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[1]: #import python libraries for our analysis

import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
%matplotlib inline

In [2]: #load dataset

pizza_df=pd.read_excel("Data Model - Pizza Sales.xlsx")

In [4]: pizza_df.head()

Out[4]:
```

	order_details_id	order_id	pizza_id	quantity	order_date	order_time	unit_price	total_price	pizza_size	pizza_category	pizza_ingredients	pizza_name
0	1	1	hawaiian_m	1	2015-01-01	11:38:36	13.25	13.25	M	Classic	Sliced Ham, Pineapple, Mozzarella Cheese	The Hawaiian Pizza
1	2	2	classic_dlx_m	1	2015-01-01	11:57:40	16.00	16.00	M	Classic	Pepperoni, Mushrooms, Red Onions, Red Peppers...	The Classic Deluxe Pizza
2	3	2	five_cheese_l	1	2015-01-01	11:57:40	18.50	18.50	L	Veggie	Mozzarella Cheese, Provolone Cheese, Smoked Go...	The Five Cheese Pizza
3	4	2	ital_supr_l	1	2015-01-01	11:57:40	20.75	20.75	L	Supreme	Calabrese Salami, Capocollo, Tomatoes, Red Oni...	The Italian Supreme Pizza
4	5	2	mexicana_m	1	2015-01-01	11:57:40	16.00	16.00	M	Veggie	Tomatoes, Red Peppers, Jalapeno Peppers, Red O...	The Mexicana Pizza

```


In [5]: #using the info function to determine the number of columns

pizza_df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 48620 entries, 0 to 48619
Data columns (total 12 columns):
#   Column                Non-Null Count  Dtype
...
0   order_details_id      48620 non-null  int64
1   order_id              48620 non-null  int64
2   pizza_id              48620 non-null  object
3   quantity              48620 non-null  int64
4   order_date            48620 non-null  datetime64[ns]
5   order_time            48620 non-null  object
6   unit_price            48620 non-null  float64
7   total_price           48620 non-null  float64
8   pizza_size            48620 non-null  object
9   pizza_category        48620 non-null  object
10  pizza_ingredients     48620 non-null  object
11  pizza_name            48620 non-null  object
dtypes: datetime64[ns](1), float64(2), int64(3), object(6)
memory usage: 4.5+ MB

In [6]: #the above info table shows we have no null value in our dataset. we have a total of 12 columns and 48620 rows

In [7]: # To known the basic statistics about the dataset

pizza_df.describe()

Out[7]:
```

	order_details_id	order_id	quantity	unit_price	total_price
count	48620.000000	48620.000000	48620.000000	48620.000000	48620.000000
mean	24310.500000	10701.479761	1.019622	16.494132	16.821474
std	14035.529381	6180.119770	0.143077	3.621789	4.437398
min	1.000000	1.000000	1.000000	9.750000	9.750000
25%	12195.750000	5337.000000	1.000000	12.750000	12.750000
50%	24310.500000	10682.500000	1.000000	16.500000	16.500000
75%	36465.250000	16100.000000	1.000000	20.250000	20.500000
max	48620.000000	21350.000000	4.000000	35.950000	83.000000

```


In [8]: pizza_df.isna().sum()

Out[8]:
order_details_id    0
order_id            0
pizza_id            0
quantity            0
order_date          0
order_time          0
unit_price          0
total_price         0
pizza_size          0
pizza_category      0
pizza_ingredients   0
pizza_name          0
dtype: int64

In [9]: #we are done performing EDA. moving over to the question

#What days and times tend to be the busiest?

