```
import pandas as pd
import numpy as np
from sklearn.preprocessing import LabelEncoder
import matplotlib.pyplot as plt
import seaborn as sns
from scipy.stats import zscore
from scipy.stats import ttest ind
from scipy.stats import pearsonr, spearmanr
from scipy.stats import kruskal
from scipy.stats import f oneway
from scipy.stats import pearsonr, spearmanr
train df= pd.read csv('TRAIN.csv')
print(train df.shape)
print(train df.info())
print(train df.describe())
(188340, 10)
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 188340 entries, 0 to 188339
Data columns (total 10 columns):
                    Non-Null Count
#
     Column
                                     Dtype
     -----
 0
     ID
                    188340 non-null object
 1
     Store id
                    188340 non-null int64
 2
     Store Type
                    188340 non-null object
 3
     Location_Type 188340 non-null object
 4
     Region Code
                    188340 non-null object
 5
                    188340 non-null object
     Date
 6
     Holiday
                    188340 non-null
                                     int64
7
                    188340 non-null object
     Discount
 8
                    188340 non-null
     0rder
                                     int64
 9
     Sales
                    188340 non-null
                                     float64
dtypes: float64(1), int64(3), object(6)
memory usage: 14.4+ MB
None
            Store id
                            Holiday
                                             0rder
                                                             Sales
       188340.000000
                      188340.000000
                                     188340.000000
                                                    188340.000000
count
          183.000000
                                         68.205692
                                                     42784.327982
                           0.131783
mean
std
          105.366308
                           0.338256
                                         30.467415
                                                      18456.708302
                           0.000000
                                          0.000000
                                                          0.000000
min
            1.000000
25%
           92.000000
                           0.000000
                                         48.000000
                                                      30426.000000
50%
          183,000000
                           0.000000
                                         63.000000
                                                      39678,000000
          274,000000
                                                      51909.000000
75%
                           0.000000
                                         82,000000
          365.000000
                           1.000000
                                        371.000000 247215.000000
max
train df['Date'] = pd.to datetime(train df['Date'], format="%d-%m-%Y")
train df['Year'] = train df['Date'].dt.year
train df['Month'] = train df['Date'].dt.month
train df['Week'] = train df['Date'].dt.isocalendar().week
```

