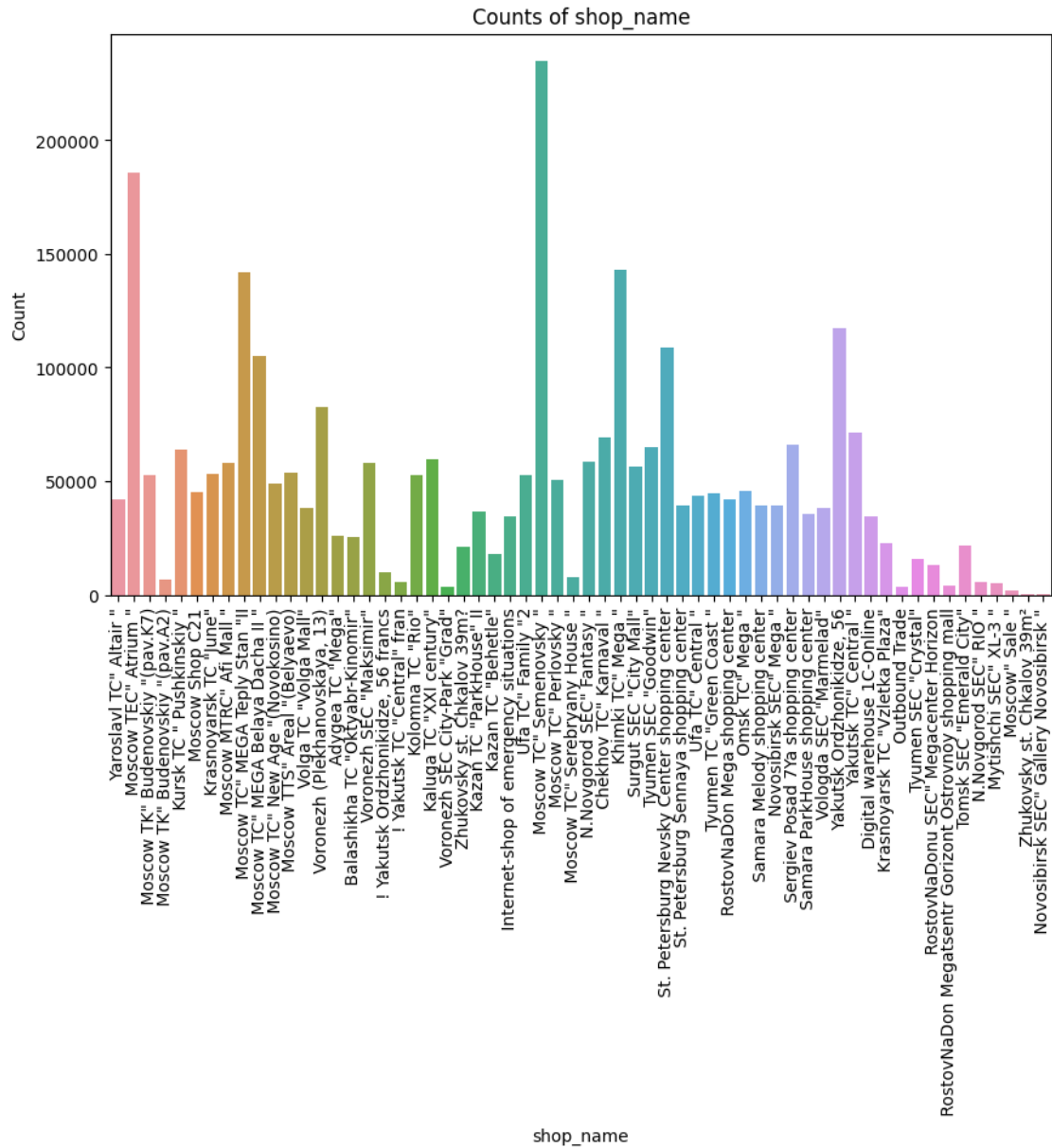


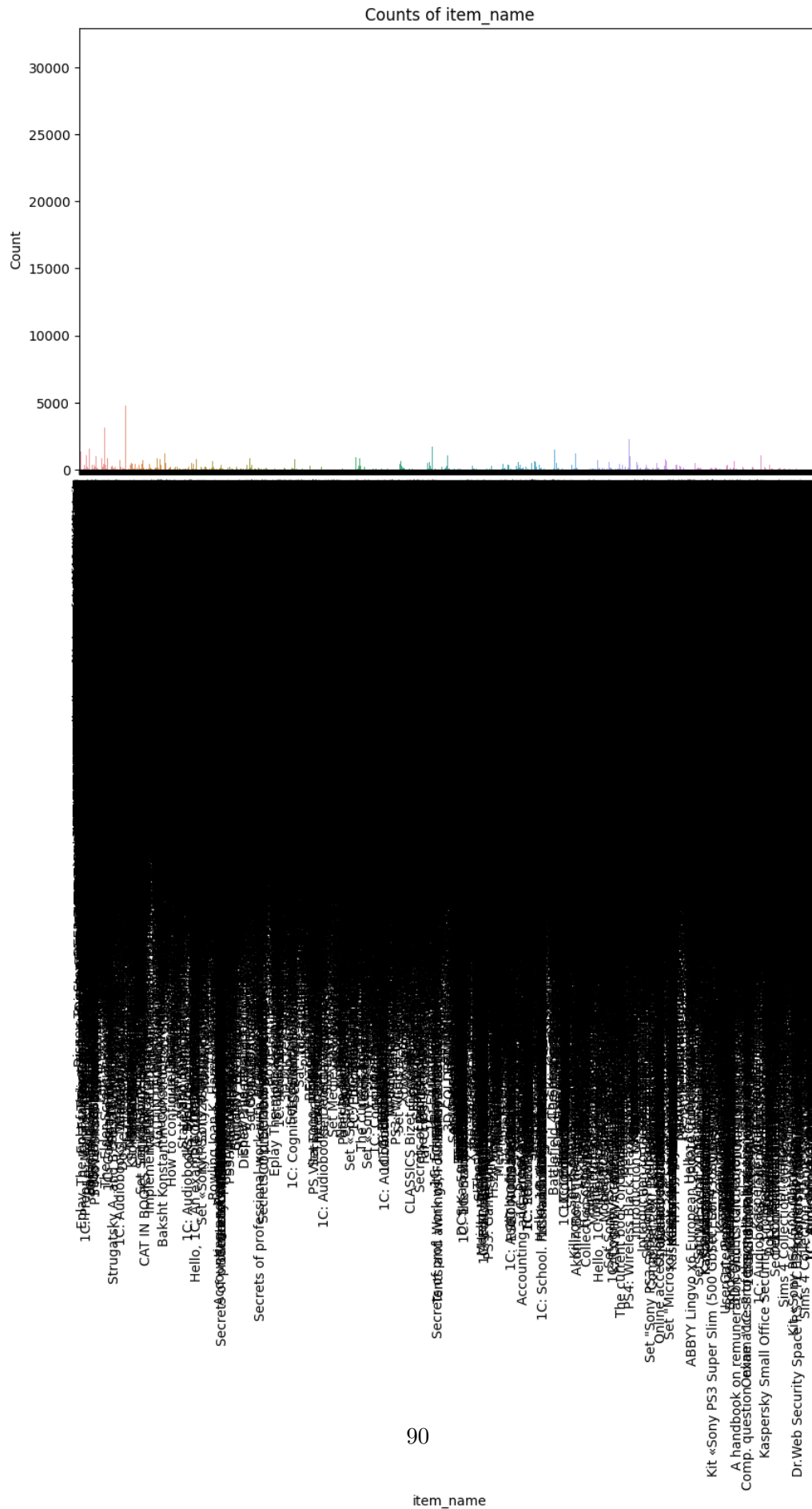


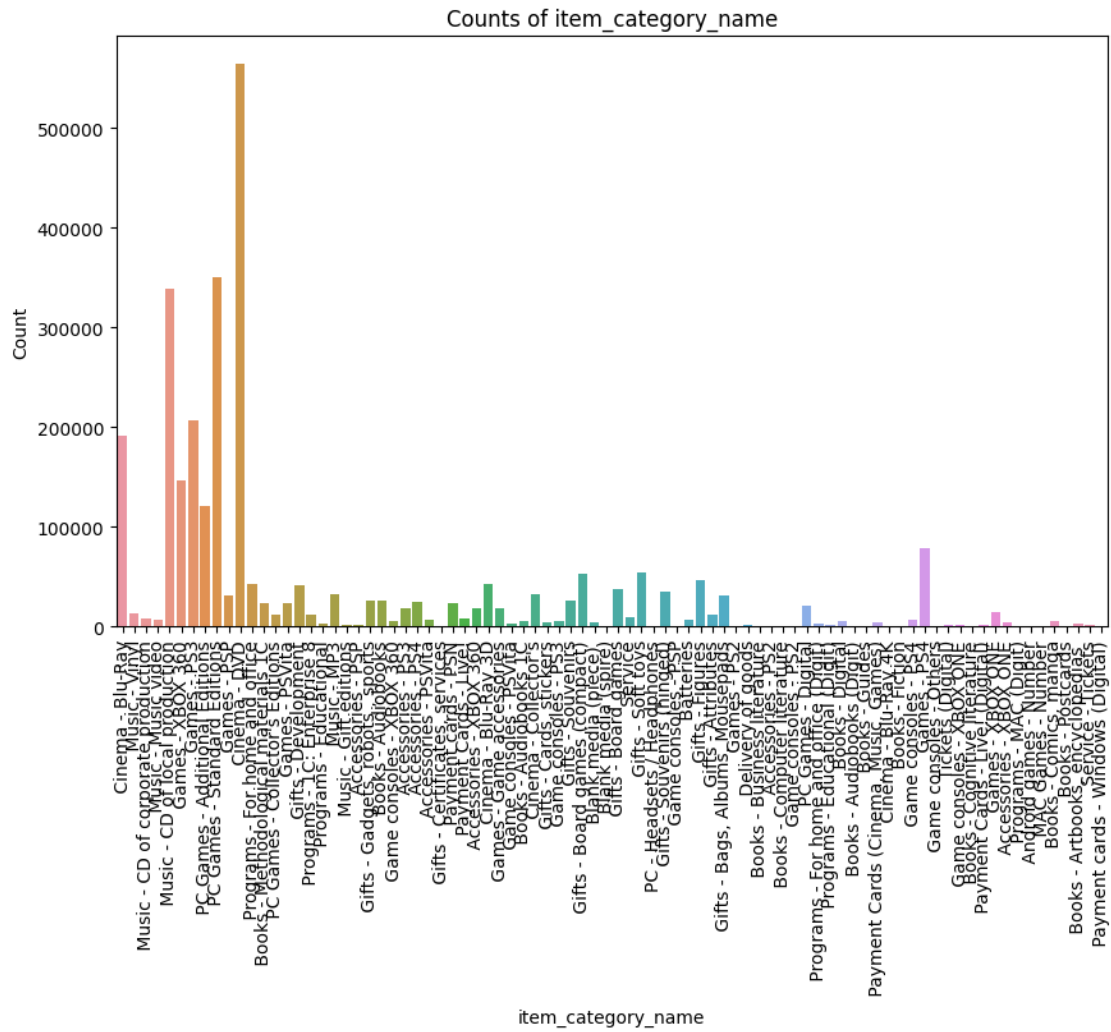


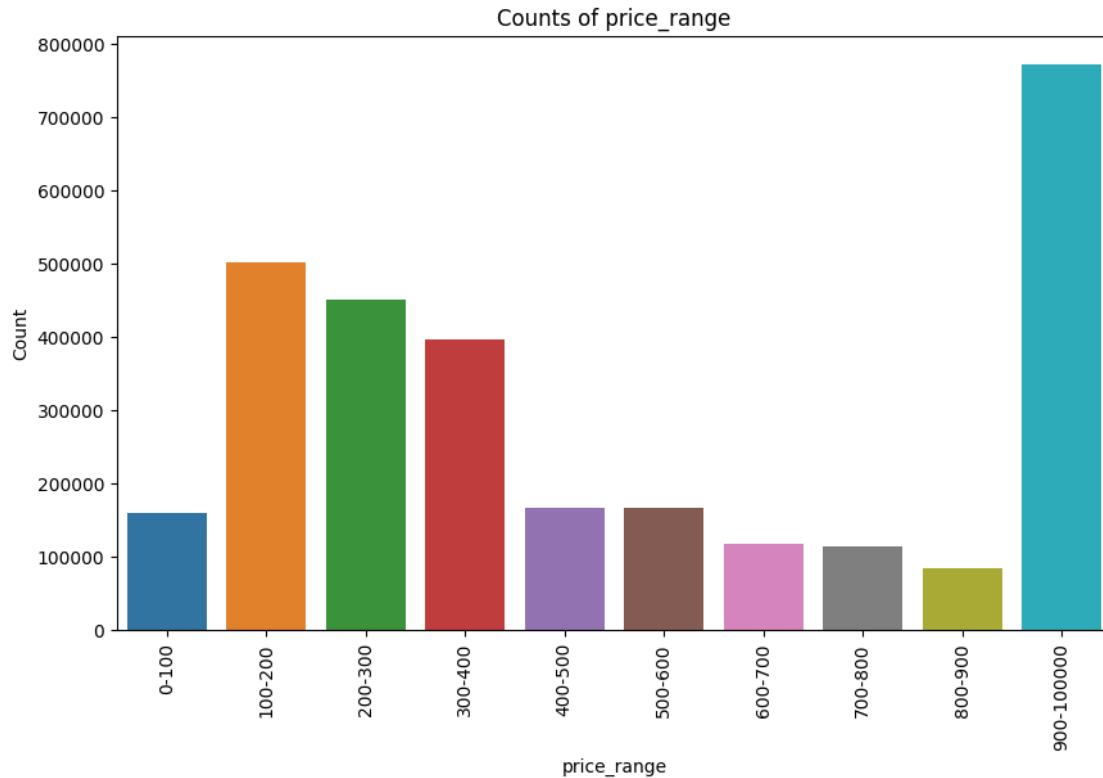












```
[ ]: #inferential analysis

np.random.seed(42)
data = np.random.normal(loc=70, scale=10, size=100)

# Create a DataFrame from the generated data
df = pd.DataFrame({'measurement': data})

# Calculate the sample mean and standard deviation
sample_mean = df['measurement'].mean()
sample_std = df['measurement'].std()

# Define a hypothetical population mean for comparison
population_mean = 75

# Perform a t-test to compare the sample mean with the population mean
t_statistic, p_value = stats.ttest_1samp(df['measurement'], population_mean)

# Print results
print(f"Sample Mean: {sample_mean:.2f}")
print(f"Sample Standard Deviation: {sample_std:.2f}")
print(f"Population Mean: {population_mean}")
```

```

print(f"T-Statistic: {t_statistic:.2f}")
print(f"P-Value: {p_value:.4f}")

# Determine statistical significance
alpha = 0.05 # Significance level (adjust as needed)
if p_value < alpha:
    print("Reject the null hypothesis: The sample mean is statistically
    ↪different from the population mean.")
else:
    print("Fail to reject the null hypothesis: There is no significant
    ↪difference between the sample mean and the population mean.")

```

Sample Mean: 68.96  
 Sample Standard Deviation: 9.08  
 Population Mean: 75  
 T-Statistic: -6.65  
 P-Value: 0.0000  
 Reject the null hypothesis: The sample mean is statistically different from the population mean.

```

[ ]: #diagnostic analytics

# Generate a hypothetical dataset
np.random.seed(42)
X = np.random.rand(100, 1) * 10
y = 3 * X + 2 + np.random.randn(100, 1)

# Create a DataFrame from the generated data
df = pd.DataFrame({'X': X.flatten(), 'y': y.flatten()})

# Diagnostic Plots
plt.figure(figsize=(12, 6))

plt.tight_layout()
plt.show()

```

<Figure size 1200x600 with 0 Axes>

```

[ ]: #qualitative analytics

category_counts = final_dataset['item_category_name'].value_counts()
print(category_counts)

cross_tab = pd.crosstab(final_dataset['shop_name'],
    ↪final_dataset['item_category_name'])
print(cross_tab)

category_frequency = (final_dataset['price_range'] == 'Low').sum()

```

```

print(f"Frequency of 'Low' price range: {category_frequency}")

average_price_per_category = final_dataset.
    ↳groupby('item_category_name')['item_price'].mean()
print(average_price_per_category)

category_counts.plot(kind='bar', figsize=(10, 6))
plt.title('Item Category Counts')
plt.xlabel('Category')
plt.ylabel('Count')
plt.xticks(rotation=90)
plt.show()

```

```

item_category_name
Cinema - DVD                563937
PC Games - Standard Editions 350787
Music - CD of local production 339127
Games - PS3                 207371
Cinema - Blu-Ray            191931

```

```

...
Books - Guides              3
Accessories - PS2          2
Books - Postcards          2
Books - Cognitive literature 1
Game consoles - PS2        1

```

Name: count, Length: 84, dtype: int64

item_category_name	Accessories - PS2 \
shop_name	
Adygea TC "Mega"	0
Balashikha TC "Oktyabr-Kinomir"	0
Chekhov TC" Karnaval "	0
Digital warehouse 1C-Online	0
Internet-shop of emergency situations	0
Kaluga TC "XXI century"	0
Kazan TC "Behetle"	0
Kazan TC "ParkHouse" II	0
Khimki TC" Mega "	0
Kolomna TC "Rio"	0
Krasnoyarsk TC "June"	0
Krasnoyarsk TC "Vzletka Plaza"	0
Kursk TC " Pushkinskiy "	0
Moscow MTRC" Afi Mall "	0
Moscow Shop C21	0
Moscow TC" MEGA Belaya Dacha II "	0
Moscow TC" MEGA Teply Stan "II	0
Moscow TC" New Age "(Novokosino)	0
Moscow TC" Perlovsky "	0
Moscow TC" Semenovskiy "	0

Moscow TC" Serebryany House "	0
Moscow TEC" Atrium "	0
Moscow TK" Budenovskiy "(pav.A2)	0
Moscow TK" Budenovskiy "(pav.K7)	0
Moscow TTS" Areal "(Belyaev)	0
Moscow" Sale "	0
Mytishchi SEC" XL-3 "	0
N.Novgorod SEC" Fantasy "	0
N.Novgorod SEC" RIO "	0
Novosibirsk SEC" Gallery Novosibirsk "	0
Novosibirsk SEC" Mega "	0
Omsk TC" Mega "	0
Outbound Trade	0
RostovNaDon Mega shopping center	0
RostovNaDon Megatsentr Gorizont Ostrovnoy shop...	0
RostovNaDonu SEC" Megacenter Horizon	0
Samara Melody shopping center	0
Samara ParkHouse shopping center	1
Sergiev Posad 7Ya shopping center	0
St. Petersburg Nevsky Center shopping center	0
St. Petersburg Sennaya shopping center	0
Surgut SEC "City Mall"	0
Tomsk SEC "Emerald City"	0
Tyumen SEC "Crystal"	0
Tyumen SEC "Goodwin"	0
Tyumen TC "Green Coast "	0
Ufa TC" Central "	0
Ufa TC" Family "2	0
Volga TC "Volga Mall"	0
Vologda SEC "Marmelad"	0
Voronezh (Plekhanovskaya, 13)	0
Voronezh SEC "Maksimir"	0
Voronezh SEC City-Park "Grad"	0
Yakutsk Ordzhonikidze, 56	0
Yakutsk TC" Central "	0
Yaroslavl TC" Altair "	0
Zhukovsky st. Chkalov 39m?	0
Zhukovsky st. Chkalov 39m <sup>2</sup>	0
! Yakutsk Ordzhonikidze, 56 francs	1
! Yakutsk TC "Central" fran	0

item_category_name	Accessories - PS3 \
shop_name	
Adygea TC "Mega"	338
Balashikha TC "Oktyabr-Kinomir"	199
Chekhov TC" Karnaval "	319
Digital warehouse 1C-Online	0
Internet-shop of emergency situations	213

Kaluga TC "XXI century"	337
Kazan TC "Behetle"	0
Kazan TC "ParkHouse" II	218
Khimki TC" Mega "	814
Kolomna TC "Rio"	418
Krasnoyarsk TC "June"	689
Krasnoyarsk TC "Vzletka Plaza"	191
Kursk TC " Pushkinskiy "	424
Moscow MTRC" Afi Mall "	244
Moscow Shop C21	119
Moscow TC" MEGA Belaya Dacha II "	868
Moscow TC" MEGA Teply Stan "II	986
Moscow TC" New Age "(Novokosino)	461
Moscow TC" Perlovsky "	360
Moscow TC" Semenovskiy "	1154
Moscow TC" Serebryany House "	38
Moscow TEC" Atrium "	972
Moscow TK" Budenovskiy "(pav.A2)	71
Moscow TK" Budenovskiy "(pav.K7)	368
Moscow TTS" Areal "(Belyaev)	307
Moscow" Sale "	0
Mytishchi SEC" XL-3 "	41
N.Novgorod SEC" Fantasy "	382
N.Novgorod SEC" RIO "	0
Novosibirsk SEC" Gallery Novosibirsk "	0
Novosibirsk SEC" Mega "	226
Omsk TC" Mega "	485
Outbound Trade	22
RostovNaDon Mega shopping center	309
RostovNaDon Megatsentr Gorizont Ostrovnoy shop...	0
RostovNaDonu SEC" Megacenter Horizon	46
Samara Melody shopping center	128
Samara ParkHouse shopping center	210
Sergiev Posad 7Ya shopping center	472
St. Petersburg Nevsky Center shopping center	548
St. Petersburg Sennaya shopping center	354
Surgut SEC "City Mall"	552
Tomsk SEC "Emerald City"	104
Tyumen SEC "Crystal"	102
Tyumen SEC "Goodwin"	665
Tyumen TC "Green Coast "	222
Ufa TC" Central "	244
Ufa TC" Family "2	329
Volga TC "Volga Mall"	301
Vologda SEC "Marmelad"	229
Voronezh (Plekhanovskaya, 13)	414
Voronezh SEC "Maksimir"	447
Voronezh SEC City-Park "Grad"	19



