

Use cttab in the Analysis

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What this Document Does

The package `cctu` has been updated to utilize the DLU file and CLU file to facilitate the analyzing process. The newly added functions will not affect previous process. But here, in this document we are going to learn some new functions provided in the package to facilitate your analysis. Before you start, you should get yourself familiarized with the `cctu` package by looking through the **Analysis Template** vignette. **Remember, the `sumby` function will not benefit from any of those new functions mentioned in this document except the `coress` reference.** You don't need to read this document if you are going to stick to the `sumby` function.

What's new

Variable and value label

If you are familiar with SAS, Stata or SPSS, you should already know there's a variable label and value label (variable format in some). The variable label gives you the description of the variable, and the value label is to explain what the values stand for. The variable will stay as a numeric but has categories attached to it. Which means, you can subset or manipulate it like a numerical variable but report the data as a categorical one. It is important to understand this, let's use `mtcars` dataset and demonstrate how it works.

```
data(mtcars)
# Assign variable label
var_lab(mtcars$am) <- "Transmission"

# Assign value label with named vector
val_lab(mtcars$am) <- c("Automatic" = 0, "Manual"=1)
str(mtcars$am)
#>  num [1:32] 1 1 1 0 0 0 0 0 0 0 ...
#> - attr(*, "label")= chr "Transmission"
#> - attr(*, "labels")= Named num [1:2] 0 1
#> ..- attr(*, "names")= chr [1:2] "Automatic" "Manual"
```

As you can above, `label` and `labels` attributes were added. But the variable is numeric. You can still summarize it as numeric variable as below:

```
summary(mtcars$am)
#>   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
#> 0.0000 0.0000  0.0000  0.4062  1.0000  1.0000
```

You can convert the value label to factor just before your final analysis. The `to_factor` function replace the value with labels and convert the variable to factor.

```
# Extract variable label
var_lab(mtcars$am)
```

```
#> [1] "Transmission"

# Convert variable to factor with labels attached to it
table(to_factor(mtcars$am))
#>
#> Automatic      Manual
#>          19          13
```

But be cautious, **some R process might drop the variable or value label attributes**. Normally it won't, but you should check it before report if you are not sure. For example, `as.numeric`, `as.character` and `as.logical` will drop the variable and value label. There are `to_numeric`, `to_character` and `to_logical` can be used to convert the data type. You can also use the `copy_lab` function to copy the variable and value labels from the other variable.

You can use `var_lab` to extract the variable label or assign one. And use `has.label` to check if the variable has any variable label with it. You may also want to use `drop_lab` to drop the variable label.

For value label, you can use `val_lab` to extract or assign value label. And use `has.labels` to check if the variable has a value label. There are also `unval` function to drop the value label and `lab2val` to replace the `data.frame` value to its corresponding value labels.

MACRO dataset utility functions

There are new functions have been added to utilize the DLU and CLU files for the data analysis. The `apply_macro_dict` function uses DLU and CLU file to assign variable and value labels to the dataset. This will also convert the dataset based on the variable type as in the DLU file. And if you want to convert the variable name to lower cases, you should set `to_lower = TRUE`. This will convert the variable name to lower cases of the dataset and DLU data internally.

The `extract_form` can extract MACRO data by form and visits. The data will be converted to `data.table` class, which is an extension of the `data.frame` and works exactly the same. It is a great package with lots of data manipulation capability, you should seek the website for more details. But one thing to remember is that all the names in the variable selection will be considered as a variable of the data.

```
vars <- c("mpg", "am")
# You can do this in the normal data.frame
mtcars[,vars]

# But you can't do this for the data.table
dat <- data.table::data.table(mtcars)
mtcars[,vars]

# You need to add with=FALSE to do that
mtcars[,vars, with=FALSE]
```

Table populate function

You might already knew how to use the `sumby` function, but now a new function called `cttab` has been added. The difference between these two functions are the latter can handle variable label and value labels. You can feed the labelled data to this function and it will populate a summary table. Report data by treatment group, stratify tables by visit. Also, you can report variable based on some conditions and group the variable in the report. It generates a missing report internally and you can dump the missing report at the end. No extra step is needed. It will also produce the summary plots of the variables. The produced plots will be arrange to 3 by 3 and new plot will be produced if the variables exceeds 9. The remaining of this document will show you with a working case.

Setting options

The `cctab` function has a good flexibility, which means it has lots of parameters you can use. You should check out the manual of the `cctab` function. But, you can use `cctu_options` to set the default value of the parameters to be used by `cctab`. Below is setting some of the options, for more details please check out the `cctu_options` and `cctab` function manuals.

```
cctu_options(digits = 3) # keep 3 digits for numerical value
cctu_options(digits_pct = 0) # keep 0 digits for percentage
cctu_options(subjid_string = "subjid") # Set subject ID string
cctu_options(print_plot = FALSE) # Don't produce plots
```

Working example

In this section, we will show how to populate tables.

