

A1-R.R

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2025-06-12

```
# --- Project Setup and Package Management ---

# Set the base directory for the project
BASE <- "C:\\Users\\ujjwa\\Documents\\VCU\\Pre-Course\\SCMA632\\Assignments\\A1\\R"
setwd(BASE) # Set the working directory to the specified base path
getwd() # Verify the current working directory

## [1] "C:/Users/ujjwa/Documents/VCU/Pre-Course/SCMA632/Assignments/A1/R"

# Define a function to install packages if they are not already installed
install <- function(pkg) {
  if (!require(pkg, character.only = T)) { # Check if the package is loaded
    # Install with dependencies if not loaded
    install.packages(pkg, dependencies = T, quiet = T, verbose = F)
  }
}

# Define a function to load packages
load <- function(pkg) {
  library(pkg, character.only = T, quietly = T, verbose = F) # Load the specified package
}

# List of required packages for data manipulation, visualization, and statistical tests
pkgs <- c("dplyr", "readr", "readxl", "tidyr", "ggplot2", "BSDA")
lapply(pkgs, install) # Apply the install function to all packages

## Loading required package: dplyr

##
## Attaching package: 'dplyr'

## The following objects are masked from 'package:stats':
##
##   filter, lag

## The following objects are masked from 'package:base':
##
##   intersect, setdiff, setequal, union

## Loading required package: readr
```

```
## Loading required package: readxl

## Loading required package: tidyr

## Loading required package: ggplot2

## Loading required package: BSDA

## Loading required package: lattice

##
## Attaching package: 'BSDA'

## The following object is masked from 'package:datasets':
##
##      Orange

## [[1]]
## NULL
##
## [[2]]
## NULL
##
## [[3]]
## NULL
##
## [[4]]
## NULL
##
## [[5]]
## NULL
##
## [[6]]
## NULL
```

```
lapply(pkgs, load) # Apply the load function to all packages
```

```
## [[1]]
## [1] "BSDA"      "lattice"   "ggplot2"   "tidyr"     "readxl"    "readr"
## [7] "dplyr"     "stats"     "graphics"  "grDevices" "utils"     "datasets"
## [13] "methods"   "base"
##
## [[2]]
## [1] "BSDA"      "lattice"   "ggplot2"   "tidyr"     "readxl"    "readr"
## [7] "dplyr"     "stats"     "graphics"  "grDevices" "utils"     "datasets"
## [13] "methods"   "base"
##
## [[3]]
## [1] "BSDA"      "lattice"   "ggplot2"   "tidyr"     "readxl"    "readr"
## [7] "dplyr"     "stats"     "graphics"  "grDevices" "utils"     "datasets"
## [13] "methods"   "base"
```

```
##
## [[4]]
## [1] "BSDA"      "lattice"    "ggplot2"    "tidyr"      "readxl"     "readr"
## [7] "dplyr"     "stats"      "graphics"   "grDevices"  "utils"      "datasets"
## [13] "methods"   "base"
##
## [[5]]
## [1] "BSDA"      "lattice"    "ggplot2"    "tidyr"      "readxl"     "readr"
## [7] "dplyr"     "stats"      "graphics"   "grDevices"  "utils"      "datasets"
## [13] "methods"   "base"
##
## [[6]]
## [1] "BSDA"      "lattice"    "ggplot2"    "tidyr"      "readxl"     "readr"
## [7] "dplyr"     "stats"      "graphics"   "grDevices"  "utils"      "datasets"
## [13] "methods"   "base"
```

```
# --- Data Loading and Initial Filtering ---
```

```
# Load the main dataset from a CSV file
data <- read.csv('./datasets/NSS068.csv')
# Filter the data to include only records for Meghalaya state (state code "17")
data_meghalaya <- data %>% filter(state == "17")

# Remove the original large dataset to free up memory
rm(data)

# Print the column names of the filtered Meghalaya dataset
print(names(data_meghalaya))
```

```
## [1] "sln"      "grp"
## [3] "Round_Centre" "FSU_number"
## [5] "Round"      "Schedule_Number"
## [7] "Sample"     "Sector"
## [9] "state"      "State_Region"
## [11] "District"   "Stratum_Number"
## [13] "Sub_Stratum" "Schedule_type"
## [15] "Sub_Round"  "Sub_Sample"
## [17] "FOD_Sub_Region" "Hamlet_Group_Sub_Block"
## [19] "t"          "X_Stage_Stratum"
## [21] "HHS_No"     "Level"
## [23] "Filler"     "hhdsz"
## [25] "NIC_2008"   "NCO_2004"
## [27] "HH_type"    "Religion"
## [29] "Social_Group" "Whether_owns_any_land"
## [31] "Type_of_land_owned" "Land_Owned"
## [33] "Land_Leased_in" "Otherwise_posessed"
## [35] "Land_Leased_out" "Land_Total_posessed"
## [37] "During_July_June_Cultivated" "During_July_June_Irrigated"
## [39] "NSS"        "NSC"
## [41] "MLT"        "land_tt"
## [43] "Cooking_code" "Lighting_code"
## [45] "Dwelling_unit_code" "Regular_salary_earner"
## [47] "Perform_Ceremony" "Meals_seved_to_non_hhld_members"
## [49] "Possess_ration_card" "Type_of_ration_card"
```

## [51]	"MPCE_URP"	"MPCE_MRP"
## [53]	"Person_Srl_No"	"Relation"
## [55]	"Sex"	"Age"
## [57]	"Marital_Status"	"Education"
## [59]	"Days_Stayed_away"	"No_of_Meals_per_day"
## [61]	"Meals_School"	"Meals_Employer"
## [63]	"Meals_Others"	"Meals_Payment"
## [65]	"Meals_At_Home"	"Item_Code"
## [67]	"Source_Code"	"ricepds_q"
## [69]	"riceos_q"	"ricetotal_q"
## [71]	"chira_q"	"khoi_q"
## [73]	"muri_q"	"ricepro_q"
## [75]	"riceGT_q"	"Wheatpds_q"
## [77]	"wheatos_q"	"wheattotal_q"
## [79]	"maida_q"	"suji_q"
## [81]	"sewai_q"	"bread_q"
## [83]	"wheatp_q"	"wheatGT_q"
## [85]	"jowarp_q"	"bajrap_q"
## [87]	"maizep_q"	"barleyp_q"
## [89]	"milletp_q"	"ragip_q"
## [91]	"cerealot_q"	"cerealtot_q"
## [93]	"cerealsub_q"	"cerealstt_q"
## [95]	"arhar_q"	"gramdal_q"
## [97]	"gramwholep_q"	"gramGT_q"
## [99]	"moong_q"	"masur_q"
## [101]	"urd_q"	"peasdal_q"
## [103]	"khesari_q"	"otpulse_q"
## [105]	"gramp_q"	"besan_q"
## [107]	"pulsep_q"	"pulsestot_q"
## [109]	"pulsestt_q"	"soyabean_q"
## [111]	"milk_q"	"babyfood_q"
## [113]	"milkcond_q"	"curd_q"
## [115]	"ghee_q"	"butter_q"
## [117]	"icecream_q"	"otmilkp_q"
## [119]	"Milktotal_q"	"milkprott_q"
## [121]	"vanas_q"	"musoil_q"
## [123]	"gnoil_q"	"cocoil_q"
## [125]	"edioilothr_q"	"edibletotal_q"
## [127]	"ediblest_q"	"eggsno_q"
## [129]	"fishprawn_q"	"goatmeat_q"
## [131]	"beef_q"	"pork_q"
## [133]	"chicken_q"	"othrbirds_q"
## [135]	"nonvegtotal_q"	"emftt_q"
## [137]	"potato_q"	"onion_q"
## [139]	"tamato_q"	"brinjal_q"
## [141]	"radish_q"	"carrot_q"
## [143]	"palak_q"	"chillig_q"
## [145]	"bhindi_q"	"parwal_q"
## [147]	"cauli_q"	"cabbage_q"
## [149]	"pumpkin_q"	"peas_q"
## [151]	"fbeans_q"	"lemonno_q"
## [153]	"otveg_q"	"veggtt_q"
## [155]	"bananano_q"	"jackfruit_q"
## [157]	"watermel_q"	"pineaplno_q"

