

Load and clean ASA24 data

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Introduction

In this tutorial, we will use mock data from the VVKAJ dataset that was created with ASA24. VVKAJ stands for Vegetarian, Vegan, Keto, American, Japanese and was designed because these different eating patterns reflect differences that are often seen in real data. This mock dataset contains dietary for 15 mock participants who report dietary data while following the 5 different dietary patterns (VVKAJ) for three days. There are a total of XX dietary records in this dataset.

ASA24 data includes the following files: `_Items.csv`, `_INS.csv`, `_Responses.csv`, `_TNS.csv`, `Totals.csv`, and `TS.csv`. Refer to the ASA24 Reserchers' website for specific explanations for each file, but for the purpose of this tutorial, we focus on using the `Items.csv`, which has all the food items reported by the participants.

In this script, you will:

1. Use Metadata 1 to filter out individuals;
2. Remove users that has only a small number of totals (days of record) - if you know which one to remove; and
3. Look for outliers in your totals by nutrient consumed on each day.

Load functions and packages

Name the path to DietR directory where input files are pulled.

```
main_wd <- "~/GitHub/DietR"
```

Load the necessary functions.

```
source("lib/specify_data_dir.R")
source("lib/load_clean_ASA24.R")
source("lib/Food_tree_scripts/format.foods_2.r")
source("lib/QCOutliers.R")
source("lib/average.by.R")
```

You can come back to the main directory by:

```
setwd(main_wd)
```

Load ASA24 Items data

Specify the directory where the data is.

```
SpecifyDataDirectory(directory.name= "eg_data/VVKAJ/")
```

Load your unprocessed (raw) food items-level data (as downloaded from the ASA24 study website). The csv file will be loaded as a dataframe in R and be named as items_raw.

```
items_raw <- read.csv("Raw_data/VVKAJ_Items.csv", sep = ",", header=T)
```

items_raw has a column called "Food_Description", but this needs to be changed to "Main.food.description". Change the column name.

```
names(items_raw)[names(items_raw) == "Food_Description"] <- "Main.food.description"
```

Check if any column names match with "Main.food.description". If there is a match, it will be printed.

```
names(items_raw)[names(items_raw) == "Main.food.description"]
```

```
## [1] "Main.food.description"
```

[NOTE] The numbers in the square brackets of the output indicate the sequential number of each element to help count the number of elements.

Save the items file as a .txt file. This command saves the object “items_raw” as a .txt file with the specified filename using the write.table function.

```
write.table(items_raw, "VVKAJ_Items.txt", sep="\t", row.names=F)
```

Special characters common in food names in dietary data such as ", ', , , % may interfere correct data loading in R; thus, we replace them with an underscore _.

Format foods so that special characters will be replaced with “_.”_f” stands for “formatted”.

```
FormatFoods(input_fn = "VVKAJ_Items.txt",  
            output_fn = "VVKAJ_Items_f.txt")
```

[NOTE] It is best practice to avoid overwriting your raw data. Always save formatted/manipulated versions as a new file as described above.

Load the Items_f.txt file to take a look at it.

You need the quote="" and colClasses="character" arguments to ignore quotation marks (do not regard them as a cell separator) and to load all the columns as characters so that FoodID will keep the trailing “0”.

```
items_f <- read.delim("VVKAJ_Items_f.txt", quote="", colClasses="character")
```

All special characters in the items data should have been replaced with an underscore in the Main.food.description column, the 3rd from the last column of the items_f. We can confirm that by using the head function, which shows the first six rows of the specified dataset by default. In this website version of tutorial, output is folded for visual clarity. You can click and expand the results.

```
head(items_f)
```

Click to expand output

```
##                               RecallRecId UserName  
## 1 41fee4cf-783f-469b-aadf-62c7e2cd33a9 VVKAJ101  
## 2 41fee4cf-783f-469b-aadf-62c7e2cd33a9 VVKAJ101  
## 3 41fee4cf-783f-469b-aadf-62c7e2cd33a9 VVKAJ101  
## 4 41fee4cf-783f-469b-aadf-62c7e2cd33a9 VVKAJ101  
## 5 41fee4cf-783f-469b-aadf-62c7e2cd33a9 VVKAJ101  
## 6 41fee4cf-783f-469b-aadf-62c7e2cd33a9 VVKAJ101  
##                               UserID RecallNo RecallAttempt RecallStatus  
## 1 7bd05142-312e-4648-b1ac-fc258540af52      1           0           2  
## 2 7bd05142-312e-4648-b1ac-fc258540af52      1           0           2  
## 3 7bd05142-312e-4648-b1ac-fc258540af52      1           0           2  
## 4 7bd05142-312e-4648-b1ac-fc258540af52      1           0           2  
## 5 7bd05142-312e-4648-b1ac-fc258540af52      1           0           2  
## 6 7bd05142-312e-4648-b1ac-fc258540af52      1           0           2  
## IntakeStartDateTime IntakeEndDateTime ReportingDate Lang Occ_No  
## 1      11/5/2021 0:00    11/5/2021 23:59    11/6/2021      1      1  
## 2      11/5/2021 0:00    11/5/2021 23:59    11/6/2021      1      1  
## 3      11/5/2021 0:00    11/5/2021 23:59    11/6/2021      1      1
```


## 2	0.00104	3.224	3.05136	0.17264	0	0	0	0	0	1.976
## 3	0	0	0	0	0	0	0	0	0	0
## 4	0	0.274815	0.164495	0.11032	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0	0	0
## 6	0.00777	10.96014	9.55599	1.21545	0	0.14319	0.00444	0	0.03441	1.998
##	CHOLN	VITE_ADD	B12_ADD	F_TOTAL	F_CITMLB	F_OTHER	F_JUICE	V_TOTAL	V_DRKGR	
## 1	0.8928	0	0	0	0	0	0	0	0	0
## 2	29.952	0	3.016	0	0	0	0	0	0	0
## 3	0.231	0	0	0	0	0	0	0	0	0
## 4	5.0235	0	0	0.65995	0.65995	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0	0	0
## 6	268.398	0	0	0	0	0	0	0	0	0
##	V_REDOR_TOTAL	V_REDOR_TOMATO	V_REDOR_OTHER	V_STARCHY_TOTAL	V_STARCHY_POTATO					
## 1	0	0	0	0	0					0
## 2	0	0	0	0	0					0
## 3	0	0	0	0	0					0
## 4	0	0	0	0	0					0
## 5	0	0	0	0	0					0
## 6	0	0	0	0	0					0
##	V_STARCHY_OTHER	V_OTHER	V_LEGUMES	G_TOTAL	G_WHOLE	G_REFINED	PF_TOTAL			
## 1	0	0	0	0	0	0	0			0
## 2	0	0	0	2.7456	2.652	0.0936	0.7384			
## 3	0	0	0	0	0	0	0			0
## 4	0	0	0	0	0	0	0			0
## 5	0	0	0	0	0	0	0			0
## 6	0	0	0	0	0	0	1.8093			
##	PF_MPS_TOTAL	PF_MEAT	PF_CUREDMEAT	PF_ORGAN	PF_POULT	PF_SEAFD_HI	PF_SEAFD_LOW			
## 1	0	0	0	0	0	0	0			0
## 2	0	0	0	0	0	0	0			0
## 3	0	0	0	0	0	0	0			0
## 4	0	0	0	0	0	0	0			0
## 5	0	0	0	0	0	0	0			0
## 6	0	0	0	0	0	0	0			0
##	PF_EGGS	PF_SOY	PF_NUTSDS	PF_LEGUMES	D_TOTAL	D_MILK	D_YOGURT	D_CHEESE	OILS	
## 1	0	0	0	0	0	0	0	0	0	0
## 2	0	0	0.7384	0	0	0	0	0	6.0736	
## 3	0	0	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0	0	0
## 6	1.8093	0	0	0	0	0	0	0	15.8508	
##	SOLID_FATS	ADD_SUGARS	A_DRINKS	FoodComp						
## 1	0	7.05312	0	1						
## 2	0	3.432	0	1						
## 3	0	2.05275	0	1						
## 4	0	0	0	1						
## 5	0	0	0	1						
## 6	4.8507	0	0	1						
##	Main.food.description									
## 1	Soft_drink_NFS									
## 2	Cereal_Post_Great_Grains_Double_Pecan_Whole_Grain_Cereal_									
## 3	Honey									
## 4	Berries_frozen_NFS									
## 5	Water_tap									
## 6	Egg_salad_made_with_mayonnaise									

```
##                                Old.Main.food.description      FoodID
## 1                                Soft drink, NFS 92400000.0
## 2 Cereal (Post Great Grains Double Pecan Whole Grain Cereal) 57231250.0
## 3                                Honey 91302010.0
## 4                                Berries, frozen, NFS 63200200.0
## 5                                Water, tap 94000100.0
## 6                                Egg salad, made with mayonnaise 32103000.0
```