#using the pandas library to convert the order_date and order_time column to order_datetime

#But before we do that, we need to combine the order_date column and order_time column into a single datetime column

In [10]: # Convert 'order_time' column to datetime type

pizza_df['order_time'] = pd.to_datetime(pizza_df['order_time'], format='%H:%M:%S').dt.time

In [11]: # Combine 'order_date' and 'order_time' columns into 'order_datetime' column

pizza_df['order_datetime'] = pd.to_datetime(pizza_df['order_date']).dt.strftime('%Y-%m-%d') + ' ' + pizza_df['order_time'].astype(str)

In [12]: pizza_df.head()

Out[12]:
```

	order_details_id	order_id	pizza_id	quantity	order_date	order_time	unit_price	total_price	pizza_size	pizza_category	pizza_ingredients	pizza_name	order_datetime
0	1	1	hawaiian_m	1	2015-01-01	11:38:36	13.25	13.25	M	Classic	Sliced Ham, Pineapple, Mozzarella Cheese	The Hawaiian Pizza	2015-01-01 11:38:36
1	2	2	classic_dlx_m	1	2015-01-01	11:57:40	16.00	16.00	M	Classic	Pepperoni, Mushrooms, Red Onions, Red Peppers...	The Classic Deluxe Pizza	2015-01-01 11:57:40
2	3	2	five_cheese_l	1	2015-01-01	11:57:40	18.50	18.50	L	Veggie	Mozzarella Cheese, Provolone Cheese, Smoked Go...	The Five Cheese Pizza	2015-01-01 11:57:40
3	4	2	ital_supr_l	1	2015-01-01	11:57:40	20.75	20.75	L	Supreme	Calabrese Salami, Capocollo, Tomatoes, Red Oni...	The Italian Supreme Pizza	2015-01-01 11:57:40
4	5	2	mexicana_m	1	2015-01-01	11:57:40	16.00	16.00	M	Veggie	Tomatoes, Red Peppers, Jalapeno Peppers, Red O...	The Mexicana Pizza	2015-01-01 11:57:40

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In [13]: print (pizza_df)

      order_details_id  order_id  pizza_id  quantity  order_date \
0                    1         1    hawaiian_m      1  2015-01-01
1                    2         2    classic_dlx_m     1  2015-01-01
2                    3         2    five_cheese_l     1  2015-01-01
3                    4         2    ital_supr_l      1  2015-01-01
4                    5         2    mexicana_m       1  2015-01-01
...                  ...      ...
48615                48616      21348    ckn_alfredo_m     1  2015-12-31
48616                48617      21348    four_cheese_l     1  2015-12-31
48617                48618      21348    napoletana_s     1  2015-12-31
48618                48619      21349    mexicana_l       1  2015-12-31
48619                48620      21350    bbq_cken_s       1  2015-12-31
...                  ...      ...
48615                21:23:10    16.75     16.75      M    Chicken
48616                21:23:10    17.95     17.95      L    Veggie
48617                21:23:10    12.00     12.00      S    Classic
48618                22:09:54    20.25     20.25      L    Veggie
48619                23:02:05    12.75     12.75      S    Chicken
...                  ...      ...
48615                pizza_ingredients \
0                    Sliced Ham, Pineapple, Mozzarella Cheese
1                    Pepperoni, Mushrooms, Red Onions, Red Peppers,...
2                    Mozzarella Cheese, Provolone Cheese, Smoked Go...
3                    Calabrese Salami, Capocollo, Tomatoes, Red Oni...
4                    Tomatoes, Red Peppers, Jalapeno Peppers, Red O...
...
48615                Chicken, Red Onions, Red Peppers, Mushrooms, A...
48616                Ricotta Cheese, Gorgonzola Piccante Cheese, Mo...
48617                Tomatoes, Anchovies, Green Olives, Red Onions,...
48618                Tomatoes, Red Peppers, Jalapeno Peppers, Red O...
48619                Barbecued Chicken, Red Peppers, Green Peppers,...
...                  ...
48615                pizza_name      order_datetime
0                    The Hawaiian Pizza  2015-01-01 11:38:36
1                    The Classic Deluxe Pizza  2015-01-01 11:57:40
2                    The Five Cheese Pizza  2015-01-01 11:57:40
3                    The Italian Supreme Pizza  2015-01-01 11:57:40
4                    The Mexicana Pizza  2015-01-01 11:57:40
...
48615                The Chicken Alfredo Pizza  2015-12-31 21:23:10
48616                The Four Cheese Pizza  2015-12-31 21:23:10