Yakutsk Ordzhonikidze, 56	592
Yakutsk TC" Central "	308
Yaroslavl TC" Altair "	218
Zhukovsky st. Chkalov 39m?	134
Zhukovsky st. Chkalov 39m <sup>2</sup>	2
! Yakutsk Ordzhonikidze, 56 francs	93
! Yakutsk TC "Central" fran	58

item_category_name	Accessories - PS4 \
shop_name	
Adygea TC "Mega"	470
Balashikha TC "Oktyabr-Kinomir"	344
Chekhov TC" Karnaval "	425
Digital warehouse 1C-Online	0
Internet-shop of emergency situations	514
Kaluga TC "XXI century"	460
Kazan TC "Behetle"	8
Kazan TC "ParkHouse" II	389
Khimki TC" Mega "	868
Kolomna TC "Rio"	469
Krasnoyarsk TC "June"	636
Krasnoyarsk TC "Vzletka Plaza"	218
Kursk TC " Pushkinskiy "	457
Moscow MTRC" Afi Mall "	581
Moscow Shop C21	322
Moscow TC" MEGA Belaya Dacha II "	1076
Moscow TC" MEGA Teply Stan "II	1229
Moscow TC" New Age "(Novokosino)	514
Moscow TC" Perlovsky "	373
Moscow TC" Semenovskiy "	1614
Moscow TC" Serebryany House "	13
Moscow TEC" Atrium "	1539
Moscow TK" Budenovskiy "(pav.A2)	11
Moscow TK" Budenovskiy "(pav.K7)	496
Moscow TTS" Areal "(Belyaev)	492
Moscow" Sale "	0
Mytishchi SEC" XL-3 "	66
N.Novgorod SEC" Fantasy "	563
N.Novgorod SEC" RIO "	83
Novosibirsk SEC" Gallery Novosibirsk "	6
Novosibirsk SEC" Mega "	357
Omsk TC" Mega "	532
Outbound Trade	34
RostovNaDon Mega shopping center	295
RostovNaDon Megatsentr Gorizont Ostrovnoy shop...	1
RostovNaDonu SEC" Megacenter Horizon	141
Samara Melody shopping center	231
Samara ParkHouse shopping center	267

Sergiev Posad 7Ya shopping center	645
St. Petersburg Nevsky Center shopping center	1117
St. Petersburg Sennaya shopping center	487
Surgut SEC "City Mall"	862
Tomsk SEC "Emerald City"	355
Tyumen SEC "Crystal"	239
Tyumen SEC "Goodwin"	638
Tyumen TC "Green Coast "	237
Ufa TC" Central "	358
Ufa TC" Family "2	426
Volga TC "Volga Mall"	317
Vologda SEC "Marmelad"	313
Voronezh (Plekhanovskaya, 13)	644
Voronezh SEC "Maksimir"	508
Voronezh SEC City-Park "Grad"	0
Yakutsk Ordzhonikidze, 56	786
Yakutsk TC" Central "	512
Yaroslavl TC" Altair "	370
Zhukovsky st. Chkalov 39m?	196
Zhukovsky st. Chkalov 39m <sup>2</sup>	4
! Yakutsk Ordzhonikidze, 56 francs	35
! Yakutsk TC "Central" fran	15

item_category_name	Accessories - PSP \
shop_name	
Adygea TC "Mega"	39
Balashikha TC "Oktyabr-Kinomir"	28
Chekhov TC" Karnaval "	45
Digital warehouse 1C-Online	0
Internet-shop of emergency situations	9
Kaluga TC "XXI century"	62
Kazan TC "Behetle"	0
Kazan TC "ParkHouse" II	19
Khimki TC" Mega "	98
Kolomna TC "Rio"	136
Krasnoyarsk TC "June"	49
Krasnoyarsk TC "Vzletka Plaza"	17
Kursk TC " Pushkinskiy "	59
Moscow MTRC" Afi Mall "	17
Moscow Shop C21	11
Moscow TC" MEGA Belaya Dacha II "	95
Moscow TC" MEGA Teply Stan "II	61
Moscow TC" New Age "(Novokosino)	52
Moscow TC" Perlovsky "	68
Moscow TC" Semenovskiy "	136
Moscow TC" Serebryany House "	9
Moscow TEC" Atrium "	45
Moscow TK" Budenovskiy "(pav.A2)	4

Moscow TK" Budenovskiy "(pav.K7)	38
Moscow TTS" Areal "(Belyaev)	21
Moscow" Sale "	0
Mytishchi SEC" XL-3 "	1
N.Novgorod SEC" Fantasy "	19
N.Novgorod SEC" RIO "	0
Novosibirsk SEC" Gallery Novosibirsk "	0
Novosibirsk SEC" Mega "	19
Omsk TC" Mega "	38
Outbound Trade	0
RostovNaDon Mega shopping center	44
RostovNaDon Megatsentr Gorizont Ostrovnoy shop...	0
RostovNaDonu SEC" Megacenter Horizon	2
Samara Melody shopping center	4
Samara ParkHouse shopping center	23
Sergiev Posad 7Ya shopping center	41
St. Petersburg Nevsky Center shopping center	35
St. Petersburg Sennaya shopping center	36
Surgut SEC "City Mall"	85
Tomsk SEC "Emerald City"	0
Tyumen SEC "Crystal"	12
Tyumen SEC "Goodwin"	90
Tyumen TC "Green Coast "	55
Ufa TC" Central "	55
Ufa TC" Family "2	68
Volga TC "Volga Mall"	90
Vologda SEC "Marmelad"	39
Voronezh (Plekhanovskaya, 13)	22
Voronezh SEC "Maksimir"	90
Voronezh SEC City-Park "Grad"	2
Yakutsk Ordzhonikidze, 56	115
Yakutsk TC" Central "	73
Yaroslavl TC" Altair "	52
Zhukovsky st. Chkalov 39m?	6
Zhukovsky st. Chkalov 39m <sup>2</sup>	0
! Yakutsk Ordzhonikidze, 56 francs	35
! Yakutsk TC "Central" fran	23
item_category_name	Accessories - PSVita \
shop_name	
Adygea TC "Mega"	49
Balashikha TC "Oktyabr-Kinomir"	107
Chekhov TC" Karnaval "	153
Digital warehouse 1C-Online	0
Internet-shop of emergency situations	100
Kaluga TC "XXI century"	66
Kazan TC "Behetle"	0
Kazan TC "ParkHouse" II	48

Khimki TC" Mega "	275
Kolomna TC "Rio"	176
Krasnoyarsk TC "June"	208
Krasnoyarsk TC "Vzletka Plaza"	64
Kursk TC " Pushkinskiy "	136
Moscow MTRC" Afi Mall "	169
Moscow Shop C21	115
Moscow TC" MEGA Belaya Dacha II "	403
Moscow TC" MEGA Teply Stan "II	306
Moscow TC" New Age "(Novokosino)	168
Moscow TC" Perlovsky "	167
Moscow TC" Semenovskiy "	508
Moscow TC" Serebryany House "	21
Moscow TEC" Atrium "	449
Moscow TK" Budenovskiy "(pav.A2)	20
Moscow TK" Budenovskiy "(pav.K7)	160
Moscow TTS" Areal "(Belyaev)	163
Moscow" Sale "	0
Mytishchi SEC" XL-3 "	15
N.Novgorod SEC" Fantasy "	148
N.Novgorod SEC" RIO "	0
Novosibirsk SEC" Gallery Novosibirsk "	0
Novosibirsk SEC" Mega "	87
Omsk TC" Mega "	65
Outbound Trade	0
RostovNaDon Mega shopping center	83
RostovNaDon Megatsentr Gorizont Ostrovnoy shop...	0
RostovNaDonu SEC" Megacenter Horizon	19
Samara Melody shopping center	66
Samara ParkHouse shopping center	102
Sergiev Posad 7Ya shopping center	136
St. Petersburg Nevsky Center shopping center	389
St. Petersburg Sennaya shopping center	120
Surgut SEC "City Mall"	167
Tomsk SEC "Emerald City"	35
Tyumen SEC "Crystal"	39
Tyumen SEC "Goodwin"	242
Tyumen TC "Green Coast "	59
Ufa TC" Central "	117
Ufa TC" Family "2	113
Volga TC "Volga Mall"	155
Vologda SEC "Marmelad"	95
Voronezh (Plekhanovskaya, 13)	169
Voronezh SEC "Maksimir"	200
Voronezh SEC City-Park "Grad"	7
Yakutsk Ordzhonikidze, 56	231
Yakutsk TC" Central "	129
Yaroslavl TC" Altair "	52

Zhukovsky st. Chkalov 39m?	48
Zhukovsky st. Chkalov 39m <sup>2</sup>	1
! Yakutsk Ordzhonikidze, 56 francs	60
! Yakutsk TC "Central" fran	24

item_category_name	Accessories - XBOX 360 \
shop_name	
Adygea TC "Mega"	317
Balashikha TC "Oktyabr-Kinomir"	243
Chekhov TC" Karnaval "	492
Digital warehouse 1C-Online	0
Internet-shop of emergency situations	132
Kaluga TC "XXI century"	443
Kazan TC "Behetle"	0
Kazan TC "ParkHouse" II	231
Khimki TC" Mega "	691
Kolomna TC "Rio"	439
Krasnoyarsk TC "June"	368
Krasnoyarsk TC "Vzletka Plaza"	125
Kursk TC " Pushkinskiy "	330
Moscow MTRC" Afi Mall "	266
Moscow Shop C21	121
Moscow TC" MEGA Belaya Dacha II "	988
Moscow TC" MEGA Teply Stan "II	963
Moscow TC" New Age "(Novokosino)	288
Moscow TC" Perlovsky "	362
Moscow TC" Semenovskiy "	1096
Moscow TC" Serebryany House "	26
Moscow TEC" Atrium "	888
Moscow TK" Budenovskiy "(pav.A2)	37
Moscow TK" Budenovskiy "(pav.K7)	430
Moscow TTS" Areal "(Belyaev)	307
Moscow" Sale "	0
Mytishchi SEC" XL-3 "	37
N.Novgorod SEC" Fantasy "	338
N.Novgorod SEC" RIO "	8
Novosibirsk SEC" Gallery Novosibirsk "	2
Novosibirsk SEC" Mega "	208
Omsk TC" Mega "	467
Outbound Trade	21
RostovNaDon Mega shopping center	334
RostovNaDon Megatsentr Gorizont Ostrovnay shop...	0
RostovNaDonu SEC" Megacenter Horizon	57
Samara Melody shopping center	108
Samara ParkHouse shopping center	211
Sergiev Posad 7Ya shopping center	569
St. Petersburg Nevsky Center shopping center	520
St. Petersburg Sennaya shopping center	349

Surgut SEC "City Mall"	355
Tomsk SEC "Emerald City"	179
Tyumen SEC "Crystal"	145
Tyumen SEC "Goodwin"	571
Tyumen TC "Green Coast "	174
Ufa TC" Central "	337
Ufa TC" Family "2	481
Volga TC "Volga Mall"	319
Vologda SEC "Marmelad"	207
Voronezh (Plekhanovskaya, 13)	352
Voronezh SEC "Maksimir"	419
Voronezh SEC City-Park "Grad"	11
Yakutsk Ordzhonikidze, 56	826
Yakutsk TC" Central "	655
Yaroslavl TC" Altair "	247
Zhukovsky st. Chkalov 39m?	138
Zhukovsky st. Chkalov 39m <sup>2</sup>	6
! Yakutsk Ordzhonikidze, 56 francs	116
! Yakutsk TC "Central" fran	57

item_category_name	Accessories - XBOX ONE \
shop_name	
Adygea TC "Mega"	68
Balashikha TC "Oktyabr-Kinomir"	60
Chekhov TC" Karnaval "	57
Digital warehouse 1C-Online	0
Internet-shop of emergency situations	161
Kaluga TC "XXI century"	87
Kazan TC "Behetle"	0
Kazan TC "ParkHouse" II	58
Khimki TC" Mega "	123
Kolomna TC "Rio"	54
Krasnoyarsk TC "June"	77
Krasnoyarsk TC "Vzletka Plaza"	28
Kursk TC " Pushkinskiy "	45
Moscow MTRC" Afi Mall "	153
Moscow Shop C21	74
Moscow TC" MEGA Belaya Dacha II "	206
Moscow TC" MEGA Teply Stan "II	238
Moscow TC" New Age "(Novokosino)	53
Moscow TC" Perlovsky "	34
Moscow TC" Semenovsky "	212
Moscow TC" Serebryany House "	0
Moscow TEC" Atrium "	289
Moscow TK" Budenovskiy "(pav.A2)	0
Moscow TK" Budenovskiy "(pav.K7)	65
Moscow TTS" Areal "(Belyaev)	119
Moscow" Sale "	0

Mytishchi SEC" XL-3 "	23
N.Novgorod SEC" Fantasy "	131
N.Novgorod SEC" RIO "	40
Novosibirsk SEC" Gallery Novosibirsk "	0
Novosibirsk SEC" Mega "	87
Omsk TC" Mega "	122
Outbound Trade	15
RostovNaDon Mega shopping center	46
RostovNaDon Megatsentr Gorizont Ostrovnoy shop...	0
RostovNaDonu SEC" Megacenter Horizon	80
Samara Melody shopping center	47
Samara ParkHouse shopping center	56
Sergiev Posad 7Ya shopping center	79
St. Petersburg Nevsky Center shopping center	271
St. Petersburg Sennaya shopping center	72
Surgut SEC "City Mall"	99
Tomsk SEC "Emerald City"	94
Tyumen SEC "Crystal"	56
Tyumen SEC "Goodwin"	71
Tyumen TC "Green Coast "	28
Ufa TC" Central "	56
Ufa TC" Family "2	73
Volga TC "Volga Mall"	45
Vologda SEC "Marmelad"	65
Voronezh (Plekhanovskaya, 13)	92
Voronezh SEC "Maksimir"	63
Voronezh SEC City-Park "Grad"	0
Yakutsk Ordzhonikidze, 56	168
Yakutsk TC" Central "	81
Yaroslavl TC" Altair "	72
Zhukovsky st. Chkalov 39m?	36
Zhukovsky st. Chkalov 39m <sup>2</sup>	1
! Yakutsk Ordzhonikidze, 56 francs	0
! Yakutsk TC "Central" fran	0

item_category_name	Android games - Number	\
shop_name		
Adygea TC "Mega"	0	
Balashikha TC "Oktyabr-Kinomir"	0	
Chekhov TC" Karnaval "	0	
Digital warehouse 1C-Online	91	
Internet-shop of emergency situations	0	
Kaluga TC "XXI century"	0	
Kazan TC "Behetle"	0	
Kazan TC "ParkHouse" II	0	
Khimki TC" Mega "	0	
Kolomna TC "Rio"	0	
Krasnoyarsk TC "June"	0	

Krasnoyarsk TC "Vzletka Plaza"	0
Kursk TC " Pushkinskiy "	0
Moscow MTRC" Afi Mall "	0
Moscow Shop C21	0
Moscow TC" MEGA Belaya Dacha II "	0
Moscow TC" MEGA Teply Stan "II	0
Moscow TC" New Age "(Novokosino)	0
Moscow TC" Perlovsky "	0
Moscow TC" Semenovskiy "	0
Moscow TC" Serebryany House "	0
Moscow TEC" Atrium "	0
Moscow TK" Budenovskiy "(pav.A2)	0
Moscow TK" Budenovskiy "(pav.K7)	0
Moscow TTS" Areal "(Belyaev)	0
Moscow" Sale "	0
Mytishchi SEC" XL-3 "	0
N.Novgorod SEC" Fantasy "	0
N.Novgorod SEC" RIO "	0
Novosibirsk SEC" Gallery Novosibirsk "	0
Novosibirsk SEC" Mega "	0
Omsk TC" Mega "	0
Outbound Trade	0
RostovNaDon Mega shopping center	0
RostovNaDon Megatsentr Gorizont Ostrovnoy shop...	0
RostovNaDonu SEC" Megacenter Horizon	0
Samara Melody shopping center	0
Samara ParkHouse shopping center	0
Sergiev Posad 7Ya shopping center	0
St. Petersburg Nevsky Center shopping center	0
St. Petersburg Sennaya shopping center	0
Surgut SEC "City Mall"	0
Tomsk SEC "Emerald City"	0
Tyumen SEC "Crystal"	0
Tyumen SEC "Goodwin"	0
Tyumen TC "Green Coast "	0
Ufa TC" Central "	0
Ufa TC" Family "2	0
Volga TC "Volga Mall"	0
Vologda SEC "Marmelad"	0
Voronezh (Plekhanovskaya, 13)	0
Voronezh SEC "Maksimir"	0
Voronezh SEC City-Park "Grad"	0
Yakutsk Ordzhonikidze, 56	0
Yakutsk TC" Central "	0
Yaroslavl TC" Altair "	0
Zhukovsky st. Chkalov 39m?	0
Zhukovsky st. Chkalov 39m <sup>2</sup>	0
! Yakutsk Ordzhonikidze, 56 francs	0



! Yakutsk TC "Central" fran

0

item_category_name	Batteries \
shop_name	
Adygea TC "Mega"	34
Balashikha TC "Oktyabr-Kinomir"	36
Chekhov TC" Karnaval "	152
Digital warehouse 1C-Online	0
Internet-shop of emergency situations	2
Kaluga TC "XXI century"	62
Kazan TC "Behetle"	0
Kazan TC "ParkHouse" II	164
Khimki TC" Mega "	169
Kolomna TC "Rio"	202
Krasnoyarsk TC "June"	122
Krasnoyarsk TC "Vzletka Plaza"	0
Kursk TC " Pushkinskiy "	96
Moscow MTRC" Afi Mall "	129
Moscow Shop C21	84
Moscow TC" MEGA Belaya Dacha II "	53
Moscow TC" MEGA Teply Stan "II	411
Moscow TC" New Age "(Novokosino)	108
Moscow TC" Perlovsky "	130
Moscow TC" Semenovskiy "	439
Moscow TC" Serebryany House "	0
Moscow TEC" Atrium "	338
Moscow TK" Budenovskiy "(pav.A2)	0
Moscow TK" Budenovskiy "(pav.K7)	44
Moscow TTS" Areal "(Belyaev)	73
Moscow" Sale "	2
Mytishchi SEC" XL-3 "	29
N.Novgorod SEC" Fantasy "	67
N.Novgorod SEC" RIO "	43
Novosibirsk SEC" Gallery Novosibirsk "	2
Novosibirsk SEC" Mega "	157
Omsk TC" Mega "	74
Outbound Trade	0
RostovNaDon Mega shopping center	75
RostovNaDon Megatsentr Gorizont Ostrovnoy shop...	0
RostovNaDonu SEC" Megacenter Horizon	73
Samara Melody shopping center	546
Samara ParkHouse shopping center	136
Sergiev Posad 7Ya shopping center	128
St. Petersburg Nevsky Center shopping center	155
St. Petersburg Sennaya shopping center	155
Surgut SEC "City Mall"	134
Tomsk SEC "Emerald City"	49
Tyumen SEC "Crystal"	36

Tyumen SEC "Goodwin"	241
Tyumen TC "Green Coast "	242
Ufa TC" Central "	183
Ufa TC" Family "2	117
Volga TC "Volga Mall"	50
Vologda SEC "Marmelad"	82
Voronezh (Plekhanovskaya, 13)	392
Voronezh SEC "Maksimir"	110
Voronezh SEC City-Park "Grad"	0
Yakutsk Ordzhonikidze, 56	374
Yakutsk TC" Central "	170
Yaroslavl TC" Altair "	347
Zhukovsky st. Chkalov 39m?	89
Zhukovsky st. Chkalov 39m <sup>2</sup>	0
! Yakutsk Ordzhonikidze, 56 francs	48
! Yakutsk TC "Central" fran	41

item_category_name	Blank media (piece)	\
shop_name		
Adygea TC "Mega"	42	
Balashikha TC "Oktyabr-Kinomir"	0	
Chekhov TC" Karnaval "	310	
Digital warehouse 1C-Online	0	
Internet-shop of emergency situations	3	
Kaluga TC "XXI century"	0	
Kazan TC "Behetle"	105	
Kazan TC "ParkHouse" II	0	
Khimki TC" Mega "	120	
Kolomna TC "Rio"	186	
Krasnoyarsk TC "June"	86	
Krasnoyarsk TC "Vzletka Plaza"	0	
Kursk TC " Pushkinskiy "	0	
Moscow MTRC" Afi Mall "	0	
Moscow Shop C21	101	
Moscow TC" MEGA Belaya Dacha II "	0	
Moscow TC" MEGA Teply Stan "II	0	
Moscow TC" New Age "(Novokosino)	285	
Moscow TC" Perlovsky "	266	
Moscow TC" Semenovskiy "	0	
Moscow TC" Serebryany House "	0	
Moscow TEC" Atrium "	254	
Moscow TK" Budenovskiy "(pav.A2)	0	
Moscow TK" Budenovskiy "(pav.K7)	0	
Moscow TTS" Areal "(Belyaev)	0	
Moscow" Sale "	0	
Mytishchi SEC" XL-3 "	0	
N.Novgorod SEC" Fantasy "	0	
N.Novgorod SEC" RIO "	0	

Novosibirsk SEC" Gallery Novosibirsk "	0
Novosibirsk SEC" Mega "	84
Omsk TC" Mega "	0
Outbound Trade	0
RostovNaDon Mega shopping center	67
RostovNaDon Megatsentr Gorizont Ostrovnoy shop...	0
RostovNaDonu SEC" Megacenter Horizon	0
Samara Melody shopping center	548
Samara ParkHouse shopping center	0
Sergiev Posad 7Ya shopping center	0
St. Petersburg Nevsky Center shopping center	0
St. Petersburg Sennaya shopping center	0
Surgut SEC "City Mall"	0
Tomsk SEC "Emerald City"	0
Tyumen SEC "Crystal"	0
Tyumen SEC "Goodwin"	0
Tyumen TC "Green Coast "	508
Ufa TC" Central "	291
Ufa TC" Family "2	0
Volga TC "Volga Mall"	208
Vologda SEC "Marmelad"	0
Voronezh (Plekhanovskaya, 13)	290
Voronezh SEC "Maksimir"	0
Voronezh SEC City-Park "Grad"	0
Yakutsk Ordzhonikidze, 56	228
Yakutsk TC" Central "	170
Yaroslavl TC" Altair "	0
Zhukovsky st. Chkalov 39m?	110
Zhukovsky st. Chkalov 39m <sup>2</sup>	0
! Yakutsk Ordzhonikidze, 56 francs	69
! Yakutsk TC "Central" fran	59

item_category_name	...	\
shop_name	...	
Adygea TC "Mega"	...	
Balashikha TC "Oktyabr-Kinomir"	...	
Chekhov TC" Karnaval "	...	
Digital warehouse 1C-Online	...	
Internet-shop of emergency situations	...	
Kaluga TC "XXI century"	...	
Kazan TC "Behetle"	...	
Kazan TC "ParkHouse" II	...	
Khimki TC" Mega "	...	
Kolomna TC "Rio"	...	
Krasnoyarsk TC "June"	...	
Krasnoyarsk TC "Vzletka Plaza"	...	
Kursk TC " Pushkinskiy "	...	
Moscow MTRC" Afi Mall "	...	

