+ Code + Text

Importing library

import pandas as pd
import numpy as np
import seaborn as sns
import plotly.express as px
import matplotlib.pyplot as plt

Importing dataset file

df=pd.read_csv(r"/content/Data Model - Pizza Sales-CSV.csv")

Printing the details in the file

df

₹

pizz	pizza_category	pizza_size	total_price(INR)	total_price(USD)	unit_price	order_time	order_date	quantity	pizza_id
Мо	Classic	М	1086.5	13.25	13.25	11:38:36	1/1/2021	1	hawaiian_m
N	Classic	М	1312.0	16.00	16.00	11:57:40	1/1/2021	1	classic_dlx_m
Moz Pro	Veggie	L	1517.0	18.50	18.50	11:57:40	1/1/2021	1	five_cheese_l
Ca	Supreme	L	1701.5	20.75	20.75	11:57:40	1/1/2021	1	ital_supr_l
Pe Pe	Veggie	М	1312.0	16.00	16.00	11:57:40	1/1/2021	1	mexicana_m
	Chicken	М	1373.5	16.75	16.75	21:23:10	12/31/2021	1	ckn_alfredo_m
Gorg	Veggie	L	1471.9	17.95	17.95	21:23:10	12/31/2021	1	four_cheese_l
Aı	Classic	S	984.0	12.00	12.00	21:23:10	12/31/2021	1	napolitana_s
Pe Pe	Veggie	L	1660.5	20.25	20.25	22:09:54	12/31/2021	1	mexicana_I
Barb Red	Chicken	S	1045.5	12.75	12.75	23:02:05	12/31/2021	1	bbq_ckn_s

Here (rows, columns)

df.shape

→ (48620, 15)

Information about the file is given like datatype, null values etc....

```
df.info()
```

```
<pr
   RangeIndex: 48620 entries, 0 to 48619
   Data columns (total 15 columns):
    # Column
                   Non-Null Count Dtype
                        -----
    0
                        48620 non-null int64
       S.no
    1
        order_id
                        48620 non-null int64
                        48620 non-null object
       pizza_id
                        48620 non-null int64
    3
       quantity
    4
       order_date
                        48620 non-null object
       order_time
                        48620 non-null object
    6
                        48620 non-null float64
       unit_price
       total_price(USD) 48620 non-null float64
    7
       total_price(INR) 48620 non-null float64
                        48620 non-null object
       pizza_size
    10 pizza_category
                        48620 non-null object
    11 pizza_ingredients 48620 non-null object
    12 pizza_name
                        48620 non-null object
    13 Unnamed: 13
                        0 non-null
                                       float64
    14 Unnamed: 14
                        1 non-null
                                       float64
   dtypes: float64(5), int64(3), object(7)
   memory usage: 5.6+ MB
```

Here At What time, how many pizzas were ordered is given

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

```
→ count
```

```
hour
 10
         17
       2672
 11
       6543
 12
       6203
 13
 14
       3521
       3170
 15
       4185
 16
 17
       5143
 18
       5359
 19
       4350
 20
       3487
 21
       2528
 22
       1370
 23
         68
 9
          4
```

Here the Total amount of sales is happened in that year

```
totalsales=df["total_price(INR)"].sum()
print(totalsales , "rupees")
```

→ 67064524.1 rupees

Here analysing the data for each days.

```
df['order_day'] = pd.to_datetime(df['order_date'], errors='coerce')

df["order_day"] = df["order_day"].dt.day_name()

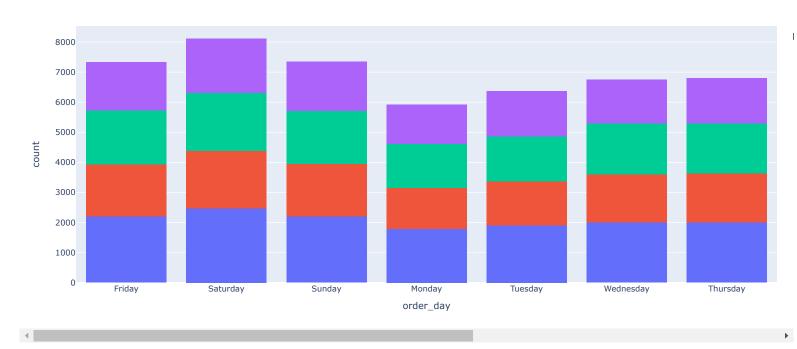
df["order_day"].value_counts().sort_index()
```