## [159]	"cocono_q"	"cocogno_q"
## [161]	"guava_q"	"sighara_q"
## [163]	"orangen_q"	"papayar_q"
## [165]	"mango_q"	"kharbooz_q"
## [167]	"pears_q"	"berries_q"
## [169]	"leechi_q"	"apple_q"
## [171]	"grapes_q"	"otfruits_q"
## [173]	"fruitstt_q"	"fruitt_total"
## [175]	"cocodf_q"	"gnutdf_q"
## [177]	"datesdf_q"	"cashewdf_q"
## [179]	"walnutdf_q"	"otnutsdf_q"
## [181]	"kishmish_q"	"otherdf_q"
## [183]	"dryfruitstotal_q"	"dftt_q"
## [185]	"sugarpds_q"	"sugaros_q"
## [187]	"sugarst_q"	"gur_q"
## [189]	"misri_q"	"honey_q"
## [191]	"sugarttotal_q"	"sugartt_q"
## [193]	"salt_q"	"ginger_q"
## [195]	"garlic_q"	"jeera_q"
## [197]	"dhanian_q"	"turnmeric_q"
## [199]	"blackpepper_q"	"drychilly_q"
## [201]	"tamarind_q"	"currypowder_q"
## [203]	"oilseeds_q"	"spicesothr_q"
## [205]	"spicetot_q"	"spicestotal_q"
## [207]	"teacupno_q"	"tealeaf_q"
## [209]	"teatotal_q"	"cofeeno_q"
## [211]	"coffeepwdr_q"	"cofeetotal_q"
## [213]	"ice_q"	"coldbvrg_q"
## [215]	"juice_q"	"othrbvrg_q"
## [217]	"bevergest_q"	"Biscuits_q"
## [219]	"preparedsweet_q"	"pickle_q"
## [221]	"sauce_jam_q"	"Othrprocessed_q"
## [223]	"Beveragestotal_q"	"ricepds_v"
## [225]	"riceos_v"	"ricetotal_v"
## [227]	"chira_v"	"khai_v"
## [229]	"muri_v"	"ricepro_v"
## [231]	"riceGT_v"	"Wheatpds_v"
## [233]	"wheatos_v"	"wheatttotal_v"
## [235]	"maida_v"	"suji_v"
## [237]	"sewai_v"	"bread_v"
## [239]	"wheatp_v"	"wheatGT_v"
## [241]	"jowarp_v"	"bajrap_v"
## [243]	"maizep_v"	"barleyp_v"
## [245]	"milletp_v"	"ragip_v"
## [247]	"cerealot_v"	"cerealtot_v"
## [249]	"cerealsub_v"	"cerealstt_v"
## [251]	"arhar_v"	"gramdal_v"
## [253]	"gramwholep_v"	"gramGT_v"
## [255]	"moong_v"	"masur_v"
## [257]	"urd_v"	"peasdal_v"
## [259]	"khesari_v"	"otpulse_v"
## [261]	"gramp_v"	"besan_v"
## [263]	"pulsep_v"	"pulsestot_v"
## [265]	"pulsestt_v"	"soyabean_v"

## [267]	"milk_v"	"babyfood_v"
## [269]	"milkcond_v"	"curd_v"
## [271]	"ghee_v"	"butter_v"
## [273]	"icecream_v"	"otmilkp_v"
## [275]	"Milktotal_v"	"milkprott_v"
## [277]	"vanas_v"	"musoil_v"
## [279]	"gnoil_v"	"cocoil_v"
## [281]	"edioilothr_v"	"edibletotal_v"
## [283]	"ediblest_v"	"eggsno_v"
## [285]	"fishprawn_v"	"goatmeat_v"
## [287]	"beef_v"	"pork_v"
## [289]	"chicken_v"	"othrbirds_v"
## [291]	"nonvegtotal_v"	"emftt_v"
## [293]	"potato_v"	"onion_v"
## [295]	"tamato_v"	"brinjal_v"
## [297]	"radish_v"	"carrot_v"
## [299]	"palak_v"	"chillig_v"
## [301]	"bhindi_v"	"parwal_v"
## [303]	"cauli_v"	"cabbage_v"
## [305]	"pumpkin_v"	"peas_v"
## [307]	"fbeans_v"	"lemonno_v"
## [309]	"otveg_v"	"veggtt_v"
## [311]	"bananano_v"	"jackfruit_v"
## [313]	"watermel_v"	"pineaplno_v"
## [315]	"cocono_v"	"cocogno_v"
## [317]	"guava_v"	"sighara_v"
## [319]	"orangenno_v"	"papayar_v"
## [321]	"mango_v"	"kharbooz_v"
## [323]	"pears_v"	"berries_v"
## [325]	"leechi_v"	"apple_v"
## [327]	"grapes_v"	"otfruits_v"
## [329]	"fruitstt_v"	"cocodf_v"
## [331]	"gnutdf_v"	"datesdf_v"
## [333]	"cashewdf_v"	"walnutdf_v"
## [335]	"otnutsdf_v"	"kishmish_v"
## [337]	"otherdf_v"	"dryfruitsttotal_v"
## [339]	"dftt_v"	"sugarpds_v"
## [341]	"sugaros_v"	"sugarst_v"
## [343]	"gur_v"	"misri_v"
## [345]	"honey_v"	"sugartotal_v"
## [347]	"sugartt_v"	"salt_v"
## [349]	"ginger_v"	"garlic_v"
## [351]	"jeera_v"	"dhanian_v"
## [353]	"turnmeric_v"	"blackpepper_v"
## [355]	"drychilly_v"	"tamarind_v"
## [357]	"currypowder_v"	"oilseeds_v"
## [359]	"spicesothr_v"	"spicetot_v"
## [361]	"spicestotal_v"	"teacupno_v"
## [363]	"tealeaf_v"	"teatotal_v"
## [365]	"cofeeno_v"	"coffeepwdr_v"
## [367]	"cofeetotal_v"	"ice_v"
## [369]	"coldbvrg_v"	"juice_v"
## [371]	"othrbvrg_v"	"bevergest_v"
## [373]	"Biscuits_v"	"preparedsweet_v"

```
## [375] "pickle_v" "sauce_jam_v"
## [377] "Othrprocessed_v" "Beveragestotal_v"
## [379] "foodtotal_v" "foodtotal_q"
## [381] "state_1" "Region"
## [383] "fruits_df_tt_v" "fv_tot"
```

```
# Print the first few rows of the Meghalaya dataset for a quick preview
print(head(data_meghalaya))
```

```
##      slno      grp Round_Centre FSU_number Round Schedule_Number Sample Sector
## 1 17686 4.46E+31      1      44610      68      10      1      2
## 2 17687 4.46E+31      1      44610      68      10      1      2
## 3 17688 4.46E+31      1      44610      68      10      1      2
## 4 17689 4.46E+31      1      44610      68      10      1      2
## 5 17690 4.46E+31      1      44610      68      10      1      2
## 6 17691 4.46E+31      1      44610      68      10      1      2
##      state State_Region District Stratum_Number Sub_Stratum Schedule_type
## 1      17      171      6      6      6      1
## 2      17      171      6      6      6      1
## 3      17      171      6      6      6      1
## 4      17      171      6      6      6      1
## 5      17      171      6      6      6      1
## 6      17      171      6      6      6      1
##      Sub_Round Sub_Sample FOD_Sub_Region Hamlet_Group_Sub_Block      t
## 1      3      2      1710      1 1.01e+13
## 2      3      2      1710      1 1.02e+13
## 3      3      2      1710      1 2.01e+13
## 4      3      2      1710      1 2.02e+13
## 5      3      2      1710      1 2.03e+13
## 6      3      2      1710      1 2.04e+13
##      X_Stage_Stratum HHS_No Level Filler hhdsz NIC_2008 NCO_2004 HH_type Religion
## 1      1      1      5      0      3      84119      419      2      3
## 2      1      2      5      0      5      32111      121      1      2
## 3      2      1      5      0      4      47221      341      2      3
## 4      2      2      5      0      3      47110      121      1      1
## 5      2      3      5      0      3      47594      121      1      2
## 6      2      4      5      0      6      85212      232      3      1
##      Social_Group Whether_owns_any_land Type_of_land_owned Land_Owned
## 1      1      1      1      1      4
## 2      9      2      2      NA      NA
## 3      1      2      2      NA      NA
## 4      9      2      2      NA      NA
## 5      9      2      2      NA      NA
## 6      9      2      2      NA      NA
##      Land_Leased_in Otherwise_posessed Land_Leased_out Land_Total_posessed
## 1      NA      NA      NA      4
## 2      NA      NA      NA      NA
## 3      3      NA      NA      3
## 4      3      NA      NA      3
## 5      3      NA      NA      3
## 6      4      NA      NA      4
##      During_July_June_Cultivated During_July_June_Irrigated NSS NSC      MLT land_tt
## 1      NA      NA      2      4 121200      0.04
## 2      NA      NA      2      4 121200      NA
```