Moscow Shop C21	...
Moscow TC" MEGA Belaya Dacha II "	...
Moscow TC" MEGA Teply Stan "II	...
Moscow TC" New Age "(Novokosino)	...
Moscow TC" Perlovsky "	...
Moscow TC" Semenovsky "	...
Moscow TC" Serebryany House "	...
Moscow TEC" Atrium "	...
Moscow TK" Budenovskiy "(pav.A2)	...
Moscow TK" Budenovskiy "(pav.K7)	...
Moscow TTS" Areal "(Belyaev)	...
Moscow" Sale "	...
Mytishchi SEC" XL-3 "	...
N.Novgorod SEC" Fantasy "	...
N.Novgorod SEC" RIO "	...
Novosibirsk SEC" Gallery Novosibirsk "	...
Novosibirsk SEC" Mega "	...
Omsk TC" Mega "	...
Outbound Trade	...
RostovNaDon Mega shopping center	...
RostovNaDon Megatsentr Gorizont Ostrovnoy shop...	...
RostovNaDonu SEC" Megacenter Horizon	...
Samara Melody shopping center	...
Samara ParkHouse shopping center	...
Sergiev Posad 7Ya shopping center	...
St. Petersburg Nevsky Center shopping center	...
St. Petersburg Sennaya shopping center	...
Surgut SEC "City Mall"	...
Tomsk SEC "Emerald City"	...
Tyumen SEC "Crystal"	...
Tyumen SEC "Goodwin"	...
Tyumen TC "Green Coast "	...
Ufa TC" Central "	...
Ufa TC" Family "2	...
Volga TC "Volga Mall"	...
Vologda SEC "Marmelad"	...
Voronezh (Plekhanovskaya, 13)	...
Voronezh SEC "Maksimir"	...
Voronezh SEC City-Park "Grad"	...
Yakutsk Ordzhonikidze, 56	...
Yakutsk TC" Central "	...
Yaroslavl TC" Altair "	...
Zhukovsky st. Chkalov 39m?	...
Zhukovsky st. Chkalov 39m <sup>2</sup>	...
! Yakutsk Ordzhonikidze, 56 francs	...
! Yakutsk TC "Central" fran	...

item\_category\_name

Programs - 1C: Enterprise 8

\	
shop_name	
Adygea TC "Mega"	165
Balashikha TC "Oktyabr-Kinomir"	62
Chekhov TC" Karnaval "	136
Digital warehouse 1C-Online	0
Internet-shop of emergency situations	1447
Kaluga TC "XXI century"	115
Kazan TC "Behetle"	8
Kazan TC "ParkHouse" II	128
Khimki TC" Mega "	402
Kolomna TC "Rio"	222
Krasnoyarsk TC "June"	102
Krasnoyarsk TC "Vzletka Plaza"	30
Kursk TC " Pushkinskiy "	113
Moscow MTRC" Afi Mall "	206
Moscow Shop C21	2134
Moscow TC" MEGA Belaya Dacha II "	833
Moscow TC" MEGA Teply Stan "II	387
Moscow TC" New Age "(Novokosino)	260
Moscow TC" Perlovsky "	239
Moscow TC" Semenovskiy "	641
Moscow TC" Serebryany House "	52
Moscow TEC" Atrium "	784
Moscow TK" Budenovskiy "(pav.A2)	49
Moscow TK" Budenovskiy "(pav.K7)	520
Moscow TTS" Areal "(Belyaev)	389
Moscow" Sale "	0
Mytishchi SEC" XL-3 "	9
N.Novgorod SEC" Fantasy "	86
N.Novgorod SEC" RIO "	19
Novosibirsk SEC" Gallery Novosibirsk "	0
Novosibirsk SEC" Mega "	171
Omsk TC" Mega "	135
Outbound Trade	0
RostovNaDon Mega shopping center	95
RostovNaDon Megatsentr Gorizont Ostrovnoy shop...	0
RostovNaDonu SEC" Megacenter Horizon	27
Samara Melody shopping center	68
Samara ParkHouse shopping center	101
Sergiev Posad 7Ya shopping center	144
St. Petersburg Nevsky Center shopping center	430
St. Petersburg Sennaya shopping center	241
Surgut SEC "City Mall"	183
Tomsk SEC "Emerald City"	26
Tyumen SEC "Crystal"	29
Tyumen SEC "Goodwin"	79
Tyumen TC "Green Coast "	49

Ufa TC" Central "	114
Ufa TC" Family "2	171
Volga TC "Volga Mall"	87
Vologda SEC "Marmelad"	47
Voronezh (Plekhanovskaya, 13)	235
Voronezh SEC "Maksimir"	129
Voronezh SEC City-Park "Grad"	6
Yakutsk Ordzhonikidze, 56	214
Yakutsk TC" Central "	120
Yaroslavl TC" Altair "	106
Zhukovsky st. Chkalov 39m?	130
Zhukovsky st. Chkalov 39m <sup>2</sup>	3
! Yakutsk Ordzhonikidze, 56 francs	20
! Yakutsk TC "Central" fran	14

item_category_name	Programs - Educational \
shop_name	
Adygea TC "Mega"	33
Balashikha TC "Oktyabr-Kinomir"	12
Chekhov TC" Karnaval "	126
Digital warehouse 1C-Online	0
Internet-shop of emergency situations	140
Kaluga TC "XXI century"	24
Kazan TC "Behetle"	5
Kazan TC "ParkHouse" II	7
Khimki TC" Mega "	118
Kolomna TC "Rio"	126
Krasnoyarsk TC "June"	60
Krasnoyarsk TC "Vzletka Plaza"	4
Kursk TC " Pushkinskiy "	22
Moscow MTRC" Afi Mall "	52
Moscow Shop C21	543
Moscow TC" MEGA Belaya Dacha II "	5
Moscow TC" MEGA Teply Stan "II	69
Moscow TC" New Age "(Novokosino)	61
Moscow TC" Perlovsky "	49
Moscow TC" Semenovskiy "	29
Moscow TC" Serebryany House "	48
Moscow TEC" Atrium "	227
Moscow TK" Budenovskiy "(pav.A2)	54
Moscow TK" Budenovskiy "(pav.K7)	66
Moscow TTS" Areal "(Belyaev)	12
Moscow" Sale "	0
Mytishchi SEC" XL-3 "	0
N.Novgorod SEC" Fantasy "	79
N.Novgorod SEC" RIO "	0
Novosibirsk SEC" Gallery Novosibirsk "	0
Novosibirsk SEC" Mega "	108

Omsk TC" Mega "	130
Outbound Trade	0
RostovNaDon Mega shopping center	59
RostovNaDon Megatsentr Gorizont Ostrovnoy shop...	0
RostovNaDonu SEC" Megacenter Horizon	1
Samara Melody shopping center	89
Samara ParkHouse shopping center	11
Sergiev Posad 7Ya shopping center	25
St. Petersburg Nevsky Center shopping center	30
St. Petersburg Sennaya shopping center	8
Surgut SEC "City Mall"	1
Tomsk SEC "Emerald City"	3
Tyumen SEC "Crystal"	1
Tyumen SEC "Goodwin"	7
Tyumen TC "Green Coast "	85
Ufa TC" Central "	153
Ufa TC" Family "2	8
Volga TC "Volga Mall"	103
Vologda SEC "Marmelad"	12
Voronezh (Plekhanovskaya, 13)	176
Voronezh SEC "Maksimir"	17
Voronezh SEC City-Park "Grad"	20
Yakutsk Ordzhonikidze, 56	324
Yakutsk TC" Central "	172
Yaroslavl TC" Altair "	4
Zhukovsky st. Chkalov 39m?	8
Zhukovsky st. Chkalov 39m <sup>2</sup>	0
! Yakutsk Ordzhonikidze, 56 francs	104
! Yakutsk TC "Central" fran	61

item_category_name	Programs - Educational
(Digit) \	
shop_name	
Adygea TC "Mega"	
0	
Balashikha TC "Oktyabr-Kinomir"	
0	
Chekhov TC" Karnaval "	
0	
Digital warehouse 1C-Online	
2346	
Internet-shop of emergency situations	
0	
Kaluga TC "XXI century"	
0	
Kazan TC "Behetle"	
0	
Kazan TC "ParkHouse" II	

0  
 Khimki TC" Mega "  
 0  
 Kolomna TC "Rio"  
 0  
 Krasnoyarsk TC "June"  
 0  
 Krasnoyarsk TC "Vzletka Plaza"  
 0  
 Kursk TC " Pushkinskiy "  
 0  
 Moscow MTRC" Afi Mall "  
 0  
 Moscow Shop C21  
 0  
 Moscow TC" MEGA Belaya Dacha II "  
 0  
 Moscow TC" MEGA Teply Stan "II  
 0  
 Moscow TC" New Age "(Novokosino)  
 0  
 Moscow TC" Perlovsky "  
 0  
 Moscow TC" Semenovskiy "  
 0  
 Moscow TC" Serebryany House "  
 0  
 Moscow TEC" Atrium "  
 0  
 Moscow TK" Budenovskiy "(pav.A2)  
 0  
 Moscow TK" Budenovskiy "(pav.K7)  
 0  
 Moscow TTS" Areal "(Belyaev)  
 0  
 Moscow" Sale "  
 0  
 Mytishchi SEC" XL-3 "  
 0  
 N.Novgorod SEC" Fantasy "  
 0  
 N.Novgorod SEC" RIO "  
 0  
 Novosibirsk SEC" Gallery Novosibirsk "  
 0  
 Novosibirsk SEC" Mega "  
 0  
 Omsk TC" Mega "



0  
 Outbound Trade  
 0  
 RostovNaDon Mega shopping center  
 0  
 RostovNaDon Megatsentr Gorizont Ostrovnoy shop...  
 0  
 RostovNaDonu SEC" Megacenter Horizon  
 0  
 Samara Melody shopping center  
 0  
 Samara ParkHouse shopping center  
 0  
 Sergiev Posad 7Ya shopping center  
 0  
 St. Petersburg Nevsky Center shopping center  
 0  
 St. Petersburg Sennaya shopping center  
 0  
 Surgut SEC "City Mall"  
 0  
 Tomsk SEC "Emerald City"  
 0  
 Tyumen SEC "Crystal"  
 0  
 Tyumen SEC "Goodwin"  
 0  
 Tyumen TC "Green Coast "  
 0  
 Ufa TC" Central "  
 0  
 Ufa TC" Family "2  
 0  
 Volga TC "Volga Mall"  
 0  
 Vologda SEC "Marmelad"  
 0  
 Voronezh (Plekhanovskaya, 13)  
 0  
 Voronezh SEC "Maksimir"  
 0  
 Voronezh SEC City-Park "Grad"  
 0  
 Yakutsk Ordzhonikidze, 56  
 0  
 Yakutsk TC" Central "  
 0  
 Yaroslavl TC" Altair "

0  
 Zhukovsky st. Chkalov 39m?  
 0  
 Zhukovsky st. Chkalov 39m<sup>2</sup>  
 0  
 ! Yakutsk Ordzhonikidze, 56 francs  
 0  
 ! Yakutsk TC "Central" fran  
 0

item_category_name	Programs - For home and
office \	
shop_name	
Adygea TC "Mega"	
481	
Balashikha TC "Oktyabr-Kinomir"	
152	
Chekhov TC" Karnaval "	
434	
Digital warehouse 1C-Online	
0	
Internet-shop of emergency situations	
417	
Kaluga TC "XXI century"	
1825	
Kazan TC "Behetle"	
38	
Kazan TC "ParkHouse" II	
328	
Khimki TC" Mega "	
1237	
Kolomna TC "Rio"	
1087	
Krasnoyarsk TC "June"	
515	
Krasnoyarsk TC "Vzletka Plaza"	
192	
Kursk TC " Pushkinskiy "	
761	
Moscow MTRC" Afi Mall "	
500	
Moscow Shop C21	
2773	
Moscow TC" MEGA Belaya Dacha II "	
1486	
Moscow TC" MEGA Teply Stan "II	
804	
Moscow TC" New Age "(Novokosino)	

1694  
 Moscow TC" Perlovsky "  
 671  
 Moscow TC" Semenovskiy "  
 3177  
 Moscow TC" Serebryany House "  
 168  
 Moscow TEC" Atrium "  
 1556  
 Moscow TK" Budenovskiy "(pav.A2)  
 501  
 Moscow TK" Budenovskiy "(pav.K7)  
 3780  
 Moscow TTS" Areal "(Belyaev)  
 909  
 Moscow" Sale "  
 0  
 Mytishchi SEC" XL-3 "  
 25  
 N.Novgorod SEC" Fantasy "  
 219  
 N.Novgorod SEC" RIO "  
 18  
 Novosibirsk SEC" Gallery Novosibirsk "  
 3  
 Novosibirsk SEC" Mega "  
 617  
 Omsk TC" Mega "  
 509  
 Outbound Trade  
 4  
 RostovNaDon Mega shopping center  
 603  
 RostovNaDon Megatsentr Gorizont Ostrovnoy shop...  
 0  
 RostovNaDonu SEC" Megacenter Horizon  
 41  
 Samara Melody shopping center  
 608  
 Samara ParkHouse shopping center  
 388  
 Sergiev Posad 7Ya shopping center  
 845  
 St. Petersburg Nevsky Center shopping center  
 630  
 St. Petersburg Sennaya shopping center  
 330  
 Surgut SEC "City Mall"

568  
 Tomsk SEC "Emerald City"  
 56  
 Tyumen SEC "Crystal"  
 115  
 Tyumen SEC "Goodwin"  
 740  
 Tyumen TC "Green Coast "  
 542  
 Ufa TC" Central "  
 464  
 Ufa TC" Family "2  
 574  
 Volga TC "Volga Mall"  
 718  
 Vologda SEC "Marmelad"  
 122  
 Voronezh (Plekhanovskaya, 13)  
 1421  
 Voronezh SEC "Maksimir"  
 856  
 Voronezh SEC City-Park "Grad"  
 35  
 Yakutsk Ordzhonikidze, 56  
 2760  
 Yakutsk TC" Central "  
 1111  
 Yaroslavl TC" Altair "  
 742  
 Zhukovsky st. Chkalov 39m?  
 920  
 Zhukovsky st. Chkalov 39m<sup>2</sup>  
 20  
 ! Yakutsk Ordzhonikidze, 56 francs  
 252  
 ! Yakutsk TC "Central" fran  
 121

item\_category\_name

office (Digit) \

shop\_name

Adygea TC "Mega"

0

Balashikha TC "Oktyabr-Kinomir"

0

Chekhov TC" Karnaval "

0

Digital warehouse 1C-Online

Programs - For home and

3746

Internet-shop of emergency situations

0

Kaluga TC "XXI century"

0

Kazan TC "Behetle"

0

Kazan TC "ParkHouse" II

0

Khimki TC" Mega "

0

Kolomna TC "Rio"

0

Krasnoyarsk TC "June"

0

Krasnoyarsk TC "Vzletka Plaza"

0

Kursk TC " Pushkinskiy "

0

Moscow MTRC" Afi Mall "

0

Moscow Shop C21

0

Moscow TC" MEGA Belaya Dacha II "

0

Moscow TC" MEGA Teply Stan "II

0

Moscow TC" New Age "(Novokosino)

0

Moscow TC" Perlovsky "

0

Moscow TC" Semenovsky "

0

Moscow TC" Serebryany House "

0

Moscow TEC" Atrium "

0

Moscow TK" Budenovskiy "(pav.A2)

0

Moscow TK" Budenovskiy "(pav.K7)

0

Moscow TTS" Areal "(Belyaev)

0

Moscow" Sale "

0

Mytishchi SEC" XL-3 "

0

N.Novgorod SEC" Fantasy "

0  
 N.Novgorod SEC" RIO "  
 0  
 Novosibirsk SEC" Gallery Novosibirsk "  
 0  
 Novosibirsk SEC" Mega "  
 0  
 Omsk TC" Mega "  
 0  
 Outbound Trade  
 0  
 RostovNaDon Mega shopping center  
 0  
 RostovNaDon Megatsentr Gorizont Ostrovnoy shop...  
 0  
 RostovNaDonu SEC" Megacenter Horizon  
 0  
 Samara Melody shopping center  
 0  
 Samara ParkHouse shopping center  
 0  
 Sergiev Posad 7Ya shopping center  
 0  
 St. Petersburg Nevsky Center shopping center  
 0  
 St. Petersburg Sennaya shopping center  
 0  
 Surgut SEC "City Mall"  
 0  
 Tomsk SEC "Emerald City"  
 0  
 Tyumen SEC "Crystal"  
 0  
 Tyumen SEC "Goodwin"  
 0  
 Tyumen TC "Green Coast "  
 0  
 Ufa TC" Central "  
 0  
 Ufa TC" Family "2  
 0  
 Volga TC "Volga Mall"  
 0  
 Vologda SEC "Marmelad"  
 0  
 Voronezh (Plekhanovskaya, 13)  
 0  
 Voronezh SEC "Maksimir"

0  
 Voronezh SEC City-Park "Grad"  
 0  
 Yakutsk Ordzhonikidze, 56  
 0  
 Yakutsk TC" Central "  
 0  
 Yaroslavl TC" Altair "  
 0  
 Zhukovsky st. Chkalov 39m?  
 0  
 Zhukovsky st. Chkalov 39m<sup>2</sup>  
 0  
 ! Yakutsk Ordzhonikidze, 56 francs  
 0  
 ! Yakutsk TC "Central" fran  
 0

item_category_name	Programs - MAC (Digit) \
shop_name	
Adygea TC "Mega"	0
Balashikha TC "Oktyabr-Kinomir"	0
Chekhov TC" Karnaval "	0
Digital warehouse 1C-Online	56
Internet-shop of emergency situations	0
Kaluga TC "XXI century"	0
Kazan TC "Behetle"	0
Kazan TC "ParkHouse" II	0
Khimki TC" Mega "	0
Kolomna TC "Rio"	0
Krasnoyarsk TC "June"	0
Krasnoyarsk TC "Vzletka Plaza"	0
Kursk TC " Pushkinskiy "	0
Moscow MTRC" Afi Mall "	0
Moscow Shop C21	0
Moscow TC" MEGA Belaya Dacha II "	0
Moscow TC" MEGA Teply Stan "II	0
Moscow TC" New Age "(Novokosino)	0
Moscow TC" Perlovsky "	0
Moscow TC" Semenovskiy "	0
Moscow TC" Serebryany House "	0
Moscow TEC" Atrium "	0
Moscow TK" Budenovskiy "(pav.A2)	0
Moscow TK" Budenovskiy "(pav.K7)	0
Moscow TTS" Areal "(Belyaev)	0
Moscow" Sale "	0
Mytishchi SEC" XL-3 "	0
N.Novgorod SEC" Fantasy "	0

