

```
In [4]: import pandas as pd
import plotly.express as px
import plotly.graph_objects as go
import plotly.io as pio
pio.templates.default = "plotly_white"
path=r'C:\Users\saisr\Downloads\retail_price.csv'
data = pd.read_csv(path)
print(data.head())
```

	product_id	product_category_name	month_year	qty	total_price	\
0	bed1	bed_bath_table	01-05-2017	1	45.95	
1	bed1	bed_bath_table	01-06-2017	3	137.85	
2	bed1	bed_bath_table	01-07-2017	6	275.70	
3	bed1	bed_bath_table	01-08-2017	4	183.80	
4	bed1	bed_bath_table	01-09-2017	2	91.90	

	freight_price	unit_price	product_name_lenght	product_description_lenght	\
0	15.100000	45.95	39	161	
1	12.933333	45.95	39	161	
2	14.840000	45.95	39	161	
3	14.287500	45.95	39	161	
4	15.100000	45.95	39	161	

	product_photos_qty	...	comp_1	ps1	fp1	comp_2	ps2	\
0	2	...	89.9	3.9	15.011897	215.000000	4.4	
1	2	...	89.9	3.9	14.769216	209.000000	4.4	
2	2	...	89.9	3.9	13.993833	205.000000	4.4	
3	2	...	89.9	3.9	14.656757	199.509804	4.4	
4	2	...	89.9	3.9	18.776522	163.398710	4.4	

	fp2	comp_3	ps3	fp3	lag_price
0	8.760000	45.95	4.0	15.100000	45.90
1	21.322000	45.95	4.0	12.933333	45.95
2	22.195932	45.95	4.0	14.840000	45.95
3	19.412885	45.95	4.0	14.287500	45.95
4	24.324687	45.95	4.0	15.100000	45.95

[5 rows x 30 columns]

```
In [5]: print(data.isnull().sum())
```

```
product_id          0
product_category_name  0
month_year          0
qty                0
total_price         0
freight_price       0
unit_price          0
product_name_lenght  0
product_description_lenght  0
product_photos_qty  0
product_weight_g     0
product_score       0
customers           0
weekday             0
weekend             0
holiday             0
month              0
year               0
s                  0
volume             0
comp_1             0
ps1                0
fp1                0
comp_2             0
ps2                0
fp2                0
comp_3             0
ps3                0
fp3                0
lag_price           0
dtype: int64
```

```
In [6]: print(data.describe())
```

	qty	total_price	freight_price	unit_price	\
count	676.000000	676.000000	676.000000	676.000000	
mean	14.495562	1422.708728	20.682270	106.496800	
std	15.443421	1700.123100	10.081817	76.182972	
min	1.000000	19.900000	0.000000	19.900000	
25%	4.000000	333.700000	14.761912	53.900000	
50%	10.000000	807.890000	17.518472	89.900000	
75%	18.000000	1887.322500	22.713558	129.990000	
max	122.000000	12095.000000	79.760000	364.000000	

	product_name_lenght	product_description_lenght	product_photos_qty	\
count	676.000000	676.000000	676.000000	
mean	48.720414	767.399408	1.994083	
std	9.420715	655.205015	1.420473	
min	29.000000	100.000000	1.000000	
25%	40.000000	339.000000	1.000000	
50%	51.000000	501.000000	1.500000	
75%	57.000000	903.000000	2.000000	
max	60.000000	3006.000000	8.000000	

	product_weight_g	product_score	customers	...	comp_1	\
count	676.000000	676.000000	676.000000	...	676.000000	
mean	1847.498521	4.085503	81.028107	...	79.452054	
std	2274.808483	0.232021	62.055560	...	47.933358	
min	100.000000	3.300000	1.000000	...	19.900000	
25%	348.000000	3.900000	34.000000	...	49.910000	
50%	950.000000	4.100000	62.000000	...	69.900000	
75%	1850.000000	4.200000	116.000000	...	104.256549	
max	9750.000000	4.500000	339.000000	...	349.900000	

	ps1	fp1	comp_2	ps2	fp2	comp_3	\
count	676.000000	676.000000	676.000000	676.000000	676.000000	676.000000	
mean	4.159467	18.597610	92.930079	4.123521	18.620644	84.182642	
std	0.121652	9.406537	49.481269	0.207189	6.424174	47.745789	
min	3.700000	0.095439	19.900000	3.300000	4.410000	19.900000	
25%	4.100000	13.826429	53.900000	4.100000	14.485000	53.785714	
50%	4.200000	16.618984	89.990000	4.200000	16.811765	59.900000	
75%	4.200000	19.732500	117.888889	4.200000	21.665238	99.990000	
max	4.500000	57.230000	349.900000	4.400000	57.230000	255.610000	