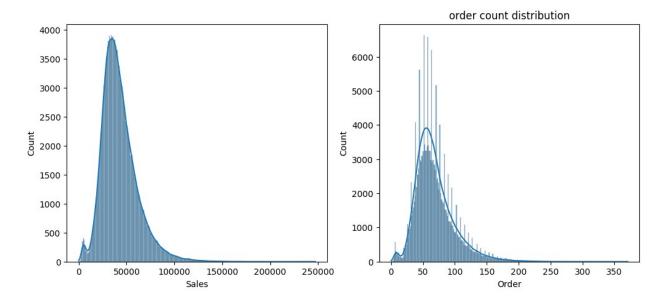
```
train df['DayOfWeek'] = train df['Date'].dt.dayofweek # Monday=0,
Sunday=6
train_df['Is_Weekend'] = train_df['DayOfWeek'].isin([5,
6]).astype(int)
train df['YearMonth'] = train df['Date'].dt.to period('M')
categorical_columns=['Store_Type','Location_Type','Region_Code','Disco
unt','Holiday']
for columns in categorical_columns:
    train_df[columns]=train_df[columns].astype('category')
duplicates = train df.duplicated(subset=['ID']).sum()
print(f"Duplicate IDs: {duplicates}")
print(train df)
Duplicate IDs: 0
                  Store id Store Type Location Type Region Code
Date \
                                                                R1 2018-
        T1000001
                          1
                                    S1
                                                   L3
01-01
        T1000002
                        253
                                    S4
                                                   L2
                                                                R1 2018-
01-01
        T1000003
                        252
                                    S3
                                                   L2
                                                                R1 2018-
01-01
                                    S2
                                                   L3
        T1000004
                        251
                                                                R1 2018-
01 - 01
        T1000005
                        250
                                    S2
                                                   L3
                                                                R4 2018-
01 - 01
. . .
       T1188336
                        149
                                    S2
                                                   L3
                                                                R2 2019-
188335
05 - 31
                                    S4
                                                   L2
                                                                R1 2019-
188336
       T1188337
                        153
05 - 31
188337
        T1188338
                        154
                                    S1
                                                   L3
                                                                R2 2019-
05 - 31
188338
       T1188339
                        155
                                    S3
                                                   L1
                                                                R2 2019-
05 - 31
                                                                R1 2019-
188339
       T1188340
                        152
                                    S2
                                                   L1
05 - 31
       Holiday Discount Order
                                    Sales
                                            Year
                                                  Month Week DayOfWeek
0
                     Yes
                              9
                                  7011.84
                                            2018
                                                      1
                                                                        0
                                                             1
1
                     Yes
                             60
                                 51789.12
                                            2018
                                                      1
                                                             1
                                                                        0
2
                     Yes
                             42
                                 36868.20
                                            2018
                                                      1
                                                             1
```

```
3
                     Yes
                              23
                                  19715.16
                                            2018
                                                       1
                                                              1
                                                                         0
              1
                                                       1
                                                             1
                                                                         0
                     Yes
                             62
                                  45614.52
                                            2018
188335
                     Yes
                                  37272.00
                                            2019
                                                       5
                                                            22
                                                                         4
              1
                              51
188336
              1
                      No
                              90
                                  54572.64
                                                       5
                                                            22
                                                                         4
                                            2019
188337
              1
                                                       5
                                                            22
                                                                         4
                      No
                              56
                                  31624.56
                                            2019
188338
              1
                     Yes
                              70
                                  49162.41
                                            2019
                                                       5
                                                            22
                                                                         4
                                                       5
                                                                         4
188339
              1
                      No
                              47
                                  37977.00 2019
                                                            22
        Is Weekend YearMonth
0
                      2018-01
                      2018-01
1
                  0
2
                  0
                      2018-01
3
                  0
                      2018-01
4
                      2018-01
                  0
                      2019-05
188335
                  0
188336
                  0
                      2019-05
                      2019-05
188337
                  0
188338
                  0
                      2019-05
188339
                  0
                      2019-05
[188340 rows x 16 columns]
fig, ax=plt.subplots(1, 2, figsize=(12, 5))
sns.histplot(train_df['Sales'],kde=True,ax=ax[0])
sns.histplot(train df['Order'],kde=True,ax=ax[1])
ax[1].set_title('order count distribution')
plt.show()
categorical_col=['Store_Type','Location_Type','Region_Code','Holiday',
'Discount'l
for col in categorical col:
sns.countplot(data=train_df,x=col,order=train_df[col].value_counts().i
ndex)
    plt.title(f'{col} Count')
    plt.xticks(rotation=45)
    plt.show()
    print(f"\n frequency table for {col}:")
    print(train df[col].value counts())
```

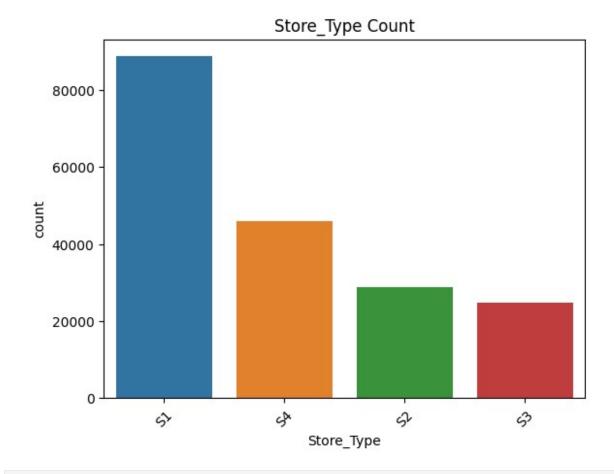
C:\Users\HP\anaconda3\Lib\site-packages\seaborn_oldcore.py:1119: FutureWarning: use_inf_as_na option is deprecated and will be removed in a future version. Convert inf values to NaN before operating instead.