N.Novgorod SEC" RIO "	0
Novosibirsk SEC" Gallery Novosibirsk "	0
Novosibirsk SEC" Mega "	0
Omsk TC" Mega "	0
Outbound Trade	0
RostovNaDon Mega shopping center	0
RostovNaDon Megatsentr Gorizont Ostrovnoy shop...	0
RostovNaDonu SEC" Megacenter Horizon	0
Samara Melody shopping center	0
Samara ParkHouse shopping center	0
Sergiev Posad 7Ya shopping center	0
St. Petersburg Nevsky Center shopping center	0
St. Petersburg Sennaya shopping center	0
Surgut SEC "City Mall"	0
Tomsk SEC "Emerald City"	0
Tyumen SEC "Crystal"	0
Tyumen SEC "Goodwin"	0
Tyumen TC "Green Coast "	0
Ufa TC" Central "	0
Ufa TC" Family "2	0
Volga TC "Volga Mall"	0
Vologda SEC "Marmelad"	0
Voronezh (Plekhanovskaya, 13)	0
Voronezh SEC "Maksimir"	0
Voronezh SEC City-Park "Grad"	0
Yakutsk Ordzhonikidze, 56	0
Yakutsk TC" Central "	0
Yaroslavl TC" Altair "	0
Zhukovsky st. Chkalov 39m?	0
Zhukovsky st. Chkalov 39m <sup>2</sup>	0
! Yakutsk Ordzhonikidze, 56 francs	0
! Yakutsk TC "Central" fran	0

item_category_name	Service \
shop_name	
Adygea TC "Mega"	161
Balashikha TC "Oktyabr-Kinomir"	179
Chekhov TC" Karnaval "	159
Digital warehouse 1C-Online	0
Internet-shop of emergency situations	0
Kaluga TC "XXI century"	185
Kazan TC "Behetle"	0
Kazan TC "ParkHouse" II	178
Khimki TC" Mega "	179
Kolomna TC "Rio"	218
Krasnoyarsk TC "June"	193
Krasnoyarsk TC "Vzletka Plaza"	62
Kursk TC " Pushkinskiy "	176



Moscow MTRC" Afi Mall "	218
Moscow Shop C21	262
Moscow TC" MEGA Belaya Dacha II "	373
Moscow TC" MEGA Teply Stan "II	248
Moscow TC" New Age "(Novokosino)	162
Moscow TC" Perlovsky "	156
Moscow TC" Semenovskiy "	478
Moscow TC" Serebryany House "	8
Moscow TEC" Atrium "	288
Moscow TK" Budenovskiy "(pav.A2)	8
Moscow TK" Budenovskiy "(pav.K7)	245
Moscow TTS" Areal "(Belyaev)	265
Moscow" Sale "	0
Mytishchi SEC" XL-3 "	49
N.Novgorod SEC" Fantasy "	279
N.Novgorod SEC" RIO "	72
Novosibirsk SEC" Gallery Novosibirsk "	1
Novosibirsk SEC" Mega "	129
Omsk TC" Mega "	274
Outbound Trade	4
RostovNaDon Mega shopping center	97
RostovNaDon Megatsentr Gorizont Ostrovnoy shop...	4
RostovNaDonu SEC" Megacenter Horizon	104
Samara Melody shopping center	90
Samara ParkHouse shopping center	145
Sergiev Posad 7Ya shopping center	226
St. Petersburg Nevsky Center shopping center	382
St. Petersburg Sennaya shopping center	165
Surgut SEC "City Mall"	212
Tomsk SEC "Emerald City"	195
Tyumen SEC "Crystal"	105
Tyumen SEC "Goodwin"	117
Tyumen TC "Green Coast "	0
Ufa TC" Central "	156
Ufa TC" Family "2	223
Volga TC "Volga Mall"	302
Vologda SEC "Marmelad"	195
Voronezh (Plekhanovskaya, 13)	330
Voronezh SEC "Maksimir"	328
Voronezh SEC City-Park "Grad"	1
Yakutsk Ordzhonikidze, 56	160
Yakutsk TC" Central "	116
Yaroslavl TC" Altair "	205
Zhukovsky st. Chkalov 39m?	0
Zhukovsky st. Chkalov 39m <sup>2</sup>	0
! Yakutsk Ordzhonikidze, 56 francs	0
! Yakutsk TC "Central" fran	0

item_category_name	Service - Tickets \
shop_name	
Adygea TC "Mega"	0
Balashikha TC "Oktyabr-Kinomir"	44
Chekhov TC" Karnaval "	36
Digital warehouse 1C-Online	3
Internet-shop of emergency situations	169
Kaluga TC "XXI century"	46
Kazan TC "Behetle"	0
Kazan TC "ParkHouse" II	18
Khimki TC" Mega "	0
Kolomna TC "Rio"	73
Krasnoyarsk TC "June"	4
Krasnoyarsk TC "Vzletka Plaza"	0
Kursk TC " Pushkinskiy "	20
Moscow MTRC" Afi Mall "	65
Moscow Shop C21	53
Moscow TC" MEGA Belaya Dacha II "	19
Moscow TC" MEGA Teply Stan "II	109
Moscow TC" New Age "(Novokosino)	1
Moscow TC" Perlovsky "	0
Moscow TC" Semenovsky "	134
Moscow TC" Serebryany House "	0
Moscow TEC" Atrium "	144
Moscow TK" Budenovskiy "(pav.A2)	0
Moscow TK" Budenovskiy "(pav.K7)	41
Moscow TTS" Areal "(Belyaev)	61
Moscow" Sale "	0
Mytishchi SEC" XL-3 "	0
N.Novgorod SEC" Fantasy "	30
N.Novgorod SEC" RIO "	18
Novosibirsk SEC" Gallery Novosibirsk "	0
Novosibirsk SEC" Mega "	2
Omsk TC" Mega "	5
Outbound Trade	0
RostovNaDon Mega shopping center	3
RostovNaDon Megatsentr Gorizont Ostrovnoy shop...	0
RostovNaDonu SEC" Megacenter Horizon	9
Samara Melody shopping center	10
Samara ParkHouse shopping center	10
Sergiev Posad 7Ya shopping center	65
St. Petersburg Nevsky Center shopping center	43
St. Petersburg Sennaya shopping center	0
Surgut SEC "City Mall"	3
Tomsk SEC "Emerald City"	0
Tyumen SEC "Crystal"	0
Tyumen SEC "Goodwin"	10
Tyumen TC "Green Coast "	0

Ufa TC" Central "	2
Ufa TC" Family "2	4
Volga TC "Volga Mall"	2
Vologda SEC "Marmelad"	11
Voronezh (Plekhanovskaya, 13)	32
Voronezh SEC "Maksimir"	13
Voronezh SEC City-Park "Grad"	0
Yakutsk Ordzhonikidze, 56	0
Yakutsk TC" Central "	0
Yaroslavl TC" Altair "	12
Zhukovsky st. Chkalov 39m?	0
Zhukovsky st. Chkalov 39m <sup>2</sup>	0
! Yakutsk Ordzhonikidze, 56 francs	0
! Yakutsk TC "Central" fran	0

item_category_name	Tickets (Digital) \
shop_name	
Adygea TC "Mega"	3
Balashikha TC "Oktyabr-Kinomir"	35
Chekhov TC" Karnaval "	41
Digital warehouse 1C-Online	507
Internet-shop of emergency situations	111
Kaluga TC "XXI century"	46
Kazan TC "Behetle"	0
Kazan TC "ParkHouse" II	9
Khimki TC" Mega "	76
Kolomna TC "Rio"	66
Krasnoyarsk TC "June"	0
Krasnoyarsk TC "Vzletka Plaza"	1
Kursk TC " Pushkinskiy "	21
Moscow MTRC" Afi Mall "	57
Moscow Shop C21	55
Moscow TC" MEGA Belaya Dacha II "	82
Moscow TC" MEGA Teply Stan "II	57
Moscow TC" New Age "(Novokosino)	44
Moscow TC" Perlovsky "	70
Moscow TC" Semenovskiy "	114
Moscow TC" Serebryany House "	0
Moscow TEC" Atrium "	98
Moscow TK" Budenovskiy "(pav.A2)	0
Moscow TK" Budenovskiy "(pav.K7)	39
Moscow TTS" Areal "(Belyaev)	62
Moscow" Sale "	0
Mytishchi SEC" XL-3 "	8
N.Novgorod SEC" Fantasy "	25
N.Novgorod SEC" RIO "	5
Novosibirsk SEC" Gallery Novosibirsk "	0
Novosibirsk SEC" Mega "	1

Omsk TC" Mega "	2
Outbound Trade	0
RostovNaDon Mega shopping center	9
RostovNaDon Megatsentr Gorizont Ostrovnoy shop...	0
RostovNaDonu SEC" Megacenter Horizon	7
Samara Melody shopping center	4
Samara ParkHouse shopping center	7
Sergiev Posad 7Ya shopping center	52
St. Petersburg Nevsky Center shopping center	26
St. Petersburg Sennaya shopping center	22
Surgut SEC "City Mall"	2
Tomsk SEC "Emerald City"	0
Tyumen SEC "Crystal"	0
Tyumen SEC "Goodwin"	2
Tyumen TC "Green Coast "	0
Ufa TC" Central "	3
Ufa TC" Family "2	8
Volga TC "Volga Mall"	2
Vologda SEC "Marmelad"	26
Voronezh (Plekhanovskaya, 13)	25
Voronezh SEC "Maksimir"	8
Voronezh SEC City-Park "Grad"	0
Yakutsk Ordzhonikidze, 56	5
Yakutsk TC" Central "	1
Yaroslavl TC" Altair "	27
Zhukovsky st. Chkalov 39m?	0
Zhukovsky st. Chkalov 39m <sup>2</sup>	0
! Yakutsk Ordzhonikidze, 56 francs	0
! Yakutsk TC "Central" fran	0
item_category_name	PC - Headsets / Headphones
shop_name	
Adygea TC "Mega"	0
Balashikha TC "Oktyabr-Kinomir"	0
Chekhov TC" Karnaval "	0
Digital warehouse 1C-Online	0
Internet-shop of emergency situations	0
Kaluga TC "XXI century"	0
Kazan TC "Behetle"	0
Kazan TC "ParkHouse" II	0
Khimki TC" Mega "	0
Kolomna TC "Rio"	0
Krasnoyarsk TC "June"	0
Krasnoyarsk TC "Vzletka Plaza"	0
Kursk TC " Pushkinskiy "	0
Moscow MTRC" Afi Mall "	0
Moscow Shop C21	0
Moscow TC" MEGA Belaya Dacha II "	0

Moscow TC" MEGA Teply Stan "II	0
Moscow TC" New Age "(Novokosino)	0
Moscow TC" Perlovsky "	0
Moscow TC" Semenovskiy "	0
Moscow TC" Serebryany House "	0
Moscow TEC" Atrium "	1
Moscow TK" Budenovskiy "(pav.A2)	0
Moscow TK" Budenovskiy "(pav.K7)	0
Moscow TTS" Areal "(Belyaev)	0
Moscow" Sale "	0
Mytishchi SEC" XL-3 "	0
N.Novgorod SEC" Fantasy "	0
N.Novgorod SEC" RIO "	0
Novosibirsk SEC" Gallery Novosibirsk "	0
Novosibirsk SEC" Mega "	0
Omsk TC" Mega "	1
Outbound Trade	0
RostovNaDon Mega shopping center	0
RostovNaDon Megatsentr Gorizont Ostrovnoy shop...	0
RostovNaDonu SEC" Megacenter Horizon	0
Samara Melody shopping center	0
Samara ParkHouse shopping center	0
Sergiev Posad 7Ya shopping center	0
St. Petersburg Nevsky Center shopping center	0
St. Petersburg Sennaya shopping center	0
Surgut SEC "City Mall"	0
Tomsk SEC "Emerald City"	0
Tyumen SEC "Crystal"	0
Tyumen SEC "Goodwin"	0
Tyumen TC "Green Coast "	0
Ufa TC" Central "	0
Ufa TC" Family "2	0
Volga TC "Volga Mall"	0
Vologda SEC "Marmelad"	0
Voronezh (Plekhanovskaya, 13)	0
Voronezh SEC "Maksimir"	0
Voronezh SEC City-Park "Grad"	0
Yakutsk Ordzhonikidze, 56	0
Yakutsk TC" Central "	0
Yaroslavl TC" Altair "	0
Zhukovsky st. Chkalov 39m?	0
Zhukovsky st. Chkalov 39m <sup>2</sup>	0
! Yakutsk Ordzhonikidze, 56 francs	1
! Yakutsk TC "Central" fran	0

[60 rows x 84 columns]

Frequency of 'Low' price range: 0

item\_category\_name

```
#stationarity analysis
```



```

[ ]: # Convert the date column to datetime format
final_dataset['date'] = pd.to_datetime(final_dataset['date'], format='%d.%m.%Y')

monthly_data = final_dataset.groupby(final_dataset['date'].dt.to_period('M')).
    ↪agg({
        'item_cnt_month': 'sum',
    }).reset_index()

def adf_test(timeseries):
    result = adfuller(timeseries, autolag='AIC')
    print('ADF Statistic:', result[0])
    print('p-value:', result[1])
    print('Critical Values:')
    for key, value in result[4].items():
        print(f'    {key}: {value}')

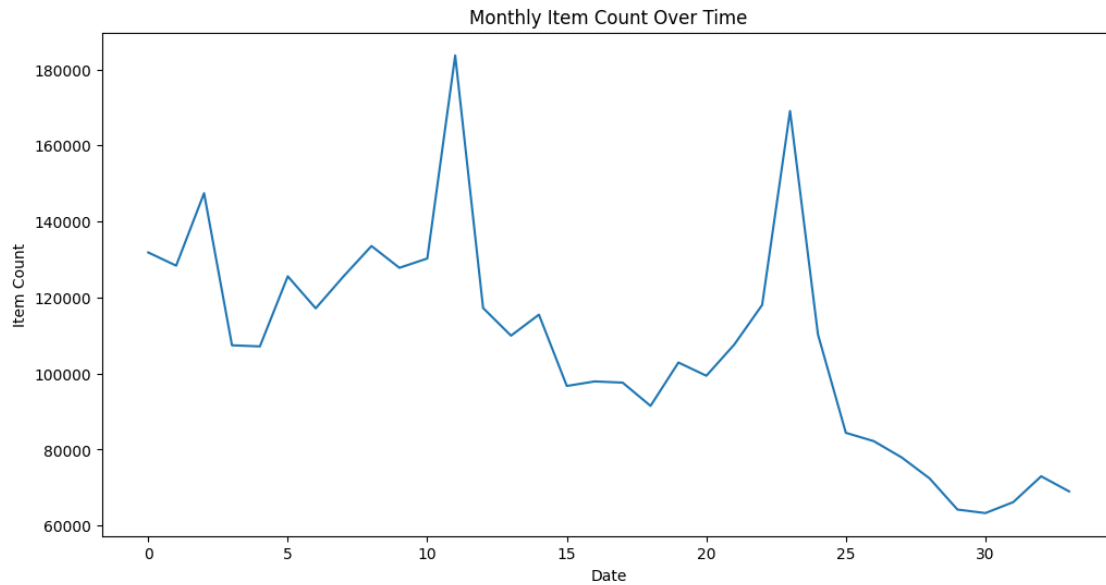
    if result[1] <= 0.05:
        print("Stationary (Reject the null hypothesis)")
    else:
        print("Non-Stationary (Fail to reject the null hypothesis)")

item_cnt_month_series = monthly_data['item_cnt_month']

plt.figure(figsize=(12, 6))
plt.plot(item_cnt_month_series)
plt.title('Monthly Item Count Over Time')
plt.xlabel('Date')
plt.ylabel('Item Count')
plt.show()

adf_test(item_cnt_month_series)

```



ADF Statistic: -2.372251519825604

p-value: 0.1497218397733845

Critical Values:

1%: -3.6461350877925254

5%: -2.954126991123355

10%: -2.6159676124885216

Non-Stationary (Fail to reject the null hypothesis)

```
[ ]: print("\n\nHead of final_dataset:\n")
      print(final_dataset.head(20))
      print(final_dataset.shape)
      print(final_dataset.info())
```

Head of final\_dataset:

	date	date_num	month_name	year_num	shop_id	\
0	2013-01-02	02	1	0	59	
1	2013-01-03	03	1	0	25	
3	2013-01-06	06	1	0	25	
4	2013-01-15	15	1	0	25	
5	2013-01-10	10	1	0	25	
6	2013-01-02	02	1	0	25	
7	2013-01-04	04	1	0	25	
8	2013-01-11	11	1	0	25	
9	2013-01-03	03	1	0	25	
10	2013-01-03	03	1	0	25	



11	2013-01-05	05	1	0	25
12	2013-01-07	07	1	0	25
13	2013-01-08	08	1	0	25
14	2013-01-10	10	1	0	25
15	2013-01-11	11	1	0	25
16	2013-01-13	13	1	0	25
17	2013-01-16	16	1	0	25
18	2013-01-26	26	1	0	25
19	2013-01-27	27	1	0	25
20	2013-01-09	09	1	0	25

	shop_name	item_id	\
0	Yaroslavl TC" Altair "	22154	
1	Moscow TEC" Atrium "	2552	
3	Moscow TEC" Atrium "	2554	
4	Moscow TEC" Atrium "	2555	
5	Moscow TEC" Atrium "	2564	
6	Moscow TEC" Atrium "	2565	
7	Moscow TEC" Atrium "	2572	
8	Moscow TEC" Atrium "	2572	
9	Moscow TEC" Atrium "	2573	
10	Moscow TEC" Atrium "	2574	
11	Moscow TEC" Atrium "	2574	
12	Moscow TEC" Atrium "	2574	
13	Moscow TEC" Atrium "	2574	
14	Moscow TEC" Atrium "	2574	
15	Moscow TEC" Atrium "	2574	
16	Moscow TEC" Atrium "	2574	
17	Moscow TEC" Atrium "	2574	
18	Moscow TEC" Atrium "	2574	
19	Moscow TEC" Atrium "	2574	
20	Moscow TEC" Atrium "	2593	

	item_name	category_id	\
0	SCENE 2012 (BD)	37	
1	DEEP PURPLE The House Of Blue Light LP	58	
3	DEEP PURPLE Who Do You Think We Are LP	58	
4	DEEP PURPLE 30 Very Best Of 2CD (Businesses).	56	
5	DEEP PURPLE Perihelion: Live In Concert DVD (C...	59	
6	DEEP PURPLE Stormbringer (firms).	56	
7	DEFTONES Koi No Yokan	55	
8	DEFTONES Koi No Yokan	55	
9	DEL REY LANA Born To Die	55	
10	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
11	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
12	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
13	DEL REY LANA Born To Die The Paradise Edition 2CD	55	
14	DEL REY LANA Born To Die The Paradise Edition 2CD	55	

15	DEL REY LANA Born To Die The Paradise Edition 2CD	55
16	DEL REY LANA Born To Die The Paradise Edition 2CD	55
17	DEL REY LANA Born To Die The Paradise Edition 2CD	55
18	DEL REY LANA Born To Die The Paradise Edition 2CD	55
19	DEL REY LANA Born To Die The Paradise Edition 2CD	55
20	DEPECHE MODE Music For The Masses	55

	item_category_name	item_price	item_cnt_month	revenue	\
0	Cinema - Blu-Ray	999.00	1	999.00	
1	Music - Vinyl	899.00	1	899.00	
3	Music - Vinyl	1709.05	1	1709.05	
4	Music - CD of corporate production	1099.00	1	1099.00	
5	Music - Music video	349.00	1	349.00	
6	Music - CD of corporate production	549.00	1	549.00	
7	Music - CD of local production	239.00	1	239.00	
8	Music - CD of local production	299.00	1	299.00	
9	Music - CD of local production	299.00	3	897.00	
10	Music - CD of local production	399.00	2	798.00	
11	Music - CD of local production	399.00	1	399.00	
12	Music - CD of local production	399.00	1	399.00	
13	Music - CD of local production	399.00	2	798.00	
14	Music - CD of local production	399.00	1	399.00	
15	Music - CD of local production	399.00	2	798.00	
16	Music - CD of local production	399.00	1	399.00	
17	Music - CD of local production	399.00	1	399.00	
18	Music - CD of local production	399.00	1	399.00	
19	Music - CD of local production	399.00	1	399.00	
20	Music - CD of local production	279.00	1	279.00	

	price_range	log_revenue	scaled_revenue
0	900-100000	6.906755	0.000546
1	800-900	6.801283	0.000491
3	900-100000	7.443693	0.000934
4	900-100000	7.002156	0.000601
5	300-400	5.855072	0.000191
6	500-600	6.308098	0.000300
7	200-300	5.476464	0.000131
8	200-300	5.700444	0.000163
9	200-300	6.799056	0.000490
10	300-400	6.682109	0.000436
11	300-400	5.988961	0.000218
12	300-400	5.988961	0.000218
13	300-400	6.682109	0.000436
14	300-400	5.988961	0.000218
15	300-400	6.682109	0.000436
16	300-400	5.988961	0.000218
17	300-400	5.988961	0.000218
18	300-400	5.988961	0.000218

```

19      300-400      5.988961      0.000218
20      200-300      5.631212      0.000152
(2928483, 16)
<class 'pandas.core.frame.DataFrame'>
Index: 2928483 entries, 0 to 2935848
Data columns (total 16 columns):
 #   Column              Dtype
---  -
 0   date                datetime64[ns]
 1   date_num            object
 2   month_name          int64
 3   year_num            int64
 4   shop_id             int64
 5   shop_name           object
 6   item_id             int64
 7   item_name           object
 8   category_id         int64
 9   item_category_name  object
10   item_price          float64
11   item_cnt_month      int64
12   revenue             float64
13   price_range         category
14   log_revenue         float64
15   scaled_revenue      float64
dtypes: category(1), datetime64[ns](1), float64(4), int64(6), object(4)
memory usage: 360.3+ MB
None

