```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sb
from sklearn.model selection import train test split
from sklearn.preprocessing import LabelEncoder, StandardScaler
from sklearn import metrics
from sklearn.svm import SVC
from xgboost import XGBRegressor
from sklearn.linear model import LinearRegression, Lasso, Ridge
from sklearn.ensemble import RandomForestRegressor
from sklearn.metrics import mean absolute error as mae
import warnings
warnings.filterwarnings('ignore')
df = pd.read csv('/content/train.csv.zip')
display(df.head())
display(df.tail())
df.shape
#column of the dataset contains which type of data.
df.info()
df.describe()
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sb
from sklearn.model selection import train test split
from sklearn.preprocessing import LabelEncoder, StandardScaler
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import warnings
warnings.filterwarnings('ignore')
df = pd.read csv('/content/train.csv.zip')
display(df.head())
```

```
display(df.tail())
df.shape
df.info()
df.describe()
parts = df["date"].str.split("-", n = 3, expand = True)
df["year"]= parts[0].astype('int')
df["month"]= parts[1].astype('int')
df["day"]= parts[2].astype('int')
df.head()
from datetime import datetime
import calendar
def weekend or weekday(year,month,day):
 d = datetime(year,month,day)
 if d.weekday()>4:
  return 1
 else:
  return 0
df['weekend'] = df.apply(lambda x:weekend or weekday(x['year'], x['month'],
x['day']), axis=1)
df.head()
plt.subplots(figsize=(12, 5))
plt.subplot(1, 2, 1)
sb.distplot(df['sales'])
plt.subplot(1, 2, 2)
sb.boxplot(df['sales'])
plt.show()
plt.figure(figsize=(10, 10))
sb.heatmap(df.corr() > 0.8,
  annot=True,
  cbar=False)
plt.show()
features = df.drop(['sales', 'year'], axis=1)
target = df['sales'].values
X train, X val, Y train, Y val = train test split(features, target,
               test size = 0.05,
               random state=22)
```

```
X train.shape, X val.shape
#for weekday
from datetime import date
import holidays
import pandas as pd
# Sample DataFrame with 'date', 'store', 'year', 'month', 'id', and 'item' columns
df = pd.DataFrame(
  'date': ['2024-01-01', '2024-02-15', '2024-04-14', '2024-04-15'],
  'store': [1, 2, 1, 2],
  'year': [2024, 2024, 2024, 2024],
  'month': [1, 2, 4, 4],
  'id': [101, 102, 103, 104],
  'item': ['A', 'B', 'A', 'B']
})
def is holiday(x):
  india holidays = holidays.CountryHoliday('IN')
  return 1 if x in india holidays else 0
df['date'] = pd.to datetime(df['date'])
df['day of week'] = df['date'].dt.day name()
df['is weekend'] = df['date'].dt.dayofweek // 5 # 1 for Saturday or Sunday, 0 for
weekdays
df['is holiday'] = df['date'].apply(is holiday)
print(df.head())
df['m1'] = np.sin(df['month'] * (2 * np.pi / 12))
df['m2'] = np.cos(df['month'] * (2 * np.pi / 12))
df.head()
from datetime import datetime
import holidays
import pandas as pd
# Assuming df is your DataFrame with 'year', 'month', 'day', and other columns
df = pd.DataFrame( {
  'year': [2024, 2024, 2024],
  'month': [1, 2, 4],
  'day': [1, 15, 14],
  'store': [1, 2, 1],
  'item': ['A', 'B', 'A'],
  'id': [101, 102, 103],
```

```
'm1': [10, 15, 20],
  'm2': [5, 8, 12]
})
def is holiday(x):
  india holidays = holidays.CountryHoliday('IN')
  return 1 if x in india holidays else 0
def which day(row):
  d = datetime(row['year'], row['month'], row['day'])
  return d.weekday()
df['date'] = pd.to datetime(df[['year', 'month', 'day']])
df['weekday'] = df.apply(which day, axis=1)
df['is weekend'] = df['date'].dt.dayofweek // 5 # 1 for Saturday or Sunday, 0 for
weekdays
df['is holiday'] = df['date'].apply(is holiday)
print(df.head())
from datetime import datetime
import holidays
import pandas as pd
# Assuming df is your DataFrame with 'year', 'month', 'day', and other columns
df = pd.DataFrame(
  'year': [2024, 2024, 2024],
  'month': [1, 2, 4],
  'day': [1, 15, 14],
  'store': [1, 2, 1],
  'item': ['A', 'B', 'A'],
  'id': [101, 102, 103],
  'm1': [10, 15, 20],
  'm2': [5, 8, 12]
})
def is holiday(x):
  india holidays = holidays.CountryHoliday('IN')
  return 1 if x in india holidays else 0
def which day(row):
  d = datetime(row['year'], row['month'], row['day'])
  return d.weekday()
```

```
df['date'] = pd.to datetime(df[['year', 'month', 'day']])
df['weekday'] = df.apply(which day, axis=1)
df['is weekend'] = df['date'].dt.dayofweek
df['is holiday'] = df['date'].apply(is holiday)
print(df.head())
from datetime import datetime
import holidays
import pandas as pd
df = pd.DataFrame( {
  'year': [2024, 2024, 2024],
  'month': [1, 2, 4],
  'day': [1, 15, 14],
  'store': [1, 2, 1],
  'item': ['A', 'B', 'A'],
  'id': [101, 102, 103],