Data reading

You should read the data as before, but you can and should read the data by setting all the columns to character. This will be handled later. Same process can be found in the [Analysis Template vignette](#).

```
# Read example data
dt <- read.csv(system.file("extdata", "pilotdata.csv", package="cctu"),
               colClasses = "character")

# Read DLU and CLU
dlu <- read.csv(system.file("extdata", "pilotdata_dlu.csv", package="cctu"))
clu <- read.csv(system.file("extdata", "pilotdata_clu.csv", package="cctu"))
```

If you have multiple datasets, you should combine them at this stage. Next, we will apply the DLU and CLU files to the dataset.

```
# Create subjid
dt$subjid <- substr(dt$USUBJID, 8, 11)

# Apply CLU and DLU files
dt <- apply_macro_dict(dt, dlu = dlu, clu = clu)

# Give new variable a label
var_lab(dt$subjid) <- "Subject ID"
```

After this, you should follow the [Analysis Template vignette](#) and setup the population etc.

```
#> safety environment created in R_GlobalEnv
#> data frames removed from R_GlobalEnv : dt
```

Next we will do some data analysis.

Data analysis

After you have attached the population, next thing you may want to do is extract a particular form from the data. For the table, assume we are reporting age, sex and BMI from the patient registration form by treatment arm. For demonstrating the filtering, we only report non-white patients' BMI.

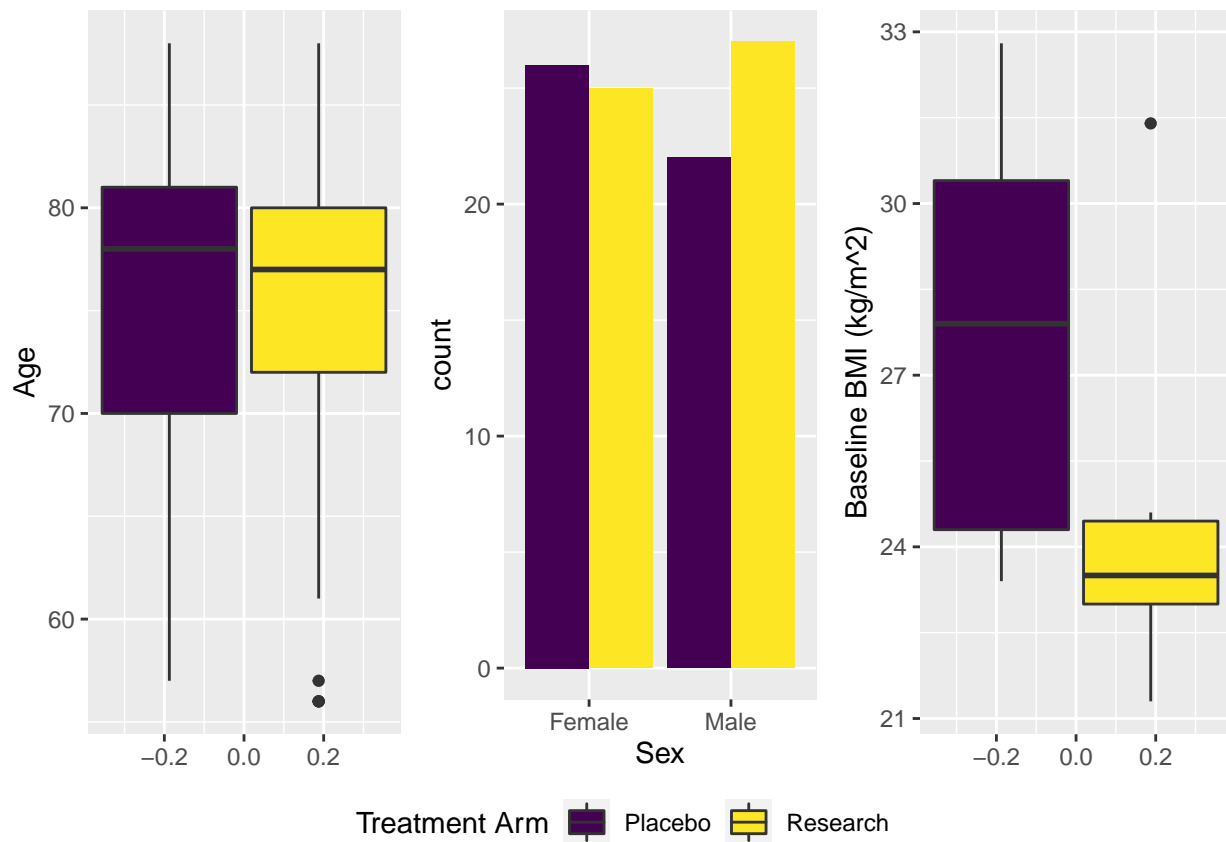
```
# Attach population
attach_pop("1.1")
#> safety attached containing: dt
```

```

# Extract patient registration form and keep subjid variable
df <- extract_form(dt, "PatientReg", vars_keep = c("subjid"))

# Now report Age, Sex and BMI. For BMI, report not white only
X <- cttab(vars = c("AGE", "SEX", "BMIBL"), # Variable to report
  group = "ARM", # Group variable
  data = df, # Data
  select = c("BMIBL" = "RACEN != 1")) # Filter for variable BMI

```



```

# Write table
X
#> +-----+
#> |           | Placebo | Research | Total   |
#> +-----+
#> | Observation |      48 |      52 |     100 |
#> +-----+
#> +-----+
#> | Age        |           |           |           |
#> +-----+
#> | Valid Obs. |      48 |      52 |     100 |
#> | Mean (SD)  |  75.5 (7.86) |  74.8 (8.04) |  75.1 (7.93) |
#> | Median [Min, Max] | 78.0 [57.0, 88.0] | 77.0 [56.0, 88.0] | 77.0 [56.0, 88.0] |
#> +-----+
#> | Sex        |           |           |           |
#> +-----+
#> | Female     |  26/48 (54.2%) |  25/52 (48.1%) |  51/100 (51.0%) |