Add a human-readable sample identifier (SampleID) with a desired prefix, and save it as a .txt file. SampleIDs are IDs unique to each combination of users and day and represent days of dietary intake in this dataset.

```
AddSampleIDtoItems(input.fn="VVKAJ_Items_f.txt", user.name="UserName", recall.no="RecallNo",
                    prefix="vvkaj.", out.fn="VVKAJ_Items_f_id.txt")
```

Load the formatted Items file with SampleID added.

```
items_f_id <- read.delim("VVKAJ_Items_f_id.txt", quote="", colClasses="character")
```

A combination of the specified prefix and sequential number (vvkaj.00001) should be added in the SampleID column, the first column of the items_f_id dataframe. You will probably need to scroll up the output a little bit in the console to view the first column.

```
head(items_f_id)
```

Click to expand output

```
##      SampleID                                RecallRecId UserName
## 1 vvkaj.00001 41fee4cf-783f-469b-aadf-62c7e2cd33a9 VVKAJ101
## 2 vvkaj.00001 41fee4cf-783f-469b-aadf-62c7e2cd33a9 VVKAJ101
## 3 vvkaj.00001 41fee4cf-783f-469b-aadf-62c7e2cd33a9 VVKAJ101
## 4 vvkaj.00001 41fee4cf-783f-469b-aadf-62c7e2cd33a9 VVKAJ101
## 5 vvkaj.00001 41fee4cf-783f-469b-aadf-62c7e2cd33a9 VVKAJ101
## 6 vvkaj.00001 41fee4cf-783f-469b-aadf-62c7e2cd33a9 VVKAJ101
##                                UserID RecallNo RecallAttempt RecallStatus
## 1 7bd05142-312e-4648-b1ac-fc258540af52          1              0          2
## 2 7bd05142-312e-4648-b1ac-fc258540af52          1              0          2
## 3 7bd05142-312e-4648-b1ac-fc258540af52          1              0          2
## 4 7bd05142-312e-4648-b1ac-fc258540af52          1              0          2
## 5 7bd05142-312e-4648-b1ac-fc258540af52          1              0          2
## 6 7bd05142-312e-4648-b1ac-fc258540af52          1              0          2
##      IntakeStartDateTime IntakeEndDateTime ReportingDate Lang Occ_No
## 1      11/5/2021 0:00      11/5/2021 23:59      11/6/2021    1      1
## 2      11/5/2021 0:00      11/5/2021 23:59      11/6/2021    1      1
## 3      11/5/2021 0:00      11/5/2021 23:59      11/6/2021    1      1
## 4      11/5/2021 0:00      11/5/2021 23:59      11/6/2021    1      1
## 5      11/5/2021 0:00      11/5/2021 23:59      11/6/2021    1      2
## 6      11/5/2021 0:00      11/5/2021 23:59      11/6/2021    1      3
##                                Occ_Time Occ_Name EatWith WatchTVuseComputer Location FoodNum FoodType
## 1 11/5/2021 7:00          1      <NA>          <NA>          1          1          1
## 2 11/5/2021 7:00          1      <NA>          <NA>          1          2          1
## 3 11/5/2021 7:00          1      <NA>          <NA>          1          3          2
## 4 11/5/2021 7:00          1      <NA>          <NA>          1          4          1
```

## 5	11/5/2021 10:00	7	<NA>	<NA>	1	5	1					
## 6	11/5/2021 12:00	3	<NA>	<NA>	1	6	1					
##	FoodSrce	CodeNum	FoodCode	ModCode	HowMany	SubCode	PortionCode	FoodAmt	KCAL			
## 1	<NA>	1	92400000	0	9.6	0	30001	297.6	124.992			
## 2	<NA>	2	57231250	0	1	0	10205	104	419.12			
## 3	<NA>	3	91302010	0	0.5	0	21000	10.5	31.92			
## 4	<NA>	4	63200200	0	0.5	0	10205	98.5	50.235			
## 5	<NA>	5	94000100	0	8.4	0	30000	252	0			
## 6	<NA>	6	32103000	0	0.5	0	10205	111	285.27			
##	PROT	TFAT	CARB	MOIS	ALC	CAFF	THEO	SUGR	FIBE	CALC		
## 1	0	0.744	30.83136	265.93536	0	26.784	0	29.58144	0	2.976		
## 2	9.36	10.92	75.92	5.408	0	0	0	16.016	10.088	40.56		
## 3	0.0315	0	8.652	1.7955	0	0	0	8.6226	0.021	0.63		
## 4	0.4137	0.6304	11.98745	85.29115	0	0	0	8.32325	2.6595	7.88		
## 5	0	0	0	251.748	0	0	0	0	0	7.56		
## 6	11.3886	25.6854	1.1211	70.9956	0	0	0	1.1211	0	46.62		
##	IRON	MAGN	PHOS	POTA	SODI	ZINC	COPP	SELE	VC	VB1		
## 1	0.05952	0	26.784	14.88	8.928	0.26784	0.020832	0.2976	0	0		
## 2	17.992	101.92	307.84	350.48	298.48	2.392	0.416	43.472	0.208	0.728		
## 3	0.0441	0.21	0.42	5.46	0.42	0.0231	0.00378	0.084	0.0525	0		
## 4	0.1773	4.925	10.835	53.19	0.985	0.06895	0.032505	0.0985	2.4625	0.03152		
## 5	0	2.52	0	0	10.08	0.0252	0.0252	0	0	0		
## 6	1.11	8.88	157.62	116.55	420.69	0.9657	0.01554	27.861	0	0.06105		
##	VB2	NIAC	VB6	FOLA	FA	FF	FDFE	VB12	VARA	RET		
## 1	0	0	0	0	0	0	0	0	0	0		
## 2	0.832	9.984	1.04	199.68	179.92	19.76	326.56	3.016	447.2	447.2		
## 3	0.00399	0.012705	0.00252	0.21	0	0.21	0.21	0	0	0		
## 4	0.036445	0.5122	0.058115	6.895	0	6.895	6.895	0	1.97	0		
## 5	0	0	0	0	0	0	0	0	0	0		
## 6	0.45954	0.05661	0.10878	39.96	0	39.96	39.96	1.0101	135.42	134.31		
##	BCAR	ACAR	CRYP	LYCD	LZ	ATOC	VK	CHOLE	SFAT	S040	S060	S080
## 1	0	0	0	0	0	0	0	0	0	0	0	
## 2	5.2	0	1.04	0	142.48	0.7696	2.288	0	1.04	0	0	0
## 3	0	0	0	0	0	0	0	0	0	0	0	
## 4	27.58	0	0	0	66.98	0.4728	16.154	0	0.052205	0	0	0
## 5	0	0	0	0	0	0	0	0	0	0	0	
## 6	11.1	0	9.99	0	316.35	1.6317	35.631	340.77	5.44344	0	0	0.00222
##	S100	S120	S140	S160	S180	MFAT	M161	M181	M201			
## 1	0	0	0	0	0	0	0	0	0			
## 2	0	0.00208	0.00208	0.80392	0.20592	6.032	0.00312	5.98936	0.03744			
## 3	0	0	0	0	0	0	0	0	0			
## 4	0	0	0	0.030535	0.010835	0.089635	0.00197	0.085695	0			
## 5	0	0	0	0	0	0	0	0	0			
## 6	0.00222	0.00222	0.04329	3.80397	1.40859	7.27938	0.29415	6.882	0.07881			
##	M221	PFAT	P182	P183	P184	P204	P205	P225	P226	VITD		
## 1	0	0	0	0	0	0	0	0	0	0		
## 2	0.00104	3.224	3.05136	0.17264	0	0	0	0	0	1.976		
## 3	0	0	0	0	0	0	0	0	0	0		
## 4	0	0.274815	0.164495	0.11032	0	0	0	0	0	0		
## 5	0	0	0	0	0	0	0	0	0	0		
## 6	0.00777	10.96014	9.55599	1.21545	0	0.14319	0.00444	0	0.03441	1.998		
##	CHOLN	VITE_ADD	B12_ADD	F_TOTAL	F_CITMLB	F_OTHER	F_JUICE	V_TOTAL	V_DRKGR			
## 1	0.8928	0	0	0	0	0	0	0	0			
## 2	29.952	0	3.016	0	0	0	0	0	0			