  'm1': [10, 15, 20],
  'm2': [5, 8, 12]
})
def is holiday(x):
  india holidays = holidays.CountryHoliday('IN')
  return 1 if x in india holidays else 0
def which day(row):
  d = datetime(row['year'], row['month'], row['day'])
  return d.weekday()
df['date'] = pd.to datetime(df[['year', 'month', 'day']])
df['weekday'] = df.apply(which_day, axis=1)
df['is weekend'] = df['date'].dt.dayofweek
df['is holiday'] = df['date'].apply(is holiday)
print(df.head())
import matplotlib.pyplot as plt
import pandas as pd
# Assuming df is your DataFrame with necessary columns including 'sales'
df = pd.DataFrame(
  'store': [1, 2, 1, 2, 1, 2],
  'year': [2024, 2024, 2024, 2024, 2024, 2024],
  'month': [1, 2, 3, 4, 5, 6],
  'weekday': [0, 1, 2, 3, 4, 5], #Assuming 0=Monday, 1=Tuesday, ..., 5=Saturday
```

```
'weekend': [0, 0, 0, 0, 0, 1], #Assuming 0=Weekday, 1=Weekend
  'holidays': [0, 0, 1, 0, 0, 0],
  'sales': [100, 150, 120, 200, 180, 250]
})
features = ['store', 'year', 'month', 'weekday', 'weekend', 'holidays']
plt.subplots(figsize=(20, 10))
for i, col in enumerate(features):
  plt.subplot(2, 3, i + 1)
  df.groupby(col).mean()['sales'].plot.bar()
  plt.title(fMean Sales by {col}')
plt.tight layout()
plt.show()
import matplotlib.pyplot as plt
import pandas as pd
# Assuming df is your DataFrame with necessary columns including 'sales'
df = pd.DataFrame(
  'date': pd.date range(start='2024-01-01', periods=30),
  'sales': [100, 150, 120, 200, 180, 250, 130, 170, 190, 220, 110, 130, 180, 210,
240,
         150, 160, 130, 170, 200, 120, 140, 180, 210, 230, 120, 140, 160, 180,
2001.
  'weekday': pd.date range(start='2024-01-01', periods=30).day name()
})
plt.figure(figsize=(10, 5))
df.groupby('weekday').mean()['sales'].plot(marker='o', linestyle='-')
plt.title('Mean Sales by Day of the Week')
plt.xlabel('Day of the Week')
plt.ylabel('Mean Sales')
plt.grid(True)
plt.show()
import pandas as pd
from sklearn.model selection import train test split
from sklearn.preprocessing import StandardScaler
# Assuming df is your DataFrame with necessary columns including 'sales' and
'year'
df = pd.DataFrame( {
  'year': [2024, 2024, 2024, 2024, 2024, 2024],
```

```
'month': [1, 2, 3, 4, 5, 6],
  'weekday': [0, 1, 2, 3, 4, 5],
  'weekend': [0, 0, 0, 0, 0, 1],
  'holidays': [0, 0, 1, 0, 0, 0],
  'sales': [100, 150, 120, 200, 180, 250]
})
# Creating features DataFrame by dropping 'sales' and 'year' columns
features = df.drop(['sales', 'year'], axis=1)
# Creating target array using the 'sales' column
target = df['sales'].values
# Splitting the data into training and validation sets
X train, X val, y train, y val = train test split(features, target, test size=0.2,
random state=42)
# Normalizing the features using StandardScaler
scaler = StandardScaler()
X train scaled = scaler.fit transform(X train)
X val scaled = scaler.transform(X val)
print("Scaled Features (Training):")
print(X train scaled)
print("\nScaled Features (Validation):")
print(X val scaled)
import pandas as pd
from sklearn.preprocessing import StandardScaler
from sklearn.model selection import train test split
from sklearn.linear model import LinearRegression, Ridge
from sklearn.ensemble import RandomForestRegressor,
GradientBoostingRegressor, AdaBoostRegressor, BaggingRegressor
from sklearn.metrics import mean absolute error
# Assuming df is your DataFrame with necessary columns including 'sales' and
'year'
df = pd.DataFrame(
  'year': [2024, 2024, 2024, 2024, 2024, 2024],
  'month': [1, 2, 3, 4, 5, 6],
  'weekday': [0, 1, 2, 3, 4, 5],
  'weekend': [0, 0, 0, 0, 0, 1],
  'holidays': [0, 0, 1, 0, 0, 0],
  'sales': [100, 150, 120, 200, 180, 250]
```

```
})
# Creating features DataFrame by dropping 'sales' and 'year' columns
features = df.drop(['sales', 'year'], axis=1)
# Creating target array using the 'sales' column
target = df['sales'].values
# Splitting the data into training and validation sets
X train, X val, Y train, Y val = train test split(features, target, test size=0.2,
random state=42)
# Normalizing the features using StandardScaler
scaler = StandardScaler()
X train = scaler.fit transform(X train)
X \text{ val} = \text{scaler.transform}(X \text{ val})
# Defining models
models = [
  LinearRegression(),
  Ridge(),
  RandomForestRegressor(),
  GradientBoostingRegressor(),
  AdaBoostRegressor(),
  BaggingRegressor()
1
for model in models:
  model.fit(X train, Y train)
  print(f'Model: {model}')
  train preds = model.predict(X train)
  print('Training Error: ', mean absolute error(Y train, train preds))
  val preds = model.predict(X val)
  print('Validation Error: ', mean absolute error(Y val, val preds))
  print()
```