```

```

[ ]: #export the final dataset to csv file
final_dataset.to_csv('./data-set/output/final_dataset_with_cleaning.csv',
    ↪index=False)

```

## 1.5 Model Development, Error Analysis & Comparison

```

[ ]: #prepare the data for modeling
df = pd.read_csv('./data-set/sales_train.csv')
#rename item_cnt_day column
df.rename(columns={'item_cnt_day': 'item_count'}, inplace=True)
#removes duplicates
df.drop_duplicates(inplace=True)
#outlier treatment
df = df[(df['item_count'] > 0) & (df['item_count'] < 307980)]
df = df[df['item_count'] < 1000]
#handles incorrect data
df = df[(df['item_price'] > 0) & (df['item_price'] < 100000)]
#converts date column to datetime format
df['date'] = pd.to_datetime(df['date'], format='%d.%m.%Y')

```

```

#convert date to year-month format
df['year-month'] = df['date'].dt.strftime('%Y-%m')
#drop date column and item_price column
df.drop(columns=['date', 'item_price'], inplace=True)
# group features
df_train_group = df.groupby(['year-month', 'shop_id', 'item_id']).sum().
    ↪reset_index()
# pivot table
df = df_train_group.pivot_table(index=['shop_id', 'item_id'],
    ↪columns='year-month', values='item_count', fill_value=0).reset_index()

print(df.head(10))
print(df.shape)
print(df.info())

```

year-month	shop_id	item_id	2013-01	2013-02	2013-03	2013-04	2013-05	\
0	0	30	0	31	0	0	0	
1	0	31	0	11	0	0	0	
2	0	32	6	10	0	0	0	
3	0	33	3	3	0	0	0	
4	0	35	1	14	0	0	0	
5	0	36	0	1	0	0	0	
6	0	40	0	1	0	0	0	
7	0	42	0	1	0	0	0	
8	0	43	1	0	0	0	0	
9	0	49	0	2	0	0	0	

year-month	2013-06	2013-07	2013-08	...	2015-01	2015-02	2015-03	\
0	0	0	0	...	0	0	0	
1	0	0	0	...	0	0	0	
2	0	0	0	...	0	0	0	
3	0	0	0	...	0	0	0	
4	0	0	0	...	0	0	0	
5	0	0	0	...	0	0	0	
6	0	0	0	...	0	0	0	
7	0	0	0	...	0	0	0	
8	0	0	0	...	0	0	0	
9	0	0	0	...	0	0	0	

year-month	2015-04	2015-05	2015-06	2015-07	2015-08	2015-09	2015-10
0	0	0	0	0	0	0	0
1	0	0	0	0	0	0	0
2	0	0	0	0	0	0	0
3	0	0	0	0	0	0	0
4	0	0	0	0	0	0	0
5	0	0	0	0	0	0	0
6	0	0	0	0	0	0	0

7	0	0	0	0	0	0	0
8	0	0	0	0	0	0	0
9	0	0	0	0	0	0	0

[10 rows x 36 columns]

(424097, 36)

<class 'pandas.core.frame.DataFrame'>

RangeIndex: 424097 entries, 0 to 424096

Data columns (total 36 columns):

#	Column	Non-Null Count	Dtype
0	shop_id	424097 non-null	int64
1	item_id	424097 non-null	int64
2	2013-01	424097 non-null	int64
3	2013-02	424097 non-null	int64
4	2013-03	424097 non-null	int64
5	2013-04	424097 non-null	int64
6	2013-05	424097 non-null	int64
7	2013-06	424097 non-null	int64
8	2013-07	424097 non-null	int64
9	2013-08	424097 non-null	int64
10	2013-09	424097 non-null	int64
11	2013-10	424097 non-null	int64
12	2013-11	424097 non-null	int64
13	2013-12	424097 non-null	int64
14	2014-01	424097 non-null	int64
15	2014-02	424097 non-null	int64
16	2014-03	424097 non-null	int64
17	2014-04	424097 non-null	int64
18	2014-05	424097 non-null	int64
19	2014-06	424097 non-null	int64
20	2014-07	424097 non-null	int64
21	2014-08	424097 non-null	int64
22	2014-09	424097 non-null	int64
23	2014-10	424097 non-null	int64
24	2014-11	424097 non-null	int64
25	2014-12	424097 non-null	int64
26	2015-01	424097 non-null	int64
27	2015-02	424097 non-null	int64
28	2015-03	424097 non-null	int64
29	2015-04	424097 non-null	int64
30	2015-05	424097 non-null	int64
31	2015-06	424097 non-null	int64
32	2015-07	424097 non-null	int64
33	2015-08	424097 non-null	int64
34	2015-09	424097 non-null	int64
35	2015-10	424097 non-null	int64

dtypes: int64(36)

memory usage: 116.5 MB  
None

```
[ ]: #export the final dataset to csv file
final_dataset.to_csv('./data-set/output/dataset_for_modeling.csv', index=False)
```

```
[ ]: # Create X and y variables for train and test sets
X = df[df.columns[:-1]]
y = df[df.columns[-1]]

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,
    ↪random_state=42)

print(X_train.shape)
print(X_test.shape)
print(y_train.shape)
print(y_test.shape)
```

```
(339277, 35)
(84820, 35)
(339277,)
(84820,)
```

```
[ ]: #creating evaluation metrics
scores_and_names = []

# Create a function to evaluate the model
def evaluate_the_model(y_true, y_pred, model_name, model):

    # Calculate the MAE
    mae = mean_absolute_error(y_true, y_pred)
    print(f"MAE for {model_name}: {mae:.5f}")

    # Calculate the MSE
    mse = mean_squared_error(y_true, y_pred)
    print(f"MSE for {model_name}: {mse:.5f}")

    # Calculate the RMSE
    rmse = np.sqrt(mse)
    print(f"RMSE for {model_name}: {rmse:.5f}")

    # Plot the predictions vs. the actual values
    plt.figure(figsize=(12, 6))
    plt.plot(y_true, label='Actual Values')
    plt.plot(y_pred, label='Predicted Values')
    plt.title(f'Predictions vs. Actual Values ({model_name})')
    plt.xlabel('Observation')
```

```
plt.ylabel('Item Count')
plt.legend()
plt.show()

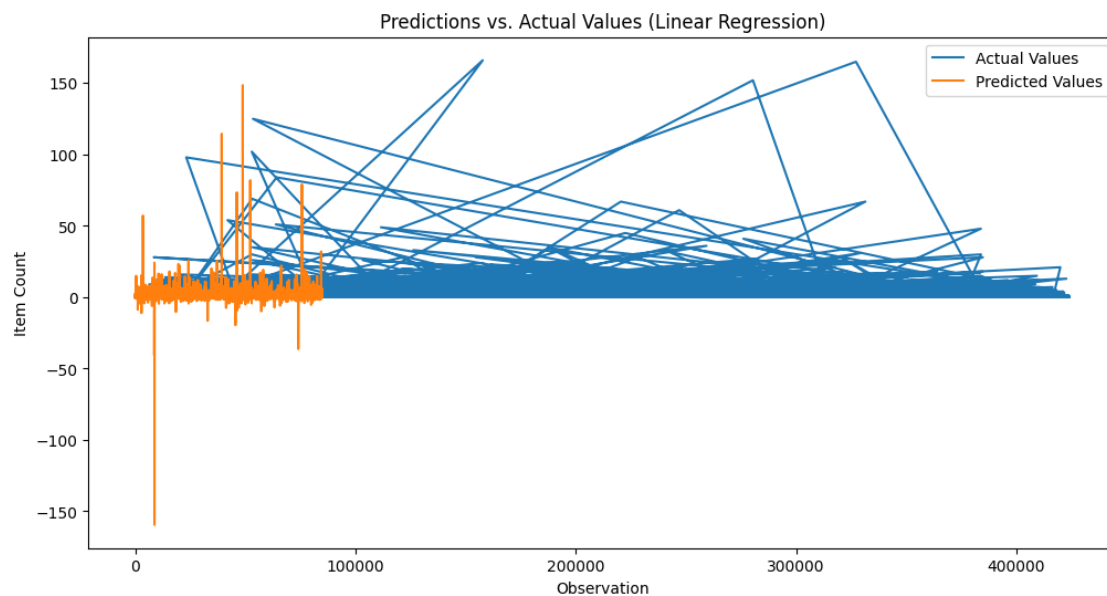
scores_and_names.append((model_name, rmse))
```

### 1.5.1 linear regression

```
[ ]: # Create a Linear Regression model
lin_reg = LinearRegression()
lin_reg.fit(X_train, y_train)
y_pred = lin_reg.predict(X_test)

evaluate_the_model(y_test, y_pred, 'Linear Regression', lin_reg)
```

MAE for Linear Regression: 0.29022  
MSE for Linear Regression: 2.39626  
RMSE for Linear Regression: 1.54798



### 1.5.2 Logistic Regression

```
[ ]: # Create a logistic regression model
log_reg = LogisticRegression()
log_reg.fit(X_train, y_train)
y_pred = log_reg.predict(X_test)
```

```
evaluate_the_model(y_test, y_pred, 'Logistic Regression', log_reg)
```

C:\Users\srumel\AppData\Roaming\Python\Python311\site-packages\sklearn\linear\_model\\_logistic.py:460: ConvergenceWarning: lbfgs failed to converge (status=1):

STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.

Increase the number of iterations (max\_iter) or scale the data as shown in:

<https://scikit-learn.org/stable/modules/preprocessing.html>

Please also refer to the documentation for alternative solver options:

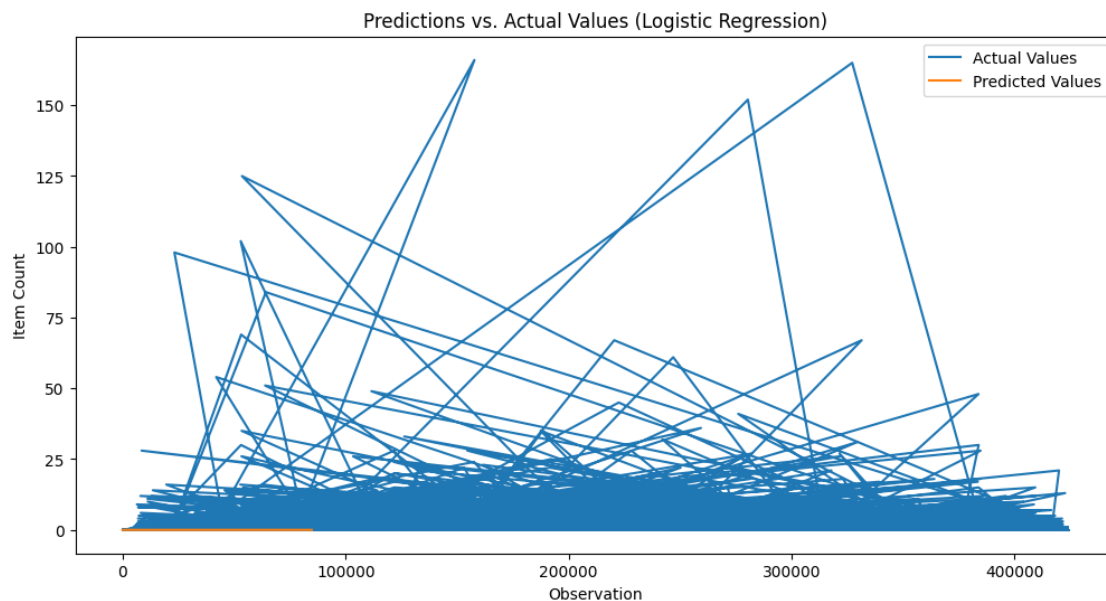
[https://scikit-learn.org/stable/modules/linear\\_model.html#logistic-regression](https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression)

```
n_iter_i = _check_optimize_result(
```

MAE for Logistic Regression: 0.16065

MSE for Logistic Regression: 2.61462

RMSE for Logistic Regression: 1.61698



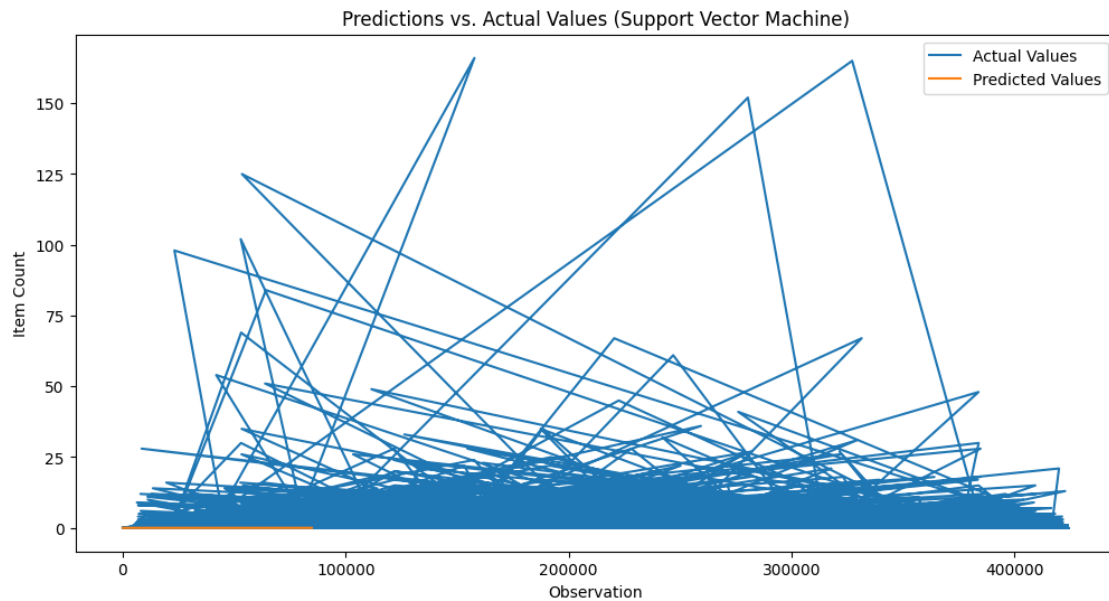
### 1.5.3 SVM

```
[ ]: # create a support vector machine model
svm = SVC()
svm.fit(X_train, y_train)
y_pred = svm.predict(X_test)

evaluate_the_model(y_test, y_pred, 'Support Vector Machine', svm)
```



MAE for Support Vector Machine: 0.16065  
MSE for Support Vector Machine: 2.61462  
RMSE for Support Vector Machine: 1.61698

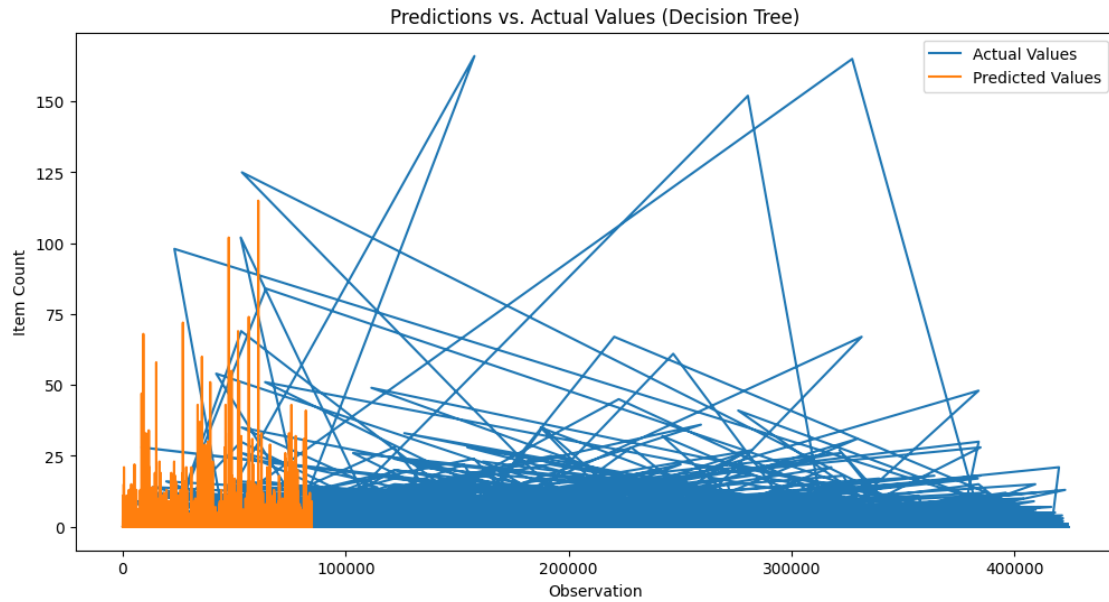


#### 1.5.4 Decision Tree

```
[ ]: # create a decision tree model
dt = DecisionTreeClassifier()
dt.fit(X_train, y_train)
y_pred = dt.predict(X_test)

evaluate_the_model(y_test, y_pred, 'Decision Tree', dt)
```

MAE for Decision Tree: 0.17817  
MSE for Decision Tree: 1.88140  
RMSE for Decision Tree: 1.37164



### 1.5.5 random forest

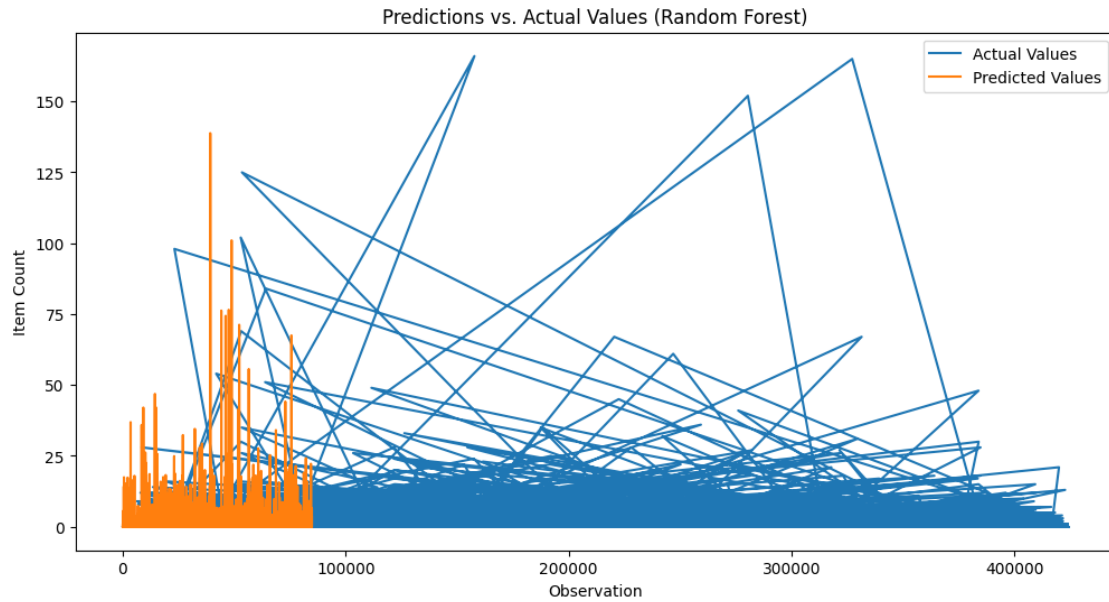
```
[ ]: # create a random forest model
rf = RandomForestRegressor()
rf.fit(X_train, y_train)
y_pred = rf.predict(X_test)

evaluate_the_model(y_test, y_pred, 'Random Forest', rf)
```

MAE for Random Forest: 0.16314

MSE for Random Forest: 1.13545

RMSE for Random Forest: 1.06558



### 1.5.6 Stochastic Gradient Descent

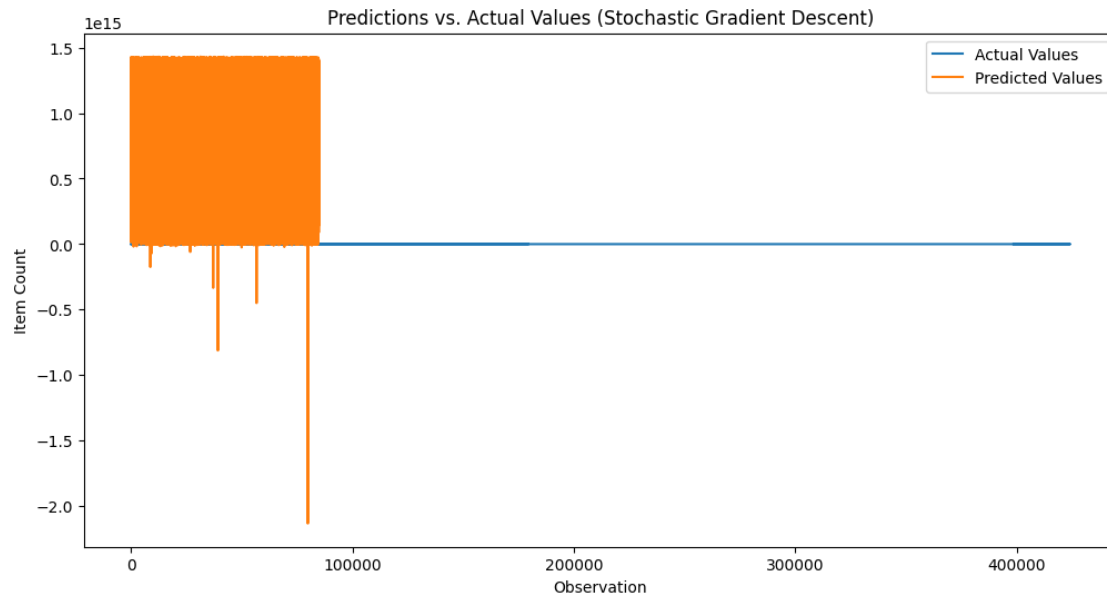
```
[ ]: #create a stochastic gradient descent model
sgd_reg = SGDRegressor()
sgd_reg.fit(X_train, y_train)
y_pred = sgd_reg.predict(X_test)

evaluate_the_model(y_test, y_pred, 'Stochastic Gradient Descent', sgd_reg)
```