## 3		NA		NA	2	4	44188	0.03
## 4		NA		NA	2	4	44188	0.03
## 5		NA		NA	2	4	44188	0.03
## 6		NA		NA	2	4	44188	0.04
##	Cooking_code	Lighting_code	Dwelling_unit_code	Regular_salary_earner				
## 1	3	5	1				1	
## 2	3	5	2				2	
## 3	3	5	2				1	
## 4	3	5	2				2	
## 5	3	5	2				2	
## 6	3	5	2				2	
##	Perform_Ceremony	Meals_seved_to_non_hhld_members	Possess_ration_card					
## 1	2		0				1	
## 2	2		NA				1	
## 3	2		0				1	
## 4	2		0				1	
## 5	2		0				1	
## 6	2		0				1	
##	Type_of_ration_card	MPCE_URP	MPCE_MRP	Person_Srl_No	Relation	Sex	Age	
## 1	3	3512.33	3186.12	1	1	2	47	
## 2	3	4178.00	3938.99	1	1	1	56	
## 3	3	2426.00	2370.18	1	1	1	70	
## 4	3	2095.00	2203.08	1	1	1	58	
## 5	3	2260.67	2372.08	1	1	1	42	
## 6	3	2146.50	2024.83	1	1	2	45	
##	Marital_Status	Education	Days_Stayed_away	No_of_Meals_per_day	Meals_School			
## 1	3	10	0		2		0	
## 2	2	10	0		2		0	
## 3	2	5	0		2		0	
## 4	2	10	0		2		0	
## 5	2	12	0		2		0	
## 6	3	12	0		2		0	
##	Meals_Employer	Meals_Others	Meals_Payment	Meals_At_Home	Item_Code	Source_Code		
## 1	0	0	0	60	101		1	
## 2	0	0	0	60	101		1	
## 3	0	0	0	60	101		1	
## 4	0	0	0	60	101		1	
## 5	0	0	0	60	101		1	
## 6	0	0	0	60	101		1	
##	ricepds_q	riceos_q	ricetotal_q	chira_q	khoi_q	muri_q	ricepro_q	riceGT_q
## 1	4.000000	6.000000	10.000000	0	0	0	0	10.000000
## 2	4.000000	3.600000	7.600000	0	0	0	0	7.600000
## 3	3.000000	5.000000	8.000000	0	0	0	0	8.000000
## 4	3.333333	1.666667	5.000000	0	0	0	0	5.000000
## 5	3.000000	6.000000	9.000000	0	0	0	0	9.000000
## 6	3.333333	4.250000	7.583333	0	0	0	0	7.583333
##	Wheatpds_q	wheatos_q	wheattotal_q	maida_q	suji_q	sewai_q	bread_q	
## 1	0	1.666667	1.666667	0.000000	0	0.0	0.666667	
## 2	0	1.200000	1.200000	0.760000	0	0.8	0.800000	
## 3	0	0.750000	0.750000	0.000000	0	0.0	0.250000	
## 4	0	4.000000	4.000000	1.000000	0	0.0	0.000000	
## 5	0	0.666667	0.666667	0.000000	0	0.0	1.000000	
## 6	0	0.500000	0.500000	0.266667	0	0.0	0.083333	
##	wheatp_q	wheatGT_q	jowarp_q	bajrap_q	maizep_q	barleyp_q	milletp_q	ragip_q

## 1	0	2.333333	0	0	0	0	0	0
## 2	0	3.560000	0	0	0	0	0	0
## 3	0	1.000000	0	0	0	0	0	0
## 4	0	5.000000	0	0	0	0	0	0
## 5	0	1.666667	0	0	0	0	0	0
## 6	0	0.850000	0	0	0	0	0	0
##	cerealot_q	cerealtot_q	cerealsub_q	cerealstt_q	arhar_q	gramdal_q		
## 1	0	12.333333	0	12.333333	0.000000	0.000000		
## 2	0	11.160000	0	11.160000	0.000000	0.400000		
## 3	0	9.000000	0	9.000000	0.000000	0.000000		
## 4	0	10.000000	0	10.000000	0.000000	0.000000		
## 5	0	10.666667	0	10.666667	0.166667	0.166667		
## 6	0	8.433333	0	8.433333	0.333333	0.000000		
##	gramwholep_q	gramGT_q	moong_q	masur_q	urd_q	peasdal_q	khesari_q	
## 1	0	0.000000	1.000000	0.666667	0.000000	0	0	
## 2	0	0.400000	0.200000	0.520000	0.000000	0	0	
## 3	0	0.000000	0.250000	0.500000	0.000000	0	0	
## 4	0	0.000000	0.333333	0.666667	0.333333	0	0	
## 5	0	0.166667	0.000000	0.333333	0.000000	0	0	
## 6	0	0.000000	0.166667	0.333333	0.166667	0	0	
##	otpulse_q	gramp_q	besan_q	pulsep_q	pulsestot_q	pulsestt_q	soyabean_q	milk_q
## 1	0	0	0.666667	0	2.333333	2.333333	NA	5.20
## 2	0	0	0.000000	0	1.120000	1.120000	NA	6.24
## 3	0	0	0.125000	0	0.875000	0.875000	NA	7.80
## 4	0	0	0.000000	0	1.333333	1.333333	NA	5.20
## 5	0	0	0.000000	0	0.666667	0.666667	NA	2.08
## 6	0	0	0.000000	0	1.000000	1.000000	NA	5.20
##	babyfood_q	milkcond_q	curd_q	ghee_q	butter_q	icecream_q	otmilkp_q	Milkttotal_q
## 1	0	0	0	0.05	0.1	0	0	0
## 2	0	0	0	0.00	0.1	0	0	0
## 3	0	0	0	0.25	0.0	0	0	0
## 4	0	0	0	0.00	0.0	0	0	0
## 5	0	0	0	0.00	0.0	0	0	0
## 6	0	0	0	0.00	0.0	0	0	0
##	milkprott_q	vanas_q	musoil_q	gnoil_q	cocoail_q	edioilothr_q	edibletotal_q	
## 1	5.35	0	0.666667	0	0	0	1.000000	
## 2	6.34	0	0.800000	0	0	0	1.200000	
## 3	8.05	0	0.500000	0	0	0	1.000000	
## 4	5.20	0	0.333333	0	0	0	0.666667	
## 5	2.08	0	0.666667	0	0	0	1.000000	
## 6	5.20	0	0.333333	0	0	0	0.500000	
##	ediblest_q	eggsno_q	fishprawn_q	goatmeat_q	beef_q	pork_q	chicken_q	
## 1	0.666667	0.000275	0.666667	0.0	1.000000	0	0.333333	
## 2	0.800000	0.000660	1.000000	0.6	0.450000	0	0.000000	
## 3	0.500000	0.0004125	0.000000	0.5	0.000000	0	0.000000	
## 4	0.333333	0.000550	0.333333	0.0	0.000000	0	0.666667	
## 5	0.666667	0.000000	0.166667	0.0	0.333333	0	0.000000	
## 6	0.333333	0.0001375	0.166667	0.0	0.000000	0	0.166667	
##	othrbirds_q	nonvegttotal_q	emftt_q	potato_q	onion_q	tamato_q	brinjal_q	
## 1	0	0	2.000275	1.000000	0.666667	1.000000	0.333333	
## 2	0	0	2.050660	1.000000	0.800000	1.200000	0.000000	
## 3	0	0	0.5004125	1.250000	0.500000	0.750000	0.250000	
## 4	0	0	1.000550	1.000000	0.666667	0.333333	0.333333	
## 5	0	0	0.500000	0.333333	0.666667	1.000000	0.666667	

```

## 6          0          0 0.3334708 1.0000000 0.3333333 0.5000000 0.0000000
##   radish_q  carrot_q  palak_q  chillig_q  bhindi_q  parwal_q  cauli_q
## 1 0.0000000 0.0000000 1.0000000 0.16666667      0      0 0.6666667
## 2 0.4000000 0.6000000 0.8000000 0.00000000      0      0 0.4000000
## 3 0.0000000 0.2500000 0.5000000 0.12500000      0      0 0.7500000
## 4 0.0000000 0.3333333 0.6666667 0.10000000      0      0 0.3333333
## 5 0.0000000 0.3333333 0.0000000 0.00000000      0      0 0.6666667
## 6 0.1666667 0.3333333 0.5000000 0.08333333      0      0 0.3333333
##   cabbage_q  pumpkin_q  peas_q  fbeans_q  lemonno_q  otveg_q  vegtt_q
## 1 0.0000000      0 0.3333333 0.3333333 1e-04      1 6.500100
## 2 0.6000000      0 0.4000000 0.8000000 0e+00      0 7.000000
## 3 0.5000000      0 0.0000000 0.2500000 0e+00      0 5.125000
## 4 0.3333333      0 0.3333333 0.3333333 0e+00      0 4.766667
## 5 0.0000000      0 0.0000000 0.3333333 0e+00      0 4.000000
## 6 0.3333333      0 0.0000000 0.3333333 0e+00      0 3.916667
##   bananano_q  jackfruit_q  watermel_q  pineaplno_q  cocono_q  cocogno_q  guava_q
## 1 0.000600000      0      0 0.001500      0      0      0
## 2 0.000280000      0      0 0.001500      0      0      0
## 3 0.000300000      0      0 0.001125      0      0      0
## 4 0.000266667      0      0 0.000500      0      0      0
## 5 0.000400000      0      0 0.001500      0      0      0
## 6 0.000200000      0      0 0.000500      0      0      0
##   sighara_q  orangenno_q  papayar_q  mango_q  kharbooz_q  pears_q  berries_q  leechi_q
## 1      0      0 0.0000000      0      0      0      0      0
## 2      0      0 0.6000000      0      0      0      0      0
## 3      0      0 0.0000000      0      0      0      0      0
## 4      0      0 0.0000000      0      0      0      0      0
## 5      0      0 0.6666667      0      0      0      0      0
## 6      0      0 0.0000000      0      0      0      0      0
##   apple_q  grapes_q  otfruits_q  fruitstt_q  fruitt_total  cocodf_q  gnutdf_q
## 1 0.0000000 0.3333333      0      0 89.00000      0 0.0
## 2 0.2000000 0.2000000      0      0 113.40000      0 0.2
## 3 0.0000000 0.0000000      0      0 39.87500      0 0.0
## 4 0.3333333 0.0000000      0      0 66.00000      0 0.0
## 5 0.0000000 0.0000000      0      0 55.66667      0 0.0
## 6 0.0000000 0.2166667      0      0 35.16667      0 0.0
##   datesdf_q  cashewdf_q  walnutdf_q  otnutsdf_q  kishmish_q  otherdf_q
## 1      0      0 0.0000000      0      0      0
## 2      0      0 0.1000000      0      0      0
## 3      0      0 0.0000000      0      0      0
## 4      0      0 0.0000000      0      0      0
## 5      0      0 0.1666667      0      0      0
## 6      0      0 0.0000000      0      0      0
##   dryfruitstotal_q  dftt_q  sugarpds_q  sugaros_q  sugarst_q  gur_q
## 1      0 0.0000000 0.5000000 0.0000000 0.5000000 0.06666667
## 2      0 0.3000000 0.0000000 1.0000000 1.0000000 0.00000000
## 3      0 0.0000000 0.5000000 0.0000000 0.5000000 0.10000000
## 4      0 0.0000000 0.0000000 0.3333333 0.3333333 0.06666667
## 5      0 0.1666667 0.2666667 0.4333333 0.7000000 0.00000000
## 6      0 0.0000000 0.3333333 0.3333333 0.6666667 0.25000000
##   misri_q  honey_q  sugartotal_q  sugartt_q  salt_q  ginger_q  garlic_q
## 1 0.0000000      0 0.6666667 0.5666667 0.1000000 0.000050000 1.00e-04
## 2 0.0000000      0 1.3200000 1.0000000 0.3200000 0.000100000 1.60e-04
## 3 0.0000000      0 0.7000000 0.6000000 0.1000000 0.000037500 6.25e-05

