

## assignment 16 pizza sales exercise

**Kartik thakur**

### PIZZA RESTAURANT SALES

This pizza sales dataset make up 12 relevant features:

- 1- Order\_id: Unique identifier for each order placed by a table
- 2- Order\_details\_id: Unique identifier for each pizza placed within each order (pizzas of the same type and size are kept in the same row, and the quantity increases)
- 3- Pizza\_id: Unique key identifier that ties the pizza ordered to its details, like size and price
- 4- Quantity: Quantity ordered for each pizza of the same type and size
- 5- Order\_date: Date the order was placed (entered into the system prior to cooking & serving)
- 6- Order\_time: Time the order was placed (entered into the system prior to cooking & serving)
- 7- Unit\_price: Price of the pizza in USD
- 8- Total\_price:  $\text{unit\_price} * \text{quantity}$
- 9- Pizza\_size: Size of the pizza (Small, Medium, Large, X Large, or XX Large)
- 10- Pizza\_type: Unique key identifier that ties the pizza ordered to its details, like size and price
- 11- Pizza\_ingredients: ingredients used in the pizza as shown in the menu (they all include Mozzarella Cheese, even if not specified; and they all include Tomato Sauce, unless another sauce is specified)
- 12- Pizza\_name: Name of the pizza as shown in the menu

Here are some questions that we'd like to be able to answer:

1. What days and times do we tend to be busiest?
2. How many pizzas are we making during peak periods?
3. What are our best and worst-selling pizzas?
4. What's our average order value?
5. How well are we utilizing our seating capacity? (we have 15 tables and 60 seats)

### *Importing Libraries*

```
In [3]: import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import datetime
import plotly.express as px
```

### Loading Dataset

```
In [4]: pizza=pd.read_csv('pizza_sales.csv')
pizza.head()
```

Out[4]:

	order_details_id	order_id	pizza_id	quantity	order_date	order_time	unit_price	total_price
0	1	1	hawaiian_m	1	1/1/2015	11:38:36	13.25	13.25
1	2	2	classic_dlx_m	1	1/1/2015	11:57:40	16.00	16.00
2	3	2	five_cheese_l	1	1/1/2015	11:57:40	18.50	18.50
3	4	2	ital_supr_l	1	1/1/2015	11:57:40	20.75	20.75
4	5	2	mexicana_m	1	1/1/2015	11:57:40	16.00	16.00



## Exploratory Some Information About Dataset

In [5]: `pizza.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  object
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: float64(2), int64(3), object(7)
memory usage: 4.5+ MB
```

In [6]: `pizza.describe()`

Out[6]:

	order_details_id	order_id	quantity	unit_price	total_price
<b>count</b>	48620.000000	48620.000000	48620.000000	48620.000000	48620.000000
<b>mean</b>	24310.500000	10701.479761	1.019622	16.494132	16.821474
<b>std</b>	14035.529381	6180.119770	0.143077	3.621789	4.437398
<b>min</b>	1.000000	1.000000	1.000000	9.750000	9.750000
<b>25%</b>	12155.750000	5337.000000	1.000000	12.750000	12.750000
<b>50%</b>	24310.500000	10682.500000	1.000000	16.500000	16.500000
<b>75%</b>	36465.250000	16100.000000	1.000000	20.250000	20.500000
<b>max</b>	48620.000000	21350.000000	4.000000	35.950000	83.000000

In [7]: `pizza.shape`

Out[7]: (48620, 12)

```
In [8]: pizza.isnull().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]: pizza.columns
```

```
Out[9]: Index(['order_details_id', 'order_id', 'pizza_id', 'quantity', 'order_date',
              'order_time', 'unit_price', 'total_price', 'pizza_size',
              'pizza_category', 'pizza_ingredients', 'pizza_name'],
              dtype='object')
```

### Change the hour object into integer

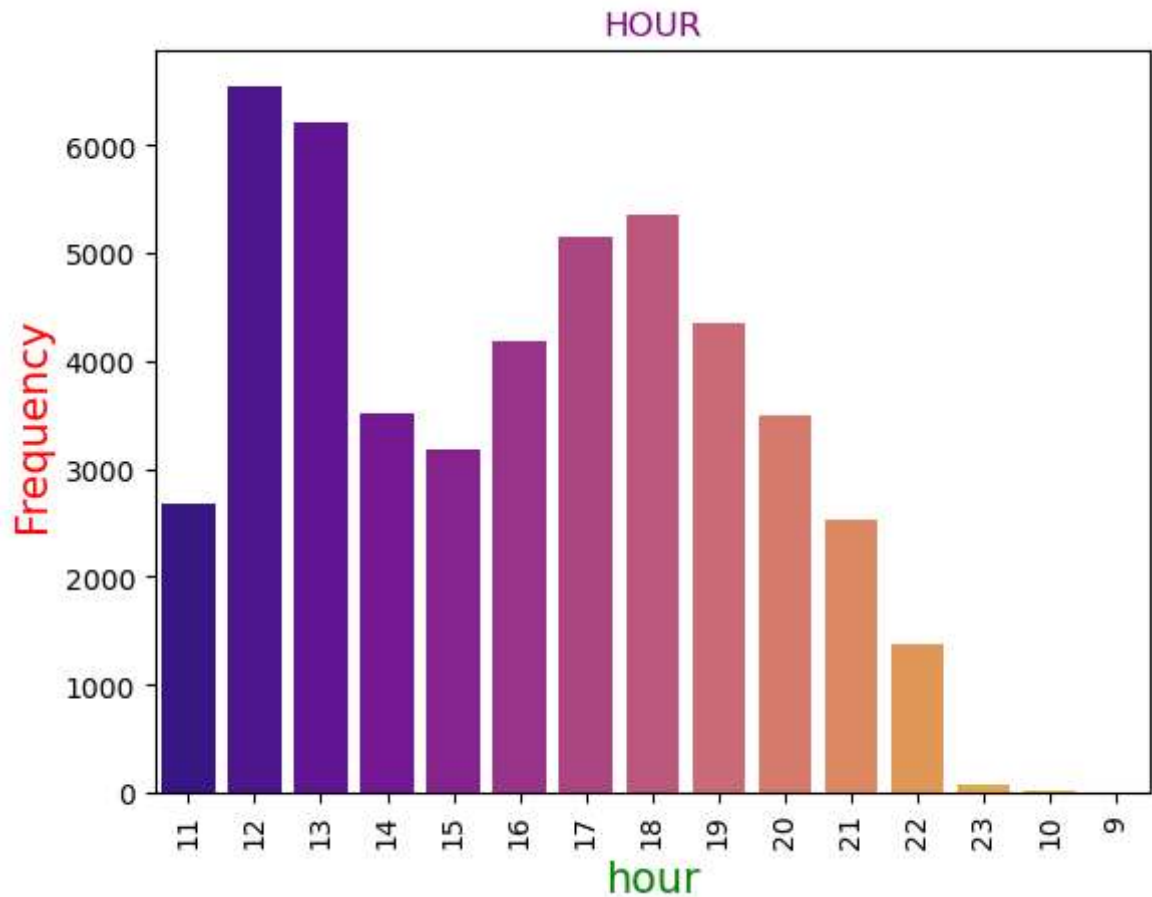
```
In [10]: pizza["order_time"] = pizza["order_time"].astype("string")
pizza[["hour", "minute", "second"]] = pizza["order_time"].str.split(":", expand=True)
pizza["hour"].value_counts()
```