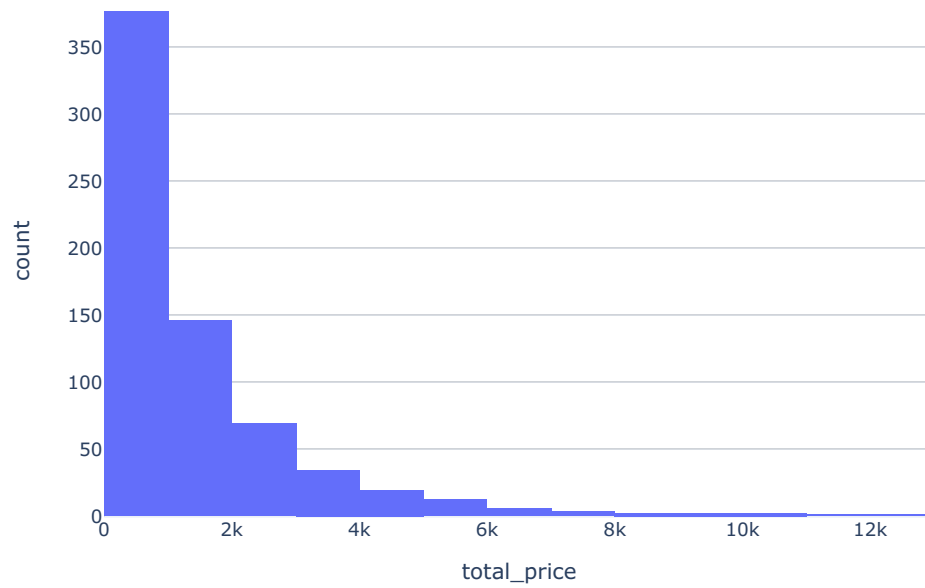
	ps3	fp3	lag_price
count	676.000000	676.000000	676.000000
mean	4.002071	17.965007	107.399684
std	0.233292	5.533256	76.974657
min	3.500000	7.670000	19.850000
25%	3.900000	15.042727	55.668750
50%	4.000000	16.517110	89.900000
75%	4.100000	19.447778	129.990000
max	4.400000	57.230000	364.000000

[8 rows x 27 columns]

```
In [7]: fig = px.histogram(data,
                        x='total_price',
                        nbins=20,
                        title='Distribution of Total Price')

fig.show()
```

Distribution of Total Price



```
In [8]: fig = px.box(data,
    y='unit_price',
    title='Box Plot of Unit Price')
fig.show()
```

```
In [9]: fig = px.scatter(data,
    x='qty',
    y='total_price',
    title='Quantity vs Total Price', trendline="ols")
fig.show()
```

```
In [10]: fig = px.bar(data, x='product_category_name',  
                    y='total_price',  
                    title='Average Total Price by Product Category')  
fig.show()
```

```
In [11]: fig = px.box(data, x='weekday',  
                    y='total_price',  
                    title='Box Plot of Total Price by Weekday')  
fig.show()
```

```
In [18]: fig = px.box(data, x='holiday',  
                    y='total_price',  
                    title='Box Plot of Total Price by Holiday')  
fig.show()
```

```
In [12]: correlation_matrix = data.corr()  
fig = go.Figure(go.Heatmap(x=correlation_matrix.columns,  
                           y=correlation_matrix.columns,  
                           z=correlation_matrix.values))  
fig.update_layout(title='Correlation Heatmap of Numerical Features')  
fig.show()
```

C:\Users\saisr\AppData\Local\Temp\ipykernel_19340\1131337044.py:1: FutureWarning:

The default value of `numeric_only` in `DataFrame.corr` is deprecated. In a future version, it will default to `False`. Select only valid columns or specify the value of `numeric_only` to silence this warning.

```
In [13]: data['comp_price_diff'] = data['unit_price'] - data['comp_1']

avg_price_diff_by_category = data.groupby('product_category_name')['comp_price_diff'].mean().reset_index()

fig = px.bar(avg_price_diff_by_category,
              x='product_category_name',
              y='comp_price_diff',
              title='Average Competitor Price Difference by Product Category')
fig.update_layout(
    xaxis_title='Product Category',
    yaxis_title='Average Competitor Price Difference'
)
fig.show()
```

```
In [14]: from sklearn.model_selection import train_test_split
from sklearn.tree import DecisionTreeRegressor
from sklearn.metrics import mean_squared_error

X = data[['qty', 'unit_price', 'comp_1',
          'product_score', 'comp_price_diff']]
```

```
y = data['total_price']

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

```
In [16]: model = DecisionTreeRegressor()
         model.fit(X_train, y_train)
```

```
Out[16]: ▼ DecisionTreeRegressor
         DecisionTreeRegressor()
```

```
In [17]: y_pred = model.predict(X_test)

fig = go.Figure()
fig.add_trace(go.Scatter(x=y_test, y=y_pred, mode='markers',
                        marker=dict(color='blue'),
                        name='Predicted vs. Actual Retail Price'))
fig.add_trace(go.Scatter(x=[min(y_test), max(y_test)], y=[min(y_test), max(y_test)],
                        mode='lines',
                        marker=dict(color='red'),
                        name='Ideal Prediction'))

fig.update_layout(
    title='Predicted vs. Actual Retail Price',
    xaxis_title='Actual Retail Price',
    yaxis_title='Predicted Retail Price'
)
fig.show()
```

```
In [ ]:
```