```

```

#> | Male | 22/48 (45.8%) | 27/52 (51.9%) | 49/100 (49.0%) |
#> +-----+
#> |Baseline BMI (kg/m^2) |
#> +-----+
#> | Valid Obs. | 5 | 6 | 11 |
#> | Mean (SD) | 27.8 (3.98) | 24.6 (3.54) | 26.0 (3.93) |
#> | Median [Min, Max]|27.9 [23.4, 32.8]|23.5 [21.3, 31.4]|24.3 [21.3, 32.8]|
#> +-----+

```

Missing data report

The `cttab` function will report the missing internally. You can use the following to get the missing report.

```

# This will save the missing report under Output folder
# Or you can set the output folder and name
dump_missing_report()

# Pull out the missing report if you want
miss_rep <- get_missing_report()

# Reset missing report
reset_missing_report()

```

After this, you can finish the remaining as in the `Analysis Template` vignette.

More to cttab

As you have seen previously, the `cttab` function can easily populate simple tables.

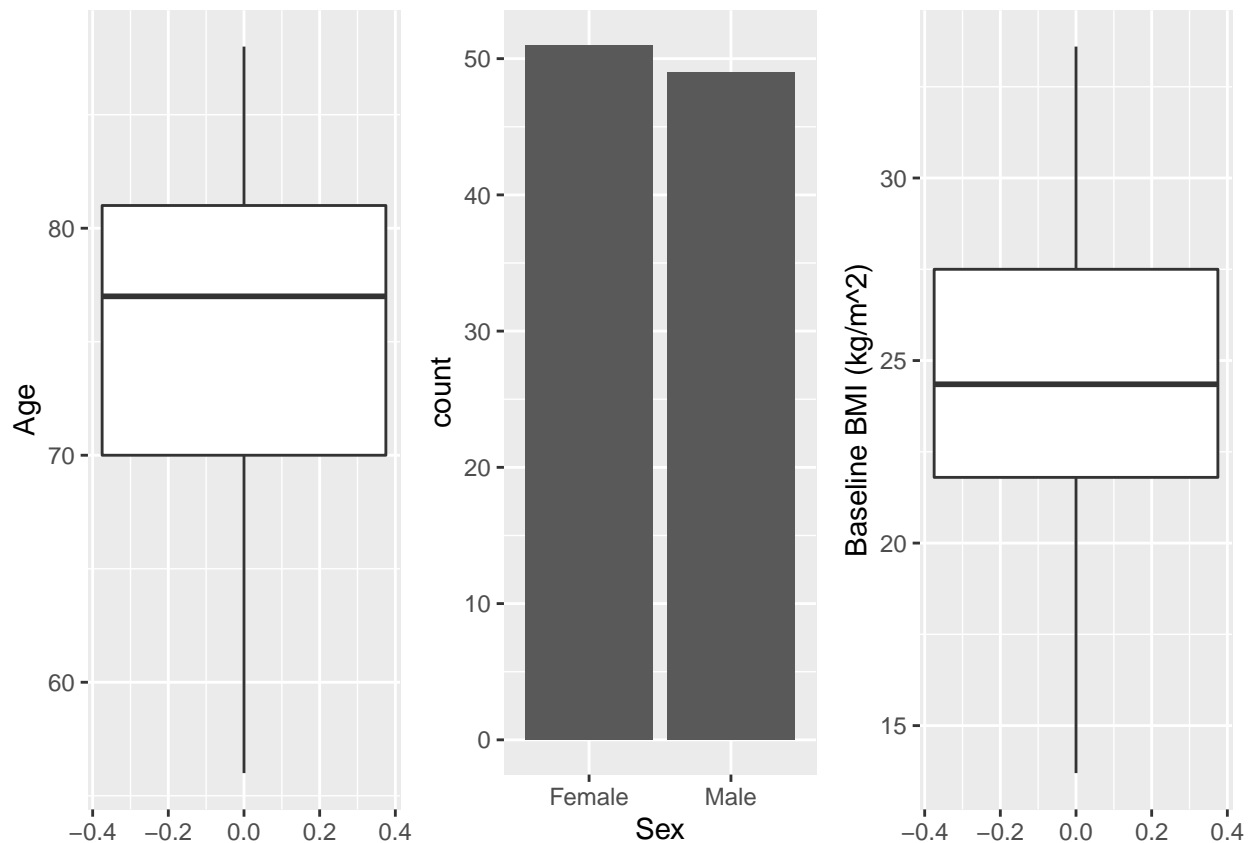
Simple table

Table only some variables, no treatment arm or variable selection.

```

X <- cttab(vars = c("AGE", "SEX", "BMIBL"), # Variable to report
           data = df)                      # Data

