1.transform

Joao Lopes

2024-10-07

Contents

1.	. NUMERIC VECTORS	3
	1.1. COUNTS	. 3
	1.2. NUMERIC TRANSFORMATION	. 4
	Arithmetic and recycling rules	
	Minimum and maximum	
	Modular arithmetic	
	Rounding	
	Cutting numbers into ranges	
	Cumulative and rolling aggregates	
	1.3. GENERAL TRANSFORMATION	. 5
	Ranks	. 5
	Offsets	. 6
	Consecutive identifiers	. 6
	1.4. SUMMARY STATISTICS	. 6
	Center	. 6
	Minimum, maximum, and quantiles	. 7
	Spread	
	Distributions	
	Positions	
2.	. FACTORS	9
	2.1. BASICS	
	2.2. DATASET gss_cat	
	2.3. MODIFYING FACTOR ORDER	
	2.4. MODIFYING FACTOR LEVELS	. 11
_	T 0 070 47 777 077 0 70	
3.	. LOGICAL VECTORS	13
	3.1. COMPARISONS	
	Missing values	
	3.2. BOOLEAN ALGEBRA	
	Boolean operations	
	Missing values	
	Operator %in%	
	3.3. SUMMARIES	
	Logical summaries	
	Numeric summaries of logical vectors	
	Logical subsetting	. 14
	3.4. CONDITIONAL TRANSFORMATIONS	
	Function if $_{\text{else}}()$	
	Function case_when() \dots	
	Compatible types	. 16

1. NUMERIC VECTORS

```
[from https://r4ds.hadley.nz/numbers]
library("nycflights13") #collection of datasets
library("skimr") #function skim() for descriptive statistics
library("tidyverse") #collection of packages for data analysis

#metadata
?flights

#data inspection
glimpse(flights)

#descriptive statistics
skim(flights)
```

1.1. COUNTS

```
#simple count
flights |> count(dest)
#sorted count
flights |> count(dest, sort = TRUE)
#count + summary statisctics
flights |>
  group_by(dest) |>
  summarize(
   n = n()
    delay = mean(arr_delay, na.rm = TRUE)
#count distinct
flights |>
  group_by(dest) |>
  summarize(carriers = n_distinct(carrier)) |>
  arrange(desc(carriers))
#sum is weighted count
flights |>
  group_by(tailnum) |>
  summarize(miles = sum(distance))
flights |> count(tailnum, wt = distance)
#count missing values
flights |>
  group_by(dest) |>
  summarize(n_cancelled = sum(is.na(dep_time)))
```

1.2. NUMERIC TRANSFORMATION

Arithmetic and recycling rules

```
x <- c(1, 2, 10, 20)

x / 5

x / c(5, 5, 5, 5)

x * c(1, 2)

x * c(1, 2, 3)
```

Minimum and maximum

```
df <- tribble(
    ~x, ~y,
    1, 3,
    5, 2,
    7, NA,
)

df |>
    mutate(
    min = pmin(x, y, na.rm = TRUE),
    max = pmax(x, y, na.rm = TRUE)
)

df |>
    mutate(
    min = min(x, y, na.rm = TRUE),
    max = max(x, y, na.rm = TRUE),
    max = max(x, y, na.rm = TRUE)
)
```

Modular arithmetic

```
1:10 %/% 3
1:10 %% 3

flights |>
   mutate(
    hour = sched_dep_time %/% 100,
    minute = sched_dep_time %% 100,
    .keep = "used"
)

flights |>
   group_by(hour = sched_dep_time %/% 100) |>
   summarize(prop_cancelled = mean(is.na(dep_time)), n = n()) |>
   filter(hour > 1) |>
   ggplot(aes(x = hour, y = prop_cancelled)) +
   geom_line(color = "grey50") +
   geom_point(aes(size = n))
```