with pd.option_context('mode.use_inf_as_na', True): C:\Users\HP\anaconda3\Lib\site-packages\seaborn_oldcore.py:1119: FutureWarning: use_inf_as_na option is deprecated and will be removed in a future version. Convert inf values to NaN before operating instead.

with pd.option context('mode.use inf as na', True):



C:\Users\HP\anaconda3\Lib\site-packages\seaborn\categorical.py:641: FutureWarning: The default of observed=False is deprecated and will be changed to True in a future version of pandas. Pass observed=False to retain current behavior or observed=True to adopt the future default and silence this warning.

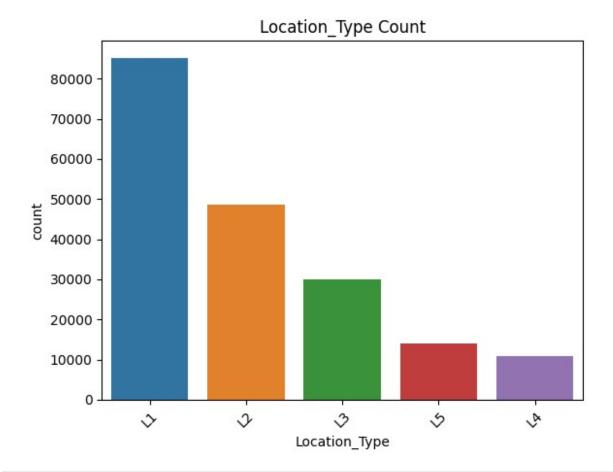


```
frequency table for Store_Type:
Store_Type
```

\$1\$8752\$445924\$228896\$324768

Name: count, dtype: int64

C:\Users\HP\anaconda3\Lib\site-packages\seaborn\categorical.py:641: FutureWarning: The default of observed=False is deprecated and will be changed to True in a future version of pandas. Pass observed=False to retain current behavior or observed=True to adopt the future default and silence this warning.



```
frequency table for Location_Type:
```

Location_Type

L1 85140 L2 48504

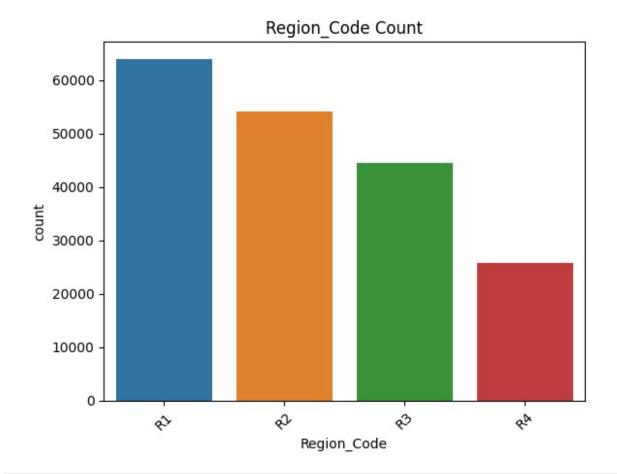
L3 29928

L5 13932

L4 10836

Name: count, dtype: int64

C:\Users\HP\anaconda3\Lib\site-packages\seaborn\categorical.py:641: FutureWarning: The default of observed=False is deprecated and will be changed to True in a future version of pandas. Pass observed=False to retain current behavior or observed=True to adopt the future default and silence this warning.



frequency table for Region Code:

Region_Code

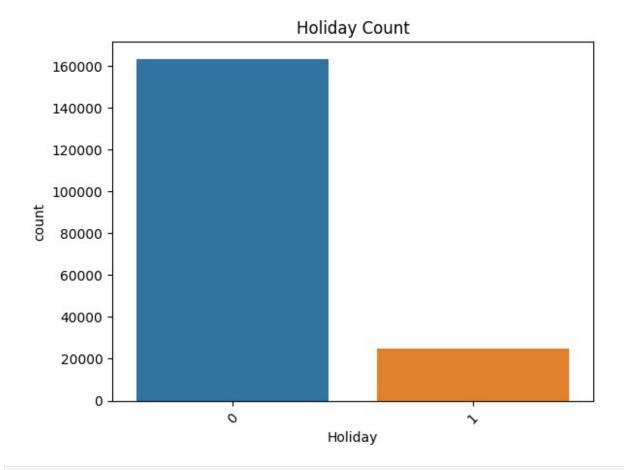
R1 63984 R2 54180 R3 44376

25800

R4

Name: count, dtype: int64

C:\Users\HP\anaconda3\Lib\site-packages\seaborn\categorical.py:641: FutureWarning: The default of observed=False is deprecated and will be changed to True in a future version of pandas. Pass observed=False to retain current behavior or observed=True to adopt the future default and silence this warning.



frequency table for Holiday:

Holiday

0 163520 1 24820

Name: count, dtype: int64

C:\Users\HP\anaconda3\Lib\site-packages\seaborn\categorical.py:641: FutureWarning: The default of observed=False is deprecated and will be changed to True in a future version of pandas. Pass observed=False to retain current behavior or observed=True to adopt the future default and silence this warning.

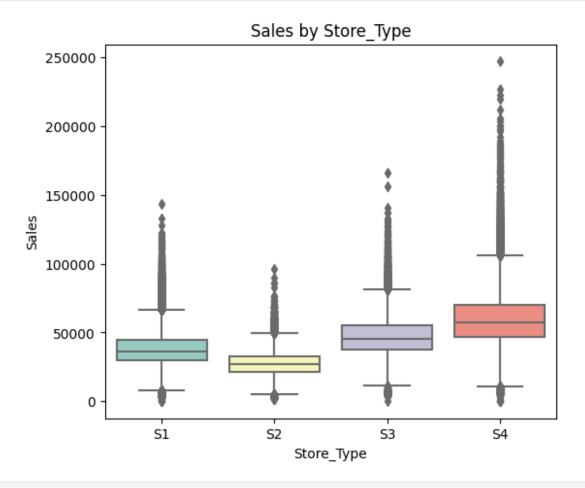


```
frequency table for Discount:
Discount
       104051
No
        84289
Yes
Name: count, dtype: int64
plt.figure(figsize=(6,5))
sns.scatterplot(data=train df,x='Order',y='Sales',hue='Discount',palet
te='coolwarm')
plt.title('order vs sales(discount highlighted)')
plt.show()
plt.figure(figsize=(6,5))
sns.scatterplot(data=train df,x='Order',y='Sales',hue='Holiday',palett
e='Set2')
plt.title('order vs sales (holiday highlighted)')
plt.show()
plt.figure(figsize=(6,5))
sns.boxplot(data=train df,x='Store Type',y='Sales',palette='Set3')
plt.title('Sales by Store_Type')
plt.show()
```

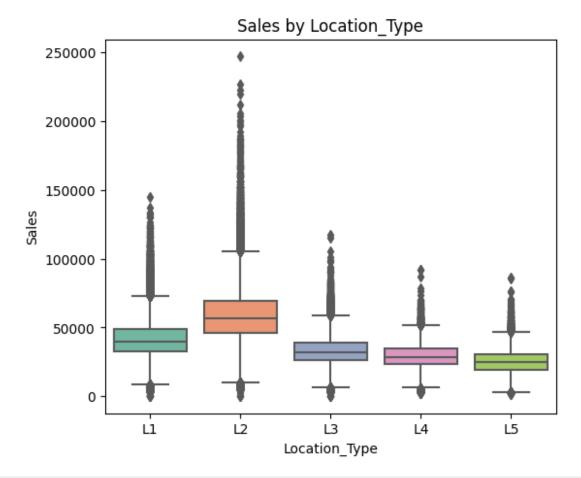
```
plt.figure(figsize=(6,5))
sns.boxplot(data=train_df,x='Location_Type',y='Sales',palette='Set2')
plt.title('Sales by Location_Type')
plt.show()
```

C:\Users\HP\anaconda3\Lib\site-packages\seaborn\categorical.py:641: FutureWarning: The default of observed=False is deprecated and will be changed to True in a future version of pandas. Pass observed=False to retain current behavior or observed=True to adopt the future default and silence this warning.

