SAS <-> R :: CHEAT SHEET

Introduction

This guide aims to familiarise SAS users with R.
R examples make use of tidyverse collection of packages.

Install tidyverse: **install.packages("tidyverse")**
Attach tidyverse packages for use: **library(tidyverse)**

R data here in 'data frames', and occasionally vectors (via **c( )**)
Other R structures (lists, matrices...) are not explored here.

Keyboard shortcuts: **<-** Alt + - **%>%** Ctrl + Shift + m

Datasets; drop, keep & rename variables

```
data new_data;
  set old_data;
run;
```

```
new_data <- old_data
```

```
data new_data (keep=id);
  set old_data (drop=job_title) ;
run;
```

```
new_data <- old_data %>%
  select(-job_title) %>%
  select(id)
```

```
data new_data (drop= temp: );
  set old_data;
run;
```

```
new_data <- old_data %>%
  select( -starts_with("temp")
```

C.f. contains( ) , ends_with( )

```
data new_data;
  set old_data;
  rename old_name = new_name;
run;
```

```
new_data <- old_data %>%
  rename(new_name = old_name)
```

Note order differs

Conditional filtering

```
data new_data;
  set old_data;
  if Sex = "M";
run;
```

```
new_data <- old_data %>%
  filter(Sex == "M")
```

```
data new_data;
  set old_data;
  if year in (2010,2011,2012);
run;
```

```
new_data <- old_data %>%
  filter(year %in% c(2010,2011,2012))
```

```
data new_data;
  set old_data;
  by id ;
  if first.id ;
run;
```

```
new_data <- old_data %>%
  group_by( id ) %>%
  slice(1)
```

Could use slice(n( )) *for last*

```
data new_data;
  set old_data;
  if dob > "25APR1990"d;
run;
```

```
new_data <- old_data %>%
  filter(dob > as.Date("1990-04-25"))
```

New variables, conditional editing

```
data new_data;
  set old_data;
  total_income = wages + benefits ;
run;
```

```
new_data <- old_data %>%
  mutate(total_income = wages + benefits)
```

```
data new_data;
  set old_data;
  if hours > 30 then full_time = "Y";
  else full_time = "N";
run;
```

```
new_data <- old_data %>%
  mutate(full_time = if_else(hours > 30 , "Y" , "N"))
```

```
data new_data;
  set old_data;
  if temp > 20 then weather = "Warm";
  else if temp > 10 then weather = "Mild";
  else weather = "Cold";
run;
```

```
new_data <- old_data %>%
  mutate(weather = case_when(
    temp > 20 ~ "Warm",
    temp > 10 ~"Mild",
    TRUE ~ "Cold" ) )
```

Counting and Summarising

```
proc freq data = old_data ;
  table job_type ;
run;
```

```
old_data %>%
  count( job_type )
```

For percent, add:
%>% mutate(percent = n*100/sum(n))

```
proc freq data = old_data ;
  table job_type*region ;
run;
```

```
old_data %>%
  count( job_type , region )
```

```
proc summary data = old_data nway ;
  class job_type region ;
  output out = new_data ;
run;
```

```
new_data <- old_data %>%
  group_by( job_type , region ) %>%
  summarise( Count = n( ) )
```

Equivalent without nway not trivially produced

```
proc summary data = old_data nway ;
  class job_type region ;
  var salary ;
  output out = new_data
    sum( salary ) = total_salaries ;
run;
```

```
new_data <- old_data %>%
  group_by( job_type , region ) %>%
  summarise(  total_salaries = sum( salary ) ,
    Count = n( ) )
```

Lots of summary functions in both languages
Swap summarise( ) *for* mutate( ) *to add summary data to original data*

Combining datasets

```
data new_data ;
  set data_1 data_2 ;
run;
```

```
new_data <- bind_rows( data_1 , data_2 )
```