```

```

## 4 0.00000000      0      0.6000000 0.4000000 0.2000000 0.000083300 8.33e-05
## 5 0.00000000      0      0.9666667 0.7000000 0.2666667 0.000166667 8.33e-05
## 6 0.03333333      0      1.0833333 0.9500000 0.1333333 0.000050000 5.00e-05
##      jeera_q dhanial_q turnmeric_q blackpepper_q drychilly_q tamarind_q
## 1 3.33e-05 0.00e+00      5.00e-05      3.33e-05      0.00e+00      0.00e+00
## 2 3.00e-05 0.00e+00      2.00e-05      2.00e-05      0.00e+00      0.00e+00
## 3 2.50e-05 3.75e-05      0.00e+00      0.00e+00      2.50e-05      0.00e+00
## 4 0.00e+00 0.00e+00      1.67e-05      1.67e-05      8.33e-05      0.00e+00
## 5 0.00e+00 0.00e+00      3.33e-05      0.00e+00      5.00e-05      0.00e+00
## 6 8.33e-06 1.67e-05      8.33e-06      0.00e+00      4.17e-05      8.33e-06
##      currypounder_q oilseeds_q spicesothr_q spicetot_q spicestotal_q teacupno_q
## 1      0      0      5.00e-05 0.000316667      0.000316667      0
## 2      0      0      3.00e-05 0.000370000      0.000360000      0
## 3      0      0      3.75e-05 0.000225000      0.000225000      0
## 4      0      0      0.00e+00 0.000283333      0.000283333      0
## 5      0      0      0.00e+00 0.000333333      0.000333333      0
## 6      0      0      2.50e-05 0.000225000      0.000208333      0
##      tealeaf_q teatotal_q cofeeno_q coffeepwdr_q cofeetotal_q ice_q coldbvrq_q
## 1      0      0      0      0      0      0      0
## 2      0      0      0      0      0      0      0
## 3      0      0      0      0      0      0      0
## 4      0      0      0      0      0      0      0
## 5      0      0      0      0      0      0      0
## 6      0      0      0      0      0      0      0
##      juice_q othrbevrg_q bevergest_q Biscuits_q preparedsweet_q pickle_q
## 1      0      0      0      0      0 3.33e-05
## 2      0      0      0      0      0 5.00e-05
## 3      0      0      0      0      0 6.25e-05
## 4      0      0      0      0      0 5.00e-05
## 5      0      0      0      0      0 0.00e+00
## 6      0      0      0      0      0 0.00e+00
##      sauce_jam_q Othrprocessed_q Beverageestotal_q ricepds_v riceos_v ricetotal_v
## 1      0      0      3.33e-05 44.00000 140.6667      184.66667
## 2      0      0      5.00e-05 44.00000 90.0000      134.00000
## 3      0      0      6.25e-05 33.00000 110.0000      143.00000
## 4      0      0      5.00e-05 36.66667 40.0000      76.66667
## 5      0      0      0.00e+00 33.00000 150.0000      183.00000
## 6      0      0      0.00e+00 35.00000 85.0000      120.00000
##      chira_v khoi_v muri_v ricepro_v riceGT_v Wheatpds_v wheatos_v wheatttotal_v
## 1      0      0      0      0 184.66667      0 23.333333      23.333333
## 2      0      0      0      0 134.00000      0 17.600000      17.600000
## 3      0      0      0      0 143.00000      0 12.000000      12.000000
## 4      0      0      0      0 76.66667      0 64.000000      64.000000
## 5      0      0      0      0 183.00000      0 10.666667      10.666667
## 6      0      0      0      0 120.00000      0 6.666667      6.666667
##      maida_v suji_v sewai_v bread_v wheatp_v wheatGT_v jowarp_v bajrap_v
## 1 0.00000      0      0.0 43.333333      0 66.66667      0      0
## 2 18.40000      0      16.8 32.000000      0 84.80000      0      0
## 3 0.00000      0      0.0 18.750000      0 30.75000      0      0
## 4 19.33333      0      0.0 0.000000      0 83.33333      0      0
## 5 0.00000      0      0.0 43.333333      0 54.00000      0      0
## 6 7.00000      0      0.0 6.666667      0 20.33333      0      0
##      maizep_v barley_p_v milletp_v ragip_v cerealot_v cerealtot_v cerealsub_v
## 1      0      0      0      0      0      251.3333      0

```

## 2	0	0	0	0	0	218.8000	0
## 3	0	0	0	0	0	173.7500	0
## 4	0	0	0	0	0	160.0000	0
## 5	0	0	0	0	0	237.0000	0
## 6	0	0	0	0	0	140.3333	0
##	cerealstt_v	arhar_v	gramdal_v	gramwholep_v	gramGT_v	moong_v	masur_v
## 1	251.3333	0.000000	0.000000		0	0.000000	41.333333
## 2	218.8000	0.000000	15.600000		0	15.600000	12.600000
## 3	173.7500	0.000000	0.000000		0	0.000000	13.500000
## 4	160.0000	0.000000	0.000000		0	0.000000	15.333333
## 5	237.0000	9.333333	8.666667		0	8.666667	0.000000
## 6	140.3333	14.666667	0.000000		0	0.000000	6.666667
##	urd_v	peasdal_v	khesari_v	otpulse_v	gramp_v	besan_v	pulsep_v
## 1	0.000000	0	0	0	0	29.33333	0
## 2	0.000000	0	0	0	0	0.00000	0
## 3	0.000000	0	0	0	0	6.00000	0
## 4	17.333333	0	0	0	0	0.00000	0
## 5	0.000000	0	0	0	0	0.00000	0
## 6	8.666667	0	0	0	0	0.00000	0
##	pulsestt_v	soyabean_v	milk_v	babyfood_v	milkcond_v	curd_v	ghee_v
## 1	106.66667	NA	130.0000	0	0	0	9
## 2	55.40000	NA	184.0800	0	0	0	0
## 3	42.00000	NA	195.0000	0	0	0	30
## 4	62.00000	NA	104.0000	0	0	0	0
## 5	34.00000	NA	52.0000	0	0	0	0
## 6	49.33333	NA	112.6667	0	0	0	0
##	icecream_v	otmilkp_v	Milkttotal_v	milkprott_v	vanas_v	musoil_v	gnoil_v
## 1	0	0	164.0000	169.0000	0	60	0
## 2	0	0	207.0000	214.0800	0	72	0
## 3	0	0	217.5000	225.0000	0	45	0
## 4	0	0	100.0000	104.0000	0	30	0
## 5	0	0	50.0000	52.0000	0	60	0
## 6	0	0	108.3333	112.6667	0	30	0
##	cocooil_v	edioilothr_v	edibletotal_v	ediblest_v	eggsno_v	fishprawn_v	
## 1	0	0	88.33333	60	1.10	113.33333	
## 2	0	0	106.00000	72	2.75	130.00000	
## 3	0	0	85.00000	45	1.65	0.00000	
## 4	0	0	61.66667	30	2.20	46.66667	
## 5	0	0	91.66667	60	0.00	20.00000	
## 6	0	0	45.83333	30	0.55	23.33333	
##	goatmeat_v	beef_v	pork_v	chicken_v	othrbirds_v	nonvegttotal_v	emftt_v
## 1	0	160.00000	0	66.66667	0	0	341.10000
## 2	150	72.00000	0	0.00000	0	0	354.75000
## 3	130	0.00000	0	0.00000	0	0	131.65000
## 4	0	0.00000	0	120.00000	0	0	168.86667
## 5	0	53.33333	0	0.00000	0	0	73.33333
## 6	0	0.00000	0	33.33333	0	0	57.21667
##	potato_v	onion_v	tamato_v	brinjal_v	radish_v	carrot_v	palak_v
## 1	15.000000	10.00000	20.000000	6.666667	0.0	0.00	15.33333
## 2	14.000000	12.00000	24.000000	0.000000	12.0	9.00	24.00000
## 3	18.750000	7.50000	17.500000	7.500000	0.0	8.75	13.75000
## 4	15.000000	13.33333	8.333333	11.666667	0.0	5.00	13.33333
## 5	3.333333	10.00000	20.000000	20.000000	0.0	5.00	0.00000
## 6	11.333333	5.00000	7.500000	0.000000	2.5	4.00	11.66667