## 3	0.231	0	0	0	0	0	0	0	0
## 4	5.0235	0	0	0.65995	0.65995	0	0	0	0
## 5	0	0	0	0	0	0	0	0	0
## 6	268.398	0	0	0	0	0	0	0	0
##	V_REDOR_TOTAL	V_REDOR_TOMATO	V_REDOR_OTHER	V_STARCHY_TOTAL	V_STARCHY_POTATO				
## 1	0	0	0	0	0				
## 2	0	0	0	0	0				
## 3	0	0	0	0	0				
## 4	0	0	0	0	0				
## 5	0	0	0	0	0				
## 6	0	0	0	0	0				
##	V_STARCHY_OTHER	V_OTHER	V_LEGUMES	G_TOTAL	G_WHOLE	G_REFINED	PF_TOTAL		
## 1	0	0	0	0	0	0	0		
## 2	0	0	0	2.7456	2.652	0.0936	0.7384		
## 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	1.8093		
##	PF_MPS_TOTAL	PF_MEAT	PF_CUREDMEAT	PF_ORGAN	PF_POULT	PF_SEAFD_HI	PF_SEAFD_LOW		
## 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		
##	PF_EGGS	PF_SOY	PF_NUTSDS	PF_LEGUMES	D_TOTAL	D_MILK	D_YOGURT	D_CHEESE	OILS
## 1	0	0	0	0	0	0	0	0	0
## 2	0	0	0.7384	0	0	0	0	0	6.0736
## 3	0	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0	0
## 6	1.8093	0	0	0	0	0	0	0	15.8508
##	SOLID_FATS	ADD_SUGARS	A_DRINKS	FoodComp					
## 1	0	7.05312	0	1					
## 2	0	3.432	0	1					
## 3	0	2.05275	0	1					
## 4	0	0	0	1					
## 5	0	0	0	1					
## 6	4.8507	0	0	1					
##	Main.food.description								
## 1	Soft_drink_NFS								
## 2	Cereal_Post_Great_Grains_Double_Pecan_Whole_Grain_Cereal_								
## 3	Honey								
## 4	Berries_frozen_NFS								
## 5	Water_tap								
## 6	Egg_salad_made_with_mayonnaise								
##	Old.Main.food.description FoodID								
## 1	Soft drink, NFS 92400000.0								
## 2	Cereal (Post Great Grains Double Pecan Whole Grain Cereal) 57231250.0								
## 3	Honey 91302010.0								
## 4	Berries, frozen, NFS 63200200.0								
## 5	Water, tap 94000100.0								
## 6	Egg salad, made with mayonnaise 32103000.0								

Ensure your items file has the expected dimensions (number of rows x number of columns, shown as number of obs. and number of variables) in the environment window of R Studio. Or by using `dim(items_f_id)` and `dim(items_raw)`. Note that `items_f_id` has 3 more columns than `items_raw` because new columns of FoodID, Old.Main.food.description, and SampleID have been added.

```
dim(items_f_id)
```

```
## [1] 779 133
```

```
dim(items_raw)
```

```
## [1] 779 130
```

Use `individuals_to_remove.txt` to filter out users marked as `Remove = yes`

Load your metadata that has information about which `UserName(s)` to remove.

```
ind_to_rm <- read.delim("individuals_to_remove.txt")
```

Metadata for this purpose (`ind_to_rm`) has `UserName` and which one to be removed.

```
ind_to_rm
```

```
##      UserName Remove
## 1  VVKAJ101
## 2  VVKAJ102
## 3  VVKAJ103
## 4  VVKAJ104
## 5  VVKAJ105
## 6  VVKAJ106
## 7  VVKAJ107
## 8  VVKAJ108
## 9  VVKAJ109
## 10 VVKAJ110
## 11 VVKAJ111
## 12 VVKAJ112
## 13 VVKAJ113
## 14 VVKAJ114
## 15 VVKAJ115
## 16 VVKAJ116      yes
## 17 VVKAJ117
```

Show which has “yes” in the “Remove” column.

```
subset(ind_to_rm, Remove == "yes")
```

```
##      UserName Remove
## 16 VVKAJ116      yes
```

As shown in the console, the user named “VVKAJ116” is marked to be removed. VVKAJ116 has only 1 day of data, which may not be complete, thus it is marked as an individual to remove. However, be careful when deleting a datapoint from your study and never remove individuals from the raw dataset, to ensure you can always go back and include them if desired.

Remove the specified individuals.

The output will be saved as a text file with the specified name. This assumes the usernames are in UserName column, and will print which user(s) will be removed.

```
RemoveRows(data=items_f_id, metadata.file= ind_to_rm,
            output.name= "VVKAJ_Items_f_id_s.txt")
```

```
## 1 row(s) below are to be removed:
##   UserName Remove
## 16 VVKAJ116    yes
```

Load the output for further processing.

```
items_f_id_s <- read.delim("VVKAJ_Items_f_id_s.txt", quote="", colClasses="character")
```

Show unique usernames in items_f_id_s and confirm “VVKAJ116” has been removed.

```
unique(items_f_id_s$UserName)
```

```
## [1] "VVKAJ101" "VVKAJ102" "VVKAJ103" "VVKAJ104" "VVKAJ105" "VVKAJ106"
## [7] "VVKAJ107" "VVKAJ108" "VVKAJ109" "VVKAJ110" "VVKAJ111" "VVKAJ112"
## [13] "VVKAJ113" "VVKAJ114" "VVKAJ115" "VVKAJ117"
```

Merge individuals’ metadata to items

ind_metadata has the participants’ gender, age, height, weight, BMI, and Waist.Circumference, etc. If desired, this individual-specific information can be added to items data.

Load ind_metadata.txt.

```
ind_metadata <- read.table("ind_metadata.txt", sep="\t", header=T)
```

Look at what the metadata has.

```
head(ind_metadata)
```

```
##   UserName      Diet Gender Age Weight Height      BMI Waist.Circumference
## 1 VVKAJ101 Vegetarian    M  31    79    186 22.83501              80
## 2 VVKAJ102    Vegan    F  60    73    163 27.47563              90
## 3 VVKAJ103    Keto    M  43    81    175 26.44898              72
## 4 VVKAJ104 American    F  25    85    169 29.76086              89
## 5 VVKAJ105 Japanese    M  71    60    169 21.00767              75
## 6 VVKAJ106 Vegetarian    F  53    68    159 26.89767              85
```

This includes information on the removed individual, VVKAJ116, but it will not be used if VVKAJ116 is not in the items data.

Add this metadata of each participant to totals or items. 'NA' will be inserted to UserNames which are not in ind_metadata.

```
items_f_id_s_m <- merge(x=items_f_id_s, y=ind_metadata, by="UserName", all.x=T)
```