```
Out[10]: 12    6543
13    6203
18    5359
17    5143
19    4350
16    4185
14    3521
20    3487
15    3170
11    2672
21    2528
22    1370
23     68
10     17
9       4
Name: hour, dtype: Int64
```

## Data Visualization

### Q1. Plot the graph between hour and sales

```
In [11]: sns.countplot(data=pizza,x="hour",palette="plasma")  
plt.xticks(rotation=90)  
plt.xlabel("hour",fontsize=15,color="green")  
plt.ylabel("Frequency",fontsize=15,color="red")  
plt.title("HOUR",color="purple")  
plt.show()
```



## Q2. Find the total order in a year

```
In [12]: pizza['order_year'] = pd.DatetimeIndex( pizza['order_date']).year
pizza['order_year'].value_counts()
or specify infer_datetime_format=True for consistent parsing.
dtarr = DatetimeArray._from_sequence_not_strict(
C:\Users\user\anaconda3\lib\site-packages\pandas\core\indexes\datetime.py:
327: UserWarning: Parsing '14/12/2015' in DD/MM/YYYY format. Provide format
or specify infer_datetime_format=True for consistent parsing.
dtarr = DatetimeArray._from_sequence_not_strict(
C:\Users\user\anaconda3\lib\site-packages\pandas\core\indexes\datetime.py:
327: UserWarning: Parsing '15/12/2015' in DD/MM/YYYY format. Provide format
or specify infer_datetime_format=True for consistent parsing.
dtarr = DatetimeArray._from_sequence_not_strict(
C:\Users\user\anaconda3\lib\site-packages\pandas\core\indexes\datetime.py:
327: UserWarning: Parsing '16/12/2015' in DD/MM/YYYY format. Provide format
or specify infer_datetime_format=True for consistent parsing.
dtarr = DatetimeArray._from_sequence_not_strict(
C:\Users\user\anaconda3\lib\site-packages\pandas\core\indexes\datetime.py:
327: UserWarning: Parsing '17/12/2015' in DD/MM/YYYY format. Provide format
or specify infer_datetime_format=True for consistent parsing.
dtarr = DatetimeArray._from_sequence_not_strict(
C:\Users\user\anaconda3\lib\site-packages\pandas\core\indexes\datetime.py:
327: UserWarning: Parsing '18/12/2015' in DD/MM/YYYY format. Provide format
```

## Q3. Find The day-wise sales

```
In [13]: pizza['order_date'] = pd.to_datetime(pizza['order_date'])
pizza['order_date']
C:\Users\user\anaconda3\lib\site-packages\pandas\core\tools\datetime.py:10
47: UserWarning: Parsing '16/12/2015' in DD/MM/YYYY format. Provide format
or specify infer_datetime_format=True for consistent parsing.
cache_array = _maybe_cache(arg, format, cache, convert_listlike)
C:\Users\user\anaconda3\lib\site-packages\pandas\core\tools\datetime.py:10
47: UserWarning: Parsing '17/12/2015' in DD/MM/YYYY format. Provide format
or specify infer_datetime_format=True for consistent parsing.
cache_array = _maybe_cache(arg, format, cache, convert_listlike)
C:\Users\user\anaconda3\lib\site-packages\pandas\core\tools\datetime.py:10
47: UserWarning: Parsing '18/12/2015' in DD/MM/YYYY format. Provide format
or specify infer_datetime_format=True for consistent parsing.
cache_array = _maybe_cache(arg, format, cache, convert_listlike)
C:\Users\user\anaconda3\lib\site-packages\pandas\core\tools\datetime.py:10
47: UserWarning: Parsing '19/12/2015' in DD/MM/YYYY format. Provide format
or specify infer_datetime_format=True for consistent parsing.
cache_array = _maybe_cache(arg, format, cache, convert_listlike)
C:\Users\user\anaconda3\lib\site-packages\pandas\core\tools\datetime.py:10
47: UserWarning: Parsing '20/12/2015' in DD/MM/YYYY format. Provide format
or specify infer_datetime_format=True for consistent parsing.
cache_array = _maybe_cache(arg, format, cache, convert_listlike)
```

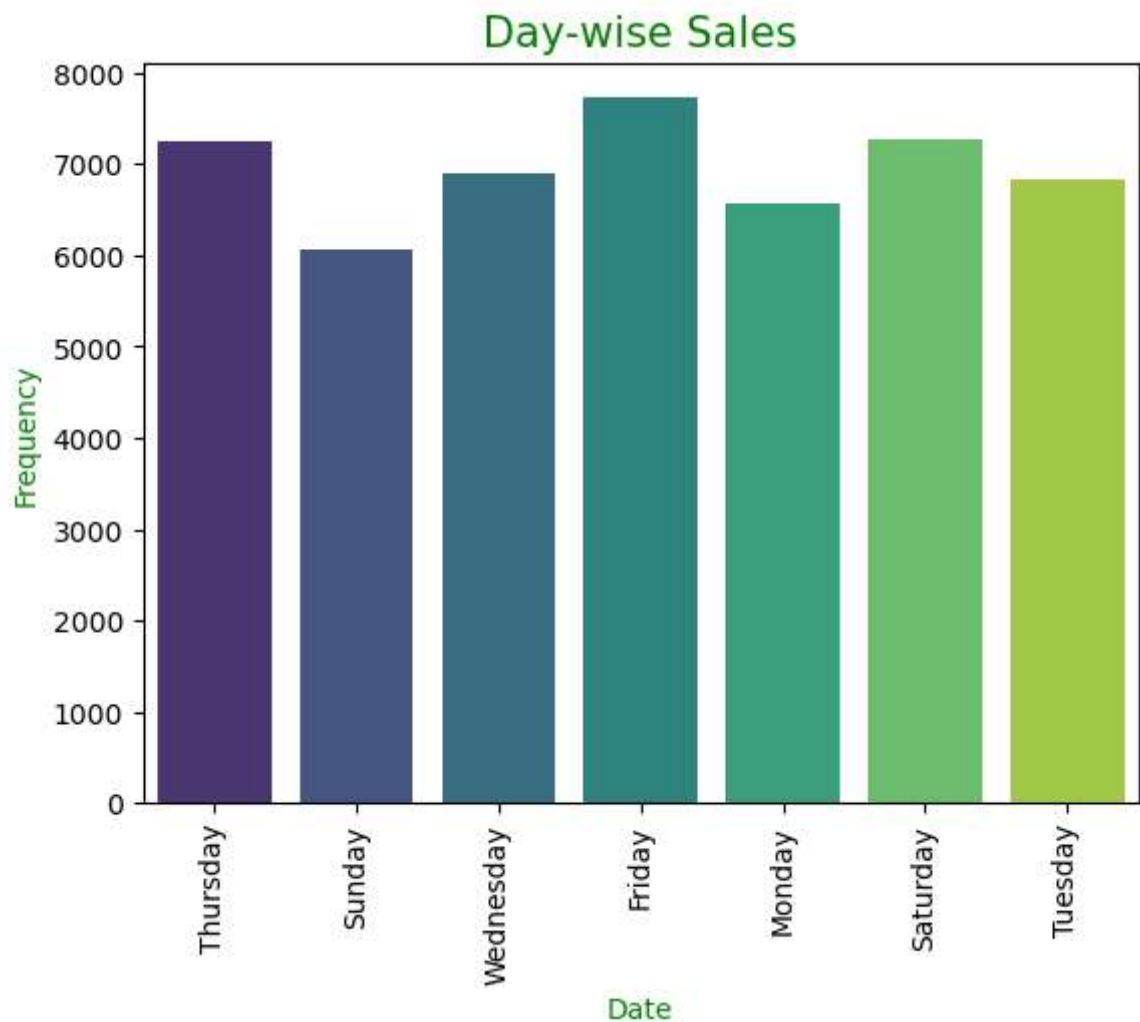
```
In [14]: pizza['order_dates'] = pizza['order_date'].dt.day_name()  
pizza['order_dates'].value_counts()
```