MAE for Stochastic Gradient Descent: 735578716284226.25000

MSE for Stochastic Gradient Descent: 698048155002567956219484438528.00000

RMSE for Stochastic Gradient Descent: 835492761789453.12500

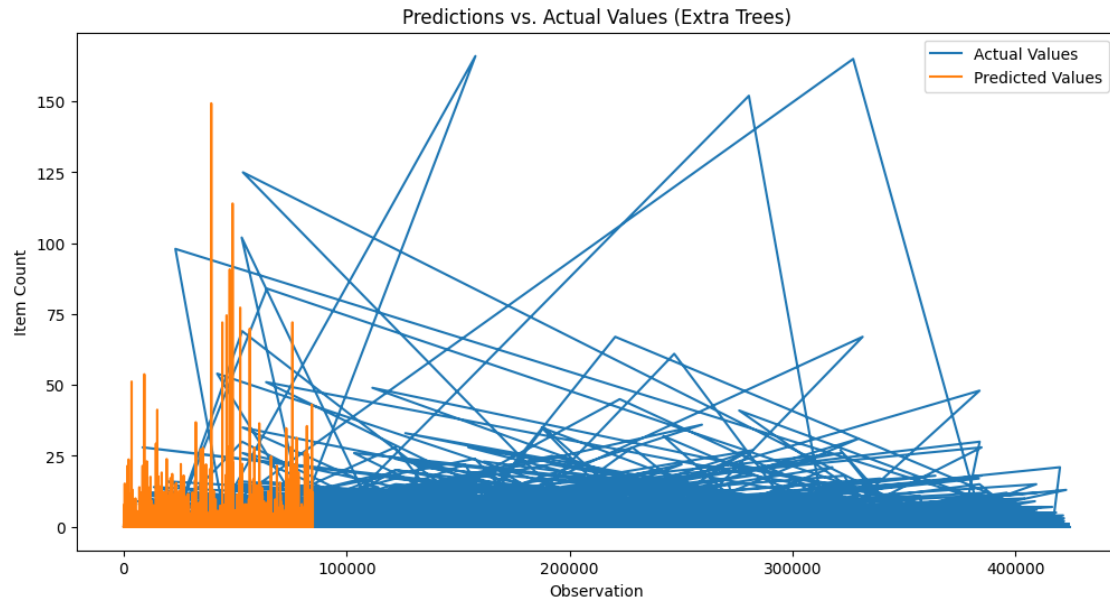


### 1.5.7 xtra tree

```
[ ]: #create a extra trees model
et = ExtraTreesRegressor()
et.fit(X_train, y_train)
y_pred = et.predict(X_test)

evaluate_the_model(y_test, y_pred, 'Extra Trees', et)
```

MAE for Extra Trees: 0.17591  
MSE for Extra Trees: 1.14391  
RMSE for Extra Trees: 1.06954



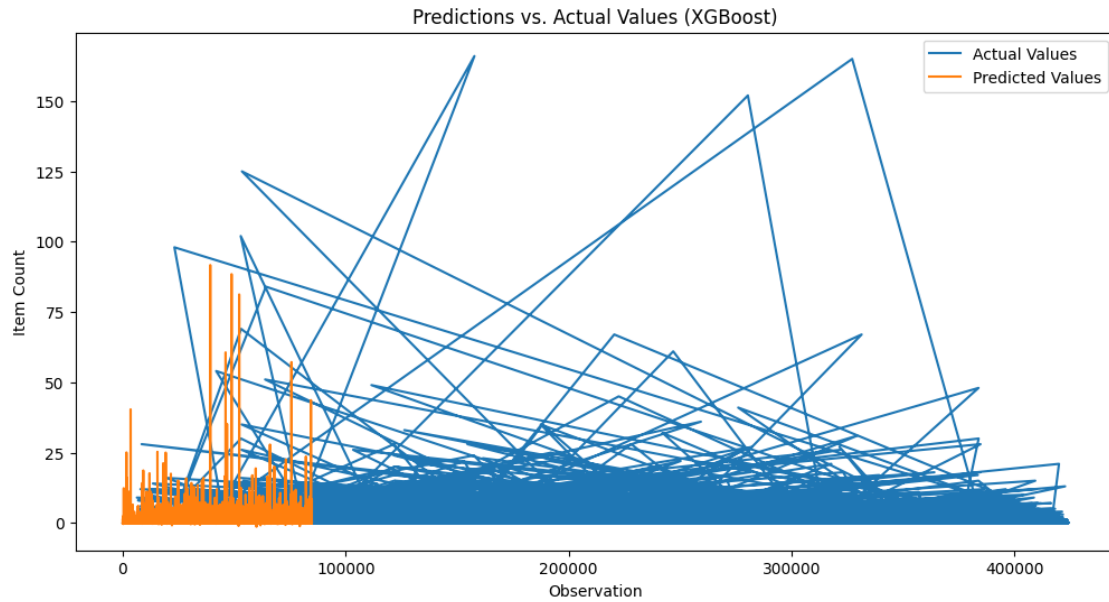
### 1.5.8 XGBoost

```
[ ]: #create a xgboost model  
xgb = XGBRegressor()  
xgb.fit(X_train, y_train)  
y_pred = xgb.predict(X_test)  
  
evaluate_the_model(y_test, y_pred, 'XGBoost', xgb)
```

MAE for XGBoost: 0.21046

MSE for XGBoost: 1.69069

RMSE for XGBoost: 1.30027



### 1.5.9 ridge regression

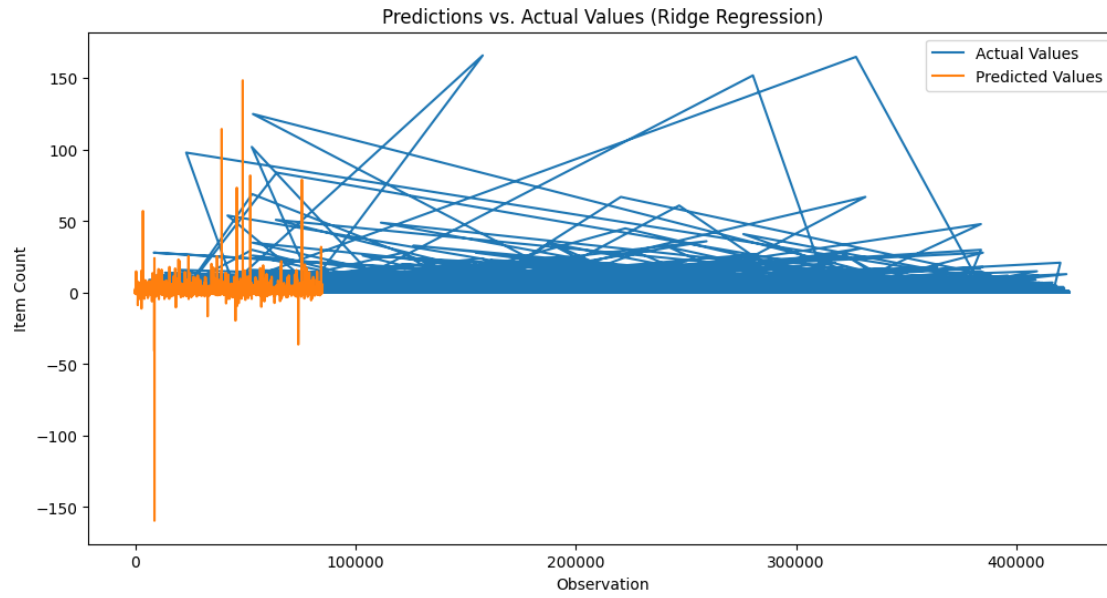
```
[ ]: #create ridge regression model
ridge = Ridge()
ridge.fit(X_train, y_train)
y_pred = ridge.predict(X_test)

evaluate_the_model(y_test, y_pred, 'Ridge Regression', ridge)
```

MAE for Ridge Regression: 0.29022

MSE for Ridge Regression: 2.39625

RMSE for Ridge Regression: 1.54798



### 1.5.10 lasso regression

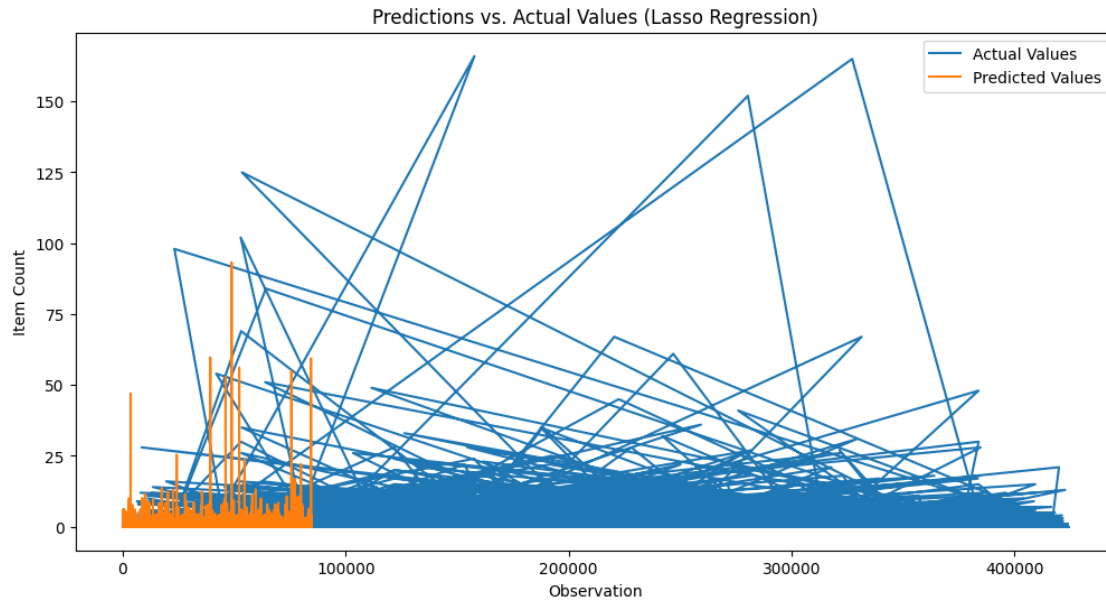
```
[ ]: #create lasso regression model
lasso = Lasso()
lasso.fit(X_train, y_train)
y_pred = lasso.predict(X_test)

evaluate_the_model(y_test, y_pred, 'Lasso Regression', lasso)
```

MAE for Lasso Regression: 0.26487

MSE for Lasso Regression: 2.11585

RMSE for Lasso Regression: 1.45460



### 1.5.11 ARIMA

```
[ ]: #create ARIMA model
arima = ARIMA(y_train, order=(1, 1, 1))
model = arima.fit()
y_pred = model.predict(start=len(y_train), end=len(y_train) + len(X_test) - 1,
                        exog=X_test)

evaluate_the_model(y_test, y_pred, 'ARIMA', arima)
```

```
c:\Users\srume\anaconda3\Lib\site-
packages\statsmodels\tsa\base\tsa_model.py:473: ValueWarning: An unsupported
index was provided and will be ignored when e.g. forecasting.
    self._init_dates(dates, freq)
c:\Users\srume\anaconda3\Lib\site-
packages\statsmodels\tsa\base\tsa_model.py:473: ValueWarning: An unsupported
index was provided and will be ignored when e.g. forecasting.
    self._init_dates(dates, freq)
c:\Users\srume\anaconda3\Lib\site-
packages\statsmodels\tsa\base\tsa_model.py:473: ValueWarning: An unsupported
index was provided and will be ignored when e.g. forecasting.
    self._init_dates(dates, freq)
c:\Users\srume\anaconda3\Lib\site-
packages\statsmodels\tsa\base\tsa_model.py:836: ValueWarning: No supported index
is available. Prediction results will be given with an integer index beginning
at `start`.
    return get_prediction_index(
```



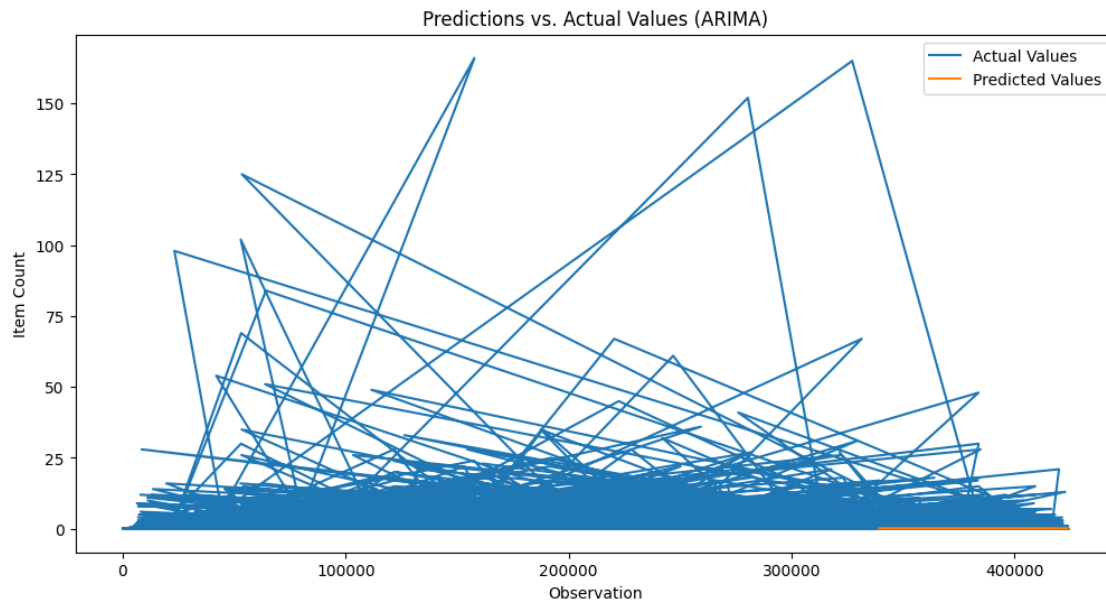
```
c:\Users\srume\anaconda3\Lib\site-  
packages\statsmodels\tsa\base\tsa_model.py:836: FutureWarning: No supported  
index is available. In the next version, calling this method in a model without  
a supported index will result in an exception.
```

```
    return get_prediction_index()
```

MAE for ARIMA: 0.30035

MSE for ARIMA: 2.58882

RMSE for ARIMA: 1.60898



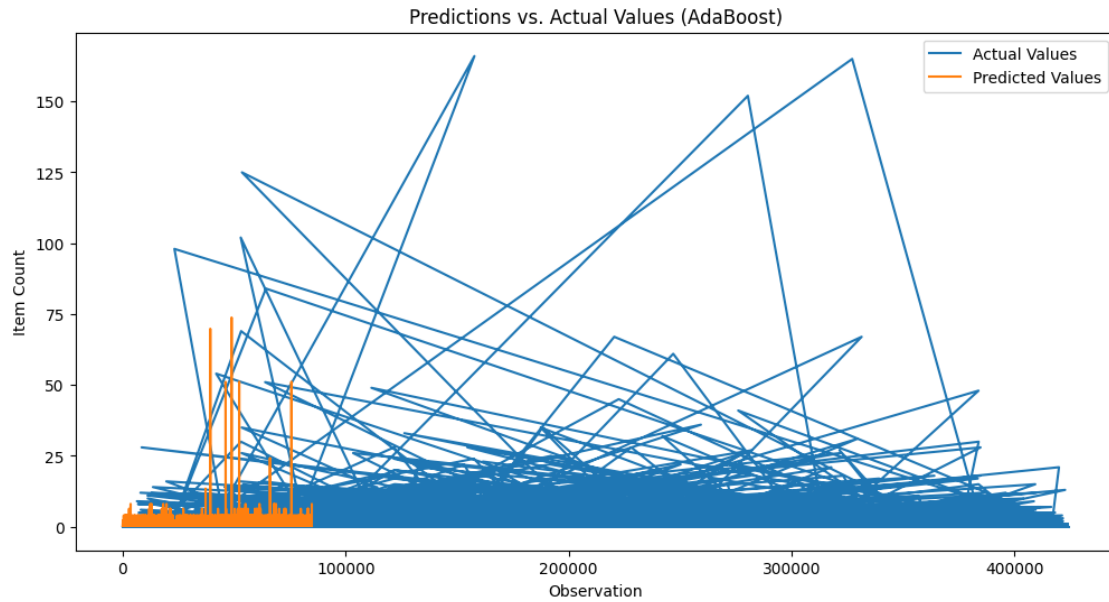
### 1.5.12 ADABOOST

```
[ ]: #create adaboost model  
ada = AdaBoostRegressor()  
ada.fit(X_train, y_train)  
y_pred = ada.predict(X_test)  
  
evaluate_the_model(y_test, y_pred, 'AdaBoost', ada)
```

MAE for AdaBoost: 0.63456

MSE for AdaBoost: 2.36423

RMSE for AdaBoost: 1.53760



### 1.5.13 BayesianRidge

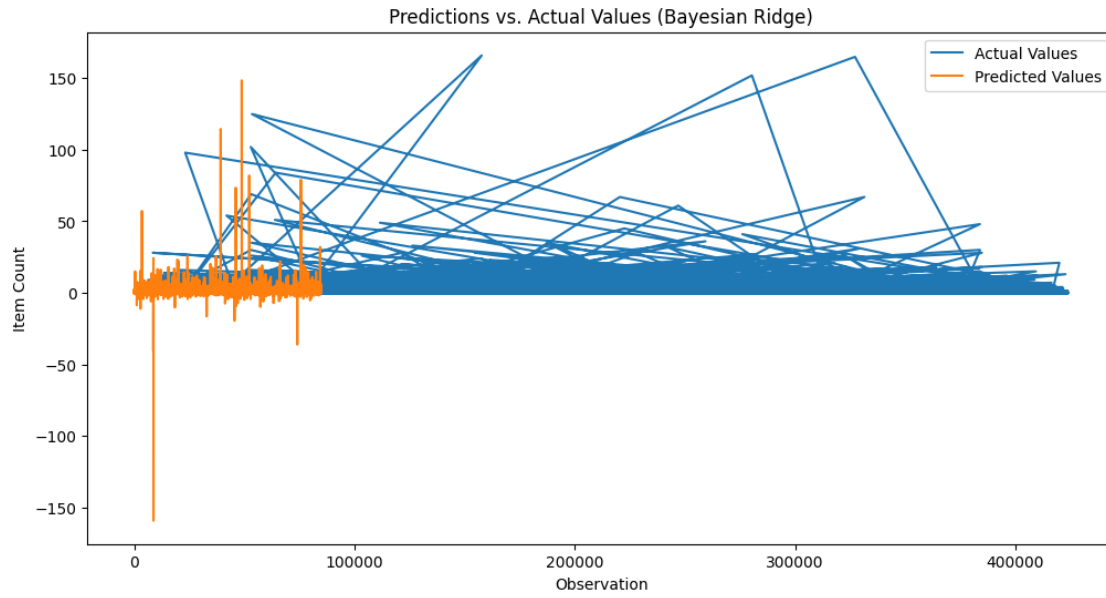
```
[ ]: # create bayesian ridge model
br = BayesianRidge()
br.fit(X_train, y_train)
y_pred = br.predict(X_test)

evaluate_the_model(y_test, y_pred, 'Bayesian Ridge', br)
```

MAE for Bayesian Ridge: 0.29018

MSE for Bayesian Ridge: 2.39567

RMSE for Bayesian Ridge: 1.54780

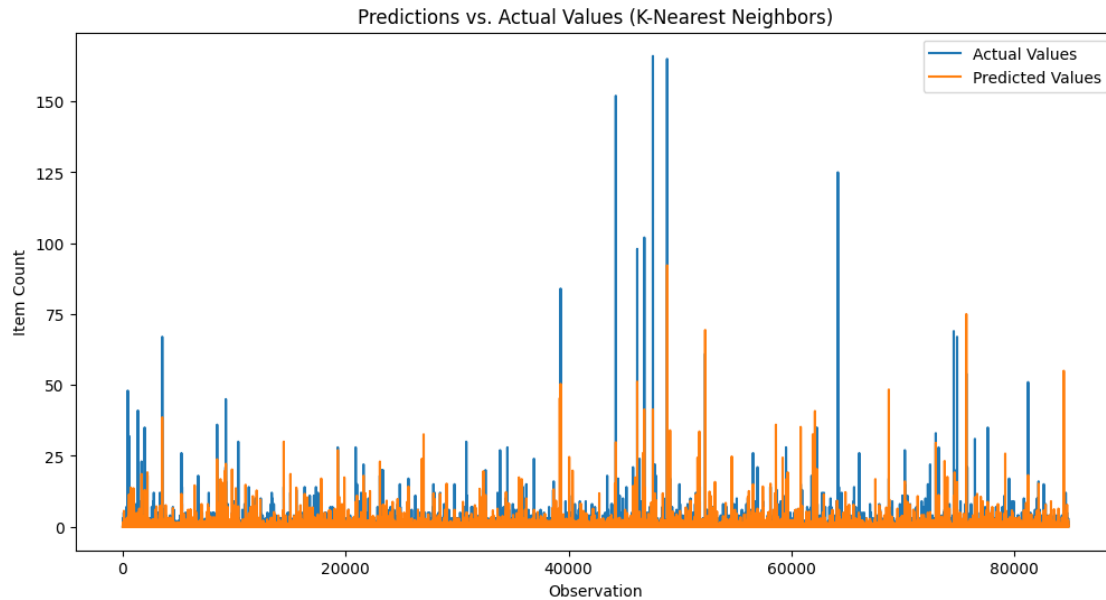


#### 1.5.14 KNN

```
[ ]: # create a knn model
knn = KNeighborsRegressor(n_neighbors=5)
knn.fit(X_train.values, y_train.values)
y_pred = knn.predict(X_test.values)

evaluate_the_model(y_test.values, y_pred, 'K-Nearest Neighbors', knn)
```