Rounding

```
#simple round
round(123.456)
#round to nearest n digit
round(123.456, 2) # two digits
round(123.456, 1) # one digit
round(123.456, -1) # round to nearest ten
round(123.456, -2) # round to nearest hundred
#floor() vs ceiling()
round(c(1.5, 2.5))
x <- 123.456
floor(x)
ceiling(x)
#round down/up to nearest n digit
floor(x / 0.01) * 0.01
ceiling(x / 0.01) * 0.01
#round to nearest multiple
round(x / 4) * 4  # round to nearest multiple of 4
round(x / 0.25) * 0.25 # round to nearest 0.25
```

Cutting numbers into ranges

```
#simple cut
x <- c(1, 2, 5, 10, 15, 20)
cut(x, breaks = c(0, 5, 10, 15, 20))
cut(x, breaks = c(0, 5, 10, 100))
cut(x,
    breaks = c(0, 5, 10, 15, 20),
    labels = c("sm", "md", "lg", "xl")
)

#cut with values outside the range
y <- c(NA, -10, 5, 10, 30)
cut(y, breaks = c(0, 5, 10, 15, 20))</pre>
```

Cumulative and rolling aggregates

```
x <- 1:10
cumsum(x)
```

1.3. GENERAL TRANSFORMATION

Ranks

```
#simple ranks
x <- c(1, 2, 2, 3, 4, NA)
min_rank(x)
```

```
min_rank(desc(x))
#more ranks
df \leftarrow tibble(x = x)
df |>
 mutate(
   row_number = row_number(x),
   dense_rank = dense_rank(x),
   percent_rank = percent_rank(x),
    cume_dist = cume_dist(x)
  )
#using ranks to divide data
df <- tibble(id = 1:10)</pre>
df |>
 mutate(
   row0 = row_number() - 1,
    three_groups = row0 %% 3,
    three_in_each_group = row0 %/% 3
```

Offsets

```
x <- c(2, 5, 11, 11, 19, 35)
lag(x)
lead(x)

x - lag(x)
x == lag(x)</pre>
```

Consecutive identifiers

```
events <- tibble(
    time = c(0, 1, 2, 3, 5, 10, 12, 15, 17, 19, 20, 27, 28, 30)
)

events <- events |>
    mutate(
        diff = time - lag(time, default = first(time)),
        has_gap = diff >= 5
    )
    events

events |> mutate(
    group = cumsum(has_gap)
)
```

1.4. SUMMARY STATISTICS

Center

```
flights |>
group_by(year, month, day) |>
```

```
summarize(
  mean = mean(dep_delay, na.rm = TRUE),
  median = median(dep_delay, na.rm = TRUE),
  n = n(),
  .groups = "drop"
) |>
  ggplot(aes(x = mean, y = median)) +
  geom_abline(slope = 1, intercept = 0, color = "white", linewidth = 2) +
  geom_point()
```

Minimum, maximum, and quantiles

```
flights |>
  group_by(year, month, day) |>
  summarize(
   max = max(dep_delay, na.rm = TRUE),
   q95 = quantile(dep_delay, 0.95, na.rm = TRUE),
   .groups = "drop"
)
```

Spread

```
flights |>
  group_by(origin, dest) |>
  summarize(
    distance_sd = IQR(distance),
    n = n(),
    .groups = "drop"
) |>
  filter(distance_sd > 0)
```

Distributions

```
flights |>
  filter(dep_delay < 120) |>
  ggplot(aes(x = dep_delay, group = interaction(day, month))) +
  geom_freqpoly(binwidth = 5, alpha = 1/5)
```

Positions

```
flights |>
  group_by(year, month, day) |>
  summarize(
    first_dep = first(dep_time, na_rm = TRUE),
    fifth_dep = nth(dep_time, 5, na_rm = TRUE),
    last_dep = last(dep_time, na_rm = TRUE)
)

flights |>
  group_by(year, month, day) |>
  mutate(r = min_rank(sched_dep_time)) |>
  filter(r %in% c(1, max(r)))
```

2. FACTORS

```
[from https://r4ds.hadley.nz/factors]
```

```
library("tidyverse") #collection of packages for data analysis
```