```

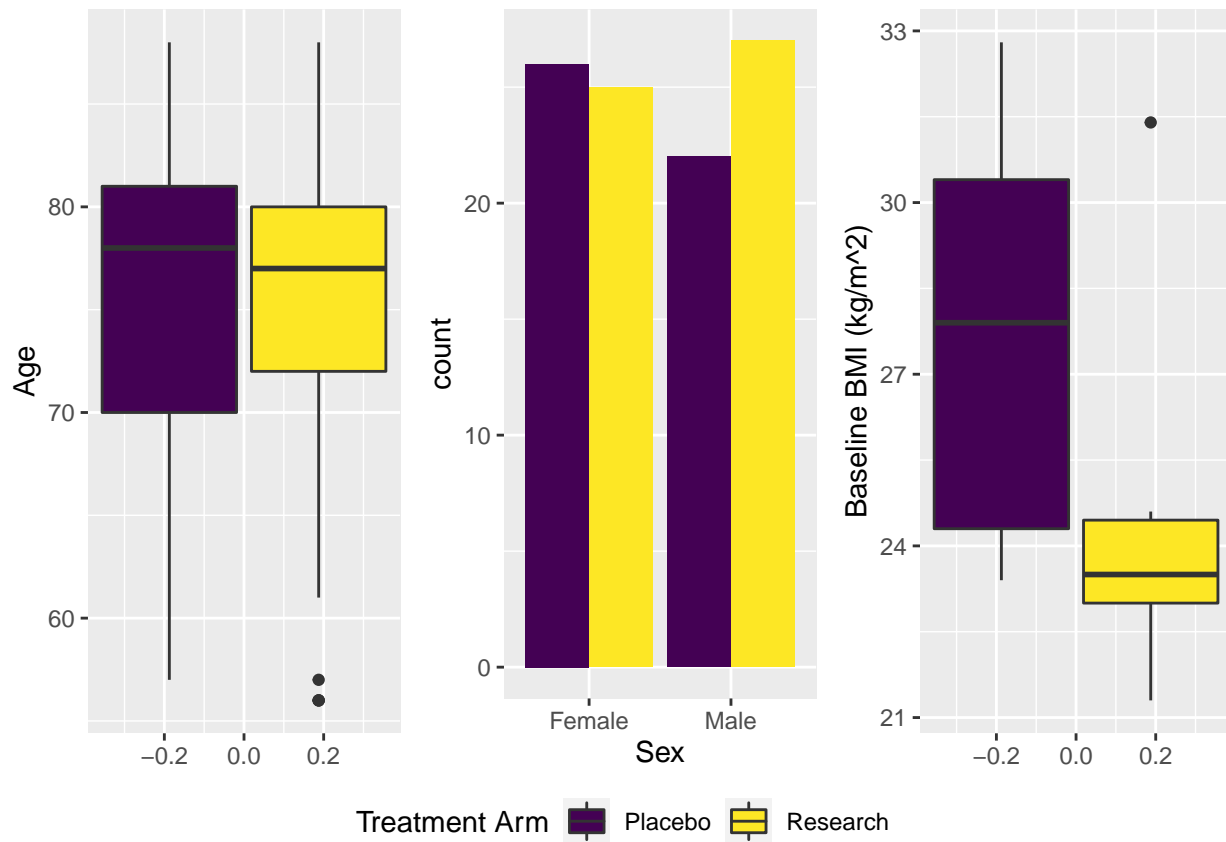


```
X
#> +-----+
#> |           |      Total      |
#> +-----+
#> |Age           |
#> +-----+
#> | Valid Obs.   |      100      |
#> | Mean (SD)    |  75.1 (7.93)  |
#> | Median [Min, Max]|77.0 [56.0, 88.0]|
#> +-----+
#> |Sex           |
#> +-----+
#> | Female       | 51/100 (51.0%) |
#> | Male        | 49/100 (49.0%) |
#> +-----+
#> |Baseline BMI (kg/m^2)
#> +-----+
#> | Valid Obs.   |      100      |
#> | Mean (SD)    |  24.5 (4.15)  |
#> | Median [Min, Max]|24.4 [13.7, 33.6]|
#> +-----+
```

By group and filter

This is what we have seen before

```
X <- cttab(vars = c("AGE", "SEX", "BMIBL"), # Variable to report
  group = "ARM", # Group variable
  data = df, # Data
  select = c("BMIBL" = "RACEN != 1")) # Filter for variable BMI
```



```
X
#> +-----+
#> |           | Placebo | Research | Total |
#> +-----+
#> |Observation |      48 |      52 |    100 |
#> +-----+
#> |Age
#> +-----+
#> | Valid Obs. |      48 |      52 |    100 |
#> | Mean (SD)   | 75.5 (7.86) | 74.8 (8.04) | 75.1 (7.93) |
#> | Median [Min, Max] | 78.0 [57.0, 88.0] | 77.0 [56.0, 88.0] | 77.0 [56.0, 88.0] |
#> +-----+
#> |Sex
#> +-----+
#> | Female      | 26/48 (54.2%) | 25/52 (48.1%) | 51/100 (51.0%) |
#> | Male        | 22/48 (45.8%) | 27/52 (51.9%) | 49/100 (49.0%) |
#> +-----+
#> |Baseline BMI (kg/m^2)
#> +-----+
#> | Valid Obs. |      5 |      6 |    11 |
```