C.f. rbind( ) *which produces error if columns are not identical*

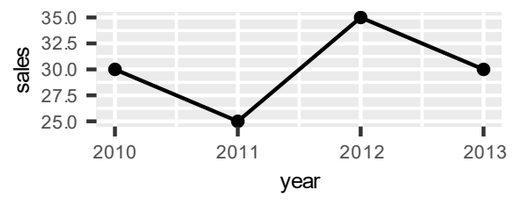
```
data new_data ;
  merge data_1 (in= in_1) data_2 ;
  by id ;
  if in_1 ;
run;
```

```
new_data <- left_join( data_1 , data_2 , by = "id")
```

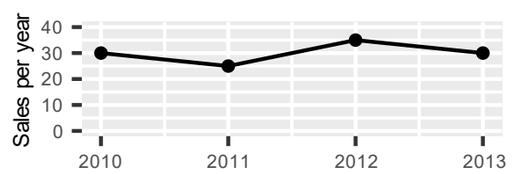
C.f. full_join( ) , right_join( ) , inner_join( )

Some plotting in R

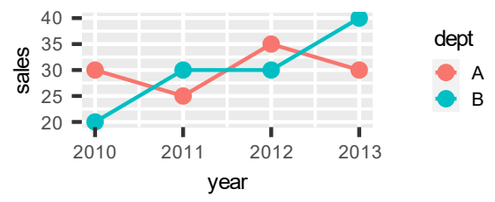
```
ggplot( my_data , aes( year , sales ) ) +
  geom_point( ) + geom_line( )
```



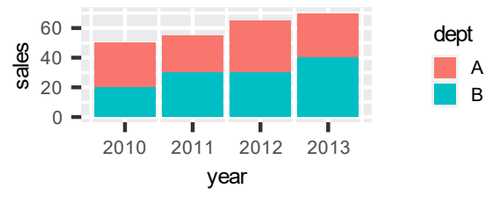

```
ggplot( my_data , aes( year , sales ) ) +
  geom_point( ) + geom_line( ) + ylim(0, 40) +
  labs(x = "" , y = "Sales per year")
```




```
ggplot(my_data, aes( year, sales, colour = dept) ) +
  geom_point( ) + geom_line( )
```

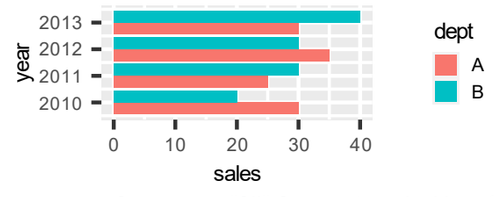



```
ggplot( my_data , aes( year, sales, fill = dept) ) +
  geom_col( )
```



Note 'colour' *for lines & points,* 'fill' *for shapes*

```
ggplot( my_data , aes( year, sales, fill = dept) ) +
  geom_col( position = "dodge" ) + coord_flip( )
```



C.f. position = "fill" *for 100% stacked bars/cols*

Sorting and Row-Wise Operations

```
proc sort data=old_data out=new_data;
  by id descending income ;
run;
```

```
new_data <- old_data %>%
  arrange( id , desc( income ) )
```

```
proc sort data=old_data nodup;
  by id job_type;
run;
```

```
old_data <- old_data %>%
  arrange( id , job_type)) %>%
  distinct( )
```

Note nodup *relies on adjacency of duplicate rows,* distinct( ) *does not*

```
proc sort data=old_data nodupkey;
  by id ;
run;
```

```
old_data <- old_data %>%
  arrange( id ) %>%
  group_by( id ) %>%
  slice( 1 )
```

```
data new_data;
  set old_data;
  by id descending income ;
  if first.id ;
run;
```

```
new_data <- old_data %>%
  group_by( id ) %>%
  slice(which.max( income ))
```

C.f.which.min( )

Swap to preserve duplicate maxima: ... slice_max(income)

Alternatively: ... filter(income==max(income))

```
data new_data;
  set old_data;
  prev_id= lag( id );
run;
```

```
new_data <- old_data %>%
  mutate( prev_id = lag( id , 1 ))
```