2.1. BASICS

```
#use just strings
x1 <- c("Dec", "Apr", "Jan", "Mar")</pre>
x2 <- c("Dec", "Apr", "Jam", "Mar")
sort(x1)
#simple factors
month_levels <- c(</pre>
 "Jan", "Feb", "Mar", "Apr", "May", "Jun",
"Jul", "Aug", "Sep", "Oct", "Nov", "Dec"
y1 <- factor(x1, levels = month_levels)</pre>
у1
sort(y1)
y2 <- factor(x2, levels = month_levels)</pre>
#use fct()
y2 <- fct(x2, levels = month_levels)</pre>
levels(y2)
factor(x1)
fct(x1)
#use col_factor()
csv <- "
month, value
Jan, 12
Feb,56
Mar, 12"
df <- read_csv(csv, col_types = cols(month = col_factor(month_levels)))</pre>
```

2.2. DATASET gss_cat

```
#General Social Survey
gss_cat
?gss_cat
#look at factors
gss_cat |>
count(race)
```

2.3. MODIFYING FACTOR ORDER

```
#reorder in agplot
relig_summary <- gss_cat |>
  group_by(relig) |>
  summarize(
   tvhours = mean(tvhours, na.rm = TRUE),
   n = n()
  )
ggplot(relig_summary, aes(x = tvhours, y = relig)) +
  geom_point()
ggplot(relig_summary, aes(x = tvhours, y = fct_reorder(relig, tvhours))) +
  geom_point()
#reorder in tibble
relig_summary |>
  mutate(
   relig = fct_reorder(relig, tvhours)
  ) |>
  ggplot(aes(x = tvhours, y = relig)) +
  geom_point()
#reorder and relevel
rincome_summary <- gss_cat |>
  group_by(rincome) |>
  summarize(
    age = mean(age, na.rm = TRUE),
    n = n()
ggplot(rincome_summary, aes(x = age, y = fct_reorder(rincome, age))) +
  geom_point()
ggplot(rincome_summary, aes(x = age, y = fct_relevel(rincome, "Not applicable"))) +
  geom_point()
by_age <- gss_cat |>
  filter(!is.na(age)) |>
  count(age, marital) |>
  group_by(age) |>
  mutate(
    prop = n / sum(n)
#reorder and legend order
ggplot(by_age, aes(x = age, y = prop, color = marital)) +
  geom_line(linewidth = 1) +
  scale color brewer(palette = "Set1")
ggplot(by_age, aes(x = age, y = prop, color = fct_reorder2(marital, age, prop))) +
  geom_line(linewidth = 1) +
  scale_color_brewer(palette = "Set1") +
```

```
labs(color = "marital")

#reorder and bar plots
gss_cat |>
  mutate(marital = marital |> fct_infreq() |> fct_rev()) |>
  ggplot(aes(x = marital)) +
  geom_bar()
```

2.4. MODIFYING FACTOR LEVELS

```
gss_cat |> count(partyid)
#ommit levels using fct_recode()
gss_cat |>
  mutate(
    partyid = fct_recode(partyid,
      "Republican, strong" = "Strong republican",
"Republican, weak" = "Not str republican",
      "Independent, near rep" = "Ind, near rep",
      "Independent, near dem" = "Ind, near dem",
      "Democrat, weak" = "Not str democrat",
      "Democrat, strong" = "Strong democrat"
    )
  ) |>
  count(partyid)
#combine levels using fct_recode()
gss_cat |>
  mutate(
    partyid = fct_recode(partyid,
      "Republican, strong" = "Strong republican",
"Republican, weak" = "Not str republican",
      "Independent, near rep" = "Ind, near rep",
      "Independent, near dem" = "Ind, near dem",
      "Democrat, weak" = "Not str democrat",

"Democrat, strong" = "Strong democrat",
      "Other"
                               = "No answer",
      "Other"
                               = "Don't know",
      "Other"
                               = "Other party"
    )
  )
#combine levels using fct_collapse()
gss_cat |>
  mutate(
    partyid = fct_collapse(partyid,
      "other" = c("No answer", "Don't know", "Other party"),
      "rep" = c("Strong republican", "Not str republican"),
      "ind" = c("Ind, near rep", "Independent", "Ind, near dem"),
      "dem" = c("Not str democrat", "Strong democrat")
    )
  ) |>
  count(partyid)
```

```
#lump unfrequent groups using fct_lump_lowfreq()
gss_cat |>
  mutate(relig = fct_lump_lowfreq(relig)) |>
  count(relig)

#lump unfrequent groups using fct_lump_n()
gss_cat |>
  mutate(relig = fct_lump_n(relig, n = 10)) |>
  count(relig, sort = TRUE)
```

