PROGRAM FOR THE PROBLEM STATEMENTS

```
import pandas as pd
# Load the dataset
file path = '/content/sample data/Super Store data.csv'
df = pd.read csv(file path,encoding='latin1')
# Display the first few rows of the dataset
print(df.head())
# Check for missing values
print(df.isnull().sum())
# Handle missing values (if any)
df = df.dropna()
# Convert categorical columns to numerical values using one-hot
encoding
df = pd.get dummies(df, drop first=True)
# Check the correlation with the target variable (assuming 'Sales' is the
target)
corr matrix = df.corr()
print(corr matrix['Sales'].sort values(ascending=False))
# Print the available columns after get dummies to see the correct
names
print(df.columns)
```

```
# Select features with a high correlation to 'Sales'
# Make sure to use the updated column names from the previous print
statement
features = df[['Row ID']] # Replace with actual feature names, 'Order
ID' and 'Customer ID' are no longer available
target = df['Sales']
# Split the data into training and testing sets
from sklearn.model selection import train test split
X train, X test, y train, y test = train test split(features, target,
test_size=0.2, random_state=42)
from sklearn.linear model import LinearRegression
from sklearn.metrics import mean squared error, r2 score
# Initialize and train the model
model = LinearRegression()
model.fit(X_train, y_train)
# Make predictions on the test set
y pred = model.predict(X test)
# Evaluate the model
mse = mean squared error(y test, y pred)
r2 = r2 score(y test, y pred)
print(f'Mean Squared Error: {mse}')
print(f'R-squared: {r2}')
```

```
# Example of making future predictions
new_data = pd.DataFrame({
    'Row ID':
[0,1,2,3,4,5,6,7,8,9,10,11,12,13,14,15,16,17,18,19,20,21,22,23,24,25] #
Replace with actual values
})

# Predict future sales
future_sales = model.predict(new_data)
print(f'Predicted future sales: {future_sales}')
```

OUTPUT FOR THE PROBLEM STATEMENTS

```
Row ID
            0
Order ID
            0
Order Date
           0
Ship Date
            0
Ship Mode
             0
Customer ID
Customer Name 0
Segment
             0
Country
            0
City
          0
          0
State
Postal Code
             0
```

```
Region
            0
Product ID
             0
Category
             0
Sub-Category
Product Name
Sales
           0
Quantity
             0
Discount
             0
Profit
           0
dtype: int64
Sales
         1.000000
Quantity 0.437464
Product Name HON 5400 Series Task Chairs for Big and Tall
                                                          0.291594
Product ID FUR-CH-10002024
                                               0.291594
Product ID FUR-BO-10004834
                                               0.267890
Product ID FUR-FU-10004270
                                -0.056008
Product Name Staple-based wall hangings -0.056054
State Illinois
                -0.059166
City Chicago
                -0.060084
Sub-Category Furnishings
                            -0.457835
Name: Sales, Length: 6209, dtype: float64
Index(['Row ID', 'Postal Code', 'Sales', 'Quantity', 'Discount', 'Profit',
   'Order ID CA-2014-100678', 'Order ID CA-2014-100706',
   'Order ID CA-2014-100916', 'Order ID CA-2014-101462',
   'Product Name Tensor Brushed Steel Torchiere Floor Lamp',
   'Product Name Tensor Computer Mounted Lamp',
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```
'Product Name_Tensor Track Tree Floor Lamp',
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'Product Name_Ultra Commercial Grade Dual Valve Door Closer',

'Product Name Ultra Door Kickplate, 8"H x 34"W',

'Product Name Ultra Door Pull Handle',

'Product Name Ultra Door Push Plate',

'Product Name_Westinghouse Clip-On Gooseneck Lamps',

'Product Name_Westinghouse Floor Lamp with Metal Mesh Shade, Black',

'Product Name_Westinghouse Mesh Shade Clip-On Gooseneck Lamp, Black'],

dtype='object', length=6209)

Mean Squared Error: 306563.5646297518

R-squared: -0.00046797502100925925

Predicted future sales: [355.89500454 355.89411216 355.89321978

355.89232739 355.89143501

355.89054262 355.88965024 355.88875785 355.88786547

355.88697309

355.8860807 355.88518832 355.88429593 355.88340355

355.88251116

355.88161878 355.88072639 355.87983401 355.87894163

355.87804924

355.87715686 355.87626447 355.87537209 355.8744797

355.87358732

355.87269493]