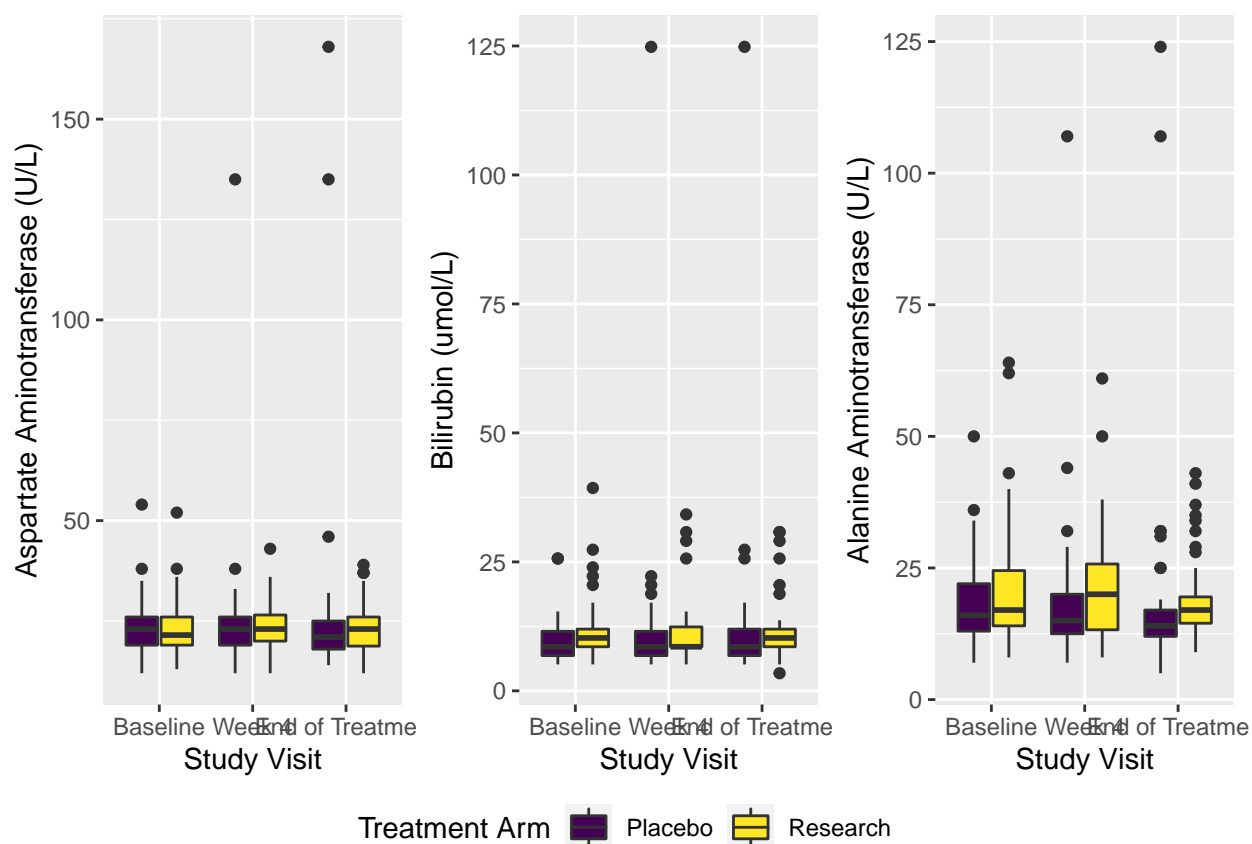
```
#> | Mean (SD)          | 27.8 (3.98) | 24.6 (3.54) | 26.0 (3.93) |
#> | Median [Min, Max] | 27.9 [23.4, 32.8] | 23.5 [21.3, 31.4] | 24.3 [21.3, 32.8] |
#> +-----+-----+-----+-----+
```

Split table row by visit

You can define `row_split` parameter to the name of visit or repeat variable.

```
attach_pop("1.1")
#> safety detached.
#> safety attached containing: dt
df <- extract_form(dt, "Lab", vars_keep = c("subjid", "ARM"))

X <- cttab(vars = c("AST", "BILI", "ALT"),
  group = "ARM",
  data = df,
  row_split = "AVISIT",          # Visit variable
  select = c("ALT" = "PERF == 1"))
```



```
X
#> +-----+-----+-----+-----+
#> |           | Placebo | Research | Total |
#> +-----+-----+-----+-----+
#> |Study Visit = Baseline|
#> +-----+-----+-----+-----+
#> |Observation          | 48      | 52      | 100    |
#> +-----+-----+-----+-----+
```



```

#> |Aspartate Aminotransferase (U/L) |
#> +-----+
#> | Valid Obs.      |      47      |      52      |      99      |
#> | Mean (SD)        | 23.4 (7.09)  | 23.6 (7.05)  | 23.5 (7.04)  |
#> | Median [Min, Max]|23.0 [12.0, 54.0]|21.5 [13.0, 52.0]|23.0 [12.0, 54.0]|
#> | Missing          |      1 (2.1%) |              |      1 (1.0%) |
#> +-----+
#> |Bilirubin (umol/L) |
#> +-----+
#> | Valid Obs.      |      46      |      51      |      97      |
#> | Mean (SD)        | 10.0 (4.38)  | 11.4 (6.01)  | 10.8 (5.32)  |
#> | Median [Min, Max]|8.55 [5.13, 25.7]|10.3 [5.13, 39.3]|10.3 [5.13, 39.3]|
#> | Missing          |      2 (4.2%) |      1 (1.9%) |      3 (3.0%) |
#> +-----+
#> |Alanine Aminotransferase (U/L) |
#> +-----+
#> | Valid Obs.      |      45      |      40      |      85      |
#> | Mean (SD)        | 18.1 (8.01)  | 22.1 (13.2)  | 20.0 (10.9)  |
#> | Median [Min, Max]|16.0 [7.00, 50.0]|17.0 [8.00, 64.0]|16.0 [7.00, 64.0]|
#> | Missing          |              |      2 (4.8%) |      2 (2.3%) |
#> +-----+
#> |Study Visit = Week 4 |
#> +-----+
#> |Observation      |      43      |      44      |      87      |
#> +-----+
#> |Aspartate Aminotransferase (U/L) |
#> +-----+
#> | Valid Obs.      |      41      |      43      |      84      |
#> | Mean (SD)        | 25.1 (18.4)  | 23.8 (6.39)  | 24.4 (13.6)  |
#> | Median [Min, Max]|23.0 [12.0, 135]|23.0 [12.0, 43.0]|23.0 [12.0, 135]|
#> | Missing          |      2 (4.7%) |      1 (2.3%) |      3 (3.4%) |
#> +-----+
#> |Bilirubin (umol/L) |
#> +-----+
#> | Valid Obs.      |      42      |      44      |      86      |
#> | Mean (SD)        | 12.7 (18.2)  | 11.4 (6.51)  | 12.0 (13.5)  |
#> | Median [Min, Max]|8.55 [5.13, 125]|8.55 [5.13, 34.2]|8.55 [5.13, 125]|
#> | Missing          |      1 (2.3%) |              |      1 (1.1%) |
#> +-----+
#> |Alanine Aminotransferase (U/L) |
#> +-----+
#> | Valid Obs.      |      39      |      38      |      77      |
#> | Mean (SD)        | 18.9 (16.1)  | 21.6 (11.0)  | 20.2 (13.8)  |
#> | Median [Min, Max]|15.0 [7.00, 107]|20.0 [8.00, 61.0]|16.0 [7.00, 107]|
#> | Missing          |              |      3 (7.3%) |      3 (3.8%) |
#> +-----+
#> |Study Visit = End of Treatment |
#> +-----+
#> |Observation      |      47      |      50      |      97      |
#> +-----+
#> |Aspartate Aminotransferase (U/L) |
#> +-----+
#> | Valid Obs.      |      45      |      48      |      93      |