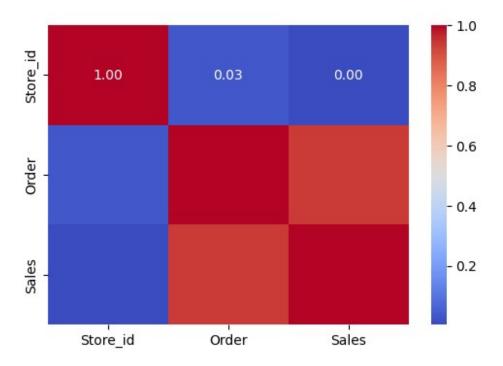
grouped vals = vals.groupby(grouper)



C:\Users\HP\anaconda3\Lib\site-packages\seaborn\categorical.py:641: FutureWarning: The default of observed=False is deprecated and will be changed to True in a future version of pandas. Pass observed=False to retain current behavior or observed=True to adopt the future default and silence this warning.

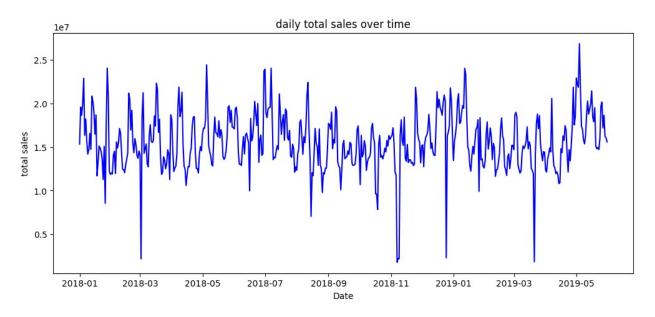


```
numerical_cols=train_df.select_dtypes(include=['int64','float64'])
plt.figure(figsize=(6,4))
sns.heatmap(numerical_cols.corr(),
annot=True,cmap='coolwarm',fmt=".2f")
plt.show()
```



```
train df['Date']=pd.to datetime(train df['Date'], format='%d-%m-%y')
daily sales=train df.groupby('Date')['Sales'].sum().reset index()
plt.figure(figsize=(12,5))
sns.lineplot(data=daily sales,x='Date',y='Sales',color='blue')
plt.title('daily total sales over time')
plt.xlabel('Date')
plt.ylabel('total sales')
plt.show()
train df['YearMonth']=train df['Date'].dt.to period('M').astype(str)
monthly sales=train df.groupby('YearMonth')
['Sales'].sum().reset index()
plt.figure(figsize=(12,5))
sns.lineplot(data=monthly sales,x='YearMonth',y='Sales',marker='o',col
or='green')
plt.xticks(rotation=45)
plt.title('monthly sales trends')
plt.xlabel('Month')
plt.ylabel('total sales')
plt.show()
train df['dayofweek']=train df['Date'].dt.day name()
dow_sales=train_df.groupby('dayofweek')['Sales'].mean().reset_index()
dow sales['dayofweek']=pd.Categorical(dow sales['dayofweek'],categorie
s=['Monday','Tuesday','Wednesday','Thursday','Friday','Saturday','Sund
ay'],ordered=True)
plt.figure(figsize=(8,5))
sns.barplot(data=dow sales,
```