Check that the items data and metadata are merged.

```
head(items_f_id_s_m)
```

Click to expand output

```
##  UserName      SampleID                      RecallRecId
## 1 VVKAJ101 vvkaj.00001 41fee4cf-783f-469b-aadf-62c7e2cd33a9
## 2 VVKAJ101 vvkaj.00001 41fee4cf-783f-469b-aadf-62c7e2cd33a9
## 3 VVKAJ101 vvkaj.00001 41fee4cf-783f-469b-aadf-62c7e2cd33a9
## 4 VVKAJ101 vvkaj.00001 41fee4cf-783f-469b-aadf-62c7e2cd33a9
## 5 VVKAJ101 vvkaj.00001 41fee4cf-783f-469b-aadf-62c7e2cd33a9
## 6 VVKAJ101 vvkaj.00001 41fee4cf-783f-469b-aadf-62c7e2cd33a9
##                               UserID RecallNo RecallAttempt RecallStatus
## 1 7bd05142-312e-4648-b1ac-fc258540af52      1      0      2
## 2 7bd05142-312e-4648-b1ac-fc258540af52      1      0      2
## 3 7bd05142-312e-4648-b1ac-fc258540af52      1      0      2
## 4 7bd05142-312e-4648-b1ac-fc258540af52      1      0      2
## 5 7bd05142-312e-4648-b1ac-fc258540af52      1      0      2
## 6 7bd05142-312e-4648-b1ac-fc258540af52      1      0      2
##  IntakeStartDateTime IntakeEndDateTime ReportingDate Lang Occ_No
## 1 11/5/2021 0:00 11/5/2021 23:59 11/6/2021 1 1
## 2 11/5/2021 0:00 11/5/2021 23:59 11/6/2021 1 1
## 3 11/5/2021 0:00 11/5/2021 23:59 11/6/2021 1 1
## 4 11/5/2021 0:00 11/5/2021 23:59 11/6/2021 1 1
## 5 11/5/2021 0:00 11/5/2021 23:59 11/6/2021 1 2
## 6 11/5/2021 0:00 11/5/2021 23:59 11/6/2021 1 3
##      Occ_Time Occ_Name EatWith WatchTVUseComputer Location FoodNum FoodType
## 1 11/5/2021 7:00      1 <NA>          <NA>      1      1      1
## 2 11/5/2021 7:00      1 <NA>          <NA>      1      2      1
## 3 11/5/2021 7:00      1 <NA>          <NA>      1      3      2
## 4 11/5/2021 7:00      1 <NA>          <NA>      1      4      1
## 5 11/5/2021 10:00      7 <NA>          <NA>      1      5      1
## 6 11/5/2021 12:00      3 <NA>          <NA>      1      6      1
##  FoodSrce CodeNum FoodCode ModCode HowMany SubCode PortionCode FoodAmt  KCAL
## 1 <NA>      1 92400000      0      9.6      0      30001      297.6 124.992
## 2 <NA>      2 57231250      0      1      0      10205      104 419.12
## 3 <NA>      3 91302010      0      0.5      0      21000      10.5 31.92
## 4 <NA>      4 63200200      0      0.5      0      10205      98.5 50.235
## 5 <NA>      5 94000100      0      8.4      0      30000      252 0
## 6 <NA>      6 32103000      0      0.5      0      10205      111 285.27
##      PROT  TFAT  CARB  MOIS  ALC  CAFF  THEO  SUGR  FIBE  CALC
## 1 0 0.744 30.83136 265.93536 0 26.784 0 29.58144 0 2.976
## 2 9.36 10.92 75.92 5.408 0 0 0 16.016 10.088 40.56
## 3 0.0315 0 8.652 1.7955 0 0 0 8.6226 0.021 0.63
```

## 4	0.4137	0.6304	11.98745	85.29115	0	0	0	8.32325	2.6595	7.88		
## 5	0	0	0	251.748	0	0	0	0	0	7.56		
## 6	11.3886	25.6854	1.1211	70.9956	0	0	0	1.1211	0	46.62		
##	IRON	MAGN	PHOS	POTA	SODI	ZINC	COPP	SELE	VC	VB1		
## 1	0.05952	0	26.784	14.88	8.928	0.26784	0.020832	0.2976	0	0		
## 2	17.992	101.92	307.84	350.48	298.48	2.392	0.416	43.472	0.208	0.728		
## 3	0.0441	0.21	0.42	5.46	0.42	0.0231	0.00378	0.084	0.0525	0		
## 4	0.1773	4.925	10.835	53.19	0.985	0.06895	0.032505	0.0985	2.4625	0.03152		
## 5	0	2.52	0	0	10.08	0.0252	0.0252	0	0	0		
## 6	1.11	8.88	157.62	116.55	420.69	0.9657	0.01554	27.861	0	0.06105		
##	VB2	NIAC	VB6	FOLA	FA	FF	FDFE	VB12	VARA	RET		
## 1	0	0	0	0	0	0	0	0	0	0		
## 2	0.832	9.984	1.04	199.68	179.92	19.76	326.56	3.016	447.2	447.2		
## 3	0.00399	0.012705	0.00252	0.21	0	0.21	0.21	0	0	0		
## 4	0.036445	0.5122	0.058115	6.895	0	6.895	6.895	0	1.97	0		
## 5	0	0	0	0	0	0	0	0	0	0		
## 6	0.45954	0.05661	0.10878	39.96	0	39.96	39.96	1.0101	135.42	134.31		
##	BCAR	ACAR	CRYP	LYCO	LZ	ATOC	VK	CHOLE	SFAT	S040	S060	S080
## 1	0	0	0	0	0	0	0	0	0	0	0	0
## 2	5.2	0	1.04	0	142.48	0.7696	2.288	0	1.04	0	0	0
## 3	0	0	0	0	0	0	0	0	0	0	0	0
## 4	27.58	0	0	0	66.98	0.4728	16.154	0	0.052205	0	0	0
## 5	0	0	0	0	0	0	0	0	0	0	0	0
## 6	11.1	0	9.99	0	316.35	1.6317	35.631	340.77	5.44344	0	0	0.00222
##	S100	S120	S140	S160	S180	MFAT	M161	M181	M201			
## 1	0	0	0	0	0	0	0	0	0	0	0	0
## 2	0	0.00208	0.00208	0.80392	0.20592	6.032	0.00312	5.98936	0.03744			
## 3	0	0	0	0	0	0	0	0	0	0	0	0
## 4	0	0	0	0.030535	0.010835	0.089635	0.00197	0.085695				0
## 5	0	0	0	0	0	0	0	0	0	0	0	0
## 6	0.00222	0.00222	0.04329	3.80397	1.40859	7.27938	0.29415	6.882	0.07881			
##	M221	PFAT	P182	P183	P184	P204	P205	P225	P226	VITD		
## 1	0	0	0	0	0	0	0	0	0	0	0	0
## 2	0.00104	3.224	3.05136	0.17264	0	0	0	0	0	1.976		
## 3	0	0	0	0	0	0	0	0	0	0	0	0
## 4	0	0.274815	0.164495	0.11032	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0	0	0	0	0
## 6	0.00777	10.96014	9.55599	1.21545	0	0.14319	0.00444	0	0.03441	1.998		
##	CHOLN	VITE_ADD	B12_ADD	F_TOTAL	F_CITMLB	F_OTHER	F_JUICE	V_TOTAL	V_DRKGR			
## 1	0.8928	0	0	0	0	0	0	0	0	0	0	0
## 2	29.952	0	3.016	0	0	0	0	0	0	0	0	0
## 3	0.231	0	0	0	0	0	0	0	0	0	0	0
## 4	5.0235	0	0	0.65995	0.65995	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0	0	0	0	0
## 6	268.398	0	0	0	0	0	0	0	0	0	0	0
##	V_REDOR_TOTAL	V_REDOR_TOMATO	V_REDOR_OTHER	V_STARCHY_TOTAL	V_STARCHY_POTATO							
## 1	0	0	0	0	0							
## 2	0	0	0	0	0							
## 3	0	0	0	0	0							
## 4	0	0	0	0	0							
## 5	0	0	0	0	0							
## 6	0	0	0	0	0							
##	V_STARCHY_OTHER	V_OTHER	V_LEGUMES	G_TOTAL	G_WHOLE	G_REFINED	PF_TOTAL					
## 1	0	0	0	0	0	0	0					

```

## 2      0      0      0 2.7456 2.652 0.0936 0.7384
## 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 1.8093
## PF_MPS_TOTAL PF_MEAT PF_CUREDMEAT PF_ORGAN PF_POULT PF_SEAFD_HI PF_SEAFD_LOW
## 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
## PF_EGGS PF_SOY PF_NUTSDS PF_LEGUMES D_TOTAL D_MILK D_YOGURT D_CHEESE OILS
## 1      0      0      0      0      0      0      0      0
## 2      0      0 0.7384      0      0      0      0      0 6.0736
## 3      0      0      0      0      0      0      0      0
## 4      0      0      0      0      0      0      0      0
## 5      0      0      0      0      0      0      0      0
## 6 1.8093      0      0      0      0      0      0      0 15.8508
## SOLID_FATS ADD_SUGARS A_DRINKS FoodComp
## 1      0 7.05312      0      1
## 2      0 3.432      0      1
## 3      0 2.05275      0      1
## 4      0      0      0      1
## 5      0      0      0      1
## 6 4.8507      0      0      1
## Main.food.description
## 1 Soft_drink_NFS
## 2 Cereal_Post_Great_Grains_Double_Pecan_Whole_Grain_Cereal_
## 3 Honey
## 4 Berries_frozen_NFS
## 5 Water_tap
## 6 Egg_salad_made_with_mayonnaise
## Old.Main.food.description FoodID
## 1 Soft drink, NFS 92400000.0
## 2 Cereal (Post Great Grains Double Pecan Whole Grain Cereal) 57231250.0
## 3 Honey 91302010.0
## 4 Berries, frozen, NFS 63200200.0
## 5 Water, tap 94000100.0
## 6 Egg salad, made with mayonnaise 32103000.0
## Diet Gender Age Weight Height BMI Waist.Circumference
## 1 Vegetarian M 31 79 186 22.83501 80
## 2 Vegetarian M 31 79 186 22.83501 80
## 3 Vegetarian M 31 79 186 22.83501 80
## 4 Vegetarian M 31 79 186 22.83501 80
## 5 Vegetarian M 31 79 186 22.83501 80
## 6 Vegetarian M 31 79 186 22.83501 80