```
Out[14]: Friday      7723  
Saturday    7274  
Thursday    7243  
Wednesday   6907  
Tuesday     6833  
Monday      6577  
Sunday      6063  
Name: order_dates, dtype: int64
```

## Plot the graph for day-wise sales

```
In [15]: palette="viridis"
```

```
In [16]: sns.countplot(data=pizza,x="order_dates",palette="viridis")  
plt.xticks(rotation=90)  
plt.xlabel("Date",fontsize=10,color="green")  
plt.ylabel("Frequency",fontsize=10,color="green")  
plt.title("Day-wise Sales",color="green",fontsize=15)  
plt.show()
```



## Q4. Find Monthly sales

```
In [17]: pizza['order_month'] =pd.DatetimeIndex (pizza['order_date']).month
pizza.loc[(pizza['order_month'] ==1), 'order_month'] = 'January'
pizza.loc[(pizza['order_month'] ==2), 'order_month'] = 'February'
pizza.loc[(pizza['order_month'] ==3), 'order_month'] = 'March'
pizza.loc[(pizza['order_month'] ==4), 'order_month'] = 'April'
pizza.loc[(pizza['order_month'] ==5), 'order_month'] = 'May'
pizza.loc[(pizza['order_month'] ==6), 'order_month'] = 'June'
pizza.loc[(pizza['order_month'] ==7), 'order_month'] = 'July'
pizza.loc[(pizza['order_month'] ==8), 'order_month'] = 'August'
pizza.loc[(pizza['order_month'] ==9), 'order_month'] = 'September'
pizza.loc[(pizza['order_month'] ==10), 'order_month'] = 'October'
pizza.loc[(pizza['order_month'] ==11), 'order_month'] = 'November'
pizza.loc[(pizza['order_month'] ==12), 'order_month'] = 'December'
pizza['order_month'].value_counts()
```

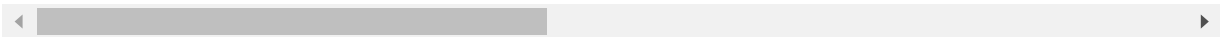
```
Out[17]: January      4288
March      4251
November   4226
July       4212
April      4182
August     4141
October    4045
June       4044
May        4008
February   3821
September  3780
December   3622
Name: order_month, dtype: int64
```



```
In [18]: pizza.head()
```

Out[18]:

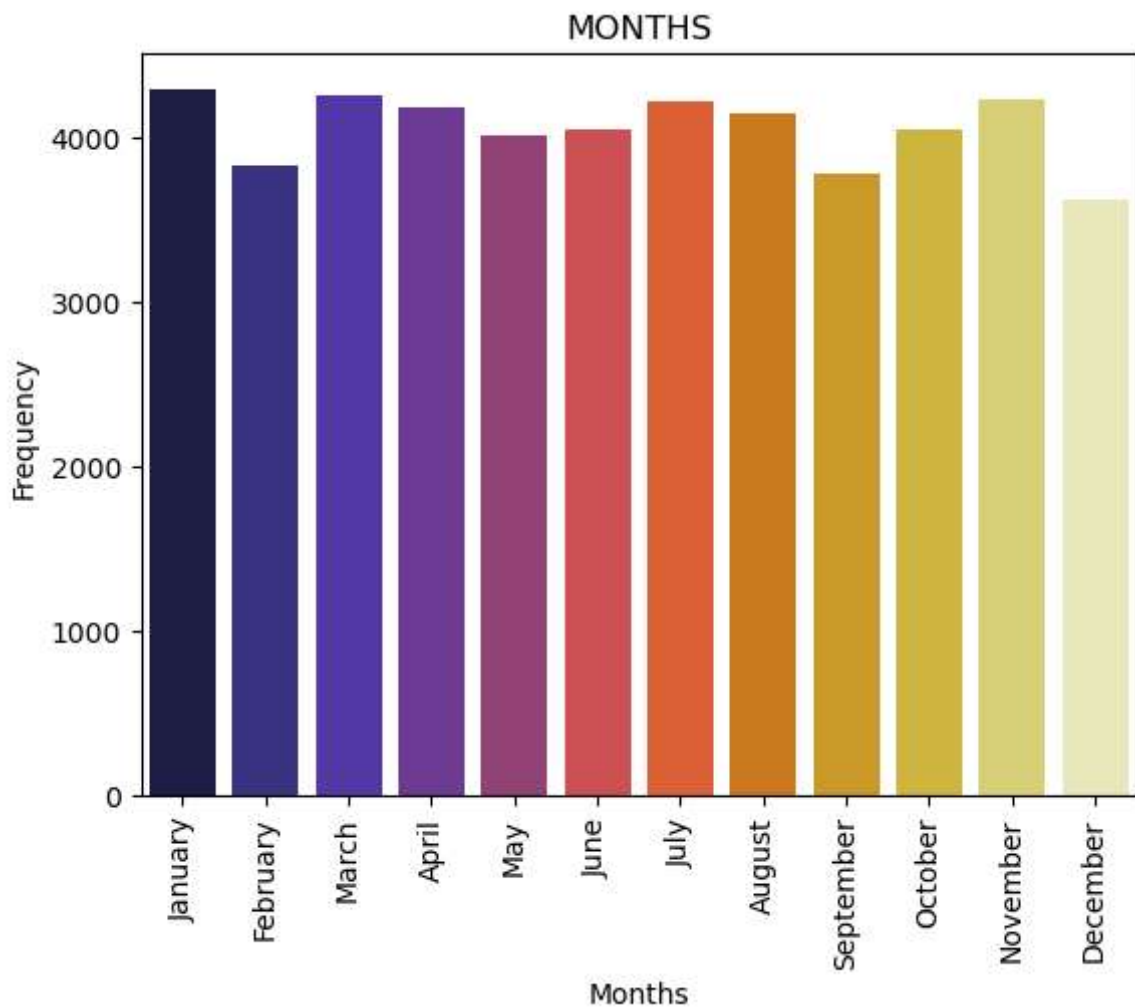
	order_details_id	order_id	pizza_id	quantity	order_date	order_time	unit_price	total_pric
0	1	1	hawaiian_m	1	2015-01-01	11:38:36	13.25	13.2
1	2	2	classic_dlx_m	1	2015-01-01	11:57:40	16.00	16.0
2	3	2	five_cheese_l	1	2015-01-01	11:57:40	18.50	18.5
3	4	2	ital_supr_l	1	2015-01-01	11:57:40	20.75	20.7
4	5	2	mexicana_m	1	2015-01-01	11:57:40	16.00	16.0



Plot graph for monthly sales