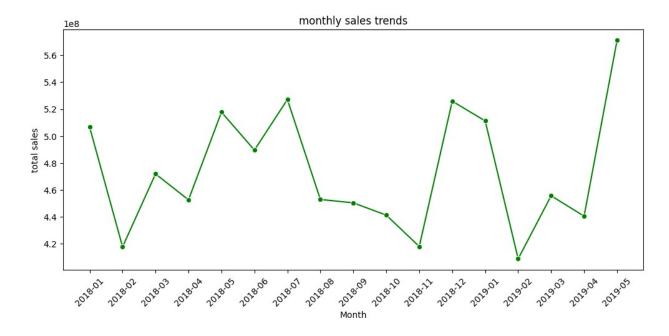
```
x='dayofweek',y='Sales',palette='coolwarm')
plt.title('Average Sales by day of week')
plt.ylabel('Average Sales')
plt.show()
holiday sales=train df.groupby(['Date', 'Holiday'])
['Sales'].sum().reset index()
plt.figure(figsize=(12,5))
sns.lineplot(data=holiday sales,x='Date',y='Sales',hue='Holiday',palet
te='Set1')
plt.title('Holiday vs Non Hoiday sales trend')
plt.ylabel('total sales')
plt.xlabel('Date')
plt.show()
C:\Users\HP\anaconda3\Lib\site-packages\seaborn\ oldcore.py:1119:
FutureWarning: use inf as na option is deprecated and will be removed
in a future version. Convert inf values to NaN before operating
instead.
 with pd.option context('mode.use inf as na', True):
C:\Users\HP\anaconda3\Lib\site-packages\seaborn\ oldcore.py:1119:
FutureWarning: use inf as na option is deprecated and will be removed
in a future version. Convert inf values to NaN before operating
instead.
 with pd.option context('mode.use inf as na', True):
```



C:\Users\HP\anaconda3\Lib\site-packages\seaborn_oldcore.py:1119: FutureWarning: use_inf_as_na option is deprecated and will be removed in a future version. Convert inf values to NaN before operating instead.

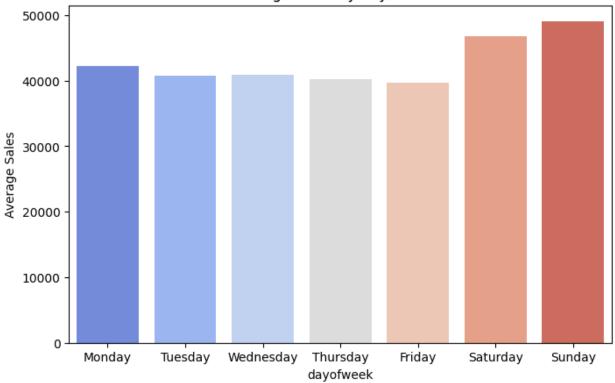
with pd.option_context('mode.use_inf_as_na', True):
C:\Users\HP\anaconda3\Lib\site-packages\seaborn_oldcore.py:1119:
FutureWarning: use_inf_as_na option is deprecated and will be removed in a future version. Convert inf values to NaN before operating instead.

with pd.option_context('mode.use_inf_as_na', True):



C:\Users\HP\anaconda3\Lib\site-packages\seaborn\categorical.py:641: FutureWarning: The default of observed=False is deprecated and will be changed to True in a future version of pandas. Pass observed=False to retain current behavior or observed=True to adopt the future default and silence this warning.

Average Sales by day of week



C:\Users\HP\AppData\Local\Temp\ipykernel_14368\3391780094.py:30: FutureWarning: The default of observed=False is deprecated and will be changed to True in a future version of pandas. Pass observed=False to retain current behavior or observed=True to adopt the future default and silence this warning.

holiday_sales=train_df.groupby(['Date','Holiday'])

['Sales'].sum().reset_index()

C:\Users\HP\anaconda3\Lib\site-packages\seaborn_oldcore.py:1119: FutureWarning: use_inf_as_na option is deprecated and will be removed in a future version. Convert inf values to NaN before operating instead.

with pd.option context('mode.use inf as na', True):

C:\Users\HP\anaconda3\Lib\site-packages\seaborn_oldcore.py:1119: FutureWarning: use_inf_as_na option is deprecated and will be removed in a future version. Convert inf values to NaN before operating instead.