```

Furthermore, as a quick way to look at the metadata of only the selected individuals, you can subset the metadata to just the usernames present in the analysis dataset (items_f_id_s) using the %in% operator.

```
ind_metadata_s <- ind_metadata[ind_metadata$username %in% items_f_id_s$username, ]
```

Use the tail function to show the last six rows of ind_metadata_s. You can see that the last individual in

this metadata is now VVKAJ117, and that VVKAJ116, which was not in items_f_id_s, has been omitted.

```
tail(ind_metadata_s)
```

Click to expand output

```
##      UserName      Diet Gender Age  Weight Height      BMI Waist.Circumference
## 11 VVKAJ111 Vegetarian      M  60 74.9000  164.1 27.81408              78.0
## 12 VVKAJ112      Vegan      F  48 69.8992  158.0 28.00000              95.5
## 13 VVKAJ113      Keto      M  43 80.0000  183.0 23.88844              98.0
## 14 VVKAJ114 American      F  62 90.7235  161.0 35.00000             110.0
## 15 VVKAJ115 Japanese      M  22 69.6348  174.0 23.00000              70.0
## 17 VVKAJ117      error      M  23 76.0000  175.0 24.81633              80.0
```

Save the merged dataframe as a .txt file.

```
write.table(items_f_id_s_m, "VVKAJ_Items_f_id_s_m.txt", sep="\t", row.names=F, quote=F)
```

Generate new totals file from the items file

Use one of the input files saved above as an input for calculating totals for. Specify which columns have usernames and Recall.No., which has the recorded days.

```
GenerateTotals(inputfn = "VVKAJ_Items_f_id_s_m.txt",
               User.Name = "UserName",
               Recall.No = "RecallNo",
               outfn = "VVKAJ_Tot.txt")
```

Load the total file generated above.

```
new_totals <- read.table("VVKAJ_Tot.txt", header=T, sep="\t")
```

The number of rows should be {No. of users x No. days}. For the example data, 16 users x 3 days = 48 rows (observations).

```
nrow(new_totals)
```

```
## [1] 48
```

View the new_totals.

```
head(new_totals)
```

Click to expand output

```
##      User_Day UserName RecallNo FoodAmt      KCAL      PROT      TFAT      CARB
## 1 VVKAJ101_1 VVKAJ101          1 2402.500 2314.9040 80.15720 82.97721 337.7973
## 2 VVKAJ101_2 VVKAJ101          2 1671.650  913.5955 35.13461 37.42791 118.5702
## 3 VVKAJ101_3 VVKAJ101          3 1920.588 1604.8050 48.29881 70.08057 215.7318
```

##	4	VVKAJ102_1	VVKAJ102	1	1957.475	1440.1995	38.54926	80.38863	152.8296	
##	5	VVKAJ102_2	VVKAJ102	2	1858.067	1508.9957	46.25199	56.89753	212.5225	
##	6	VVKAJ102_3	VVKAJ102	3	2049.400	1834.2306	58.89500	87.53381	211.8413	
##		MOIS	ALC	CAFF	THEO	SUGR	FIBE	CALC	IRON	MAGN
##	1	1879.619	0	26.784	0.00	144.20847	52.62550	889.8455	31.282000	589.3865
##	2	1467.035	0	76.800	0.00	47.67748	21.04413	619.2521	7.366001	273.2332
##	3	1572.133	0	30.240	3.36	111.81870	23.27856	1427.6826	14.628896	320.9569
##	4	1669.286	0	50.400	5.04	25.75794	33.90378	912.4538	11.797247	284.4450
##	5	1521.262	0	22.680	2.52	55.29288	31.25067	1144.7673	14.180575	366.8420
##	6	1666.272	0	0.000	0.00	31.41001	29.35169	700.0389	14.009471	373.1694
##		PHOS	POTA	SODI	ZINC	COPP	SELE	VC	VB1	
##	1	1787.4540	4615.400	2298.662	11.776515	1.935984	113.91525	209.0321	1.882480	
##	2	794.2740	2093.175	1058.073	5.470231	1.325496	58.42284	147.7427	0.983826	
##	3	1249.7137	2571.690	2069.723	8.953749	1.565804	83.81562	137.9377	1.316055	
##	4	808.2387	2974.354	2550.088	5.641533	1.278044	42.76070	179.4785	1.137953	
##	5	1043.4290	4686.336	2840.423	6.408068	1.427889	38.26618	157.1404	2.909776	
##	6	1164.2706	4653.953	4285.981	6.745011	2.066325	72.05487	117.7118	4.570952	
##		VB2	NIAC	VB6	FOLA	FA	FF	FDFE	VB12	
##	1	2.9713065	18.72793	3.829823	719.3660	179.920	539.4460	846.2460	5.17202	
##	2	1.2894892	12.24912	1.386882	372.7493	85.070	287.6793	432.4393	2.23350	
##	3	2.0125987	14.49594	4.120610	918.5919	601.800	316.7919	1340.1919	13.23145	
##	4	0.9494982	10.92889	1.637870	459.3215	52.025	410.3265	495.5415	1.80990	
##	5	1.3127483	13.76033	2.560617	534.1780	20.200	515.9980	548.3180	3.97461	
##	6	1.2672856	18.25922	2.540221	506.4956	26.700	481.8156	525.2356	3.70270	
##		VARA	RET	BCAR	ACAR	CRYP	LYCO	LZ	ATOC	
##	1	1261.6665	673.1850	7004.547	106.7165	56.11800	13.350	14273.582	9.851310	
##	2	1412.8761	82.3500	13464.676	4921.4081	81.65188	1174.626	2444.209	4.628156	
##	3	1111.7093	504.9250	6476.761	1614.4750	19.37550	4.704	5437.084	16.765260	
##	4	310.6182	165.1115	1551.208	344.1073	68.67350	7582.125	1340.894	18.987955	
##	5	1265.3238	222.6810	10784.821	3226.1183	198.74500	3380.685	4032.658	16.267538	
##	6	344.2824	144.1200	2307.391	143.9169	12.66863	3374.064	3388.637	9.870323	
##		VK	CHOLE	SFAT	S040	S060	S080	S100	S120	
##	1	674.6513	375.967	18.696057	0.310902	0.11065100	0.1176030	0.2303450	0.18606100	
##	2	176.1320	18.300	8.035444	0.137250	0.13725000	0.1401794	0.1440906	0.16560063	
##	3	356.2514	64.265	21.762310	0.461520	0.35581000	0.6559880	0.7907570	3.08172600	
##	4	193.4643	63.277	12.393455	0.063991	0.04301325	0.0834395	0.1062550	0.06083225	
##	5	271.3043	53.980	10.073571	0.086720	0.06900500	0.2074300	0.2042550	0.43357300	
##	6	270.4181	65.830	17.110502	0.318160	0.16680000	0.2767274	0.3239814	0.26321963	
##		S140	S160	S180	MFAT	M161	M181	M201	M221	
##	1	1.2467585	11.448329	4.130060	28.14091	0.7393160	26.99568	0.2349650	0.00881000	
##	2	0.5821610	4.955292	1.532696	15.32206	0.3639200	14.79923	0.1110759	0.00346125	
##	3	2.6981170	9.237105	3.415553	24.67066	0.4838810	23.65191	0.1867220	0.00864600	
##	4	0.1955330	8.961791	2.439567	34.62131	0.9667318	33.33719	0.2755755	0.00450000	
##	5	0.5196547	6.309740	1.897680	22.20331	0.4462590	21.47338	0.2112680	0.02609700	
##	6	0.8637470	10.355075	3.764050	33.49530	0.3996655	32.60066	0.3411139	0.00658725	
##		PFAT	P182	P183	P184	P204	P205	P225	P226	VITD
##	1	28.95861	25.201469	3.525870	0	0.157937	0.00444	0.00000	0.03441	4.881200
##	2	10.24234	9.135821	1.092069	0	0.001315	0.00000	0.00000	0.00000	2.379000
##	3	18.75467	16.930229	1.803171	0	0.006573	0.00000	0.00000	0.00000	10.193000
##	4	28.41416	24.796558	3.534361	0	0.038890	0.00000	0.00000	0.00909	3.375425
##	5	20.44748	17.914195	2.412754	0	0.026967	0.00000	0.00000	0.01881	4.985000
##	6	31.17042	27.616109	3.344330	0	0.038623	0.00032	0.00208	0.00670	2.102000
##		CHOLN	VITE_ADD	B12_ADD	F_TOTAL	F_CITMLB	F_OTHER	F_JUICE	V_TOTAL	
##	1	572.5637	0.00000	3.01600	4.037445	0.65995	3.035495	0.3420	3.656365	

```

## 2 143.2537 0.00000 1.41000 0.640500 0.30625 0.323750 0.0105 2.727494
## 3 195.7137 8.34700 10.79370 1.530100 1.01160 0.505000 0.0135 1.906206
## 4 182.4953 6.85640 1.53720 0.000000 0.00000 0.000000 0.0000 4.280535
## 5 212.2748 7.19922 3.21321 0.000000 0.00000 0.000000 0.0000 7.783067
## 6 219.7563 0.00000 2.83410 0.000000 0.00000 0.000000 0.0000 7.084731
##      V_DRKGR V_REDOR_TOTAL V_REDOR_TOMATO V_REDOR_OTHER V_STARCHY_TOTAL
## 1 1.6658400      0.000000      0.00000      0.000000      0.0000000
## 2 0.4840000      1.177100      0.04860      1.128500      0.0000000
## 3 1.0143750      0.300500      0.00000      0.300500      0.0000000
## 4 0.0099900      1.485350      1.48535      0.000000      0.1288625
## 5 0.6186667      1.857575      0.84840      1.009175      3.3200000
## 6 0.7412500      0.848400      0.84840      0.000000      4.6146000
##      V_STARCHY_POTATO V_STARCHY_OTHER      V_OTHER V_LEGUMES G_TOTAL G_WHOLE
## 1      0.0000      0.0000000 1.9905250 1.256750 5.4384 2.6520
## 2      0.0000      0.0000000 1.0663938 0.105000 2.6500 1.7669
## 3      0.0000      0.0000000 0.5913312 0.135000 4.4072 4.2542
## 4      0.0000      0.1288625 2.6563325 1.085300 5.0421 1.5522
## 5      3.3200      0.0000000 1.9868250 0.285000 2.2405 1.0285
## 6      4.6146      0.0000000 0.8804812 0.192375 1.6050 0.0000
##      G_REFINED PF_TOTAL PF_MPS_TOTAL PF_MEAT PF_CUREDMEAT PF_ORGAN PF_POULT
## 1      2.7864 3.111700      0.0000      0      0.0000      0      0
## 2      0.8831 3.316731      0.0000      0      0.0000      0      0
## 3      0.1530 2.501000      0.0000      0      0.0000      0      0
## 4      3.4899 0.644600      0.0000      0      0.0000      0      0
## 5      1.2120 2.226680      0.0000      0      0.0000      0      0
## 6      1.6050 4.697931      0.2016      0      0.2016      0      0
##      PF_SEAFD_HI PF_SEAFD_LOW PF_EGGS PF_SOY PF_NUTSDS PF_LEGUMES D_TOTAL D_MILK
## 1      0      0 1.8093 0.0000 1.3024000 4.935150 1.34064 0.21168
## 2      0      0 0.0000 1.9200 1.3967312 0.420000 0.75030 0.75030
## 3      0      0 0.0000 0.0000 2.5010000 0.540000 2.39525 0.88850
## 4      0      0 0.3030 0.0000 0.3416000 4.351050 0.15150 0.15150
## 5      0      0 0.2020 1.6660 0.3586800 1.155000 0.57980 0.57980
## 6      0      0 0.2020 3.3558 0.9385313 0.772875 0.70760 0.58900
##      D_YOGURT D_CHEESE      OILS SOLID_FATS ADD_SUGARS A_DRINKS
## 1 0.00000 0.5544 46.76440 14.582140 12.53787      0
## 2 0.00000 0.0000 19.10275 4.932700 3.95306      0
## 3 1.50675 0.0000 36.48778 18.798000 10.32484      0
## 4 0.00000 0.0000 69.07104 2.439577 1.72710      0
## 5 0.00000 0.0000 44.07012 3.130800 5.21406      0
## 6 0.00000 0.0140 65.58897 9.271000 0.80064      0

