

```
import pandas as pd
import numpy as np
```

```
data = pd.read_csv("sales_data.csv")
print(data.head())
```

	Item_Identifier	Item_Weight	...	Outlet_Type	Item_Outlet_Sales
0	FDA15	9.30	...	Supermarket Type1	3735.1380
1	DRC01	5.92	...	Supermarket Type2	443.4228
2	FDN15	17.50	...	Supermarket Type1	2097.2700
3	FDX07	19.20	...	Grocery Store	732.3800
4	NCD19	8.93	...	Supermarket Type1	994.7052

```
[5 rows x 12 columns]
```

```
#drop irrelevant columns
data.drop(["Item_Identifier", "Outlet_Identifier"], axis=1, inplace=True)
```

```
#replace null values with median
data = data.fillna(data.median())
```

```
#Clean Item Fat Content Column
print(data['Item_Fat_Content'].unique())
data['Item_Fat_Content'] = data['Item_Fat_Content'].replace(['Low Fat', 'LF', 'low fat'], 0)
data['Item_Fat_Content'] = data['Item_Fat_Content'].replace(['Regular', 'reg'], 1)
print(data['Item_Fat_Content'].unique())
```

```
['Low Fat' 'Regular' 'low fat' 'LF' 'reg']
[0 1]
```

```
#Clean Item Type Column
print(data['Item_Type'].unique())
perishable = ["Breads", "Breakfast", "Dairy", "Fruits and Vegetables", "Meat", "Seafood"]
non_perishable = ["Baking Goods", "Canned", "Frozen Foods", "Hard Drinks", "Health and
                  "Soft Drinks", "Snack Foods", "Starchy Foods", "Others"]
data["Item_Type"] = data["Item_Type"].replace(to_replace=perishable, value="perishable")
data["Item_Type"] = data["Item_Type"].replace(to_replace=non_perishable, value="non_")
data["Item_Type"] = data["Item_Type"].replace('perishable', 0)
data["Item_Type"] = data["Item_Type"].replace('non_perishable', 1)
print(data['Item_Type'].unique())
```

```
['Dairy' 'Soft Drinks' 'Meat' 'Fruits and Vegetables' 'Household'
 'Baking Goods' 'Snack Foods' 'Frozen Foods' 'Breakfast'
 'Health and Hygiene' 'Hard Drinks' 'Canned' 'Breads' 'Starchy Foods'
 'Others' 'Seafood']
[0 1]
```

```
#Clean Outlet Size Column
```

```
print(data['Outlet_Size'].unique())
data["Outlet_Size"] = data["Outlet_Size"].replace('High', 3)
data["Outlet_Size"] = data["Outlet_Size"].replace('Medium', 2)
data["Outlet_Size"] = data["Outlet_Size"].replace('Small', 1)
data["Outlet_Size"] = data["Outlet_Size"].replace(np.nan, 0)
print(data['Outlet_Size'].unique())
```

```
['Medium' nan 'High' 'Small']
[2. 0. 3. 1.]
```

#Clean Outlet Location Column

```
print(data['Outlet_Location_Type'].unique())
data["Outlet_Location_Type"] = data["Outlet_Location_Type"].replace("Tier 1", 1)
data["Outlet_Location_Type"] = data["Outlet_Location_Type"].replace("Tier 2", 2)
data["Outlet_Location_Type"] = data["Outlet_Location_Type"].replace("Tier 3", 3)
print(data['Outlet_Location_Type'].unique())
```

```
['Tier 1' 'Tier 3' 'Tier 2']
[1 3 2]
```

#Clean Outlet Type Column

```
print(data['Outlet_Type'].unique())
data["Outlet_Type"] = data["Outlet_Type"].replace("Grocery Store", 0)
data["Outlet_Type"] = data["Outlet_Type"].replace("Supermarket Type1", 1)
data["Outlet_Type"] = data["Outlet_Type"].replace("Supermarket Type2", 1)
data["Outlet_Type"] = data["Outlet_Type"].replace("Supermarket Type3", 3)
print(data['Outlet_Type'].unique())
```

```
['Supermarket Type1' 'Supermarket Type2' 'Grocery Store'
 'Supermarket Type3']
[1 0 3]
```

#print cleaned data

```
print(data.head(10))
```

	Item_Weight	Item_Fat_Content	...	Outlet_Type	Item_Outlet_Sales
0	9.300	0 ...	1		3735.1380
1	5.920	1 ...	1		443.4228
2	17.500	0 ...	1		2097.2700
3	19.200	1 ...	0		732.3800
4	8.930	0 ...	1		994.7052
5	10.395	1 ...	1		556.6088
6	13.650	1 ...	1		343.5528
7	12.600	0 ...	3		4022.7636
8	16.200	1 ...	1		1076.5986
9	19.200	1 ...	1		4710.5350

```
[10 rows x 10 columns]
```

```
y = data['Item_Outlet_Sales']
x = data.drop(['Item_Outlet_Sales'], axis=1)
```

```
from sklearn.model_selection import train_test_split
x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.20, random_state=42)

from sklearn.linear_model import LinearRegression
linearRegression = LinearRegression()
linearRegression.fit(x_train, y_train)
prediction = linearRegression.predict(x_test)

final = pd.DataFrame(list(zip(y_test, prediction)), columns=['Actual', 'Predicted'])
print("Actual Sales VS Predicted Sales")
print(final.head(10))
```

Actual Sales VS Predicted Sales

	Actual	Predicted
0	1743.0644	961.028092
1	356.8688	662.454505
2	377.5086	949.707263
3	5778.4782	4456.067478
4	2356.9320	3224.061636
5	865.5400	747.585254
6	4613.9940	4851.021897
7	2410.8618	2457.356477
8	1948.1308	1772.296631
9	1937.4780	3181.449336

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