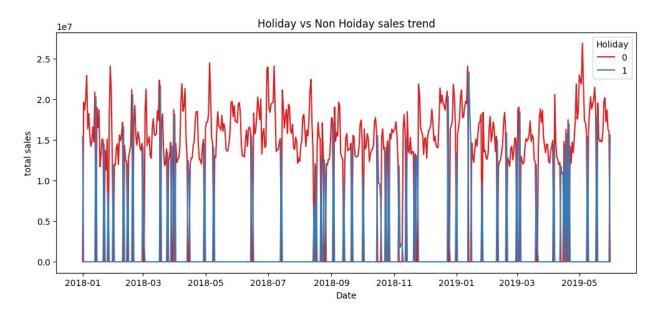
with pd.option context('mode.use inf as na', True):

C:\Users\HP\anaconda3\Lib\site-packages\seaborn_oldcore.py:1057: FutureWarning: The default of observed=False is deprecated and will be changed to True in a future version of pandas. Pass observed=False to retain current behavior or observed=True to adopt the future default and silence this warning.

grouped data = data.groupby(

C:\Users\HP\anaconda3\Lib\site-packages\seaborn_oldcore.py:1075:
FutureWarning: When grouping with a length-1 list-like, you will need

```
to pass a length-1 tuple to get_group in a future version of pandas.
Pass `(name,)` instead of `name` to silence this warning.
  data_subset = grouped_data.get_group(pd_key)
C:\Users\HP\anaconda3\Lib\site-packages\seaborn\_oldcore.py:1075:
FutureWarning: When grouping with a length-1 list-like, you will need to pass a length-1 tuple to get_group in a future version of pandas.
Pass `(name,)` instead of `name` to silence this warning.
  data_subset = grouped_data.get_group(pd_key)
```



```
g1=train df['Sales'].guantile(0.25)
q3=train df['Sales'].quantile(0.75)
iqr = q3-q1
lower bound=q1-1.5*iqr
upper bound=q3+1.5*igr
outliers igr=train df[(train df['Sales']<lower bound)|</pre>
(train df['Sales']>upper bound)]
print(f"Number of outliers(IQR method):{len(outliers iqr)}")
train_df['Sales_z']=zscore(train df['Sales'])
outliers z=train df[abs(train df['Sales z'])>3]
print(f"Number of outliers(z-score method):{len(outliers_z)}")
train df['Sales'] = np.where(train df['Sales'] > upper bound,
upper bound, train df['Sales'])
train df
Number of outliers(IOR method):5843
Number of outliers(z-score method):2421
```