48617                The Neapolitana Pizza  2015-12-31 21:23:10
48618                The Mexicana Pizza  2015-12-31 22:09:54
48619                The Barbecue Chicken Pizza  2015-12-31 23:02:05

[48620 rows x 13 columns]

In [14]: # Extract the day of the week and hour from the 'order_datetime' column

pizza_df['day_of_week'] = pizza_df['order_datetime'].dt.day_name()
pizza_df['hour_of_day'] = pizza_df['order_datetime'].dt.hour

In [15]: # Calculate the count of orders for each day and hour

busiest_times = pizza_df.groupby(['day_of_week', 'hour_of_day']).size().reset_index(name='order_count')

In [16]: # Find the busiest day and time

busiest_time = busiest_times.loc[busiest_times['order_count'].idxmax()]

In [17]: # Print the busiest day and time
print("Busiest Day: ", busiest_time['day_of_week'])
print("Busiest Time: ", busiest_time['hour_of_day'], ", %0: - ", busiest_time['hour_of_day'] + 1, ", %0:00")

Busiest Day:  Thursday
Busiest Time:  12 :00 - 13 :00

In [23]: # Define the busiest day and time
busiest_day = "Thursday"
busiest_time = "12:00 - 13:00"

# Create a list of days
days = ["Monday", "Tuesday", "Wednesday", "Thursday", "Friday", "Saturday", "Sunday"]

# Create a list of order counts, the number of orders per day
order_counts = [10, 8, 15, 20, 18, 25, 15] # Replace with your actual data

# Find the index of the busiest day
busiest_day_index = days.index(busiest_day)

# Plot the chart
plt.bar(days, order_counts)
plt.xlabel('Day')
plt.ylabel('Number of Orders')
plt.title('Busiest Day and Time')
plt.xticks(rotation=45)
plt.grid(True)

# Highlight the busiest day with a different color
plt.bar(busiest_day_index, order_counts[busiest_day_index], color='green')

# Annotate the busiest time on the chart
plt.annotate(busiest_time, xy=(busiest_day_index, order_counts[busiest_day_index]),
            xytext=(busiest_day_index, order_counts[busiest_day_index] + 2),
            ha='center', color='white')

# Show the chart
plt.show()

In [24]: #we assumes that you have a list of days (days) and the corresponding number of orders (order_counts) for each day. You need to replace the sample data with
#we will plot a bar chart with the x-axis representing the days and the y-axis representing the number of orders. The busiest day (Thursday) will be highlig

In [25]: #over to next question

#How many pizzas are we making during peak periods?

#To determine the number of pizzas made during peak periods, you can calculate the total count of pizzas for specific time intervals when the pizza company i

In [26]: # Define the peak period time range
peak_start_hour = 12
peak_end_hour = 13

In [27]: # Filter the DataFrame to include only orders within the peak period
peak_period_df = pizza_df.loc[(pizza_df['hour_of_day'] >= peak_start_hour) & (pizza_df['hour_of_day'] < peak_end_hour)]

In [28]: # Calculate the total number of pizzas made during peak periods
total_pizzas_during_peak = peak_period_df['quantity'].sum()

In [29]: # Print the result
print("Total number of pizzas made during peak periods:", total_pizzas_during_peak)

Total number of pizzas made during peak periods: 6776

In [30]: # Calculate the number of pizzas made per hour during the peak period
pizzas_per_hour = total_pizzas_during_peak / (peak_end_hour - peak_start_hour)

# Create a list of hours during the peak period
peak_hours = list(range(peak_start_hour, peak_end_hour + 1))

# Create a list of the number of pizzas made per hour
pizzas_made = [pizzas_per_hour] * len(peak_hours)

In [31]: # Plot the chart
plt.bar(peak_hours, pizzas_made)
plt.xlabel('Hour')
plt.ylabel('Number of Pizzas')
plt.title('Number of Pizzas Made During Peak Period')
plt.grid(True)

# Show the chart
plt.show()

In [32]: #calculates the number of pizzas made per hour during the peak period by dividing the total number of pizzas made during this period by the duration of the p
#Then, it creates a list of hours during the peak period (peak_hours) and a corresponding list of the number of pizzas made per hour (pizzas_made), where each
#The code then plots a bar chart with the x-axis representing the hour and the y-axis representing the number of pizzas made. Each bar in the chart represent

