



VIRGINIA COMMONWEALTH UNIVERSITY

Statistical analysis and modelling (SCMA 632)

A1a: Preliminary preparation and analysis of data- Descriptive statistics

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INTRODUCTION

This project leverages multi-linear regression analysis. We'll construct a model to predict total food consumption based on factors like how often meals are eaten at home, possession of a government ration card, age, and economic indicators potentially reflecting food prices. To ensure the model's credibility, we'll meticulously evaluate its assumptions through diagnostics. This iterative process will culminate in a more robust understanding of the significant differences observed, ultimately unveiling the key socio-economic factors shaping dietary patterns across the population of Odisha.

OBJECTIVES

Perform Multiple regression analysis, carry out the regression diagnostics, and explain your findings. Correct them and revisit your results and explain the significant differences you observe. [data "NSSO68.csv"]Extract data specific to R.

- . Finding the missing values and assigning it with mean
- . Conducting Multiple Regression

.

RESULTS AND INTERPRETATION

USING R

```
install.packages("dplyr")
library(dplyr)
install.packages("tidyverse")
library(tidyverse)
install.packages("car")
library(car)
install.packages("lmtest")
library(lmtest)
setwd("C:\\\\Users\\\\HP\\\\Documents\\\\ns")
getwd()
data = read.csv("NSSO68 new.csv")
str(data)
wb_data <- data %>%
  filter(state_1 == "WB")
relevant_columns <- c("foodtotal_q", "Meals_At_Home", "Possess_ration_card",
"Age", "MPCE_URP", "MPCE_MRP")
westbengal_data <- westbengal_data %>%
  select(all_of(relevant_columns)) %>%
  print(westbengal_data)
str(westbengal_data)
sum(is.na(westbengal_data$Meals_At_Home))
sum(is.na(westbengal_data$Possess_ration_card))
sum(is.na(westbengal_data$Age))
sum(is.na(westbengal_data$MPCE_URP))
sum(is.na(westbengal_data$MPCE_MRP))
complete_rows <- complete.cases(westbengal_data$foodtotal_q,
westbengal_data$Possess_ration_card)
```

```

filtered_data <- westbengal_data[complete_rows, ]
model <- lm(foodtotal_q ~ Possess_ration_card, data = filtered_data)

input_with_mean <- function(data,column) {
  data %>%
    mutate(across(all_of(columns), ~ ifelse(is.na(.), mean(., na.rm = TRUE), .)))
}

sum(is.na(data$foodtotal_q))
cleaned_data <- na.omit(data)
odisha_data <- cleaned_data %>%
  filter(state_1 == "WB")
nrow(westbengal_data)
model <- lm(foodtotal_q ~ Meals_At_Home + Possess_ration_card + Age +
MPCE_URP + MPCE_MRP, data = data)
summary(model)

```

RESULT

Residual standard error: 8.747 on 100424 degrees of freedom

(1232 observations deleted due to missingness)

Multiple R-squared: 0.1853, Adjusted R-squared: 0.1852

F-statistic: 4567 on 5 and 100424 DF, p-value: < 2.2e-16

CODE USING PYTHON

```

import pandas as pd
import numpy as np
import statsmodels.api as sm

```

```
data = pd.read_csv("/content/NSSO68 new.csv")
```

```
print(data.info())
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 8538 entries, 0 to 8537
Columns: 384 entries, slno to fv tot
dtypes: float64(303), int64(80), object(1)
memory usage: 25.0+ MB
None
```

```
westbengal_data = data[data['state_1'] == "WB"].copy()
```

```
relevant_columns = ["foodtotal_q", "Meals_At_Home",
"Possess_ration_card", "Age", "MPCE_URP", "MPCE_MRP"]
westbengal_data = westbengal_data[relevant_columns].copy()
```

```
print(westbengal_data.head())
```

	foodtotal_q	Meals_At_Home	Possess_ration_card	Age	MPCE_URP	MPCE_MRP
741	33.110413		60.0	2.0	31	3455.503844.95
742	31.683645		60.0	1.0	42	2572.672377.28
743	25.575244		60.0	1.0	53	1792.752039.86
744	24.920166		60.0	1.0	60	880.00970.04
745	24.742780		90.0	1.0	35	854.50935.56

```
print(westbengal_data.isnull().sum())
```

```
foodtotal_q      0
Meals_At_Home    40
Possess_ration_card  0
Age              0
MPCE_URP         0
MPCE_MRP         0
dtype: int64
```

```
cleaned_data = westbengal_data.dropna()
```

```
print(cleaned_data.shape[0])
```

1013

```
X = cleaned_data[["Meals_At_Home", "Possess_ration_card", "Age",  
"MPCE_URP", "MPCE_MRP"]]  
y = cleaned_data["foodtotal_q"]
```

```
X = sm.add_constant(X)
```

```
model = sm.OLS(y, X).fit()
```

```
print(model.summary())
```

```

OLS Regression Results
=====
Dep. Variable:          foodtotal_q    R-squared:
0.334
Model:                  OLS          Adj. R-squared:
0.331
Method:                 Least Squares    F-statistic:
101.0
Date:                   Sun, 23 Jun 2024    Prob (F-statistic):
2.10e-86
Time:                   10:07:29    Log-Likelihood:
-3404.2
No. Observations:      1013    AIC:
6820.
Df Residuals:          1007    BIC:
6850.
Df Model:               5
Covariance Type:       nonrobust
=====
=====

```

	coef	std err	t	P> t
[0.025	0.975]			
const	4.6109	1.497	3.081	0.002
1.674	7.548			
Meals_At_Home	0.1545	0.012	12.774	0.000
0.131	0.178			
Possess_ration card	-0.0047	0.473	-0.010	0.992
0.933	0.924			
Age	0.0752	0.018	4.274	0.000
0.041	0.110			

```

MPCE URP                0.0011    0.000    5.160    0.000
0.001    0.001
MPCE MRP                0.0017    0.000    8.449    0.000
0.001    0.002
=====
=====
Omnibus:                211.654    Durbin-Watson:
1.512
Prob(Omnibus):          0.000    Jarque-Bera (JB):
1935.790
Skew:                   0.679    Prob(JB):
0.00
Kurtosis:               9.635    Cond. No.
2.51e+04
=====
=====

```

```

Notes:
[1] Standard Errors assume that the covariance matrix of the errors is
correctly specified.
[2] The condition number is large, 2.51e+04. This might indicate that
there are

strong multicollinearity or other numerical problems.

```

FINDING

The code analyzes data on food expenditure in Odisha, India. It finds a moderate correlation between food expenditure and factors like frequency of meals at home, age, and monthly expenditure. Interestingly, having a ration card doesn't seem to significantly affect food expenditure in this dataset.