```
In [19]: palette="CMRmap"
```

```
In [47]: sns.countplot(data=pizza,x="order_month",palette="CMRmap")
plt.xticks(rotation=90)
plt.xlabel("Months",fontsize=10,color="black")
plt.ylabel("Frequency",fontsize=10,color="black")
plt.title("MONTHS",color="black")
plt.show()
```



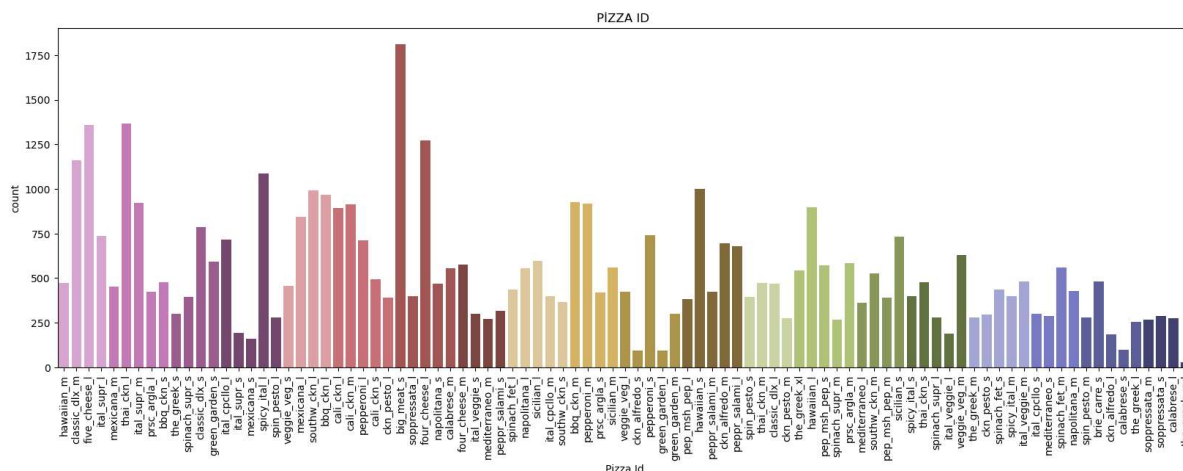
### Find which pizza has highest sales

```
In [48]: pizza.pizza_id.value_counts()
```

```
Out[48]: big_meat_s      1811
thai_chn_l      1365
five_cheese_l   1359
four_cheese_l   1273
classic_dlx_m   1159
...
mexicana_s      160
calabrese_s      99
ckn_alfredo_s    96
green_garden_l   94
the_greek_xxl    28
Name: pizza_id, Length: 91, dtype: int64
```

```
In [49]: palette="tab20b_r"
```

```
In [23]: fig, ax = plt.subplots(figsize=(20, 6))
sns.countplot(data=pizza,x="pizza_id",palette="tab20b_r",ax=ax)
plt.xticks(rotation=90)
plt.xlabel("Pizza Id",fontsize=10,color="black")
plt.title("PIZZA ID ",color="black")
plt.show()
```



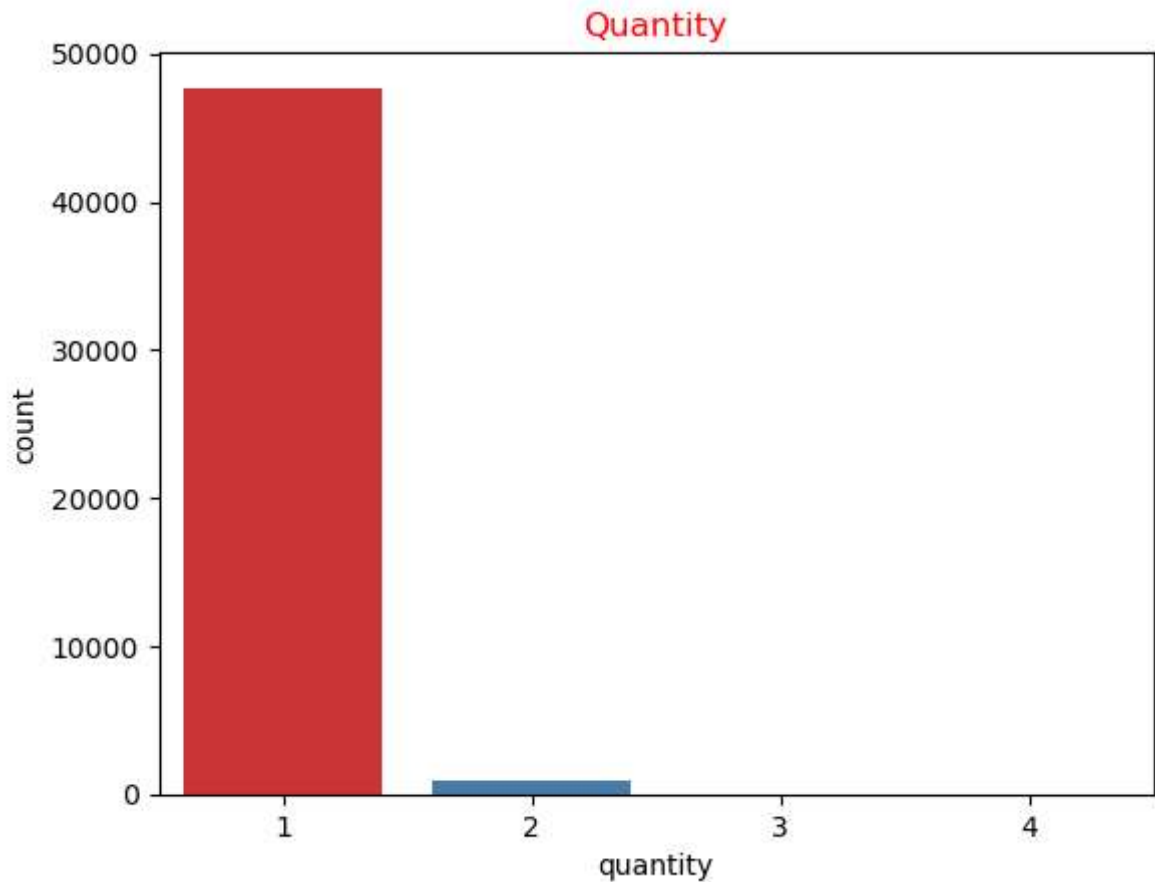
How much quantity of pizzas ordered more.