```

Add the participants' metadata back to totals

Load `ind_metadata.txt` if you have not done so.

```
ind_metadata <- read.table("ind_metadata.txt", sep="\t", header=T)
```

Add this metadata of each participant to totals. 'NA' will be inserted to UserNames which are not in `ind_metadata`.

```
new_totals_m <- merge(x=new_totals, y=ind_metadata, by="UserName", all.x=T)
```


Check that the items data and metadata are merged.

```
head(new_totals_m)
```

Click to expand output

```
##   UserName   User_Day RecallNo  FoodAmt      KCAL      PROT      TFAT      CARB
## 1 VVKAJ101 VVKAJ101_1      1 2402.500 2314.9040 80.15720 82.97721 337.7973
## 2 VVKAJ101 VVKAJ101_2      2 1671.650  913.5955 35.13461 37.42791 118.5702
## 3 VVKAJ101 VVKAJ101_3      3 1920.588 1604.8050 48.29881 70.08057 215.7318
## 4 VVKAJ102 VVKAJ102_1      1 1957.475 1440.1995 38.54926 80.38863 152.8296
## 5 VVKAJ102 VVKAJ102_2      2 1858.067 1508.9957 46.25199 56.89753 212.5225
## 6 VVKAJ102 VVKAJ102_3      3 2049.400 1834.2306 58.89500 87.53381 211.8413
##           MOIS  ALC   CAFF  THEO      SUGR   FIBE    CALC    IRON   MAGN
## 1 1879.619    0 26.784 0.00 144.20847 52.62550 889.8455 31.282000 589.3865
## 2 1467.035    0 76.800 0.00 47.67748 21.04413 619.2521 7.366001 273.2332
## 3 1572.133    0 30.240 3.36 111.81870 23.27856 1427.6826 14.628896 320.9569
## 4 1669.286    0 50.400 5.04 25.75794 33.90378 912.4538 11.797247 284.4450
## 5 1521.262    0 22.680 2.52 55.29288 31.25067 1144.7673 14.180575 366.8420
## 6 1666.272    0  0.000 0.00 31.41001 29.35169 700.0389 14.009471 373.1694
##           PHOS    POTA   SODI      ZINC    COPP    SELE    VC    VB1
## 1 1787.4540 4615.400 2298.662 11.776515 1.935984 113.91525 209.0321 1.882480
## 2 794.2740 2093.175 1058.073 5.470231 1.325496 58.42284 147.7427 0.983826
## 3 1249.7137 2571.690 2069.723 8.953749 1.565804 83.81562 137.9377 1.316055
## 4 808.2387 2974.354 2550.088 5.641533 1.278044 42.76070 179.4785 1.137953
## 5 1043.4290 4686.336 2840.423 6.408068 1.427889 38.26618 157.1404 2.909776
## 6 1164.2706 4653.953 4285.981 6.745011 2.066325 72.05487 117.7118 4.570952
##           VB2    NIAC    VB6    FOLA    FA    FF    FDFE    VB12
## 1 2.9713065 18.72793 3.829823 719.3660 179.920 539.4460 846.2460 5.17202
## 2 1.2894892 12.24912 1.386882 372.7493 85.070 287.6793 432.4393 2.23350
## 3 2.0125987 14.49594 4.120610 918.5919 601.800 316.7919 1340.1919 13.23145
## 4 0.9494982 10.92889 1.637870 459.3215 52.025 410.3265 495.5415 1.80990
## 5 1.3127483 13.76033 2.560617 534.1780 20.200 515.9980 548.3180 3.97461
## 6 1.2672856 18.25922 2.540221 506.4956 26.700 481.8156 525.2356 3.70270
##           VARA    RET    BCAR    ACAR    CRYP    LYCO    LZ    ATOC
## 1 1261.6665 673.1850 7004.547 106.7165 56.11800 13.350 14273.582 9.851310
## 2 1412.8761 82.3500 13464.676 4921.4081 81.65188 1174.626 2444.209 4.628156
## 3 1111.7093 504.9250 6476.761 1614.4750 19.37550 4.704 5437.084 16.765260
## 4 310.6182 165.1115 1551.208 344.1073 68.67350 7582.125 1340.894 18.987955
## 5 1265.3238 222.6810 10784.821 3226.1183 198.74500 3380.685 4032.658 16.267538
## 6 344.2824 144.1200 2307.391 143.9169 12.66863 3374.064 3388.637 9.870323
##           VK   CHOLE    SFAT    S040    S060    S080    S100    S120
## 1 674.6513 375.967 18.696057 0.310902 0.11065100 0.1176030 0.2303450 0.18606100
## 2 176.1320 18.300 8.035444 0.137250 0.13725000 0.1401794 0.1440906 0.16560063
## 3 356.2514 64.265 21.762310 0.461520 0.35581000 0.6559880 0.7907570 3.08172600
## 4 193.4643 63.277 12.393455 0.063991 0.04301325 0.0834395 0.1062550 0.06083225
## 5 271.3043 53.980 10.073571 0.086720 0.06900500 0.2074300 0.2042550 0.43357300
## 6 270.4181 65.830 17.110502 0.318160 0.16680000 0.2767274 0.3239814 0.26321963
##           S140    S160    S180    MFAT    M161    M181    M201    M221
## 1 1.2467585 11.448329 4.130060 28.14091 0.7393160 26.99568 0.2349650 0.00881000
## 2 0.5821610 4.955292 1.532696 15.32206 0.3639200 14.79923 0.1110759 0.00346125
## 3 2.6981170 9.237105 3.415553 24.67066 0.4838810 23.65191 0.1867220 0.00864600
## 4 0.1955330 8.961791 2.439567 34.62131 0.9667318 33.33719 0.2755755 0.00450000
## 5 0.5196547 6.309740 1.897680 22.20331 0.4462590 21.47338 0.2112680 0.02609700
```