##	bhindi_v	parwal_v	cauli_v	cabbage_v	pumpkin_v	peas_v	fbeans_v	lemonno_v
## 1	0	0	16.666667	0.000000	0	15.000000	13.333333	0.5
## 2	0	0	10.000000	12.000000	0	16.000000	28.000000	0.0
## 3	0	0	18.750000	7.500000	0	0.000000	10.000000	0.0
## 4	0	0	8.333333	6.666667	0	13.333333	16.666667	0.0
## 5	0	0	20.000000	0.000000	0	0.000000	13.333333	0.0
## 6	0	0	8.333333	6.666667	0	0.000000	11.666667	0.0
##	otveg_v	veggt_v	bananano_v	jackfruit_v	watermel_v	pineaplno_v	cocono_v	
## 1	20	139.16667	4.000000	0	0	45.000	0	
## 2	0	161.00000	1.800000	0	0	42.000	0	
## 3	0	120.00000	1.750000	0	0	28.125	0	
## 4	0	116.66667	1.833333	0	0	17.500	0	
## 5	0	91.66667	2.333333	0	0	30.000	0	
## 6	0	72.83333	1.166667	0	0	14.000	0	
##	cocogno_v	guava_v	sighara_v	orangen_v	papayar_v	mango_v	kharbooz_v	pears_v
## 1	0	0	0	0	0.00000	0	0	0
## 2	0	0	0	0	5.60000	0	0	0
## 3	0	0	0	0	0.00000	0	0	0
## 4	0	0	0	0	0.00000	0	0	0
## 5	0	0	0	0	23.33333	0	0	0
## 6	0	0	0	0	0.00000	0	0	0
##	berries_v	leechi_v	apple_v	grapes_v	otfruits_v	fruitstt_v	cocodf_v	gnutdf_v
## 1	0	0	0.00000	40	0	0	0	0
## 2	0	0	36.00000	28	0	0	0	13
## 3	0	0	0.00000	0	10	0	0	0
## 4	0	0	46.66667	0	0	0	0	0
## 5	0	0	0.00000	0	0	0	0	0
## 6	0	0	0.00000	20	0	0	0	0
##	datesdf_v	cashewdf_v	walnutdf_v	otnutsdf_v	kishmish_v	otherdf_v		
## 1	0	0	0.00000	0	0	0		
## 2	0	0	32.00000	0	0	0		
## 3	0	0	0.00000	0	0	0		
## 4	0	0	0.00000	0	0	0		
## 5	0	0	56.33333	0	0	0		
## 6	0	0	0.00000	0	0	0		
##	dryfruitsttotal_v	dftt_v	sugarpds_v	sugaros_v	sugarst_v	gur_v	misri_v	
## 1	0	0.00000	8.000000	0.00000	8.00000	2.333333	0	
## 2	0	45.00000	0.000000	36.00000	36.00000	0.000000	0	
## 3	0	0.00000	6.000000	0.00000	6.00000	2.500000	0	
## 4	0	0.00000	0.000000	18.66667	18.66667	2.000000	0	
## 5	0	56.33333	4.000000	16.00000	20.00000	0.000000	0	
## 6	0	0.00000	5.333333	12.00000	17.33333	18.000000	1	
##	honey_v	sugartotal_v	sugartt_v	salt_v	ginger_v	garlic_v	jeera_v	
## 1	0	11.66667	10.33333	1.333333	0.005000000	0.008333333	0.006666667	
## 2	0	39.20000	36.00000	3.200000	0.011000000	0.021400000	0.004000000	
## 3	0	10.00000	8.50000	1.500000	0.005000000	0.007500000	0.003250000	
## 4	0	24.00000	20.66667	3.333333	0.008333333	0.006666667	0.000000000	
## 5	0	23.66667	20.00000	3.666667	0.016666667	0.010000000	0.000000000	
## 6	0	38.00000	36.33333	1.666667	0.004166667	0.003333333	0.001666667	
##	dhania_v	turnmeric_v	blackpepper_v	drychilly_v	tamarind_v	currypowder_v		
## 1	0.000000000	0.005000000	0.002666667	0.000000000	0.000000000		0	
## 2	0.000000000	0.006000000	0.003000000	0.000000000	0.000000000		0	
## 3	0.003750000	0.000000000	0.000000000	0.005000000	0.000000000		0	
## 4	0.000000000	0.003333333	0.005000000	0.010000000	0.000000000		0	

```
## 5 0.000000000 0.006666667 0.000000000 0.006666667 0.000000000 0
## 6 0.001666667 0.000833333 0.000000000 0.004166667 0.001666667 0
## oilseeds_v spicesothr_v spicetot_v spicestotal_v teacupno_v tealeaf_v
## 1 0 0.002333333 0.03000000 0.03000000 0 0
## 2 0 0.004000000 0.05300000 0.04940000 0 0
## 3 0 0.005000000 0.02950000 0.02950000 0 0
## 4 0 0.000000000 0.03333333 0.03333333 0 0
## 5 0 0.000000000 0.04000000 0.04000000 0 0
## 6 0 0.002500000 0.02250000 0.02000000 0 0
## teatotal_v coffeeno_v coffeepwdr_v coffeetotal_v ice_v coldbvrg_v juice_v
## 1 0 0 0 0 0 0 0
## 2 0 0 0 0 0 0 0
## 3 0 0 0 0 0 0 0
## 4 0 0 0 0 0 0 0
## 5 0 0 0 0 0 0 0
## 6 0 0 0 0 0 0 0
## othrbevrg_v bevergest_v Biscuits_v preparedsweet_v pickle_v sauce_jam_v
## 1 0 0 40.00000 0.000000 0.020000000 0
## 2 0 0 18.00000 12.000000 0.014000000 0
## 3 0 0 0.00000 12.500000 0.011250000 0
## 4 0 0 23.33333 20.000000 0.006666667 0
## 5 0 0 40.00000 3.333333 0.000000000 0
## 6 0 0 25.00000 33.333333 0.000000000 0
## Othrprocessed_v Beveragestotal_v foodtotal_v foodtotal_q state_1 Region
## 1 0.00000 40.02000 1118.9833 29.85073 MEG 1
## 2 0.00000 30.01400 1190.2934 30.09107 MEG 1
## 3 0.00000 12.51125 759.9407 24.75070 MEG 1
## 4 0.00000 43.34000 708.9067 23.23422 MEG 1
## 5 43.33333 86.66667 714.7067 19.71367 MEG 1
## 6 0.00000 58.33333 558.7367 20.30035 MEG 1
## fruits_df_tt_v fv_tot
## 1 89.00000 228.1667
## 2 158.40000 319.4000
## 3 39.87500 159.8750
## 4 66.00000 182.6667
## 5 112.00000 203.6667
## 6 35.16667 108.0000
```

```
# Print the dimensions (number of rows and columns) of the Meghalaya dataset
print(dim(data_meghalaya))
```

```
## [1] 1259 384
```

```
# Save the filtered Meghalaya data to a new CSV file for future use
write.csv(data_meghalaya, './datasets/meghalaya_NSS068.csv', row.names = FALSE) # Prevent writing row n

# Select a subset of relevant columns for focused analysis
data_meghalaya <- data_meghalaya %>% select(
  state,
  state_1,
  District,
  Region,
  Sector,
```

```

State_Region,
Meals_At_Home,
ricetotal_v,
wheattotal_v,
Milktotal_v,
pulsestot_v,
nonvegtotal_v,
fruitstt_v,
No_of_Meals_per_day
)

# --- Handling Missing Values (NA) ---

# Check which columns (if any) still have missing values (NA) before imputation
print(colSums(is.na(data_meghalaya)) > 0)

```

```

##          state          state_1          District          Region
##          FALSE          FALSE          FALSE          FALSE
##          Sector          State_Region          Meals_At_Home          ricetotal_v
##          FALSE          FALSE          TRUE          FALSE
##          wheattotal_v          Milktotal_v          pulsestot_v          nonvegtotal_v
##          FALSE          FALSE          FALSE          FALSE
##          fruitstt_v No_of_Meals_per_day
##          FALSE          TRUE