	date	store	item	sales
0	2013-01-01	1	1	13
1	2013-01-02	1	1	11
2	2013-01-03	1	1	14
3	2013-01-04	1	1	13
4	2013-01-05	1	1	10

	date	store	item	sales
912995	2017-12-27	10	50	63
912996	2017-12-28	10	50	59
912997	2017-12-29	10	50	74
912998	2017-12-30	10	50	62
912999	2017-12-31	10	50	82

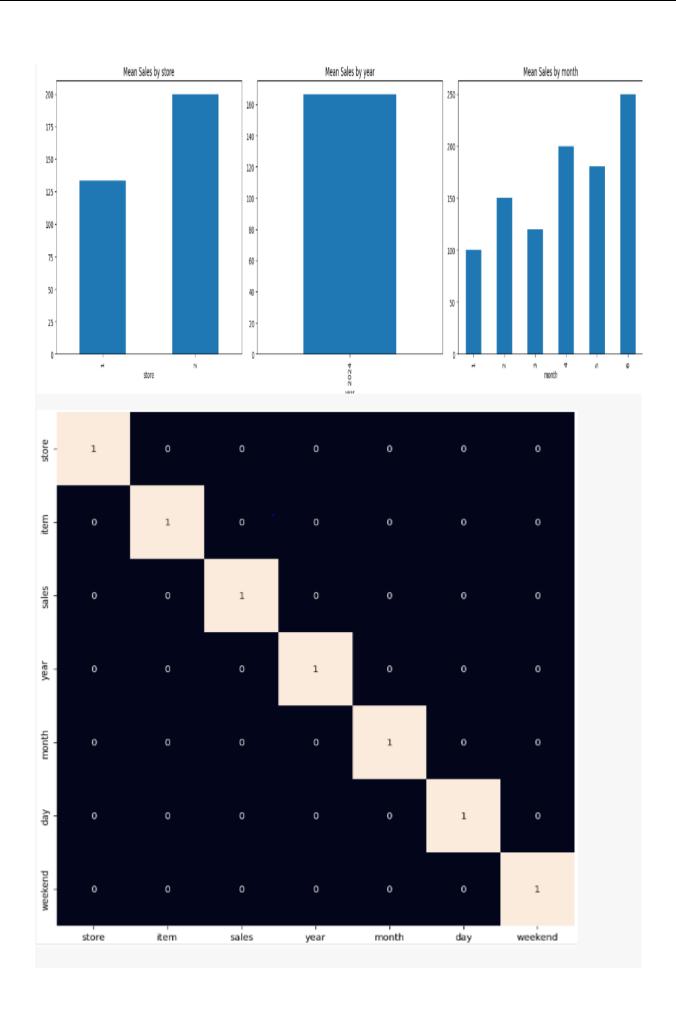
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 913000 entries, 0 to 912999