MAE for K-Nearest Neighbors: 0.18599  
MSE for K-Nearest Neighbors: 1.60036  
RMSE for K-Nearest Neighbors: 1.26505



### 1.5.15 Compare Models

```
[ ]: results = pd.DataFrame(scores_and_names, columns=['Model', 'RMSE'])
```

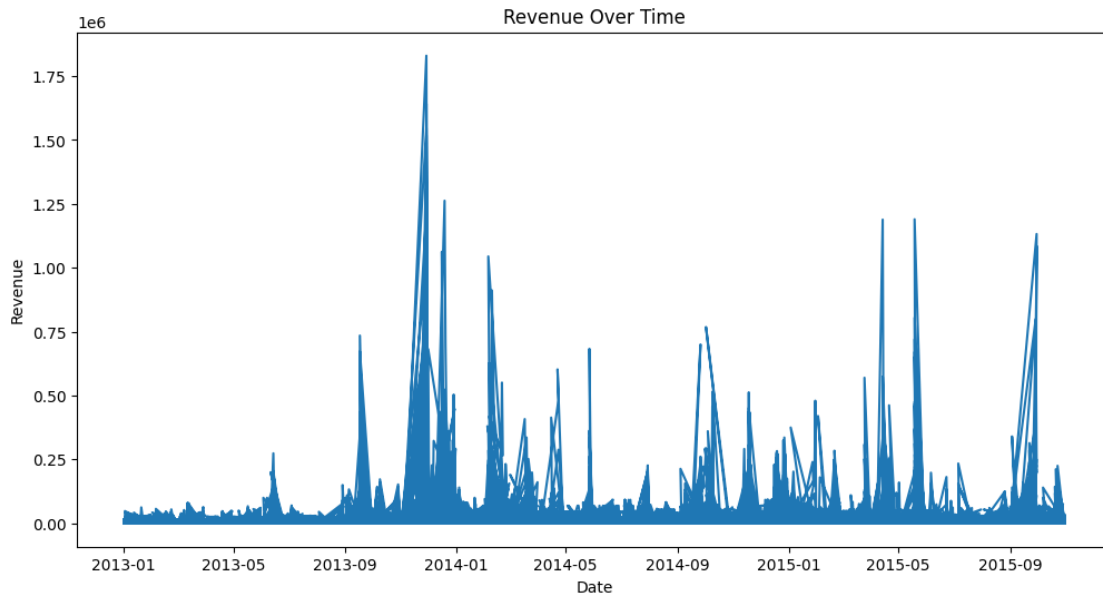
```
[ ]: #sort the results
results.sort_values(by='RMSE', ascending=True, inplace=True)
```

```
[ ]: #print the results in tabel format
print(results)
```

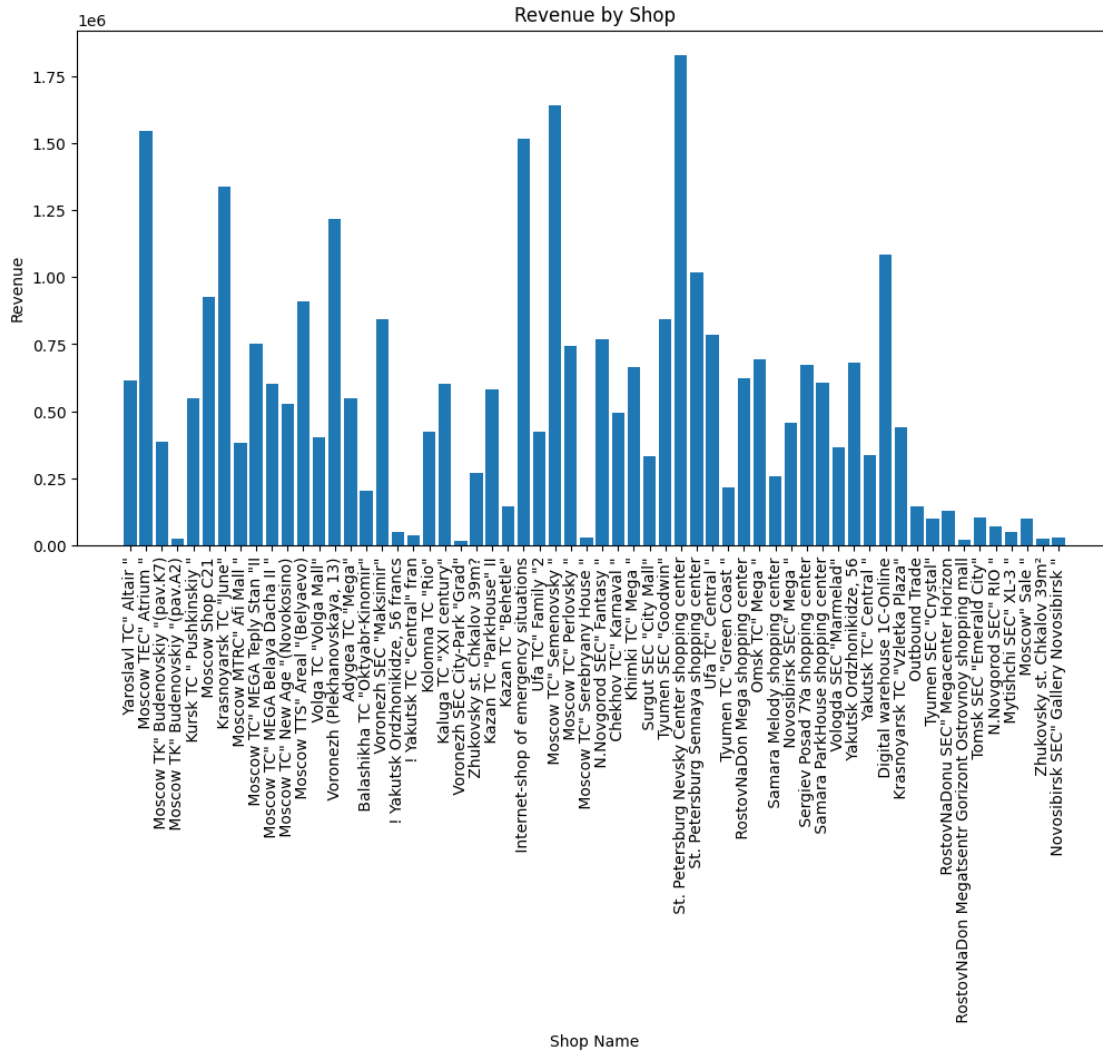
	Model	RMSE
4	Random Forest	1.065577e+00
6	Extra Trees	1.069535e+00
13	K-Nearest Neighbors	1.265051e+00
7	XGBoost	1.300267e+00
3	Decision Tree	1.371640e+00
9	Lasso Regression	1.454596e+00
11	AdaBoost	1.537605e+00
12	Bayesian Ridge	1.547795e+00
8	Ridge Regression	1.547984e+00
0	Linear Regression	1.547985e+00
10	ARIMA	1.608982e+00
1	Logistic Regression	1.616978e+00
2	Support Vector Machine	1.616978e+00
5	Stochastic Gradient Descent	8.354928e+14

## 1.6 Data Visualization

```
[ ]: #line chart
plt.figure(figsize=(12, 6))
plt.plot(final_dataset['date'], final_dataset['revenue'])
plt.title('Revenue Over Time')
plt.xlabel('Date')
plt.ylabel('Revenue')
plt.show()
```

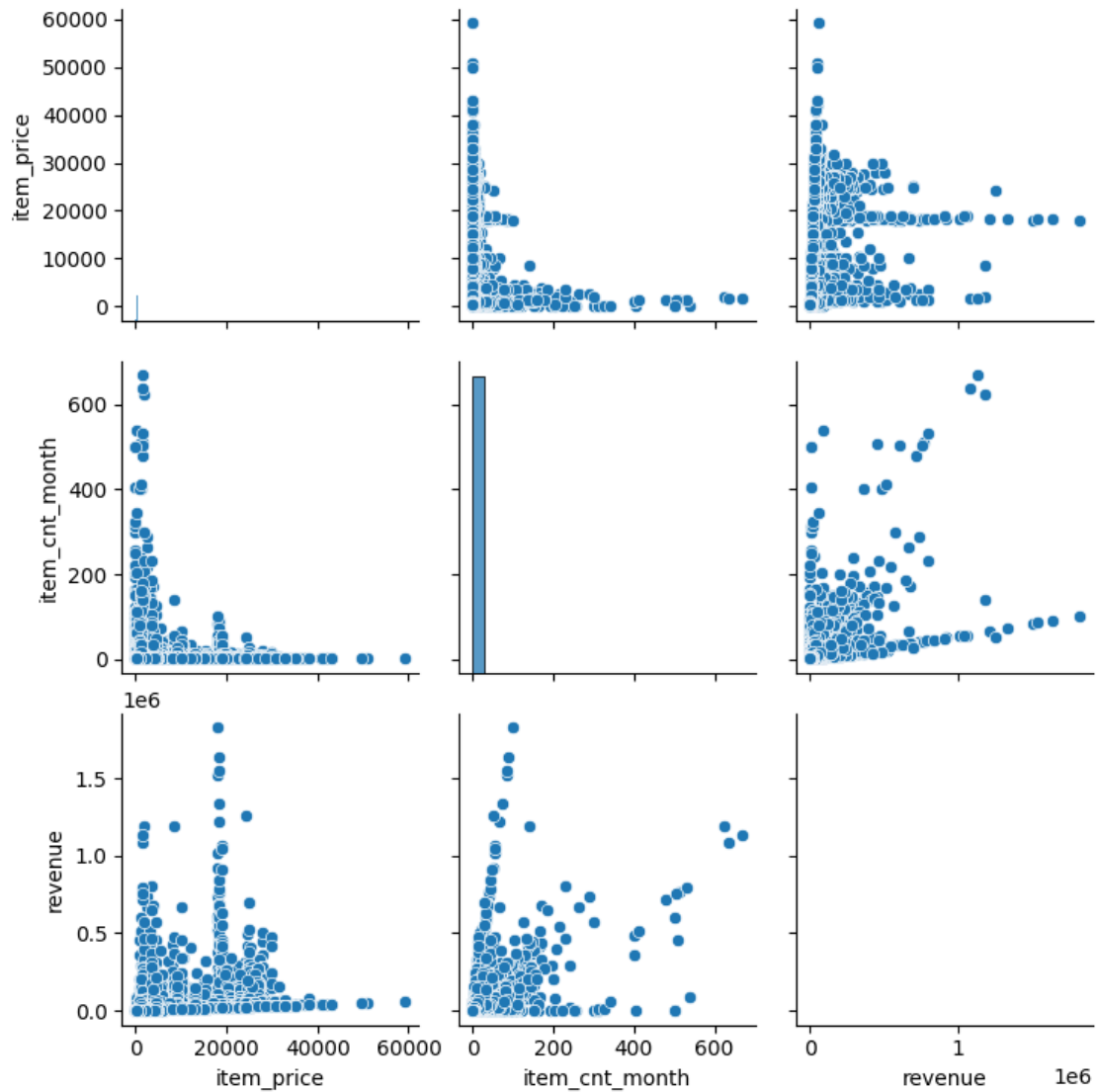


```
[ ]: #bar chart
plt.figure(figsize=(12, 6))
plt.bar(final_dataset['shop_name'], final_dataset['revenue'])
plt.title('Revenue by Shop')
plt.xlabel('Shop Name')
plt.ylabel('Revenue')
plt.xticks(rotation=90)
plt.show()
```

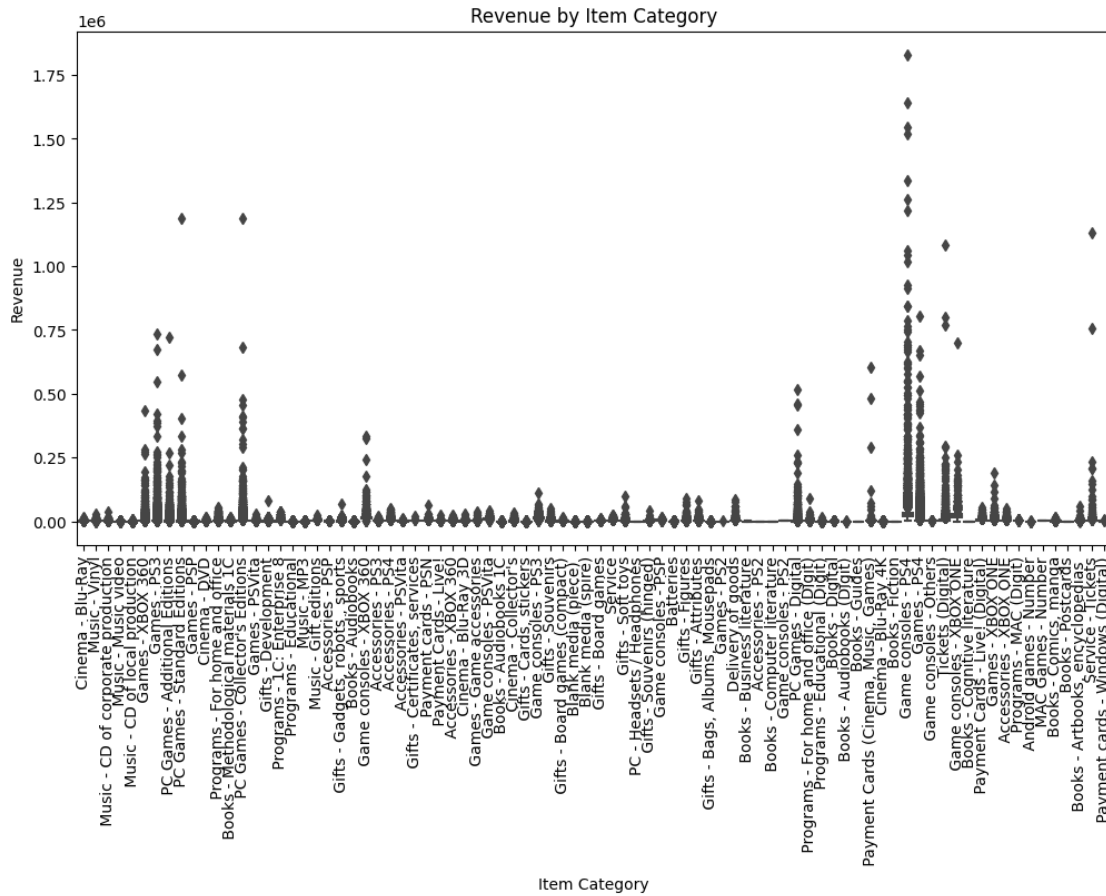


```
[ ]: #pairplot
sns.pairplot(final_dataset[['item_price', 'item_cnt_month', 'revenue']])
plt.show()
```

c:\Users\srumc\anaconda3\Lib\site-packages\seaborn\axisgrid.py:118: UserWarning:  
The figure layout has changed to tight  
self.\_figure.tight\_layout(\*args, \*\*kwargs)

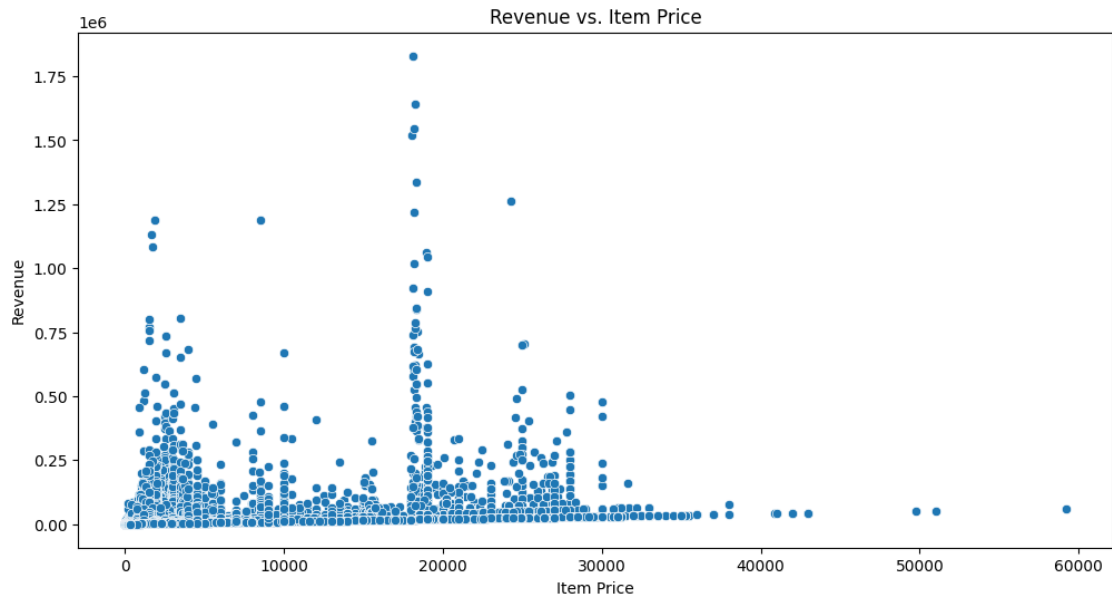


```
[ ]: #boxplot
plt.figure(figsize=(12, 6))
sns.boxplot(data=final_dataset, x='item_category_name', y='revenue')
plt.title('Revenue by Item Category')
plt.xlabel('Item Category')
plt.ylabel('Revenue')
plt.xticks(rotation=90)
plt.show()
```

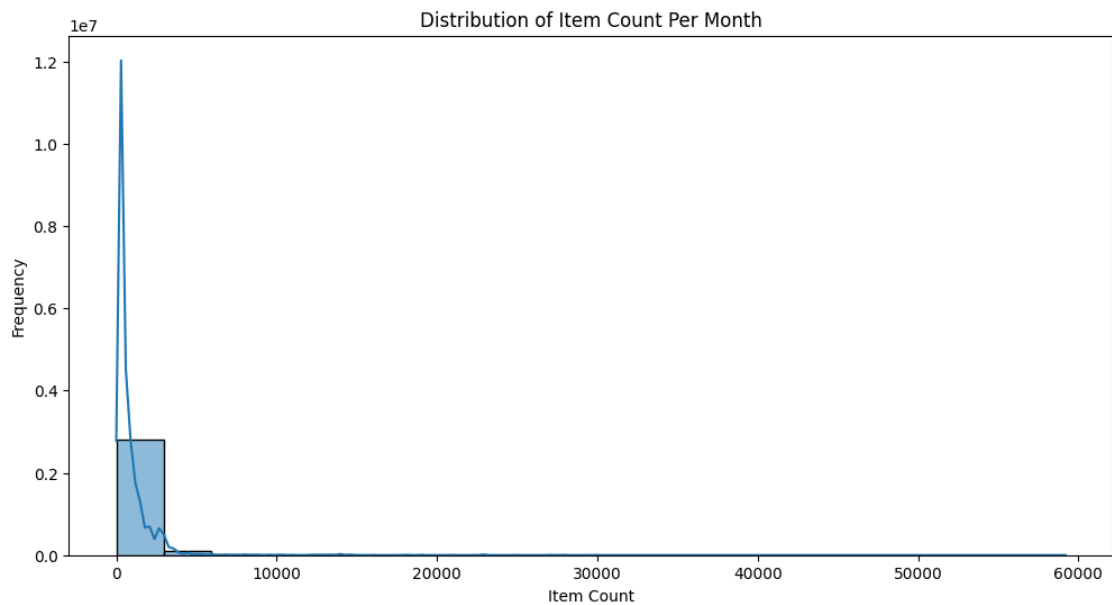


```
[ ]: #scatter chart
plt.figure(figsize=(12, 6))
sns.scatterplot(data=final_dataset, x='item_price', y='revenue')
plt.title('Revenue vs. Item Price')
plt.xlabel('Item Price')
plt.ylabel('Revenue')
plt.show()
```