```
In [24]: pizza.quantity.value_counts()
```

```
Out[24]: 1    47693
         2     903
         3      21
         4       3
         Name: quantity, dtype: int64
```

```
In [25]: palette="Set1"
```

```
In [26]: sns.countplot(data=pizza,x="quantity",palette="Set1")  
plt.xlabel("quantity",fontsize=10)  
plt.title("Quantity",color="red")  
plt.show()
```



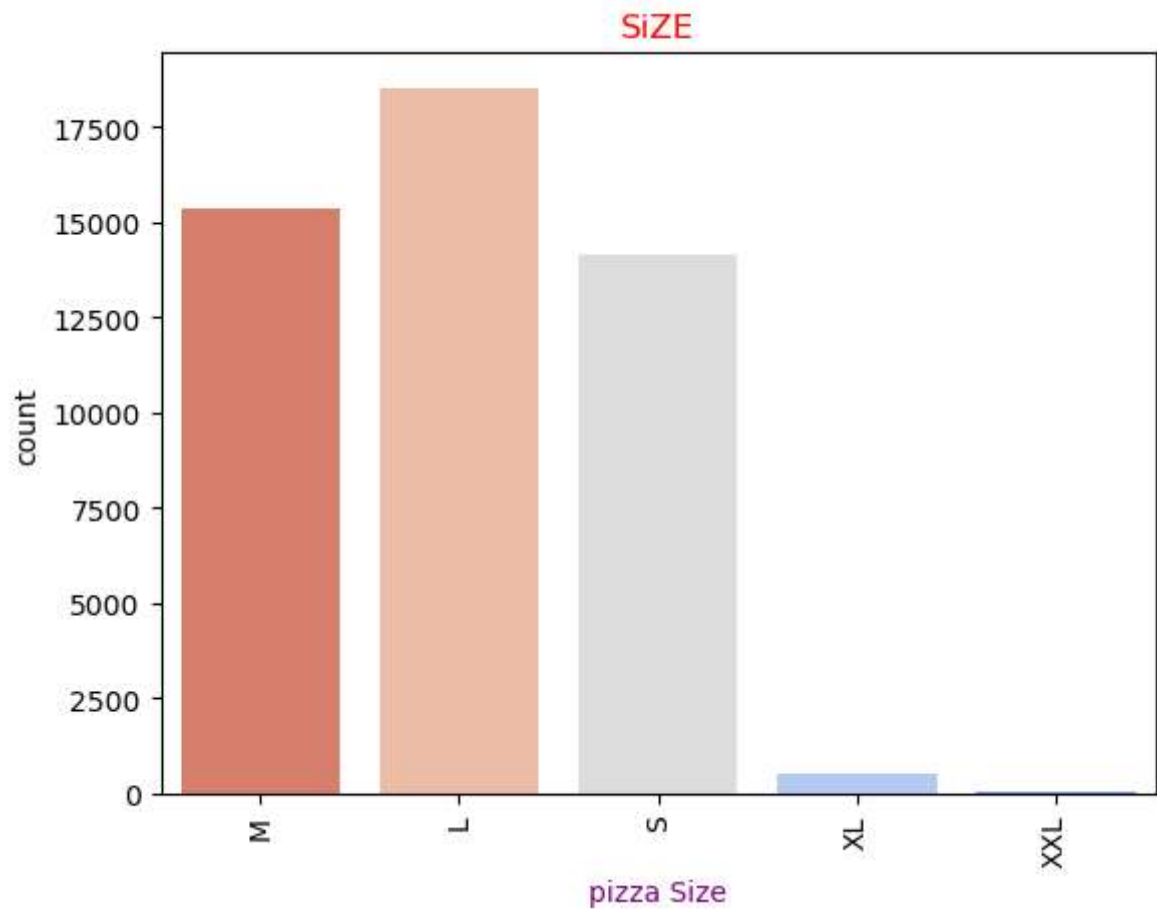
**Which pizza size sales has less in number**

```
In [27]: pizza.pizza_size.value_counts().sort_values()
```

```
Out[27]: XXL      28  
XL        544  
S       14137  
M       15385  
L       18526  
Name: pizza_size, dtype: int64
```