## 6	0.8637470	10.355075	3.764050	33.49530	0.3996655	32.60066	0.3411139	0.00658725	
##	PFAT	P182	P183	P184	P204	P205	P225	P226	VITD
## 1	28.95861	25.201469	3.525870	0	0.157937	0.00444	0.00000	0.03441	4.881200
## 2	10.24234	9.135821	1.092069	0	0.001315	0.00000	0.00000	0.00000	2.379000
## 3	18.75467	16.930229	1.803171	0	0.006573	0.00000	0.00000	0.00000	10.193000
## 4	28.41416	24.796558	3.534361	0	0.038890	0.00000	0.00000	0.00909	3.375425
## 5	20.44748	17.914195	2.412754	0	0.026967	0.00000	0.00000	0.01881	4.985000
## 6	31.17042	27.616109	3.344330	0	0.038623	0.00032	0.00208	0.00670	2.102000
##	CHOLN	VITE_ADD	B12_ADD	F_TOTAL	F_CITMLB	F_OTHER	F_JUICE	V_TOTAL	
## 1	572.5637	0.00000	3.01600	4.037445	0.65995	3.035495	0.3420	3.656365	
## 2	143.2537	0.00000	1.41000	0.640500	0.30625	0.323750	0.0105	2.727494	
## 3	195.7137	8.34700	10.79370	1.530100	1.01160	0.505000	0.0135	1.906206	
## 4	182.4953	6.85640	1.53720	0.000000	0.00000	0.000000	0.0000	4.280535	
## 5	212.2748	7.19922	3.21321	0.000000	0.00000	0.000000	0.0000	7.783067	
## 6	219.7563	0.00000	2.83410	0.000000	0.00000	0.000000	0.0000	7.084731	
##	V_DRKGR	V_REDOR_TOTAL	V_REDOR_TOMATO	V_REDOR_OTHER	V_STARCHY_TOTAL				
## 1	1.6658400	0.000000	0.00000	0.000000	0.0000000				
## 2	0.4840000	1.177100	0.04860	1.128500	0.0000000				
## 3	1.0143750	0.300500	0.00000	0.300500	0.0000000				
## 4	0.0099900	1.485350	1.48535	0.000000	0.1288625				
## 5	0.6186667	1.857575	0.84840	1.009175	3.3200000				
## 6	0.7412500	0.848400	0.84840	0.000000	4.6146000				
##	V_STARCHY_POTATO	V_STARCHY_OTHER	V_OTHER	V_LEGUMES	G_TOTAL	G_WHOLE			
## 1	0.0000	0.0000000	1.9905250	1.256750	5.4384	2.6520			
## 2	0.0000	0.0000000	1.0663938	0.105000	2.6500	1.7669			
## 3	0.0000	0.0000000	0.5913312	0.135000	4.4072	4.2542			
## 4	0.0000	0.1288625	2.6563325	1.085300	5.0421	1.5522			
## 5	3.3200	0.0000000	1.9868250	0.285000	2.2405	1.0285			
## 6	4.6146	0.0000000	0.8804812	0.192375	1.6050	0.0000			
##	G_REFINED	PF_TOTAL	PF_MPS_TOTAL	PF_MEAT	PF_CUREDMEAT	PF_ORGAN	PF_POULT		
## 1	2.7864	3.111700	0.0000	0	0.0000	0	0		
## 2	0.8831	3.316731	0.0000	0	0.0000	0	0		
## 3	0.1530	2.501000	0.0000	0	0.0000	0	0		
## 4	3.4899	0.644600	0.0000	0	0.0000	0	0		
## 5	1.2120	2.226680	0.0000	0	0.0000	0	0		
## 6	1.6050	4.697931	0.2016	0	0.2016	0	0		
##	PF_SEAFD_HI	PF_SEAFD_LOW	PF_EGGS	PF_SOY	PF_NUTSDS	PF_LEGUMES	D_TOTAL	D_MILK	
## 1	0	0	1.8093	0.0000	1.3024000	4.935150	1.34064	0.21168	
## 2	0	0	0.0000	1.9200	1.3967312	0.420000	0.75030	0.75030	
## 3	0	0	0.0000	0.0000	2.5010000	0.540000	2.39525	0.88850	
## 4	0	0	0.3030	0.0000	0.3416000	4.351050	0.15150	0.15150	
## 5	0	0	0.2020	1.6660	0.3586800	1.155000	0.57980	0.57980	
## 6	0	0	0.2020	3.3558	0.9385313	0.772875	0.70760	0.58900	
##	D_YOGURT	D_CHEESE	OILS	SOLID_FATS	ADD_SUGARS	A_DRINKS	Diet	Gender	
## 1	0.00000	0.5544	46.76440	14.582140	12.53787	0	Vegetarian	M	
## 2	0.00000	0.0000	19.10275	4.932700	3.95306	0	Vegetarian	M	
## 3	1.50675	0.0000	36.48778	18.798000	10.32484	0	Vegetarian	M	
## 4	0.00000	0.0000	69.07104	2.439577	1.72710	0	Vegan	F	
## 5	0.00000	0.0000	44.07012	3.130800	5.21406	0	Vegan	F	
## 6	0.00000	0.0140	65.58897	9.271000	0.80064	0	Vegan	F	
##	Age	Weight	Height	BMI	Waist.Circumference				
## 1	31	79	186	22.83501	80				
## 2	31	79	186	22.83501	80				
## 3	31	79	186	22.83501	80				

```
## 4 60 73 163 27.47563 90
## 5 60 73 163 27.47563 90
## 6 60 73 163 27.47563 90
```

Save the merged dataframe as a .txt file.

```
write.table(new_totals_m, "VVKAJ_Tot_m.txt", sep="\t", row.names=F, quote=F)
```

Calculate the mean of totals/participant

Calculate the mean of the totals data across all the days for each participant.

```
AverageBy(data= new_totals, by= "UserName", start.col= "FoodAmt", end.col= "A_DRINKS",
           outfn="VVKAJ_Tot_mean.txt")
```

Load the output for further processing.

```
new_totals_mean <- read.table("VVKAJ_Tot_mean.txt", header=T, sep="\t")
```

The number of rows should be equal to the number of users. This example data has 16 users, so there should be 16 rows of mean totals.

```
nrow(new_totals_mean)
```

```
## [1] 16
```

Add the participants' metadata to the mean totals

Load ind_metadata.txt if you have not done so.

```
ind_metadata <- read.table("ind_metadata.txt", sep="\t", header=T)
```

Add this metadata of each participant in the mean totals. 'NA' will be inserted to UserNames which are not in ind_metadata.

```
new_totals_mean_m <- merge(x=new_totals_mean, y=ind_metadata, by="UserName", all.x=T)
```