C.f. lead( ) *for subsequent rows*

```
data new_data;
  set old_data;
  by id;
  counter +1 ;
  if first.id then counter = 1;
run;
```

```
new_data <- old_data %>%
  group_by( id ) %>%
  mutate( counter = row_number( ) )
```

Converting and Rounding

```
data new_data;
  set old_data ;
  num_var = input("5" , 8. );
  text_var = put( 5 , 8. );
run;
```

```
new_data <- old_data %>%
  mutate(num_var = as.numeric("5" )) %>%
  mutate(text_var = as.character( 5 ))
```

```
data new_data  ;
  set old_data;
  nearest_5 = round( x , 5 )
  two_decimals = round( x , 0.01)
run;
```

```
new_data <- old_data %>%
  mutate(nearest_5 = round(x/5)*5) %>%
  mutate(two_decimals = round( x , digits = 2)
```

Creating functions to modify datasets

```
%macro add_variable(dataset_name);
data &dataset_name;
  set &dataset_name;
  new_variable = 1;
run;
%mend;
%add_variable( my_data );
```

```
add_variable <- function( dataset_name ){
  dataset_name <- dataset_name %>%
    mutate(new_variable = 1)
  return( dataset_name )
}
my_data <- add_variable( my_data )
```

Note SAS can modify within the macro, whereas R creates a copy within the function

String Manipulation

```
data new_data;
  set old_data;
  if find( job_title , "Health" );
run;
```

```
new_data <- old_data %>%
  filter( str_detect( job_title , "Health" ))
```

```
data new_data;
  set old_data;
  substring = substr( big_string , 3 , 4 );
run;
```

```
new_data <- old_data %>%
  mutate( substring = str_sub( big_string , 3 , 6 ))
```

Returns characters 3 to 6. Note SAS uses <start>, <length>, R uses <start>, <end>

```
data new_data;
  set old_data;
  address = tranwrd( address , "Street" , "St" );
run;
```

```
new_data <- old_data %>%
  mutate( address = str_replace_all( address , "Street" , "St" ))
```

C.f. str_replace( ) *for first instance of pattern only*

```
data new_data;
  set old_data;
  house_number = compress( address , , "dk" );
run;
```

```
new_data <- old_data %>%
  mutate( house_number = str_extract( address , "\\d+" ))
```

Wide range of regexps in both languages, this example extracts digits only

Transpose/Pivot

```
proc transpose data=long_data out=wide_data;
  by student ;
  id subject ;
  var grade ;
run;
```

Add NOTSORTED *if* long_data *is not sorted by* student

```
wide_data <- long_data %>%
  pivot_wider(names_from = subject , values_from = grade)
```

```
proc transpose data=wide_data
  out=long_data(rename=(col1=grade)) name=subject;
  by student ;
  var English Irish Maths;
run;
```

```
long_data <- wide_data %>%
  pivot_longer(c(English, Irish, Maths) ,
    names_to = "subject", values_to = "grade")
```

File operations

Operate in 'Work' library.
Use **libname** to define file locations

```
libname library_name "file_location";
data library_name.saved_data;
  set data_in_use;
run;
```

Operate in a particular 'working directory' (identify using **getwd( )**)
Move to other locations using **setwd( )**

```
saveRDS(data_in_use , file="file_location/saved_data.rds")
or
setwd("file_location")
saveRDS( data_in_use , file = "saved_data.rds")
```

```
libname library_name "file_location";
data data_in_use ;
  set library_name.saved_data ;
run;
```

```
data_in_use <- readRDS("file_location/saved_data.rds" )
or
setwd("file_location")
data_in_use <- readRDS("saved_data.rds")
```

```
proc export data = my_data
  outfile = "my_file.csv" dbms = csv replace;
run;
```

```
write_csv(my_data , "my_file.csv")
```

```
proc import datafile = "my_file.csv"
  out = my_data dbms = csv;
run;
```

```
my_data <- read_csv("my_file.csv")
```

Both examples assume column headers in csv file