```
In [28]: palette="coolwarm_r"
```

```
In [29]: sns.countplot(data=pizza,x="pizza_size",palette="coolwarm_r")
plt.xticks(rotation=90)
plt.xlabel("pizza Size",fontsize=10,color='purple')
plt.title("SiZE",color="red")
plt.show()
```

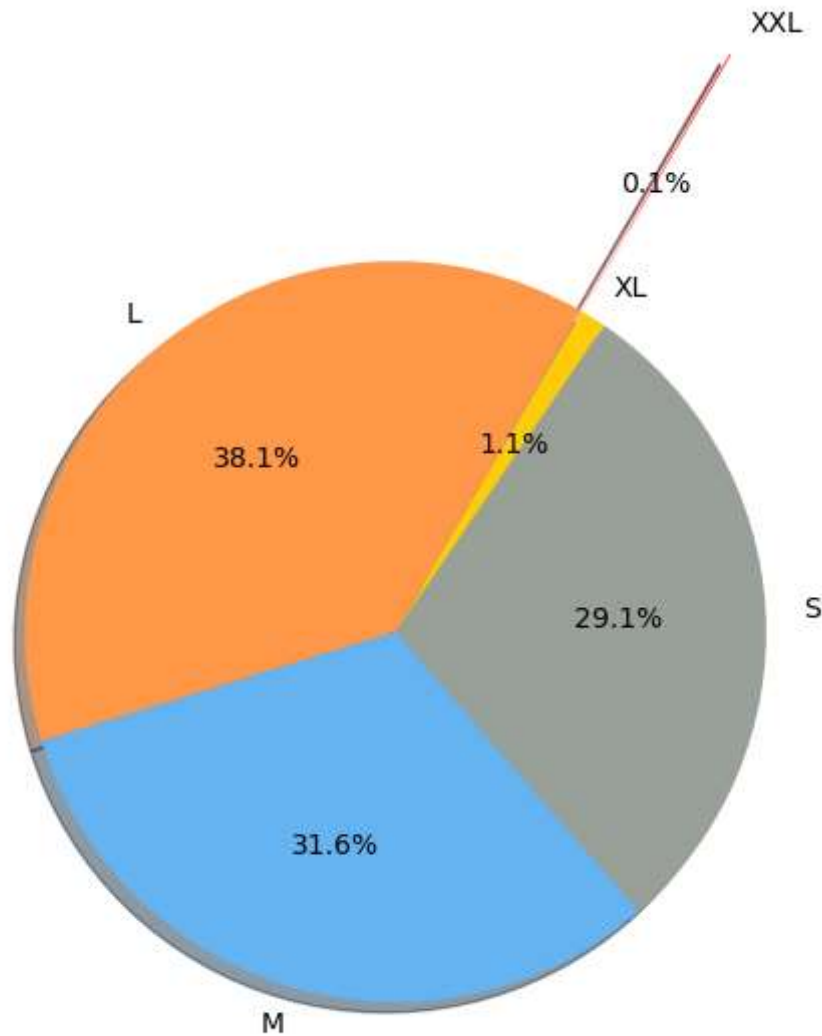


**Plot the piechart with percentage sales of pizza size**

```
In [30]: colors = ['#ff9749', '#66b3f1', '#999f99', '#ffcc09', "RED", "cyan"]
```

```
In [31]: labels = pizza["pizza_size"].value_counts().index
        sizes = pizza["pizza_size"].value_counts()
        plt.figure(figsize = (6,6))

        colors = ['#ff9749', '#66b3f1', '#999f99', '#ffcc09', "RED", "cyan"]
        plt.pie(sizes, labels=labels, rotatelabels=False, autopct='%1.1f%%', colors=colors,
                startangle=60, explode=(0,0,0,0,0.8))
        plt.show()
```



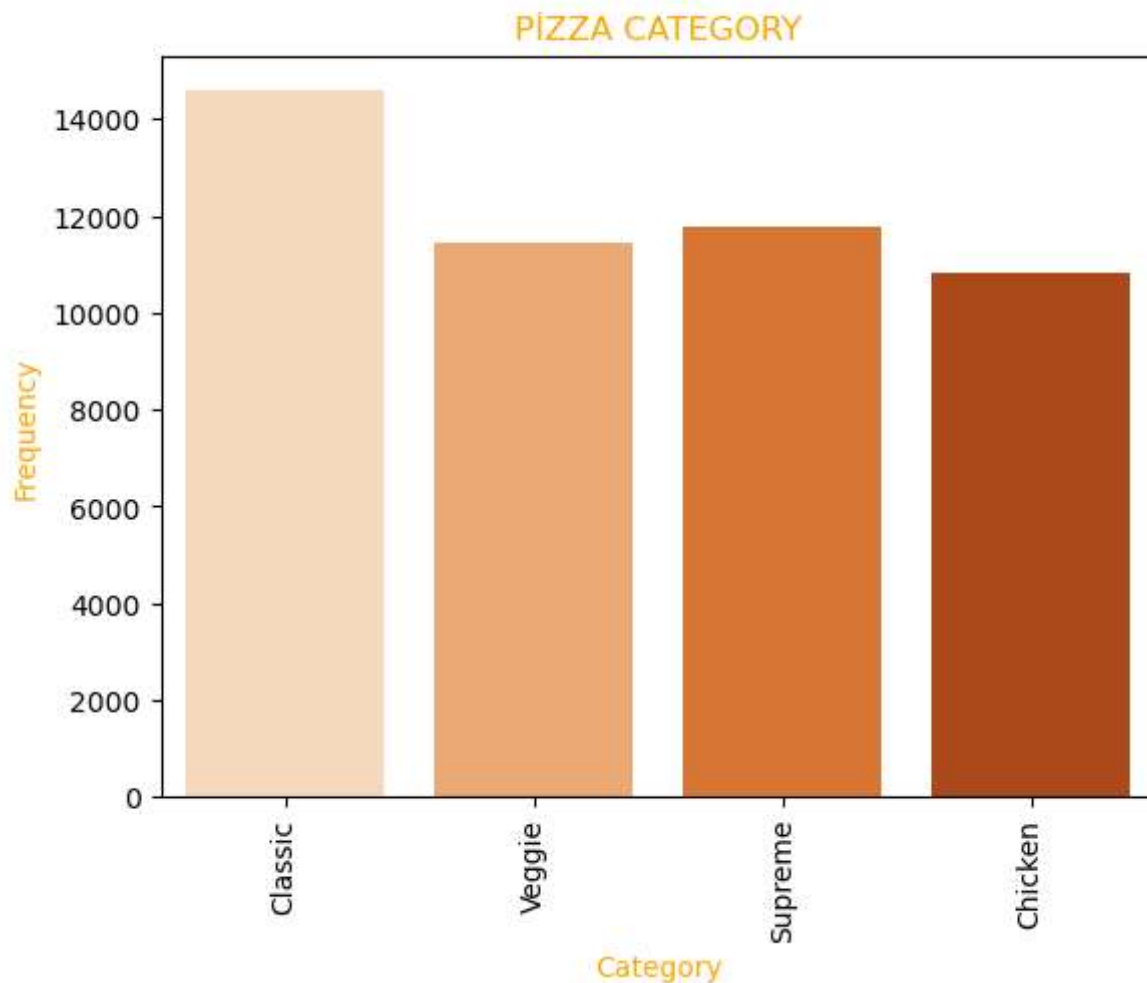
### Find which category of pizza ordered most

```
In [32]: pizza.pizza_category.value_counts()
```

```
Out[32]: Classic      14579
         Supreme      11777
         Veggie       11449
         Chicken      10815
         Name: pizza_category, dtype: int64
```