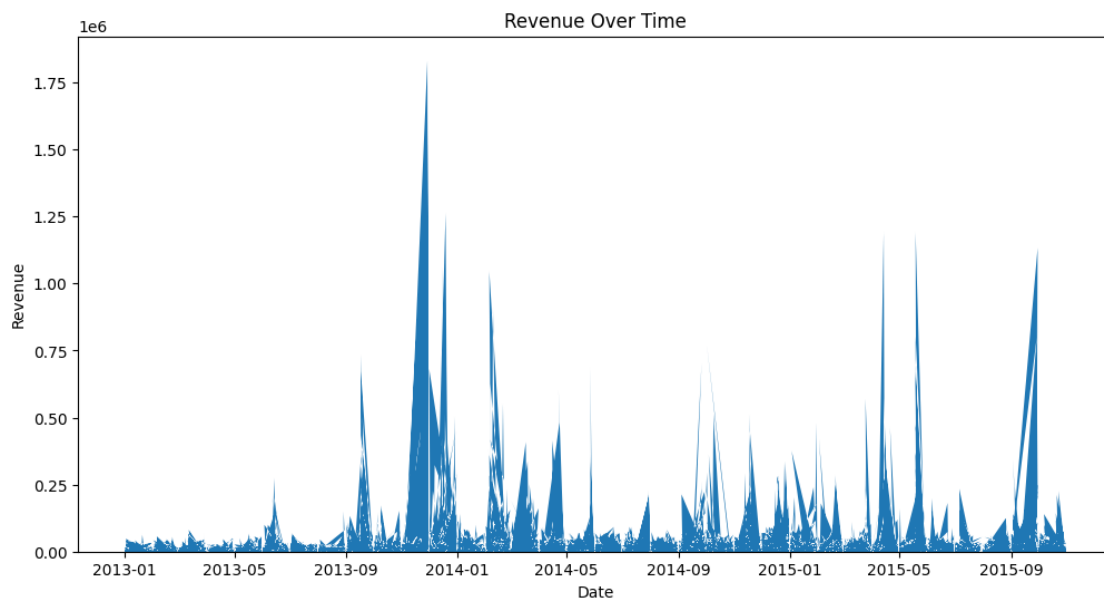




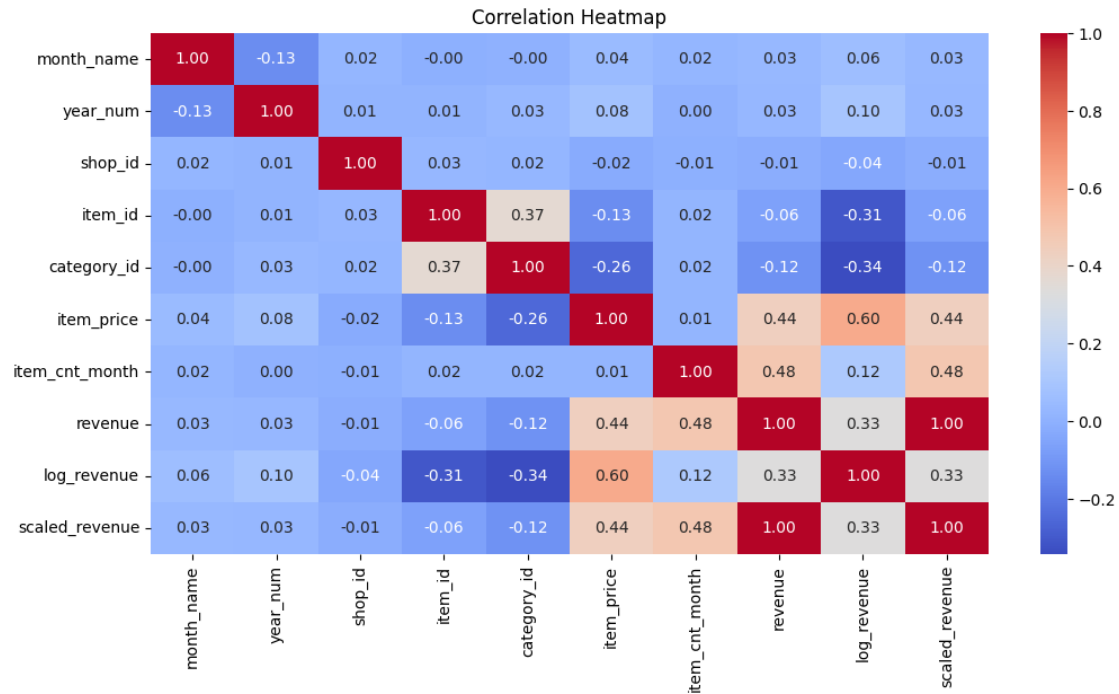
```
[ ]: #histogram
plt.figure(figsize=(12, 6))
sns.histplot(data=final_dataset, x='item_price', kde=True, bins=20)
plt.title('Distribution of Item Count Per Month')
plt.xlabel('Item Count')
plt.ylabel('Frequency')
plt.show()
```



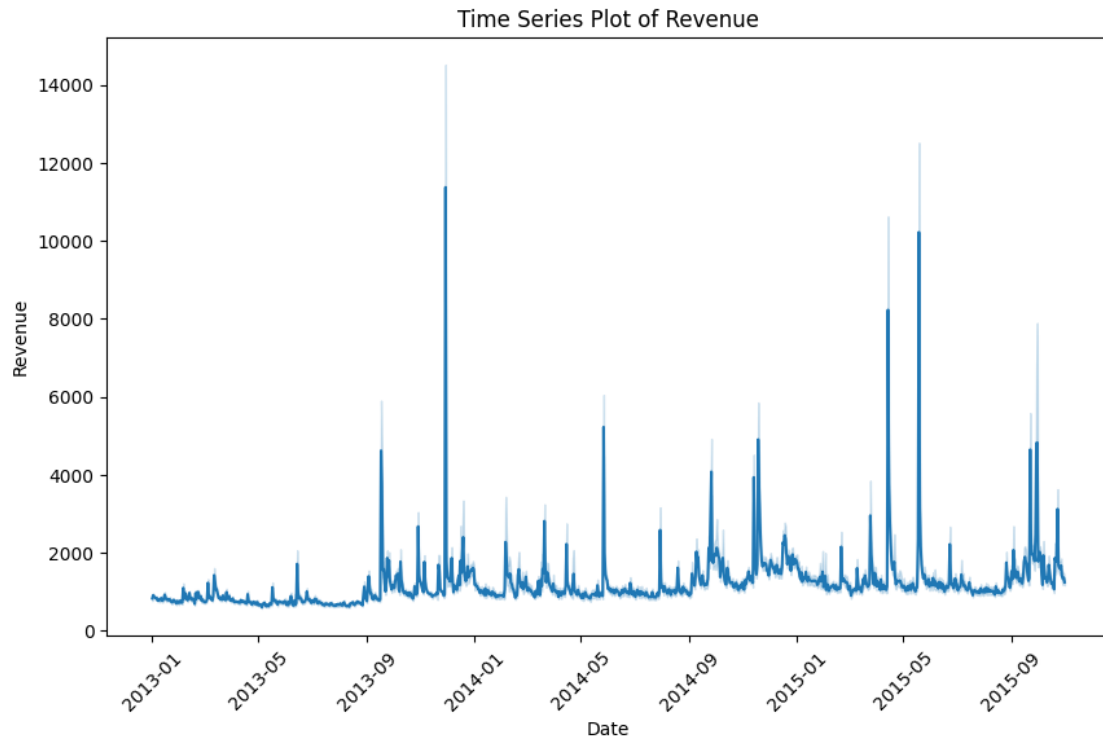
```
[ ]: #area plot
plt.figure(figsize=(12, 6))
plt.stackplot(final_dataset['date'], final_dataset['revenue'])
plt.title('Revenue Over Time')
plt.xlabel('Date')
plt.ylabel('Revenue')
plt.show()
```



```
[ ]: #heatmap
plt.figure(figsize=(12, 6))
sns.heatmap(numeric_columns.corr(), annot=True, cmap='coolwarm', fmt='%.2f')
plt.title('Correlation Heatmap')
plt.show()
```



```
[ ]: # Time Series Plot
plt.figure(figsize=(10, 6))
final_dataset['date'] = pd.to_datetime(final_dataset['date'])
sns.lineplot(x='date', y='revenue', data=final_dataset)
plt.title('Time Series Plot of Revenue')
plt.xlabel('Date')
plt.ylabel('Revenue')
plt.xticks(rotation=45)
plt.show()
```



## 1.7 Dashboard Creation

```
[ ]: print(final_dataset.head(10))
      print(final_dataset.shape)
      print(final_dataset.info())
```

	date	date_num	month_name	year_num	shop_id	\
0	2013-01-02	02	1	0	59	
1	2013-01-03	03	1	0	25	
3	2013-01-06	06	1	0	25	
4	2013-01-15	15	1	0	25	
5	2013-01-10	10	1	0	25	
6	2013-01-02	02	1	0	25	
7	2013-01-04	04	1	0	25	
8	2013-01-11	11	1	0	25	
9	2013-01-03	03	1	0	25	
10	2013-01-03	03	1	0	25	

	shop_name	item_id	\
0	Yaroslavl TC" Altair "	22154	
1	Moscow TEC" Atrium "	2552	
3	Moscow TEC" Atrium "	2554	
4	Moscow TEC" Atrium "	2555	

5	Moscow TEC" Atrium "	2564
6	Moscow TEC" Atrium "	2565
7	Moscow TEC" Atrium "	2572
8	Moscow TEC" Atrium "	2572
9	Moscow TEC" Atrium "	2573
10	Moscow TEC" Atrium "	2574

	item_name	category_id \
0	SCENE 2012 (BD)	37
1	DEEP PURPLE The House Of Blue Light LP	58
3	DEEP PURPLE Who Do You Think We Are LP	58
4	DEEP PURPLE 30 Very Best Of 2CD (Businesses).	56
5	DEEP PURPLE Perihelion: Live In Concert DVD (C...	59
6	DEEP PURPLE Stormbringer (firms).	56
7	DEFTONES Koi No Yokan	55
8	DEFTONES Koi No Yokan	55
9	DEL REY LANA Born To Die	55
10	DEL REY LANA Born To Die The Paradise Edition 2CD	55

	item_category_name	item_price	item_cnt_month	revenue \
0	Cinema - Blu-Ray	999.00	1	999.00
1	Music - Vinyl	899.00	1	899.00
3	Music - Vinyl	1709.05	1	1709.05
4	Music - CD of corporate production	1099.00	1	1099.00
5	Music - Music video	349.00	1	349.00
6	Music - CD of corporate production	549.00	1	549.00
7	Music - CD of local production	239.00	1	239.00
8	Music - CD of local production	299.00	1	299.00
9	Music - CD of local production	299.00	3	897.00
10	Music - CD of local production	399.00	2	798.00

	price_range	log_revenue	scaled_revenue
0	900-100000	6.906755	0.000546
1	800-900	6.801283	0.000491
3	900-100000	7.443693	0.000934
4	900-100000	7.002156	0.000601
5	300-400	5.855072	0.000191
6	500-600	6.308098	0.000300
7	200-300	5.476464	0.000131
8	200-300	5.700444	0.000163
9	200-300	6.799056	0.000490
10	300-400	6.682109	0.000436

(2928483, 16)

<class 'pandas.core.frame.DataFrame'>

Index: 2928483 entries, 0 to 2935848

Data columns (total 16 columns):

#	Column	Dtype
---	-----	-----

```

0    date                datetime64[ns]
1    date_num            object
2    month_name          int64
3    year_num            int64
4    shop_id             int64
5    shop_name           object
6    item_id             int64
7    item_name           object
8    category_id         int64
9    item_category_name  object
10   item_price          float64
11   item_cnt_month      int64
12   revenue            float64
13   price_range         category
14   log_revenue         float64
15   scaled_revenue      float64
dtypes: category(1), datetime64[ns](1), float64(4), int64(6), object(4)
memory usage: 360.3+ MB
None

```

```

[ ]: #decoding year_num column and month_name column
final_dataset['year_num'] = final_dataset['year_num'].replace({0: '2013', 1: '2014', 2: '2015'})
final_dataset['month_name'] = final_dataset['month_name'].replace({1: 'January', 2: 'February', 3: 'March', 4: 'April', 5: 'May', 6: 'June', 7: 'July', 8: 'August', 9: 'September', 10: 'October', 11: 'November', 12: 'December'})

print(final_dataset.head(10))
print(final_dataset.tail(10))

```

	date	date_num	month_name	year_num	shop_id	shop_name \
0	2013-01-02	02	January	2013	59	Yaroslavl TC" Altair "
1	2013-01-03	03	January	2013	25	Moscow TEC" Atrium "
3	2013-01-06	06	January	2013	25	Moscow TEC" Atrium "
4	2013-01-15	15	January	2013	25	Moscow TEC" Atrium "
5	2013-01-10	10	January	2013	25	Moscow TEC" Atrium "
6	2013-01-02	02	January	2013	25	Moscow TEC" Atrium "
7	2013-01-04	04	January	2013	25	Moscow TEC" Atrium "
8	2013-01-11	11	January	2013	25	Moscow TEC" Atrium "
9	2013-01-03	03	January	2013	25	Moscow TEC" Atrium "
10	2013-01-03	03	January	2013	25	Moscow TEC" Atrium "

	item_id	item_name	category_id \
0	22154	SCENE 2012 (BD)	37
1	2552	DEEP PURPLE The House Of Blue Light LP	58
3	2554	DEEP PURPLE Who Do You Think We Are LP	58
4	2555	DEEP PURPLE 30 Very Best Of 2CD (Businesses).	56

5	2564	DEEP PURPLE Perihelion: Live In Concert DVD (C...	59
6	2565	DEEP PURPLE Stormbringer (firms).	56
7	2572	DEFTONES Koi No Yokan	55
8	2572	DEFTONES Koi No Yokan	55
9	2573	DEL REY LANA Born To Die	55
10	2574	DEL REY LANA Born To Die The Paradise Edition 2CD	55

		item_category_name	item_price	item_cnt_month	revenue	\
0		Cinema - Blu-Ray	999.00	1	999.00	
1		Music - Vinyl	899.00	1	899.00	
3		Music - Vinyl	1709.05	1	1709.05	
4	Music - CD of corporate production		1099.00	1	1099.00	
5	Music - Music video		349.00	1	349.00	
6	Music - CD of corporate production		549.00	1	549.00	
7	Music - CD of local production		239.00	1	239.00	
8	Music - CD of local production		299.00	1	299.00	
9	Music - CD of local production		299.00	3	897.00	
10	Music - CD of local production		399.00	2	798.00	

	price_range	log_revenue	scaled_revenue
0	900-100000	6.906755	0.000546
1	800-900	6.801283	0.000491
3	900-100000	7.443693	0.000934
4	900-100000	7.002156	0.000601
5	300-400	5.855072	0.000191
6	500-600	6.308098	0.000300
7	200-300	5.476464	0.000131
8	200-300	5.700444	0.000163
9	200-300	6.799056	0.000490
10	300-400	6.682109	0.000436

	date	date_num	month_name	year_num	shop_id	\
2935839	2015-10-24	24	October	2015	25	
2935840	2015-10-31	31	October	2015	25	
2935841	2015-10-11	11	October	2015	25	
2935842	2015-10-10	10	October	2015	25	
2935843	2015-10-09	09	October	2015	25	
2935844	2015-10-10	10	October	2015	25	
2935845	2015-10-09	09	October	2015	25	
2935846	2015-10-14	14	October	2015	25	
2935847	2015-10-22	22	October	2015	25	
2935848	2015-10-03	03	October	2015	25	

	shop_name	item_id	\
2935839	Moscow TEC" Atrium "	7315	
2935840	Moscow TEC" Atrium "	7409	
2935841	Moscow TEC" Atrium "	7393	
2935842	Moscow TEC" Atrium "	7384	
2935843	Moscow TEC" Atrium "	7409	

2935844	Moscow TEC" Atrium "	7409
2935845	Moscow TEC" Atrium "	7460
2935846	Moscow TEC" Atrium "	7459
2935847	Moscow TEC" Atrium "	7440
2935848	Moscow TEC" Atrium "	7460

	item_name	category_id	\
2935839	V/A Dance Kick! 2CD (digipack)	55	
2935840	V/A Nu Jazz Selection (digipack)	55	
2935841	V/A Lounge Del Mar 3 2CD (digipack)	55	
2935842	V/A Ladies Sing The Blues 3CD	55	
2935843	V/A Nu Jazz Selection (digipack)	55	
2935844	V/A Nu Jazz Selection (digipack)	55	
2935845	V/A The Golden Jazz Collection 1 2CD	55	
2935846	V/A The Best Of The 3 Tenors	55	
2935847	V/A Relax Collection Planet MP3 (mp3-CD) (jewel)	57	
2935848	V/A The Golden Jazz Collection 1 2CD	55	

	item_category_name	item_price	item_cnt_month	revenue	\
2935839	Music - CD of local production	399.0	1	399.0	
2935840	Music - CD of local production	299.0	1	299.0	
2935841	Music - CD of local production	349.0	1	349.0	
2935842	Music - CD of local production	749.0	1	749.0	
2935843	Music - CD of local production	299.0	1	299.0	
2935844	Music - CD of local production	299.0	1	299.0	
2935845	Music - CD of local production	299.0	1	299.0	
2935846	Music - CD of local production	349.0	1	349.0	
2935847	Music - MP3	299.0	1	299.0	
2935848	Music - CD of local production	299.0	1	299.0	

	price_range	log_revenue	scaled_revenue
2935839	300-400	5.988961	0.000218
2935840	200-300	5.700444	0.000163
2935841	300-400	5.855072	0.000191
2935842	700-800	6.618739	0.000409
2935843	200-300	5.700444	0.000163
2935844	200-300	5.700444	0.000163
2935845	200-300	5.700444	0.000163
2935846	300-400	5.855072	0.000191
2935847	200-300	5.700444	0.000163
2935848	200-300	5.700444	0.000163

```
[ ]: #make data frame for the dashboard interactive
df_interactive = final_dataset.interactive()
```

```
[ ]: #creates menu button for the dashboard to select years
```



```

year_num_menu = pn.widgets.Select(name='Year',
    ↪options=df_interactive['year_num'].unique().tolist(), value='2015')

updated_year_df = df_interactive[df_interactive['year_num'] == year_num_menu]

```

```

[ ]: #monthly revenue
update_monthly_revenue = (updated_year_df.groupby('month_name')['revenue'].
    ↪mean().to_frame().reset_index().sort_values(by='month_name').
    ↪reset_index(drop=True))
update_monthly_revenue_plot = update_monthly_revenue.hvplot.bar(x='month_name',
    ↪y='revenue', rot=90, title='Average Revenue Per Month by year')

```

```

[ ]: #monthly item count
update_monthly_item_count = (updated_year_df.
    ↪groupby('month_name')['item_cnt_month'].mean().to_frame().reset_index().
    ↪sort_values(by='month_name').reset_index(drop=True))
update_monthly_item_count_pie = update_monthly_item_count.plot(kind='pie',
    ↪y='item_cnt_month', label='month_name', autopct='%1.1f%%', title='Average
    ↪Item Count Per Month by year')

```

```

[ ]: #sales by category
update_sales_by_category = (updated_year_df.
    ↪groupby('item_category_name')['revenue'].mean().to_frame().reset_index().
    ↪sort_values(by='revenue').reset_index(drop=True))
update_sales_by_category_plot = update_sales_by_category.hvplot.
    ↪area(x='item_category_name', y='revenue', rot=90, title='Average Revenue Per
    ↪Category by year', height=500, width=1000)

```

```

[ ]: #sales by shop
update_sales_by_shop = (updated_year_df.groupby('shop_name')['revenue'].mean().
    ↪to_frame().reset_index().sort_values(by='revenue').reset_index(drop=True))
update_sales_by_shop_plot = update_sales_by_shop.hvplot.area(x='shop_name',
    ↪y='revenue', rot=90, title='Average Revenue Per Shop by year', height=500,
    ↪width=1000)

```

```

[ ]: #best selling category in all year
update_best_selling_category = (df_interactive.
    ↪groupby('item_category_name')['item_cnt_month'].mean().to_frame().
    ↪reset_index().sort_values(by='item_cnt_month', ascending=False).
    ↪reset_index(drop=True))
update_best_selling_category_plot = update_best_selling_category.hvplot.
    ↪barh(x='item_category_name', y='item_cnt_month', rot=90, title='Average Item
    ↪Count Per Category', height=1000, width=1000)

```

```

[ ]: #creates echarts bar for the dashboard to display revenue by year

```

```

revenue_by_year_bar = { 'title' : { 'text' : 'Revenue by Year' }, 'tooltip' : {
    ↪ 'trigger': 'axis' }, 'legend': { 'data': ['Revenue'] }, 'xAxis' : { 'data' :
    ↪ final_dataset['year_num'].unique().tolist() }, 'yAxis' : { }, 'series' : [{
    ↪ 'name' : 'Revenue', 'type' : 'bar', 'data' : final_dataset.
    ↪ groupby('year_num')['revenue'].sum().tolist() }] }
revenue_by_year_echart_pane = pn.pane.ECharts(revenue_by_year_bar, width=400,
    ↪ height=400)

```

```

[ ]: template = pn.template.FastListTemplate(
    title='Grocery Sales Dashboard',
    sidebar=[pn.pane.Markdown("# Revenue by Year"),
        revenue_by_year_echart_pane,
        pn.pane.Markdown("#### Select Year"),
        year_num_menu],
    main=[pn.Row(pn.Column(update_monthly_revenue_plot.panel(width=500)),
        pn.Column(update_monthly_item_count_pie.panel(width=500),
    ↪ margin=(0,25)),
        ),
        pn.Row(update_sales_by_category_plot),
        pn.Row(update_sales_by_shop_plot),
        pn.Row(update_best_selling_category_plot)],

    accent_base_color="#88d8b0",
    header_background="#88d8b0",
)

template.show()

```

Launching server at <http://localhost:53043>

[ ]: <panel.io.server.Server at 0x1270fb93150>

WARNING:bokeh.core.validation.check:W-1005 (FIXED\_SIZING\_MODE): 'fixed' sizing mode requires width and height to be set:  
 Column(id='75fa204f-25bf-4df4-950f-c52628c0f79e', ...)