3. LOGICAL VECTORS

```
[from https://r4ds.hadley.nz/logicals]
```

```
library("nycflights13") #collection of datasets
library("tidyverse") #collection of packages for data analysis
```

3.1. COMPARISONS

```
#create logical vector inline
flights |>
  filter(dep_time > 600 & dep_time < 2000 & abs(arr_delay) < 20)

#create logical vector outside logical condition
flights |>
  mutate(
    daytime = dep_time > 600 & dep_time < 2000,
    approx_ontime = abs(arr_delay) < 20,
) |>
  filter(daytime & approx_ontime)
```

Missing values

```
#logical conditions
NA > 5
10 == NA
NA == NA
flights |>
   filter(dep_time == NA)

#is.na()
is.na(c(TRUE, NA, FALSE))
is.na(c(1, NA, 3))
is.na(c("a", NA, "b"))
flights |>
   filter(is.na(dep_time))
```

3.2. BOOLEAN ALGEBRA

Boolean operations

Missing values

```
df <- tibble(x = c(TRUE, FALSE, NA))

df |>
  mutate(
   and = x & NA,
   or = x | NA
)
```

Operator %in%

```
flights |>
  filter(month == 11 | month == 12)
flights |>
  filter(month %in% c(11, 12))

c(1, 2, NA) == NA
c(1, 2, NA) %in% NA
```

3.3. SUMMARIES

Logical summaries

```
flights |>
  group_by(year, month, day) |>
  summarize(
   all_delayed = all(dep_delay <= 60, na.rm = TRUE),
   any_long_delay = any(arr_delay >= 300, na.rm = TRUE),
   .groups = "drop"
)
```

Numeric summaries of logical vectors

```
flights |>
  group_by(year, month, day) |>
  summarize(
  all_delayed = mean(dep_delay <= 60, na.rm = TRUE),
  any_long_delay = sum(arr_delay >= 300, na.rm = TRUE),
  .groups = "drop"
)
```

Logical subsetting

```
flights |>
  filter(arr_delay > 0) |>
  group_by(year, month, day) |>
  summarize(
    behind = mean(arr_delay),
    n = n(),
    .groups = "drop"
)
flights |>
```

```
group_by(year, month, day) |>
summarize(
  behind = mean(arr_delay[arr_delay > 0], na.rm = TRUE),
  ahead = mean(arr_delay[arr_delay < 0], na.rm = TRUE),
  n = n(),
  .groups = "drop"
)</pre>
```

3.4. CONDITIONAL TRANSFORMATIONS

Function if_else()

Function case_when()

```
case_when(
 x == 0 ~~"0",
 x < 0 ~ "-ve",
 x > 0 "+ve",
 is.na(x) ~ "???"
case_when(
x < 0 \sim "-ve",
 x > 0 ~ "+ve"
case_when(
 x < 0 \sim "-ve",
 x > 0 ~ "+ve",
 TRUE ~ "???"
)
case_when(
x > 0 \sim "+ve",
 x > 2 ~ "big"
flights |>
 mutate(
   status = case_when(
```

Compatible types

```
if_else(TRUE, "a", 1)

case_when(
    x < -1 ~ TRUE,
    x > 0 ~ now()
)
```