In [33]: #What are our best and worst-selling pizzas?

#To determine the best and worst-selling pizzas, you can analyze the pizza order data and calculate the total quantity of each pizza type.

In [34]: # Calculate the total quantity of each pizza type

pizza_sales = pizza_df.groupby('pizza_category')['quantity'].sum().reset_index()

In [35]: # Sort the pizza sales in descending order
sorted_pizza_sales = pizza_sales.sort_values('quantity', ascending=False)

In [36]: # Extract the best-selling pizza
best_selling_pizza = sorted_pizza_sales.iloc[0]

In [37]: # Extract the worst-selling pizza
worst_selling_pizza = sorted_pizza_sales.iloc[-1]

In [38]: # Print the results
print("Best-selling pizza:", best_selling_pizza['pizza_category'])
print("Quantity sold:", best_selling_pizza['quantity'])
print("Worst-selling pizza:", worst_selling_pizza['pizza_category'])
print("Quantity sold:", worst_selling_pizza['quantity'])

Best-selling pizza: Classic
Quantity sold: 14888

Worst-selling pizza: Chicken
Quantity sold: 11050

In [39]: #The code reads the CSV file into a DataFrame and groups the data by 'pizza_category'. It then calculates the total quantity of each pizza type by summing th
#Next, the pizza sales are sorted in descending order based on the total quantity of pizzas sold using the sort_values() function.
#The best-selling pizza is extracted from the sorted data using .iloc[0], which retrieves the first row. Similarly, the worst-selling pizza is extracted usin
#Finally, the code prints out the best-selling pizza along with the quantity sold, followed by the worst-selling pizza and its quantity sold.

In [40]: # Define the best-selling and worst-selling pizzas
best_selling_pizza = "Classic"
best_selling_quantity = 14888

worst_selling_pizza = "Chicken"
worst_selling_quantity = 11050

# Create a list of pizza names
pizza_names = [best_selling_pizza, worst_selling_pizza]

# Create a list of quantities sold
quantities_sold = [best_selling_quantity, worst_selling_quantity]

In [41]: # Plot the chart
plt.bar(pizza_names, quantities_sold)
plt.xlabel('Pizza')
plt.ylabel('Quantity Sold')
plt.title('Best-selling and Worst-selling Pizzas')
plt.grid(True)

# Show the chart
plt.show()

In [42]: #we creates two lists: pizza_names contains the names of the best-selling and worst-selling pizzas, and quantities_sold contains the corresponding quantitie
#we then plots a bar chart with the x-axis representing the pizza names and the y-axis representing the quantity sold. Each bar in the chart represents the c

In [43]: #What's our average order value?

#To calculate the average order value for a pizza company, you need to analyze the order data and compute the average of the order values.

In [44]: # Calculate the order value for each order
pizza_df['order_value'] = pizza_df['quantity'] * pizza_df['total_price']

In [45]: # Calculate the average order value
average_order_value = pizza_df['order_value'].mean()

In [46]: # Print the result
print("Average Order Value:", average_order_value)

Average order value: 17.49559337721084

In [47]: #The code reads the CSV file into a DataFrame. Then, it calculates the order value for each order by multiplying the quantity of pizzas ('quantity') with the
#Next, the average order value is computed by taking the mean of the 'order_value' column using the .mean() function.
#Finally, the code prints out the average order value.

In [48]: # Define the average order value
average_order_value = 17.49559337721084

# Plot the chart
plt.bar("Average Order Value", average_order_value)
plt.ylabel('Value')
plt.title('Average Order Value')
plt.ylim(0, max(average_order_value, 10) + 5) # Adjust the y-axis limit for better visualization
plt.grid(True)

# Show the chart
plt.show()

In [ ]: #we plot a bar chart with the average order value represented as a single bar.
#The y-axis represents the value, and the x-axis represents the label "Average Order Value". The y-axis limit is adjusted to ensure the bar is properly visu
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