```
In [33]: palette="Oranges"
```

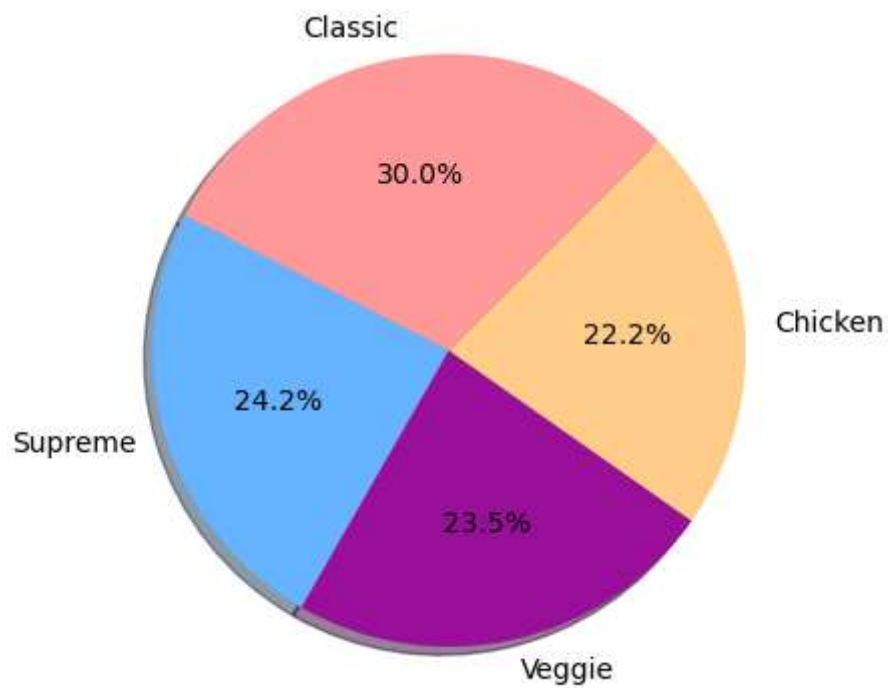
```
In [34]: sns.countplot(data=pizza,x="pizza_category",palette="Oranges")
plt.xticks(rotation=90)
plt.xlabel("Category",fontsize=10,color="orange")
plt.ylabel("Frequency",fontsize=10,color="orange")
plt.title("PIZZA CATEGORY",color="orange")
plt.show()
```



**Plot the piechart which shows the percentage of sales**

```
In [35]: colors = ['#ff9999', '#66b3ff', '#990f99', '#ffcc89', "pink", "yellow"]
```

```
In [36]: labels = pizza["pizza_category"].value_counts().index  
         sizes = pizza["pizza_category"].value_counts()  
         colors = ['#ff9999', '#66b3ff', '#990f99', '#ffcc89', "pink", "yellow"]  
         plt.pie(sizes, labels=labels, rotatelabels=False, autopct='%1.1f%%', colors=colors)  
         plt.show()
```





## Get the value counts for pizza ingredients

```
In [37]: pizza.pizza_ingredients.value_counts()
```

```

Out[37]: Pepperoni, Mushrooms, Red Onions, Red Peppers, Bacon
2416
Barbecued Chicken, Red Peppers, Green Peppers, Tomatoes, Red Onions, Barbecue
Sauce 2372
Sliced Ham, Pineapple, Mozzarella Cheese
2370
Mozzarella Cheese, Pepperoni
2369
Chicken, Pineapple, Tomatoes, Red Peppers, Thai Sweet Chilli Sauce
2315
Chicken, Artichoke, Spinach, Garlic, Jalapeno Peppers, Fontina Cheese, Gouda
Cheese 2302
Coarse Sicilian Salami, Tomatoes, Green Olives, Luganega Sausage, Onions, Gar
lic 1887
Capocollo, Tomatoes, Goat Cheese, Artichokes, Peperoncini verdi, Garlic
1887
Chicken, Tomatoes, Red Peppers, Red Onions, Jalapeno Peppers, Corn, Cilantro,
Chipotle Sauce 1885
Ricotta Cheese, Gorgonzola Piccante Cheese, Mozzarella Cheese, Parmigiano Reg
giano Cheese, Garlic 1850
Calabrese Salami, Capocollo, Tomatoes, Red Onions, Green Olives, Garlic
1849
Bacon, Pepperoni, Italian Sausage, Chorizo Sausage
1811
Mushrooms, Tomatoes, Red Peppers, Green Peppers, Red Onions, Zucchini, Spinac
h, Garlic 1510
Tomatoes, Red Peppers, Jalapeno Peppers, Red Onions, Cilantro, Corn, Chipotle
Sauce, Garlic 1456
Tomatoes, Anchovies, Green Olives, Red Onions, Garlic
1451
Spinach, Mushrooms, Red Onions, Feta Cheese, Garlic
1432
Prosciutto di San Daniele, Arugula, Mozzarella Cheese
1428
Genoa Salami, Capocollo, Pepperoni, Tomatoes, Asiago Cheese, Garlic
1422
Capocollo, Red Peppers, Tomatoes, Goat Cheese, Garlic, Oregano
1414
Kalamata Olives, Feta Cheese, Tomatoes, Garlic, Beef Chuck Roast, Red Onions
1406
Mozzarella Cheese, Provolone Cheese, Smoked Gouda Cheese, Romano Cheese, Blue
Cheese, Garlic 1359
Pepperoni, Mushrooms, Green Peppers
1342
Spinach, Mushrooms, Tomatoes, Green Olives, Feta Cheese
987
Chicken, Red Onions, Red Peppers, Mushrooms, Asiago Cheese, Alfredo Sauce
980
Eggplant, Artichokes, Tomatoes, Zucchini, Red Peppers, Garlic, Pesto Sauce
975
Chicken, Tomatoes, Red Peppers, Spinach, Garlic, Pesto Sauce
961
Spinach, Artichokes, Tomatoes, Sun-dried Tomatoes, Garlic, Pesto Sauce
957
Soppressata Salami, Fontina Cheese, Mozzarella Cheese, Mushrooms, Garlic
957
Spinach, Red Onions, Pepperoni, Tomatoes, Artichokes, Kalamata Olives, Garli

```

```
c, Asiago Cheese          940
Mushrooms, Pepperoni, Mozzarella, Red Onions, Kalamata Olives, Sun-dried Tomatoes, Feta Cheese, Plum Tomatoes, Red Onions 927
Spinach, Artichokes, Kalamata Olives, Sun-dried Tomatoes, Feta Cheese, Plum Tomatoes, Red Onions 923
Brie Carre Cheese, Prosciutto, Caramelized Onions, Pears, Thyme, Garlic 480
Name: pizza_ingredients, dtype: int64
```

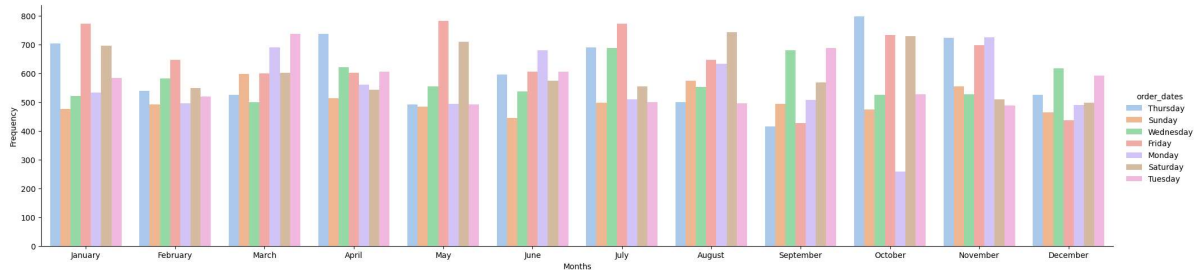
## Find Total Money Earned From Pizza Categories and also plot pie chart