→		count
	order_day	
	Friday	7323
	Monday	5917
	Saturday	8106
	Sunday	7355
	Thursday	6797
	Tuesday	6369
	Wednesday	6753
	dr	

Here we have 4 categories of pizzas

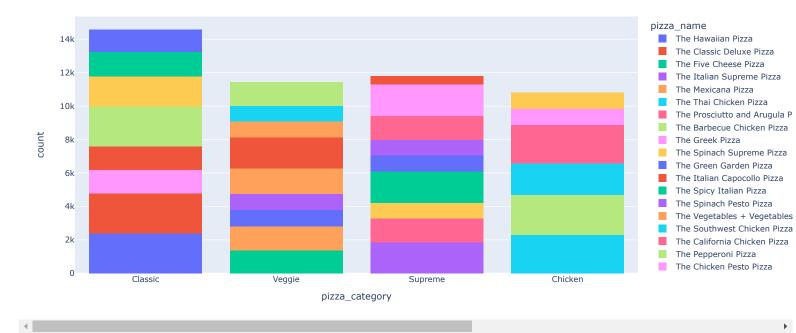
px.histogram(df,x="order_day",color='pizza_category')





Each of the categories there were some types

 $\verb|px.histogram(df,x="pizza_category",color="pizza_name")|\\$



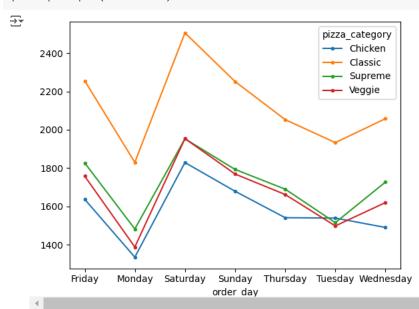
Top selling pizza's

top5 = df.groupby("pizza_name")["quantity"].count().sort_values(ascending=False).head(5)
top5

₹		quantity
	pizza_name	
	The Classic Deluxe Pizza	2416
	The Barbecue Chicken Pizza	2372
	The Hawaiian Pizza	2370
	The Pepperoni Pizza	2369
	The Thai Chicken Pizza	2315
	14	

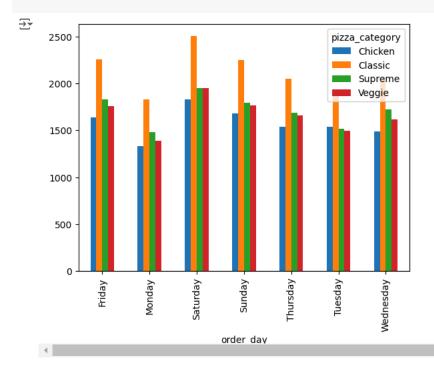
Order day VS quantity of pizza's

spread=df.groupby(["order_day","pizza_category"])["quantity"].sum().unstack()
spread=spread.plot(marker = ".")