```

```

#> | Mean (SD)          | 27.4 (27.9) | 23.1 (6.04) | 25.2 (19.9) |
#> | Median [Min, Max] |21.0 [14.0, 168] |23.0 [12.0, 39.0]|22.0 [12.0, 168] |
#> | Missing           | 2 (4.3%)   | 2 (4.0%)   | 4 (4.1%)   |
#> +-----+
#> |Bilirubin (umol/L)
#> +-----+
#> | Valid Obs.         | 45         | 46         | 91         |
#> | Mean (SD)          | 12.5 (17.7) | 11.5 (6.39) | 12.0 (13.2) |
#> | Median [Min, Max] |8.55 [5.13, 125] |10.3 [3.42, 30.8]|8.55 [3.42, 125] |
#> | Missing           | 2 (4.3%)   | 4 (8.0%)   | 6 (6.2%)   |
#> +-----+
#> |Alanine Aminotransferase (U/L)
#> +-----+
#> | Valid Obs.         | 45         | 43         | 88         |
#> | Mean (SD)          | 19.6 (21.8) | 19.7 (8.41) | 19.6 (16.6) |
#> | Median [Min, Max] |14.0 [5.00, 124] |17.0 [9.00, 43.0]|16.0 [5.00, 124] |
#> | Missing           | 1 (2.3%)   | 1 (1.1%)   | 1 (1.1%)   |
#> +-----+