```

```

# Define a function to impute missing values with the mean of the column
impute <- function(col) {
  if(any(is.na(col))) { # Check if there are any NAs in the column
    col[is.na(col)] <- mean(col, na.rm = T) # Replace NAs with the mean, ignoring NAs in mean calculation
  }
  return(col) # Return the column with imputed values
}

# Apply the imputation function to specific columns identified to have NAs
data_meghalaya$Meals_At_Home <- impute(data_meghalaya$Meals_At_Home)
data_meghalaya$No_of_Meals_per_day <- impute(data_meghalaya$No_of_Meals_per_day)

# Re-check for missing values after imputation to confirm removal
print(colSums(is.na(data_meghalaya)) > 0)

```

```

##          state          state_1          District          Region
##          FALSE          FALSE          FALSE          FALSE
##          Sector          State_Region          Meals_At_Home          ricetotal_v
##          FALSE          FALSE          FALSE          FALSE
##          wheattotal_v          Milktotal_v          pulsestot_v          nonvegtotal_v
##          FALSE          FALSE          FALSE          FALSE
##          fruitstt_v No_of_Meals_per_day
##          FALSE          FALSE

```

```

# --- Outlier Removal ---

# Define a function to remove outliers using the Interquartile Range (IQR) method

```

```

remove_outliers <- function(df, col) {
  q1 <- quantile(df[, col], 0.25) # Calculate the first quartile (25th percentile)
  q3 <- quantile(df[, col], 0.75) # Calculate the third quartile (75th percentile)
  iqr <- q3 - q1 # Calculate the Interquartile Range
  lower_threshold <- q1 - (1.5 * iqr) # Define the lower bound for outlier detection
  upper_threshold <- q3 + (1.5 * iqr) # Define the upper bound for outlier detection
  # Subset the dataframe to include only rows where the column's value is within the defined thresholds
  df <- subset(df, df[, col] >= lower_threshold & df[, col] <= upper_threshold)
  return(df) # Return the dataframe with outliers removed from the specified column
}

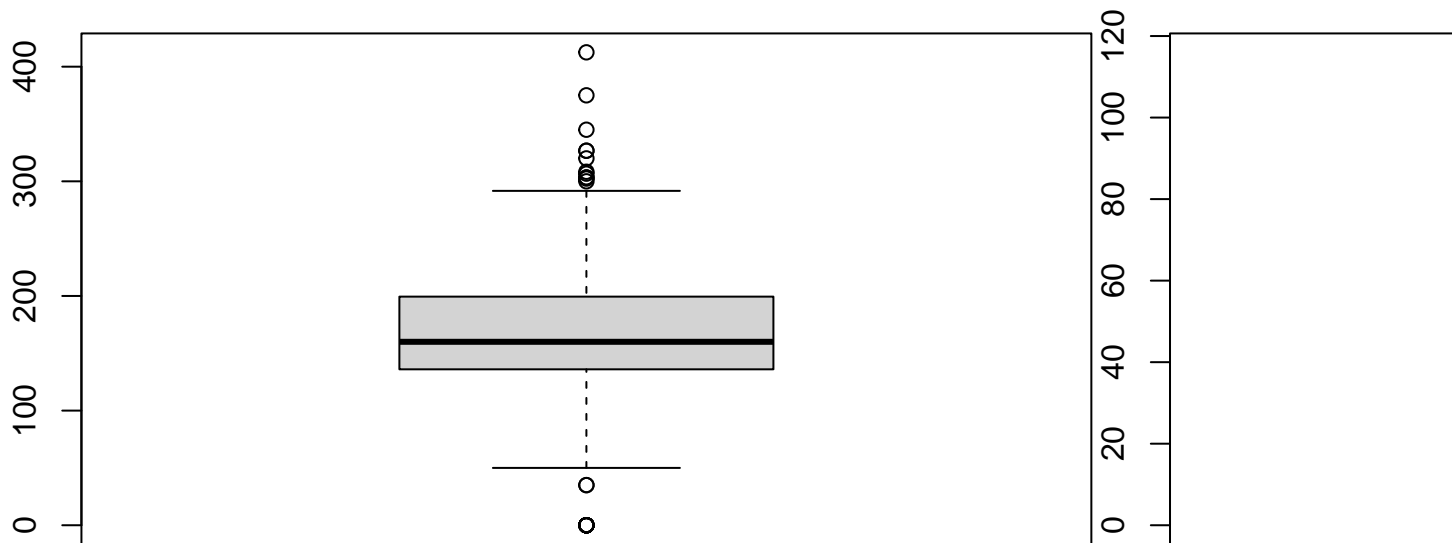
# Columns to check for outliers (various consumption variables)
outlier_check_cols <- c("ricetotal_v", "wheattotal_v", "Milkttotal_v", "pulsestot_v", "nonvegttotal_v", "nonvegtotal_v")

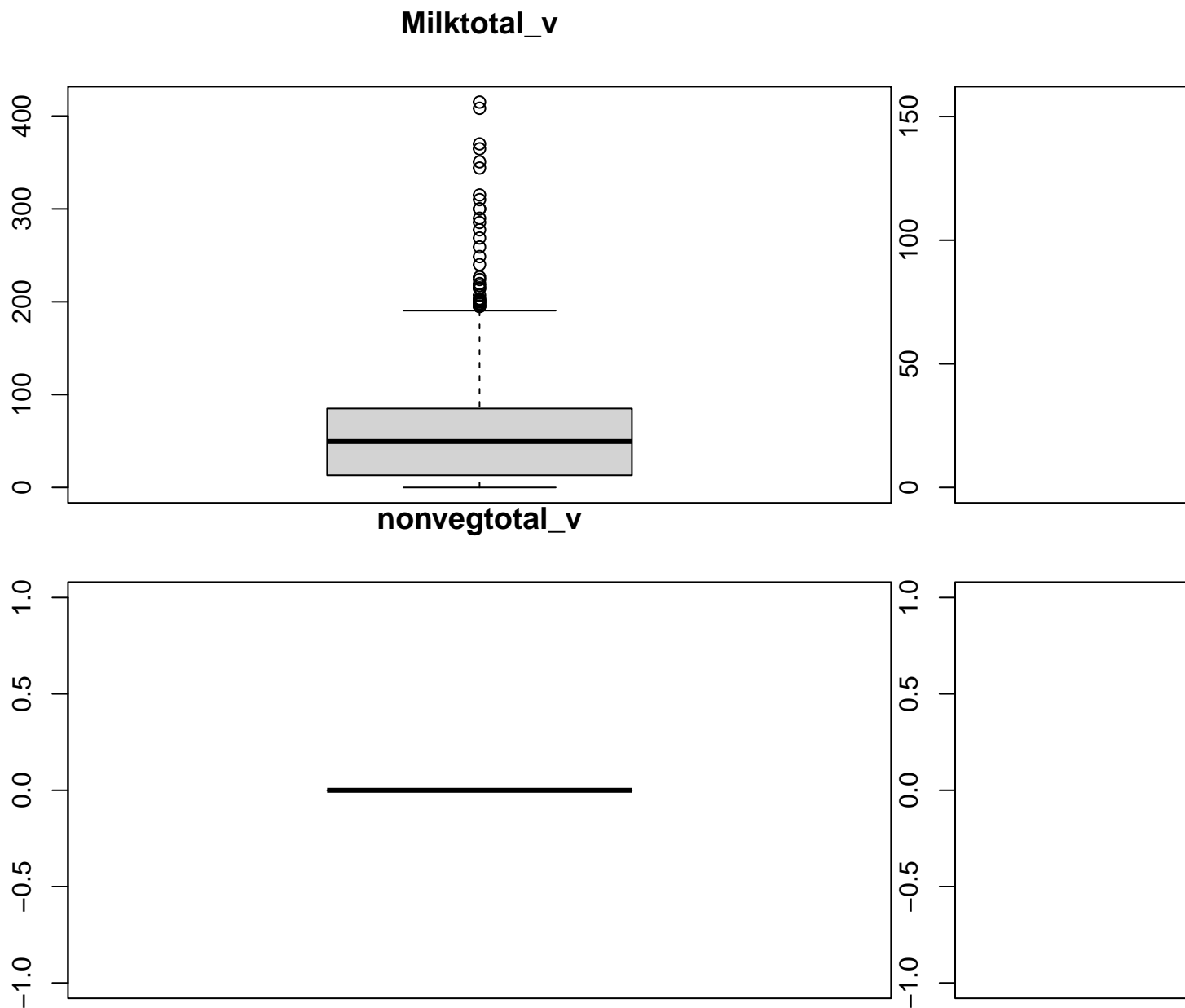
# Define a function to plot boxplots for outlier visualization
plot_outliers <- function(cols) {
  for (outlier_check_col in cols) {
    boxplot(data_meghalaya[, outlier_check_col], main = outlier_check_col) # Create a boxplot for each
  }
}

# Plot initial boxplots to visualize outliers before removal
plot_outliers(outlier_check_cols)