Check that the mean totals and the users' metadata are merged.

```
head(new_totals_mean_m, 1)
```

Click to expand output

```
##   UserName  FoodAmt      KCAL      PROT      TFAT      CARB      MOIS ALC      CAFF THEO
## 1 VVKAJ101 1998.246 1611.101 54.5302 63.49523 224.0331 1639.596 0 44.608 1.12
##      SUGR      FIBE      CALC      IRON      MAGN      PHOS      POTA      SODI
## 1 101.2349 32.31606 978.9267 17.75897 394.5255 1277.147 3093.422 1808.819
##      ZINC      COPP      SELE      VC      VB1      VB2      NIAC      VB6
## 1 8.733498 1.609095 85.38457 164.9042 1.39412 2.091131 15.15766 3.112438
##      FOLA      FA      FF      FDFE      VB12      VARA      RET      BCAR      ACAR
## 1 670.2357 288.93 381.3057 872.959 6.87899 1262.084 420.1533 8981.995 2214.2
##      CRYP      LYCO      LZ      ATOC      VK      CHOLE      SFAT      S040
## 1 52.38179 397.5601 7384.958 10.41491 402.3449 152.844 16.1646 0.303224
##      S060      S080      S100      S120      S140      S160      S180      MFAT
## 1 0.201237 0.3045901 0.3883975 1.144463 1.509012 8.546908 3.026103 22.71121
##      M161      M181      M201      M221      PFAT      P182      P183 P184
## 1 0.529039 21.81561 0.1775876 0.006972417 19.31854 17.08917 2.14037 0
##      P204      P205 P225      P226      VITD      CHOLN VITE_ADD B12_ADD F_TOTAL
## 1 0.055275 0.00148 0 0.01147 5.817733 303.8437 2.782333 5.073233 2.069348
##      F_CITMLB F_OTHER F_JUICE V_TOTAL V_DRKGR V_REDOR_TOTAL V_REDOR_TOMATO
## 1 0.6592667 1.288082 0.122 2.763355 1.054738 0.4925333 0.0162
##      V_REDOR_OTHER V_STARCHY_TOTAL V_STARCHY_POTATO V_STARCHY_OTHER V_OTHER
## 1 0.4763333 0 0 0 1.216083
##      V_LEGUMES G_TOTAL G_WHOLE G_REFINED PF_TOTAL PF_MPS_TOTAL PF_MEAT
## 1 0.4989167 4.1652 2.891033 1.274167 2.976477 0 0
##      PF_CUREDMEAT PF_ORGAN PF_POULT PF_SEAFD_HI PF_SEAFD_LOW PF_EGGS PF_SOY
## 1 0 0 0 0 0.6031 0.64
##      PF_NUTSDS PF_LEGUMES D_TOTAL D_MILK D_YOGURT D_CHEESE OILS SOLID_FATS
## 1 1.733377 1.96505 1.495397 0.6168267 0.50225 0.1848 34.11831 12.77095
##      ADD_SUGARS A_DRINKS Diet Gender Age Weight Height BMI
## 1 8.93859 0 Vegetarian M 31 79 186 22.83501
##      Waist.Circumference
## 1 80
```

Save the merged dataframe as a .txt file.

```
write.table(new_totals_mean_m, "VVKAJ_Tot_mean_m.txt", sep="\t", row.names=F, quote=F)
```

Quality Control (QC) for the mean totals data

Totals data may contain outliers due to errors in dietary reporting. These errors may be due to omission or inaccurate over- or under-estimation of portion size, leading to improbable nutrient totals. ASA24 provides General Guidelines for Reviewing & Cleaning Data for identifying and removing suspicious records.

Here, we will identify records that contain values that fall outside typically observed ranges of kilocalories (KCAL), protein (PROT), total fat (TFAT), and vitamin C (VC). The ASA24 guide provides ranges of beta carotene (BCAR), too, however, outlier checking for BCAR is omitted in this tutorial but can be considered if you identify it as a nutrient that has a high variance in your study dataset.

Please note that your input dataframe (QCtotals) will be overwritten after each outlier removal.

Load your totals if necessary - to be used as input for QC.

```
new_totals_mean_m <- read.table("VVKAJ_Tot_mean_m.txt", sep="\t", header=T)
```

Define your totals dataset to be used as input.

```
QCtotals <- new_totals_mean_m
```

Flag if KCAL is <600 or >5700 → ask remove or not → if yes, remove those rows.

```
QCOutliers(input.data = QCtotals, target.colname = "KCAL", min = 600, max = 5700)
```

This function will print out rows that fall outside the specified min-max range, and a dialogue box will appear outside the R Studio (shown below), asking whether to remove them. You should make sure to review these records carefully to double-check if the removal is warranted. It is possible to have a valid record that could meet the threshold for removal. Only you will know if you can trust the record when working with real data.



If you find potential outlier(s) here, click “No”, and view those total(s) with their other nutrient intake information by running the following;

```
KCAL_outliers <- subset(QCtotals, KCAL < 600 | KCAL > 5700)
```

```
# Sort the rows by KCAL and show only the specified variables.
KCAL_outliers[order(KCAL_outliers$KCAL, decreasing = T),
               c('UserName', 'KCAL', 'FoodAmt', 'PROT', 'TFAT', 'CARB')]
```

```
##   UserName      KCAL  FoodAmt    PROT    TFAT     CARB
## 16 VVKAJ117 5789.476 5639.525 194.1051 263.8755 674.0867
```

If you think it is a true outlier, then run the QCOutliers command for KCAL again, and click “Yes” to remove the outlier. Here for this tutorial, we will remove this individual.

```
QCOutliers(input.data = QCtotals, target.colname = "KCAL", min = 600, max = 5700)
```

```
## There are 1 observations with < 600 or > 5700 .
## Remove? (Yes/no/cancel)
## Outlier rows were removed; the cleaned data is saved as an object called "QCtotals".
## 15 rows remained.
```

Continue the QC process with other variables.

Flag if PROT is <10 or >240 → ask remove or not → if yes, remove those rows

```
QCOutliers(input.data = QCtotals, target.colname = "PROT", min = 10, max = 240)
```

```
## There are 0 observations with < 10 or > 240 .  
## There are no outlier rows, but the input data was renamed as QCtotals.  
## 15 rows remained.
```

Flag if TFAT is <15 or >230 → ask remove or not → if yes, remove those rows

```
QCOutliers(input.data = QCtotals, target.colname = "TFAT", min = 15, max = 230)
```

```
## There are 0 observations with < 15 or > 230 .  
## There are no outlier rows, but the input data was renamed as QCtotals.  
## 15 rows remained.
```

Flag if VC (Vitamin C) is <5 or >400 → ask remove or not → if yes, remove those rows.

```
QCOutliers(input.data = QCtotals, target.colname = "VC", min = 5, max = 400)
```

```
## There are 0 observations with < 5 or > 400 .  
## There are no outlier rows, but the input data was renamed as QCtotals.  
## 15 rows remained.
```

Save as a .txt file.

```
write.table(QCtotals, "VVKAJ_Tot_mean_m_QCed.txt", sep="\t", quote=F, row.names=F)
```

Adjust totals and items after QC

Remove the QC-ed individual(s) from the totals to be consistent

In the previous section, we have removed individual(s) that did not pass the QC from mean total data. We will remove those individual(s) from the totals (before taking means of days), so that we will have the same individuals in the mean_total and total.

Among the individuals in new_totals_m, retain only those in QCtotals.

```
new_totals_m_QCed <- new_totals_m[ new_totals_m$UserName %in% QCtotals$UserName, ]
```

Save as a .txt file. This will be the total for each of the “QC-ed” individuals for each day, to be used for clustering analyses.

```
write.table(new_totals_m_QCed, "VVKAJ_Tot_m_QCed.txt", sep="\t", quote=F, row.names=F)
```

Similarly, remove the QC-ed individual(s) from the items to be consistent with the QC-ed averaged totals

Among the individuals in new_totals_m, pick up only those in QCtotals.

```
items_f_id_s_m_QCed <- items_f_id_s_m[ items_f_id_s_m$UserName %in% QCtotals$UserName, ]
```

Save as a .txt file. This will be the items for each of the “QC-ed” individuals for each day, to be used for ordination etc.

```
write.table(items_f_id_s_m_QCed, "VVKAJ_Items_f_id_s_m_QCed.txt", sep="\t", quote=F, row.names=F)
```

Come back to the main directory before you start running another script.

```
setwd(main_wd)
```