	ID	Store id	Stor	e_Type Loc	cation	Type Re	gion C	ode	
Date \ 0	T1000001	_ 1		S1	_	L3	_		2018-
01-01									
1 01-01	T1000002	253		S4		L2		R1 2	2018-
2 01-01	T1000003	252		S 3		L2		R1 2	2018-
3 01-01	T1000004	251		S2		L3		R1 2	2018-
4	T1000005	250		S2		L3		R4 2	2018-
01-01									
188335	T1188336	149		S2		L3		R2 2	2019-
05-31 188336	T1188337	153		S4		L2		R1 2	2019-
05-31 188337	T1188338	154		S1		L3		R2 2	2019-
05-31 188338	T1188339	155		S 3		L1		R2 2	2019-
05-31 188339	T1188340	152		S 2		L1		R1 2	2019-
05-31									
	Holiday Di	scount Or	der	Sales	Year	Month	Week	Day	OfWeek
0	1	Yes	9	7011.84	2018	1	1		0
1	1	Yes	60	51789.12	2018	1	1		0
2	1	Yes	42	36868.20	2018	1	1		0
3	1	Yes	23	19715.16	2018	1	1		Θ
4	1	Yes	62	45614.52	2018	1	1		0
188335	1	Yes	51	37272.00	2019	5	22		4
188336	1	No	90	54572.64	2019	5	22		4
188337	1	No	56	31624.56	2019	5	22		4
100557									
188338	1	Yes	70	49162.41	2019	5	22		4
	1 1	Yes No	70 47	49162.41 37977.00	2019 2019	5 5	22 22		4
188338	1		47	37977.00		5			

```
0
                     2018-01
                                Monday -1.938189
                                Monday 0.487889
1
                 0
                     2018-01
2
                 0
                     2018-01
                                Monday -0.320542
3
                 0
                     2018-01
                                Monday -1.249910
4
                 0
                     2018-01
                                Monday 0.153343
                         . . .
. . .
                                Friday -0.298663
                     2019-05
188335
                 0
                                Friday 0.638702
188336
                 0
                     2019-05
                                Friday -0.604647
188337
                 0
                     2019-05
188338
                                Friday 0.345571
                 0
                     2019-05
                 0
                                Friday -0.260466
188339
                     2019-05
[188340 rows x 18 columns]
sales discount=train df[train df['Discount']!='Yes']['Sales']
sales no discount=train df[train df['Discount']=='No']['Sales']
t stat,p val=ttest ind(sales discount,sales no discount,equal var=Fals
print(" Impact of Discounts on Sales:")
if p val<0.05:
    print("Reject Null discounts significantly impact sales.")
    print("fail to reject null Discounts Don't significantly impact
sales.")
sales holiday= train df[train df['Holiday']==1]['Sales']
sales nonholiday=train df[train df['Holiday']==0]['Sales']
t stat, p val= ttest ind(sales holiday,
sales nonholiday,equal var=False)
print("Effect of Holidays on Sales:")
if p val<0.05:
    print("Reject Null holidays significantly impact sales.")
else:
    print("fail to reject null holidays significantly impact sales")
groups=[group['Sales'].values for name, group in
train df.groupby('Store Type')]
f_stat,p_val= f_oneway(*groups)
print("Sales Differences Across Store Types:")
if p val<0.05:
    print("Reject Null store_type significantly impact sales.")
    print("fail to reject null store type significantly impact sales")
```

```
groups=[group['Sales'].values for name, group in
train df.groupby('Region Code')]
kruskal(*groups)
print("Regional Sales Variability:")
if p val<0.05:
    print("Reject Null Region code significantly impact sales.")
    print("fail to reject null Region code significantly impact
sales")
corr, p val = pearsonr(train df['Order'], train df['Sales'])
print("Pearsonr Correlation between Orders and Sales:")
if p val<0.05:
    print("Reject Null order significantly impact sales.")
    print("fail to reject null order significantly impact sales")
# If data is skewed:
print("spearmanr Correlation between Orders and Sales:")
corr, p val = spearmanr(train df['Order'], train df['Sales'])
if p val<0.05:
    print("Reject Null order significantly impact sales.")
else:
    print("fail to reject null order significantly impact sales")
Impact of Discounts on Sales:
fail to reject null Discounts Don't significantly impact sales.
Effect of Holidays on Sales:
Reject Null holidays significantly impact sales.
C:\Users\HP\AppData\Local\Temp\ipykernel 14368\3417656146.py:25:
FutureWarning: The default of observed=False is deprecated and will be
changed to True in a future version of pandas. Pass observed=False to
retain current behavior or observed=True to adopt the future default
and silence this warning.
  groups=[group['Sales'].values for name, group in
train df.groupby('Store Type')]
C:\Users\HP\AppData\Local\Temp\ipykernel 14368\3417656146.py:34:
FutureWarning: The default of observed=False is deprecated and will be
changed to True in a future version of pandas. Pass observed=False to
retain current behavior or observed=True to adopt the future default
and silence this warning.
  groups=[group['Sales'].values for name, group in
train df.groupby('Region Code')]
```

Sales Differences Across Store Types:
Reject Null store_type significantly impact sales.
Regional Sales Variability:
Reject Null Region_code significantly impact sales.
Pearsonr Correlation between Orders and Sales:
Reject Null order significantly impact sales.
spearmanr Correlation between Orders and Sales:
Reject Null order significantly impact sales.