```

ricetotal_v

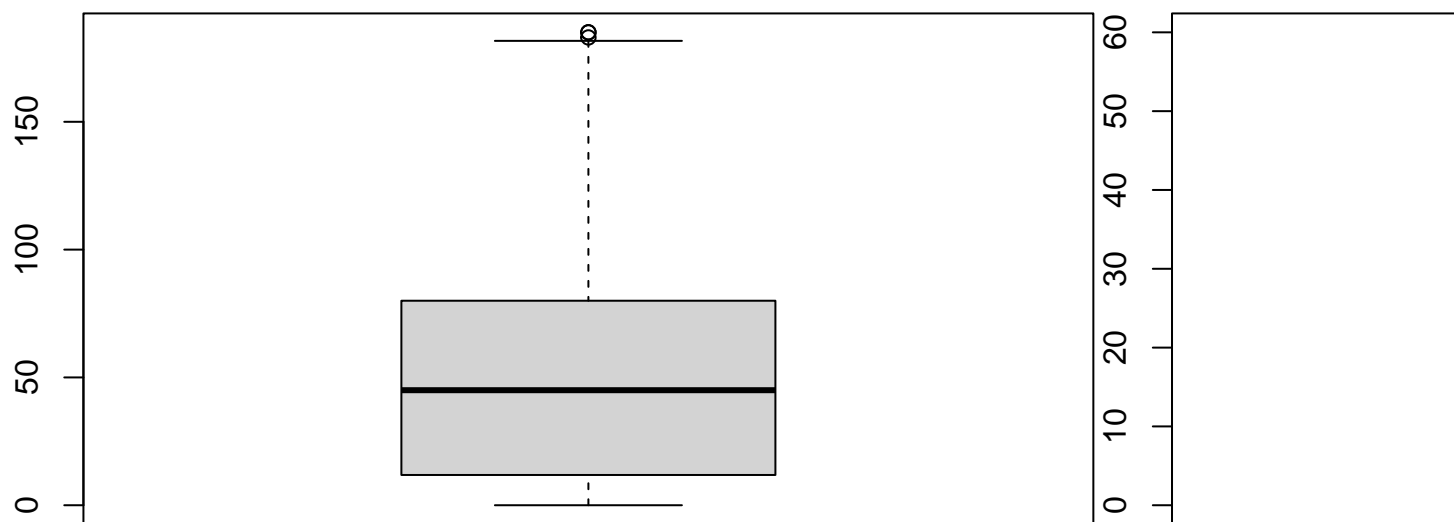
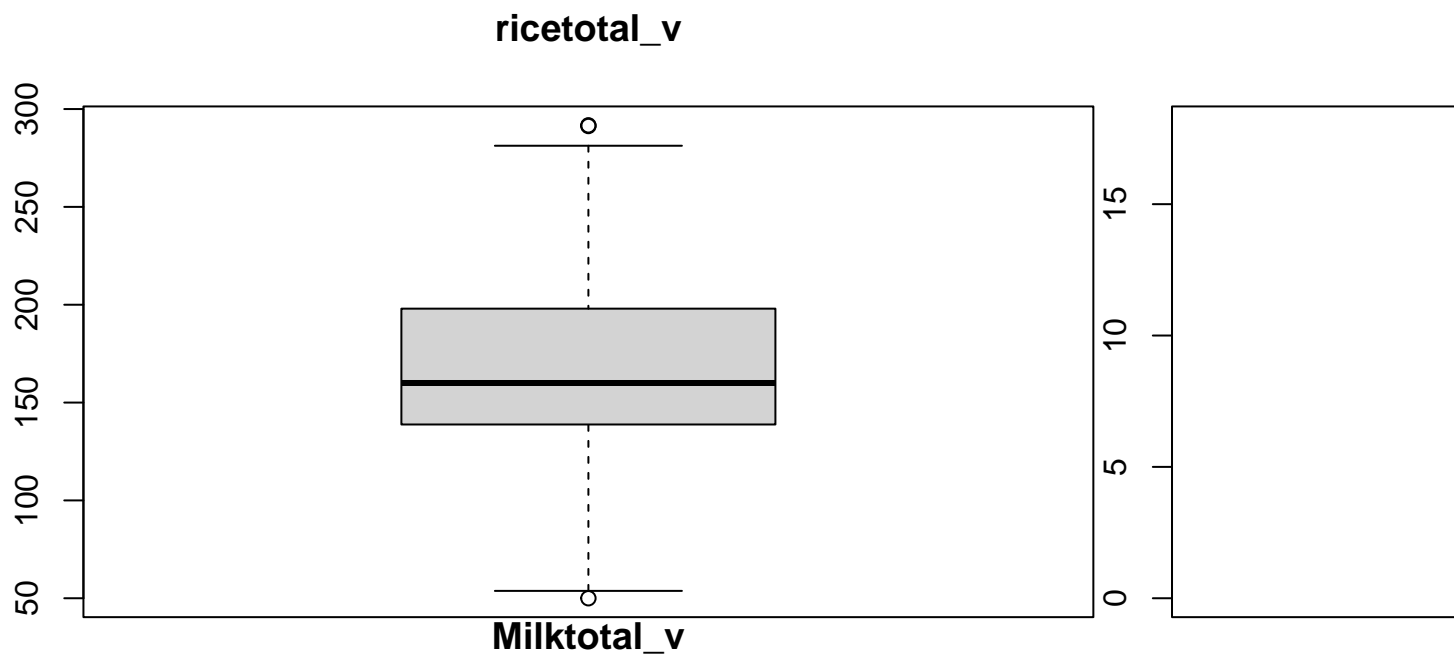




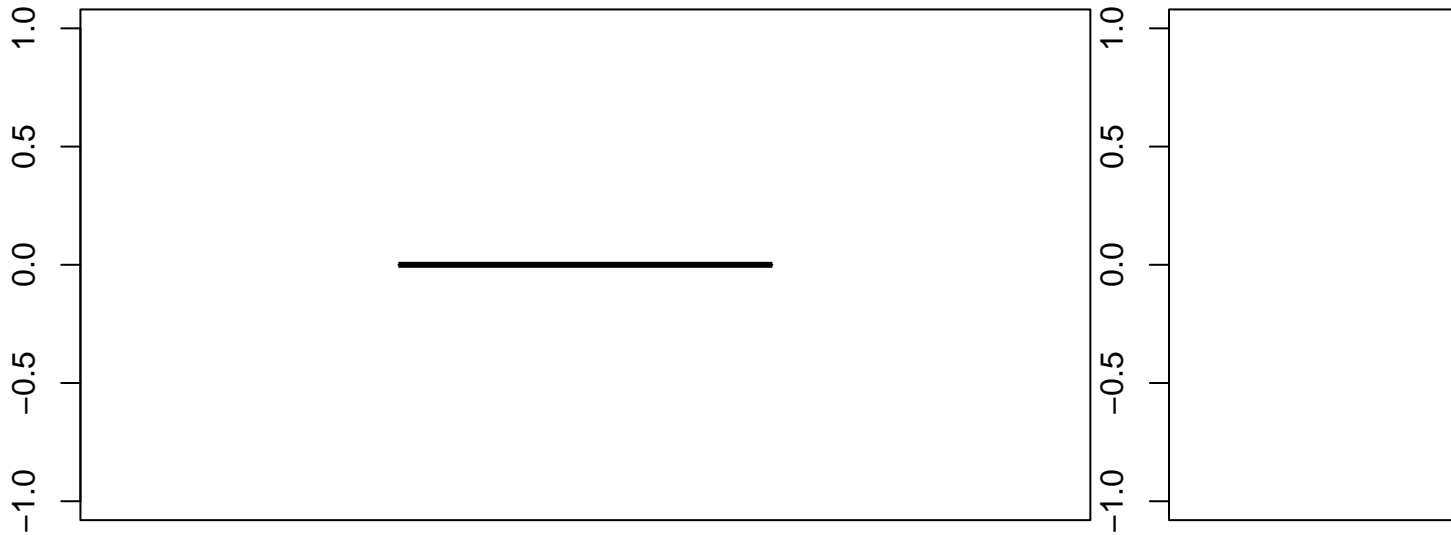
```
# Columns to apply outlier removal to (a subset of the check columns)
outlier_cols <- c("ricetotal_v", "wheattotal_v", "Milktotal_v", "pulsestot_v")

# Loop through specified columns and apply the outlier removal function
for (outlier_col in outlier_cols) {
  data_meghalaya <- remove_outliers(data_meghalaya, outlier_col)
}

# Plot boxplots again to visualize the effect of outlier removal
plot_outliers(outlier_check_cols)
```