```
In [67]: z=pizza['pizza_category'].value_counts()
fig=px.bar(z,x=z.index,y=z.values,color=z.index,text=z.values,labels={'index':
fig.show()
fig=px.pie(z,names=z.index,values=z.values,labels={'index':'pizza_category','y
fig.show()
```

## Day-wise orders placed in every month

In [51]: `palette="pastel"`

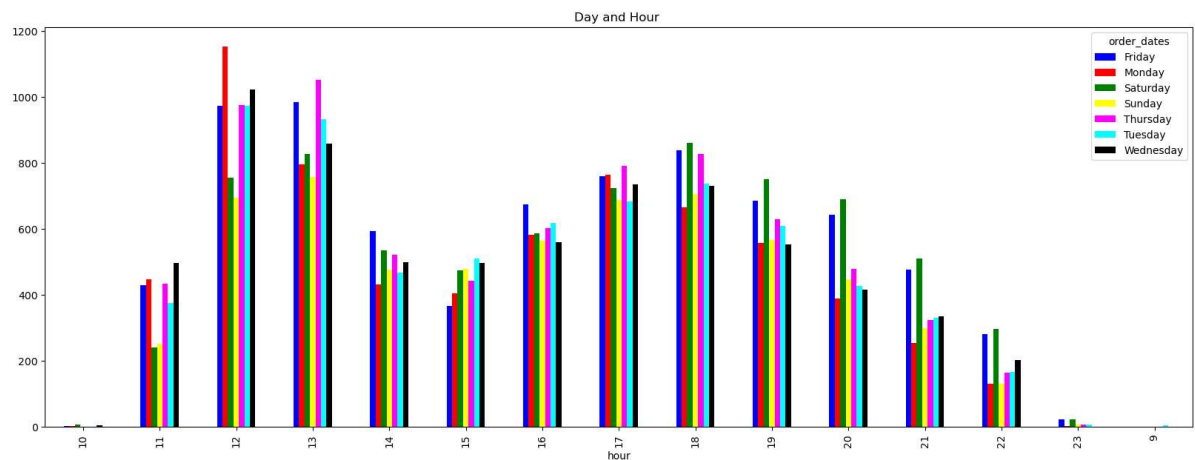
```
In [52]: g = sns.catplot(x="order_month", data=pizza, aspect=4.0, kind='count', hue='order_dates')
g.set_ylabels('Frequency')
g.set_xlabels('Months')
plt.show()
```



## Daily hour-wise sales of pizza

In [53]: `color=["blue", "red", "green", "yellow", "magenta", "cyan", "black", "orange"]`

```
In [54]: pd.crosstab(pizza["hour"], pizza["order_dates"]).plot(kind="bar", figsize=(20, 10),
color=["blue", "red", "green", "yellow", "magenta", "cyan", "black", "orange"],
title="Day and Hour ")
plt.show()
```



## Which pizza size earns more money

```
In [55]: pizzaEarn = pizza.groupby('pizza_size')['total_price'].sum()  
pizzaEarn
```

```
Out[55]: pizza_size  
L      375318.70  
M      249382.25  
S      178076.50  
XL      14076.00  
XXL      1006.60  
Name: total_price, dtype: float64
```

```
In [56]: profitable_Pizza = pizzaEarn.idxmax()  
print("The most profitable pizza size is:", profitable_Pizza)
```

The most profitable pizza size is: L

## Answers of the following question

1. What days and times do we tend to be busiest?
2. How many pizzas are we making during peak periods?
3. What are our best and worst-selling pizzas?
4. What's our average order value?
5. How well are we utilizing our seating capacity? (we have 15 tables and 60 seats)

## Ques 1. What days and times do we tend to be busiest?

Answer: The day in which we tend to be busiest is :FRIDAY and the time is between 12-13 PM

## Ques 2. How many pizzas are we making during peak periods?

```
In [57]: # Taking busiest times as Peak Periods  
peakPeriods=pizza["hour"].value_counts().head(2)  
total_pizzas_ordered=peakPeriods.values.sum()  
print(f'During peak periods, we are making {total_pizzas_ordered} pizzas in to
```

During peak periods, we are making 12746 pizzas in total.

**Answer During peak periods, we are making 37054 pizzas in total**

### Ques 3. What are our best and worst-selling pizzas?

```
In [58]: c = pizza.groupby('pizza_name')['quantity'].sum().reset_index()

best = c.loc[c['quantity'].idxmax()]['pizza_name']
worst = c.loc[c['quantity'].idxmin()]['pizza_name']
print(f'Our best-selling pizza is {best}, and our worst-selling pizza is {worst}')
```

Our best-selling pizza is The Classic Deluxe Pizza, and our worst-selling pizza is The Brie Carre Pizza.

**Answer Our best-selling pizza is The Classic Deluxe Pizza, and our worst-selling pizza is The Brie Carre Pizza.**

### Ques 4. What's our average order value?

```
In [59]: average_order_value = pizza['total_price'].sum() / pizza['order_id'].nunique()
print(f'Our average order value is ${average_order_value:.2f}.')
```

Our average order value is \$38.31.

**Answer Our average order value is \$38.31.**

### Ques 5. How well are we utilizing our seating capacity? (we have 15 tables and 60 seats)

```
In [60]: table=15
seats=60
```

```
In [61]: seats_per_table=seats/table
seats_per_table
```

Out[61]: 4.0

```
In [62]: z=pizza['hour'].value_counts()  
z
```

```
Out[62]: 12    6543  
        13    6203  
        18    5359  
        17    5143  
        19    4350  
        16    4185  
        14    3521  
        20    3487  
        15    3170  
        11    2672  
        21    2528  
        22    1370  
        23     68  
        10     17  
         9      4  
        Name: hour, dtype: Int64
```

```
In [63]: x=z.values/seats  
        util=(x.max()/seats)*100  
        util
```

```
Out[63]: 181.75
```

**Hence we can say that At Peak periods 109 people per seat are eating. So, At such a time seat Utilization is : 181.75%**