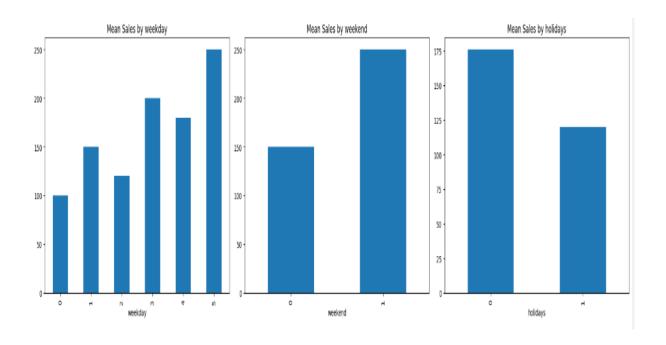
Data columns (total 4 columns):

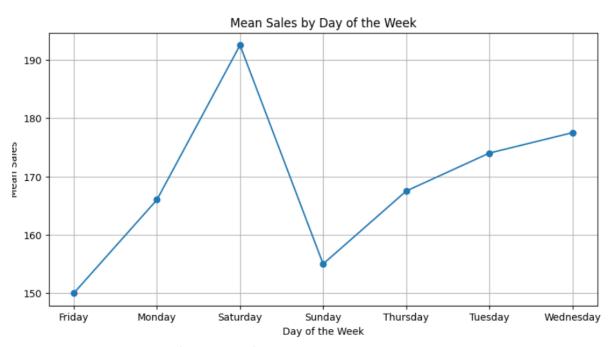
#	Column	Non-Null Count	Dtype
0	date	913000 non-null	object
1	store	913000 non-null	int64
2	item	913000 non-null	int64
3	sales	913000 non-null	int64

dtypes: int64(3), object(1)
memory usage: 27.9+ MB

	store	item	sales
count	913000.000000	913000.000000	913000.000000
mean	5.500000	25.500000	52.250287
std	2.872283	14.430878	28.801144
min	1.000000	1.000000	0.000000
25%	3.000000	13.000000	30.000000
50%	5.500000	25.500000	47.000000
75%	8.000000	38.000000	70.000000







```
Scaled Features (Training):
```

Scaled Features (Validation):

```
[[-3.13049517 -3.13049517 -0.57735027 -0.57735027]
```

[-2.23606798 -2.23606798 -0.57735027 -0.57735027]]

^[-1.34164079 -1.34164079 -0.57735027 1.73205081]

^[-0.4472136 -0.4472136 -0.57735027 -0.57735027]]

Model: LinearRegression()

Training Error: 1.4210854715202004e-14 Validation Error: 125.0000000000002

Model: Ridge()

Training Error: 11.621863799283148 Validation Error: 30.33968719452585

Model: GradientBoostingRegressor()
Training Error: 0.000996052458294372

Validation Error: 25.0

Model: AdaBoostRegressor()

Training Error: 0.0 Validation Error: 55.0

Model: BaggingRegressor()

Training Error: 14.25 Validation Error: 39.0