Using bar graph

```
spread2 = df.groupby(["order_day", "pizza_category"])["quantity"].sum().unstack()
spread2 = spread2.plot.bar()
```



Least selling pizza's

```
lowest=df.loc[df['pizza_category'] == "Chicken"]
lowest=lowest.groupby("pizza_name")["quantity"].sum().sort_values(ascending=True)
lowest
```

-	→	$\overline{}$	
	÷	_	

rizza_name The Chicken Pesto Pizza 973 The Chicken Alfredo Pizza 987 The Southwest Chicken Pizza 1917 The California Chicken Pizza 2370 The Thai Chicken Pizza 2371 The Barbecue Chicken Pizza 2432

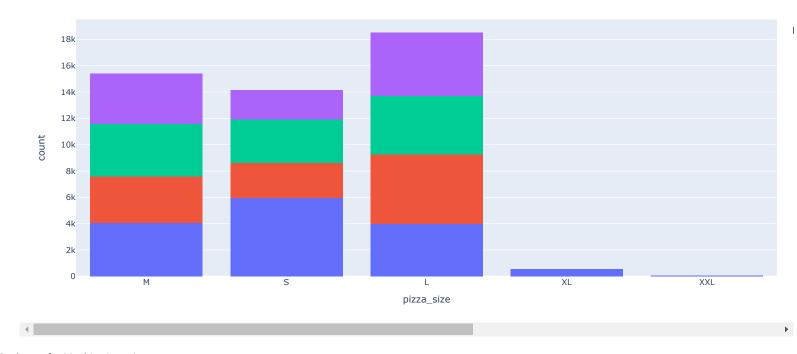
Size of the pizza

df.pizza_size.value_counts()

_		count
	pizza_size	
	L	18526
	M	15385
	S	14137
	XL	544
	XXL	28

px.histogram(df,x="pizza_size",color="pizza_category")





Packages for Machine Learning

```
import pandas as pd
import numpy as np
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import LabelEncoder
from sklearn.linear_model import LinearRegression
from sklearn.metrics import mean_squared_error, r2_score

df = pd.read_csv(r"/content/Data Model - Pizza Sales-CSV.csv")
```

Data preprocessing

```
df["order_day"] = pd.to_datetime(df['order_date'], errors='coerce')
df["order_day"] = df["order_day"].dt.day_name()
```

Convert categorical columns to numerical

```
label_encoder = LabelEncoder()
df['pizza_size'] = label_encoder.fit_transform(df['pizza_size'])
df['pizza_category'] = label_encoder.fit_transform(df['pizza_category'])
df['order_day'] = label_encoder.fit_transform(df['order_day'])
df['pizza_name'] = label_encoder.fit_transform(df['pizza_name'])
```

features and target

```
X = df[['pizza_size', 'pizza_category', 'order_day', 'pizza_name']]
y = df['total_price(INR)']
```

data into train and test sets

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

Model selection and training

```
model = LinearRegression()
model.fit(X_train, y_train)
     ▼ LinearRegression
     LinearRegression()
Make predictions
y_pred = model.predict(X_test)
Evaluate the model
mse = mean_squared_error(y_test, y_pred)
r2 = r2_score(y_test, y_pred)
MSE and R2 were printed
print(f"Mean Squared Error: {mse}")
print(f"R2 Score: {r2}")
→ Mean Squared Error: 83378.68771180959
     R2 Score: 0.3803807341108836
Evaluation based on MSE and R2
# Evaluation based on MSE and R2
if mse < 5000 and r2 > 0.8:
    print("The model is performing well with a low MSE and high R<sup>2</sup> score.")
    print("Recommendation: This model is suitable for making predictions.")
elif mse < 10000 and r2 > 0.6:
    print("The model is performing decently with a moderate MSE and acceptable R^2 score.")
    print("Recommendation: This model is acceptable, but there is room for improvement.")
    print("The model is performing poorly with a high MSE and/or low {\bf R^2} score.")
    print("Recommendation: Consider improving the model by trying other algorithms or tuning hyperparameters.")
The model is performing poorly with a high MSE and/or low R² score.
     Recommendation: Consider improving the model by trying other algorithms or tuning hyperparameters.
Prediction for new data
new_data = [[2, 1, 3, 5]] # Modify this based on actual values
predicted_sales = model.predict(new_data)
print(f"Predicted Sales: {predicted_sales[0]} INR")
```