```

Group variable

In this example, we will report demographic variable, lab results and lab abnormality. Variables will be grouped, no group name will be given to demographic variables, “Blood” to lab results and “Pts with Abnormal” to lab abnormality. Here, we count the number of patients with abnormal lab results. The `cttab` will report the count and percentage of TRUE. This is useful if you want to report patient numbers for different condition that belong to one category. Below is how to do it:

```

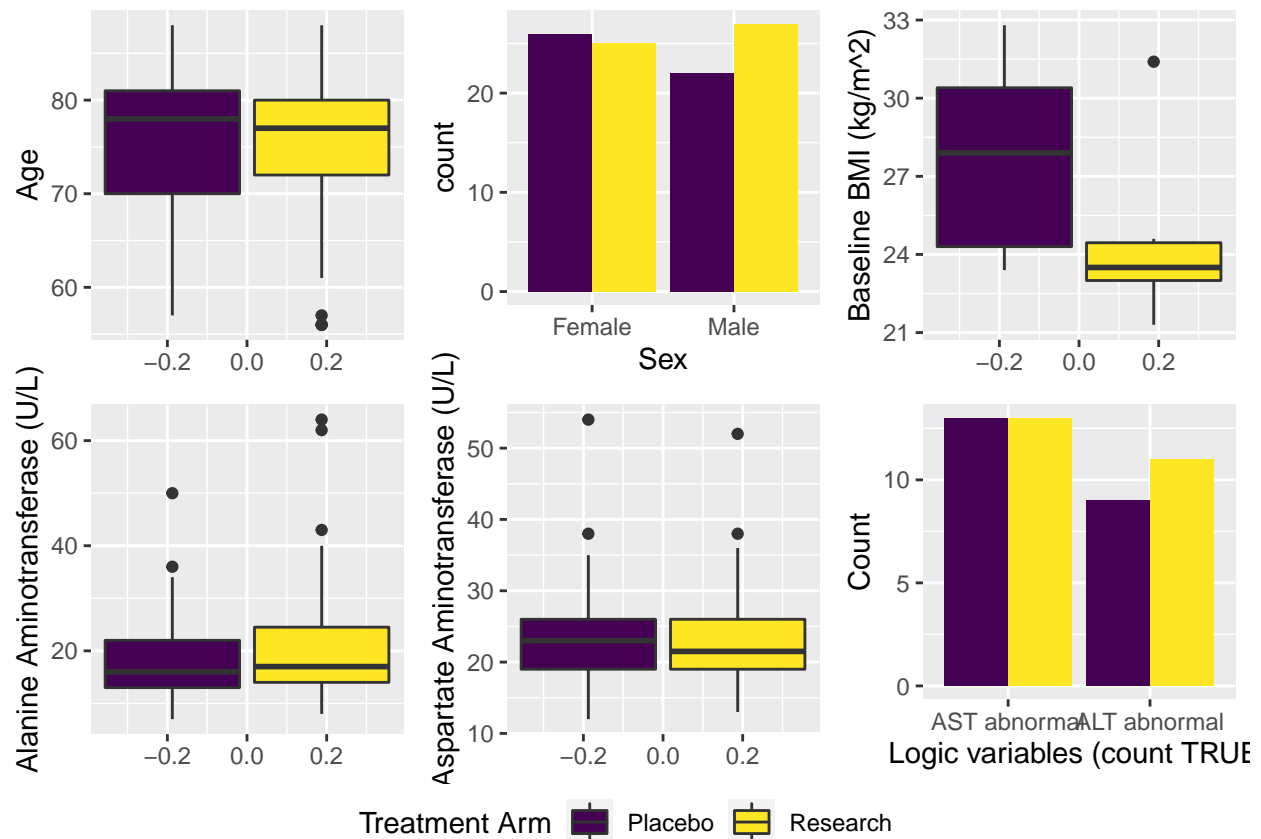
# Prepare data as before
attach_pop("1.1")
#> safety detached.
#> safety attached containing: dt
df <- extract_form(dt, "PatientReg", vars_keep = c("subjid"))
base_lab <- extract_form(dt, "Lab", visit = "SCREENING",
                        vars_keep = c("subjid"))

# Define abnormal
base_lab$ABNORMALT <- base_lab$ALT > 22.5
var_lab(base_lab$ABNORMALT) <- "ALT abnormal"
base_lab$ABNORMAST <- base_lab$AST > 25.5
var_lab(base_lab$ABNORMAST) <- "AST abnormal"

df <- merge(df, base_lab, by = "subjid")

# Table
X <- cttab(vars = list(c("AGE", "SEX", "BMIBL"),
                      # Group lab variable
                      "Blood" = c("ALT", "AST"),
                      # Group abnormal variable
                      "Pts with Abnormal" = c("ABNORMAST", "ABNORMALT")),
           group = "ARM",
           data = df,
           # Add some filtering
           select = c("BMIBL" = "RACEN != 1",
                     "ALT" = "PERF == 1"))

```



```

X
#> +-----+
#> |           | Placebo | Research | Total |
#> +-----+
#> |Observation |      48 |      52 |    100 |
#> +-----+
#> |Age
#> +-----+
#> | Valid Obs. |      48 |      52 |    100 |
#> | Mean (SD)  |  75.5 (7.86) |  74.8 (8.04) |  75.1 (7.93) |
#> | Median [Min, Max] | 78.0 [57.0, 88.0] | 77.0 [56.0, 88.0] | 77.0 [56.0, 88.0] |
#> +-----+
#> |Sex
#> +-----+
#> | Female      | 26/48 (54.2%) | 25/52 (48.1%) | 51/100 (51.0%) |
#> | Male        | 22/48 (45.8%) | 27/52 (51.9%) | 49/100 (49.0%) |
#> +-----+
#> |Baseline BMI (kg/m^2)
#> +-----+
#> | Valid Obs. |      5 |      6 |    11 |
#> | Mean (SD)  |  27.8 (3.98) |  24.6 (3.54) |  26.0 (3.93) |
#> | Median [Min, Max] | 27.9 [23.4, 32.8] | 23.5 [21.3, 31.4] | 24.3 [21.3, 32.8] |
#> +-----+
#> |Blood
#> +-----+

```

```

#> +-----+
#> |Alanine Aminotransferase (U/L) |
#> +-----+
#> | Valid Obs.      |      45      |      40      |      85      |
#> | Mean (SD)       | 18.1 (8.01)  | 22.1 (13.2)  | 20.0 (10.9)  |
#> | Median [Min, Max]|16.0 [7.00, 50.0]|17.0 [8.00, 64.0]|16.0 [7.00, 64.0]|
#> | Missing        |              | 2 (4.8%)    | 2 (2.3%)    |
#> +-----+
#> |Aspartate Aminotransferase (U/L) |
#> +-----+
#> | Valid Obs.      |      47      |      52      |      99      |
#> | Mean (SD)       | 23.4 (7.09)  | 23.6 (7.05)  | 23.5 (7.04)  |
#> | Median [Min, Max]|23.0 [12.0, 54.0]|21.5 [13.0, 52.0]|23.0 [12.0, 54.0]|
#> | Missing        | 1 (2.1%)    |              | 1 (1.0%)    |
#> +-----+
#> |Pts with Abnormal |
#> +-----+
#> | AST abnormal    | 13/48 (27.1%) | 17/52 (32.7%) | 30/100 (30.0%) |
#> | ALT abnormal    | 9/48 (18.8%)  | 11/52 (21.2%) | 20/100 (20.0%) |
#> +-----+