nonvegtotal_v



```
# --- Total Consumption Calculation and Summaries ---

# Define columns representing various consumption categories to be summed
consumption_cols <- c("ricetotal_v", "wheattotal_v", "Milktotal_v", "pulsestot_v", "nonvegtotal_v", "fr")

# Calculate the total consumption for each row by summing up the specified consumption columns
data_meghalaya$total_consumption_v <- rowSums(data_meghalaya[, consumption_cols], na.rm = T)

# Define a function to summarize total consumption by a given grouping column
summarize_consumption <- function(col) {
  summary <- data_meghalaya %>%
    group_by(across(all_of(col))) %>% # Group data by the specified column
    summarize(total = sum(total_consumption_v)) %>% # Calculate the sum of total consumption for each g
    arrange(desc(total)) # Arrange the summary table in descending order of total consumption
  return(summary) # Return the summary table
}

# Summarize consumption by District, Region, and Sector
district_summary <- summarize_consumption("District")
region_summary <- summarize_consumption("Region")
sector_summary <- summarize_consumption("Sector")

# Print the top 3 and bottom 3 consuming districts based on the summary
cat("Top 3 Consuming Districts:\n")
```

```
## Top 3 Consuming Districts:
```

```
print(head(district_summary, 3))
```

```
## # A tibble: 3 x 2
##   District total
##   <int>   <dbl>
```

```
## 1      6 70280.
## 2      1 53582.
## 3      4 36429.
```

```
cat("Bottom 3 Consuming Districts:\n")
```

```
## Bottom 3 Consuming Districts:
```

```
print(tail(district_summary, 3))
```

```
## # A tibble: 3 x 2
##   District total
##   <int>   <dbl>
## 1      5 29470.
## 2      2 28229.
## 3      3 19920.
```

```
# Print the consumption summary for regions and sectors
```

```
cat("Region Consumption Summary:\n")
```

```
## Region Consumption Summary:
```

```
print(region_summary)
```

```
## # A tibble: 1 x 2
##   Region total
##   <int>   <dbl>
## 1      1 269421.
```

```
cat("Sector Consumption Summary:\n")
```

```
## Sector Consumption Summary:
```

```
print(sector_summary)
```

```
## # A tibble: 2 x 2
##   Sector total
##   <int>   <dbl>
## 1      1 184565.
## 2      2  84856.
```

```
# --- Data Transformation (Mapping Codes to Names) ---
```

```
# Create a mapping from numeric District codes to their full names
```

```
district_map <- c(
  "1" = 'West Garo Hills',
  "2" = 'East Garo Hills',
  "3" = 'South Garo Hills',
  "4" = 'West Khasi Hills',
```

```

"5" = 'Ri Bhoi',
"6" = 'East Khasi Hills',
"7" = 'Jaintia Hills'
)

# Convert District column to character type before applying the map for consistency
data_meghalaya$District <- as.character(data_meghalaya$District)
# Replace numeric District codes with their corresponding names using the map
data_meghalaya$District <- ifelse(data_meghalaya$District %in% names(district_map), district_map[data_meghalaya$District], district_map[1])

# Create a mapping from numeric Sector codes to descriptive names
sector_map <- c(
  "1" = "Rural",
  "2" = "Urban"
)

# Convert Sector column to character type before applying the map
data_meghalaya$Sector <- as.character(data_meghalaya$Sector)
# Replace numeric Sector codes with "Rural" or "Urban"
data_meghalaya$Sector <- ifelse(data_meghalaya$Sector %in% names(sector_map), sector_map[data_meghalaya$Sector], sector_map[1])

# Re-summarize consumption by District, Region, and Sector after mapping names
# This ensures summaries use the descriptive names rather than numeric codes
district_summary <- summarize_consumption("District")
region_summary <- summarize_consumption("Region")
sector_summary <- summarize_consumption("Sector")

# --- Z-test for Rural vs. Urban Consumption ---

# Filter total consumption data for rural areas
rural <- data_meghalaya %>%
  filter(Sector == "Rural") %>%
  select(total_consumption_v)

# Filter total consumption data for urban areas
urban <- data_meghalaya %>%
  filter(Sector == "Urban") %>%
  select(total_consumption_v)

# Calculate the mean total consumption for rural and urban areas, ignoring NAs
mean_rural <- mean(rural$total_consumption_v, na.rm = T)
mean_urban <- mean(urban$total_consumption_v, na.rm = T)

# Calculate the standard deviation for rural and urban consumption, ignoring NAs
sd_rural <- sd(rural$total_consumption_v, na.rm = T)
sd_urban <- sd(urban$total_consumption_v, na.rm = T)

# Perform a two-sample Z-test to compare mean consumptions between rural and urban areas
# Using calculated sample standard deviations as estimates for population standard deviations
z_test_result_sector <- z.test(
  x = rural$total_consumption_v, # Data for the first group (rural)
  y = urban$total_consumption_v, # Data for the second group (urban)
  alternative = "two.sided", # Test for a difference in either direction

```

```

mu = 0, # Null hypothesis: difference in means is 0
sigma.x = sd_rural, # Standard deviation of the rural group
sigma.y = sd_urban, # Standard deviation of the urban group
conf.level = 0.95 # 95% confidence level
)

# Interpret the results of the Z-test for rural vs. urban consumption
cat("\n--- Z-test Results: Rural vs. Urban Consumption ---\n")

##
## --- Z-test Results: Rural vs. Urban Consumption ---

if (z_test_result_sector$p.value < 0.05) {
  cat("P value is < 0.05 (", round(z_test_result_sector$p.value, 5), "). Therefore, we reject the null hypothesis.\n")
  cat("There is a significant difference between mean consumptions of urban and rural areas.\n")
  cat("The mean consumption in Rural areas is ", round(mean_rural, 2), " and in Urban areas it's ", round(mean_urban, 2), "\n")
} else {
  cat("P value is >= 0.05 (", round(z_test_result_sector$p.value, 5), "). Therefore, we fail to reject the null hypothesis.\n")
  cat("There is no significant difference between mean consumptions of urban and rural areas.\n")
  cat("The mean consumption in Rural areas is ", round(mean_rural, 2), " and in Urban areas it's ", round(mean_urban, 2), "\n")
}

```

```

## P value is < 0.05 ( 0 ). Therefore, we reject the null hypothesis.
## There is a significant difference between mean consumptions of urban and rural areas.
## The mean consumption in Rural areas is 228.99 and in Urban areas it's 271.97 .

```

```

# --- Z-test for Top vs. Bottom Consuming Districts ---

# Get the name of the top consuming district from the summarized data
top_district_name <- head(district_summary, 1)$District
# Get the name of the bottom consuming district from the summarized data
bottom_district_name <- tail(district_summary, 1)$District

# Filter total consumption data for the top consuming district
top_district_data <- data_meghalaya %>%
  filter(District == top_district_name) %>%
  select(total_consumption_v)

# Filter total consumption data for the bottom consuming district
bottom_district_data <- data_meghalaya %>%
  filter(District == bottom_district_name) %>%
  select(total_consumption_v)

# Calculate the mean total consumption for the top and bottom districts, ignoring NAs
mean_top_district <- mean(top_district_data$total_consumption_v, na.rm = TRUE)
mean_bottom_district <- mean(bottom_district_data$total_consumption_v, na.rm = TRUE)

# Calculate the standard deviation for the top and bottom districts' consumption, ignoring NAs
sd_top_district <- sd(top_district_data$total_consumption_v, na.rm = TRUE)
sd_bottom_district <- sd(bottom_district_data$total_consumption_v, na.rm = TRUE)

# Perform a two-sample Z-test to compare mean consumptions between the top and bottom districts

```

```

# Using calculated sample standard deviations as estimates for population standard deviations
z_test_result_district <- z.test(
  x = top_district_data$total_consumption_v, # Data for the top district
  y = bottom_district_data$total_consumption_v, # Data for the bottom district
  alternative = "two.sided", # Test for a difference in either direction
  mu = 0, # Null hypothesis: difference in means is 0
  sigma.x = sd_top_district, # Standard deviation for the top district
  sigma.y = sd_bottom_district, # Standard deviation for the bottom district
  conf.level = 0.95 # 95% confidence level
)

# Interpret the results of the Z-test for top vs. bottom districts
cat("\n--- Z-test Results: Top Consuming District (", top_district_name, ") vs. Bottom Consuming District (", bottom_district_name, ")")

##
## --- Z-test Results: Top Consuming District ( East Khasi Hills ) vs. Bottom Consuming District ( South Garo Hills )

if (z_test_result_district$p.value < 0.05) {
  cat("P value is < 0.05 (", round(z_test_result_district$p.value, 5), ").\n")
  cat("Therefore, we reject the null hypothesis.\n")
  cat("There is a significant difference between mean consumptions of ", top_district_name, " and ", bottom_district_name, ".\n")
  cat("The mean consumption in ", top_district_name, " is ", round(mean_top_district, 2), " and in ", bottom_district_name, " is ", round(mean_bottom_district, 2), ".\n")
} else {
  cat("P value is >= 0.05 (", round(z_test_result_district$p.value, 5), ").\n")
  cat("Therefore, we fail to reject the null hypothesis.\n")
  cat("There is no significant difference between mean consumptions of ", top_district_name, " and ", bottom_district_name, ".\n")
  cat("The mean consumption in ", top_district_name, " is ", round(mean_top_district, 2), " and in ", bottom_district_name, " is ", round(mean_bottom_district, 2), ".\n")
}

## P value is >= 0.05 ( 0.37814 ).
## Therefore, we fail to reject the null hypothesis.
## There is no significant difference between mean consumptions of East Khasi Hills and South Garo Hills.
## The mean consumption in East Khasi Hills is 252.8 and in South Garo Hills it's 245.93 .

```