Step 1: Import Libraries

```
import pandas as pd
from sklearn.linear_model import LinearRegression
from sklearn.impute import SimpleImputer
from statsmodels.stats.outliers_influence import variance_inflation_factor
# Step 1: Import Libraries
print("Step 1: Importing Libraries")

    Step 1: Importing Libraries
```

Step 2: Load the Dataset

```
# Step 2: Load the Dataset
print("Step 2: Loading the Dataset")
data = pd.read_csv(r"NSSO68.csv")
data
```

	slno	grp	Round_Centre	FSU_number	Round
0	1	40999999999999992652495293775872.0	1	41000	68
1	2	4099999999999992652495293775872.0	1	41000	68
2	3	4099999999999992652495293775872.0	1	41000	68
3	4	4099999999999992652495293775872.0	1	41000	68
4	5	4099999999999992652495293775872.0	1	41000	68
101657	101658	7999999999999997087170359721984.0	1	79998	68
101658	101659	7999999999999997087170359721984.0	1	79998	68
101659	101660	7999999999999997087170359721984.0	1	79998	68
101660	101661	7999999999999997087170359721984.0	1	79998	68
101661	101662	7999999999999997087170359721984.0	1	79998	68

101662 rows × 384 columns

Step 3: Display Unique Values in a Column

Step 3: Display Unique Values in a Column

```
print("Step 3: Displaying Unique Values in 'state_1' Column")
print(data['state_1'].unique())
     Step 3: Displaying Unique Values in 'state_1' Column
['GUJ' 'ORI' 'CHTSD' 'MP' 'JRKD' 'WB' 'AP' 'MH' 'D&D' 'D&NH' 'MIZ' 'TRPR'
'MANPR' 'ASSM' 'MEG' 'NAG' 'A&N' 'PNDCRY' 'TN' 'GOA' 'KA' 'KE' 'LKSDP'
'SKM' 'Bhr' 'UP' 'RJ' 'ARP' 'DL' 'HR' 'Pun' 'HP' 'UT' 'Chandr' 'J$K']
Step 4: Subset Data and Select Columns
# Step 4: Subset Data and Select Columns
print("Step 4: Subsetting Data and Selecting Columns")
subset_data = data[data['state_1'] == 'AP'][['foodtotal_q', 'MPCE_URP', 'MPCE_URP', 'Age', 'Meals_At_Home', 'Possess_ration_card', 'Edu
print(subset_data)
Step 4: Subsetting Data and Selecting Columns
               foodtotal_q MPCE_MRP MPCE_URP Age Meals_At_Home
                                3818.86 3780.50 35
4100.08 4322.00 24
      6777
                 18.308732
                                                                           60.0
      6778
                 29.781670
                                                                           60.0
                                2333.55 1769.40 45
2284.85 1986.25 36
      6779
                 18.412530
                                                                           60.0
      6780
                 24.025527
                                                                           60.0
      6781
                 22.070518 1952.06 1224.20 36
                                                                           84.0
      79719
                 17.800203
                                 1304.09
                                             1344.50 26
```

60.0

60.0

90.0

0.0

Possess_ration_card Education No_of_Meals_per_day

2613.50

1820.50

1900.50

1348.00

50

32

18 14

2581.20

1374.35

1391.12

1536.62

79720

79721

79722

79723

47.825207

13.250154

21.725427

0.000000

```
7 0
     6777
                            1.0
     6778
                            2.0
                                      10.0
                                                            2.0
     6779
                            1.0
                                       8.0
                                                            2.0
     6780
                            1.0
                                      12.0
                                                            2.0
     6781
                            1.0
                                                            3.0
                                       1.0
     79719
                            1.0
                                      6.0
                                                            3.0
     79720
                            2.0
                                                            2.0
                                      8.0
     79721
                            2.0
                                      12.0
                                                            2.0
     79722
                            1.0
                                       1.0
                                                            3.0
     79723
                            2.0
                                      6.0
                                                            2.0
     [6899 rows x 8 columns]
Step 5: Check for Missing Values
# Step 5: Check for Missing Values
print("Step 5: Checking for Missing Values")
print(subset_data.isna().sum())

    Step 5: Checking for Missing Values

     foodtotal_q
     MPCE_MRP
     MPCE_URP
                              0
     Age
                              0
     Meals At Home
                            122
     Possess_ration_card
                             0
     Education
                              0
     No_of_Meals_per_day
                              0
     dtype: int64
Step 6: Impute Missing Values
# Step 6: Impute Missing Values
print("Step 6: Imputing Missing Values")
subset_data = subset_data.dropna()

→ Step 6: Imputing Missing Values

Step 7: Fit the Regression Model
# Step 7: Fit the Regression Model
print("Step 7: Fitting the Regression Model")
model = LinearRegression()
X = subset_data[['MPCE_MRP', 'MPCE_URP', 'Age', 'Meals_At_Home', 'Possess_ration_card', 'Education']]
y = subset_data['foodtotal_q']
model.fit(X, y)
     Step 7: Fitting the Regression Model
     ▼ LinearRegression
     LinearRegression()
Step 8: Print Regression Results
# Step 8: Print Regression Results
print("Step 8: Printing Regression Results")
print("Intercept:", model.intercept_)
print("Coefficients:", model.coef_)

    Step 8: Printing Regression Results

     Intercept: 7.6648416046440175
     Coefficients: [ 2.16935253e-03 6.59504317e-04 1.10990344e-01 9.25106872e-02
      -1.57629375e+00 2.56726961e-03]
Step 9: Check for Multicollinearity (VIF)
# Step 9: Check for Multicollinearity (VIF)
print("Step 9: Checking for Multicollinearity (VIF)")
vif_data = pd.DataFrame()
vif_data['feature'] = X.columns
vif_data['VIF'] = [variance_inflation_factor(X.values, i) for i in range(X.shape[1])]
print(vif_data)
Step 9: Checking for Multicollinearity (VIF)
                    feature
                             6.576028
                   MPCE MRP
```

```
1 MPCE_URP 4.770529
2 Age 9.723373
3 Meals_At_Home 10.504278
4 Possess_ration_card 9.006736
5 Education 3.836287
```

Step 10: Construct and Print the Regression Equation

```
# Step 10: Construct and Print the Regression Equation
print("Step 10: Constructing and Printing the Regression Equation")
equation = f"y = {model.intercept_:.2f}"
for i, coef in enumerate(model.coef_):
    equation += f" + {coef:.6f} * x{i+1}"
print(equation)

→ Step 10: Constructing and Printing the Regression Equation

     y = 7.66 + 0.002169 * x1 + 0.000660 * x2 + 0.110990 * x3 + 0.092511 * x4 + -1.576294 * x5 + 0.002567 * x6
Step 11: Display Head of Selected Columns
# Step 11: Display Head of Selected Columns
print("Step 11: Displaying Head of Selected Columns")
print(subset_data[['MPCE_MRP', 'MPCE_URP', 'Age', 'Meals_At_Home', 'Possess_ration_card', 'Education', 'foodtotal_q']].head(1))
Step 11: Displaying Head of Selected Columns

MPCE_MRP MPCE_URP Age Meals_At_Home Possess_ration_card Education \
6777 3818.86 3780.5 35 60.0 1.0 7.0
            \verb|foodtotal_q|
     6777
           18.308732
```

Start coding or generate with AI.