```

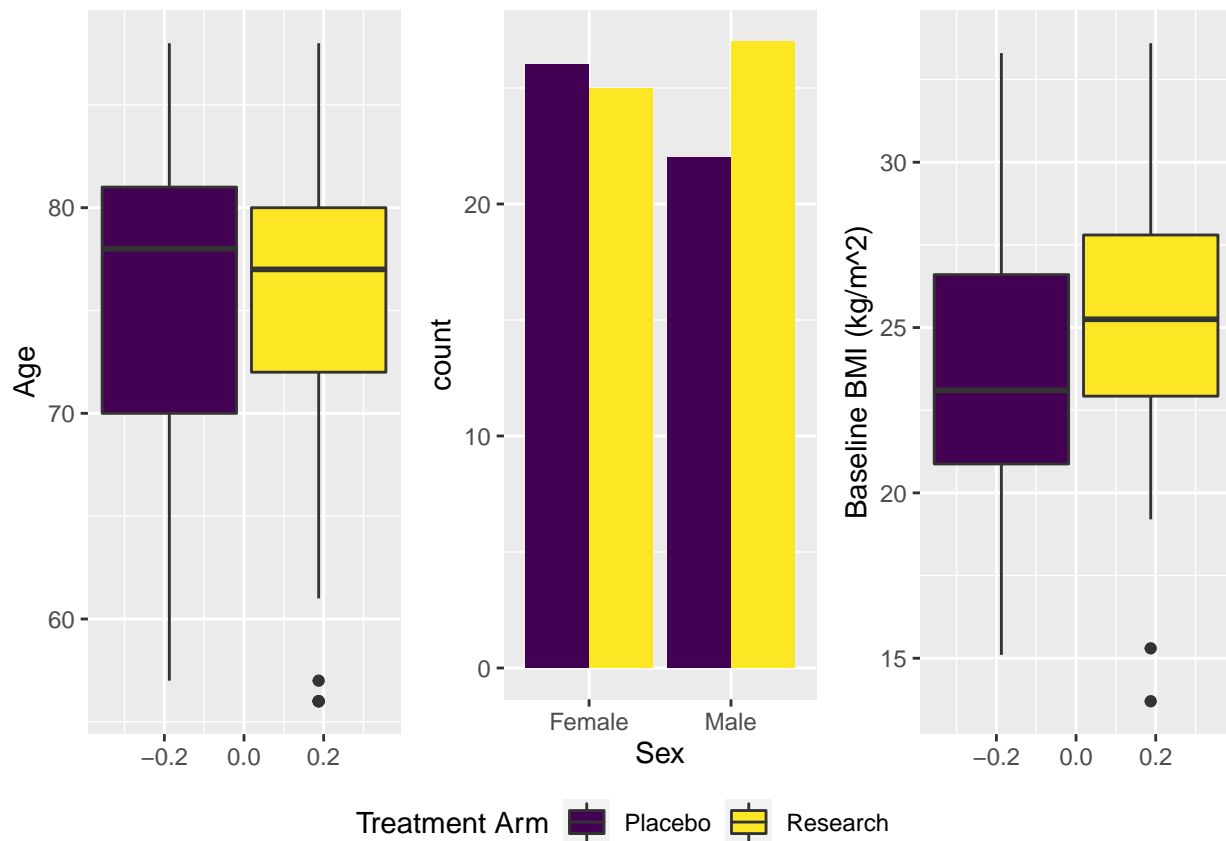
Rounding

The default behaviour of this function is to keep one digits for the percentage and 3 significant value for the numerical values. The default rounding function is `signif_pad`, you can also use `round` or `round_pad` to keep digits in the summary. The `format_percent` function is used to format the percentage values. There is `format_pval` might be useful to you.

```

X <- cttab(vars = c("AGE", "SEX", "BMIBL"), # Variable to report
            group = "ARM",                  # Group variable
            data = df,                      # Data
            digits = 2,                     # Keep 2 digits for numerical
            digits_pct = 1,                 # Keep 1 digits for percentage
            rounding_fn = round)             # Use function round for rounding

```



```
x
#> +-----+
#> |           | Placebo | Research | Total |
#> +-----+
#> |Observation |      48 |      52 |    100 |
#> +-----+
#> |Age
#> +-----+
#> | Valid Obs. |      48 |      52 |    100 |
#> | Mean (SD)  | 75.52 (7.86) | 74.75 (8.04) | 75.12 (7.93) |
#> | Median [Min, Max]| 78 [57, 88] | 77 [56, 88] | 77 [56, 88] |
#> +-----+
#> |Sex
#> +-----+
#> | Female      | 26/48 (54.2%) | 25/52 (48.1%) | 51/100 (51.0%) |
#> | Male        | 22/48 (45.8%) | 27/52 (51.9%) | 49/100 (49.0%) |
#> +-----+
#> |Baseline BMI (kg/m^2)
#> +-----+
#> | Valid Obs. |      48 |      52 |    100 |
#> | Mean (SD)  | 23.55 (4.05) | 25.29 (4.1) | 24.45 (4.15) |
#> | Median [Min, Max]|23.1 [15.1, 33.3]|25.25 [13.7, 33.6]|24.35 [13.7, 33.6]|
#> +-----+
```

Cross reference

Tables can be cross-referenced in the word. This is very convenient if you are writing narratives and want to reference tables or figures. You can find how to cross reference tables or figures [here](#). If you can changed the order of the tables by cut and paste or by dragging the headings in the navigation pane of the word the reference will still be there. You can press **Ctrl+A** select all, and press **F9**. This will change numbering of the table and figure, this change will also be applied to the references you made in the narratives.