problemset6

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GitHub

```
https://github.com/juny1z/Problemset6.git
a
library(DBI)
library(parallel)
library(future)
library(furrr)
library(dplyr)
Attaching package: 'dplyr'
The following objects are masked from 'package:stats':
    filter, lag
The following objects are masked from 'package:base':
    intersect, setdiff, setequal, union
lahman <- dbConnect(RSQLite::SQLite(), "/Users/zjyyy/Desktop/lahman_1871-2022.sqlite")</pre>
fielding <- dbGetQuery(lahman, "SELECT * FROM Fielding") %>%
  mutate(RF = 3 * (PO + A) / InnOuts) %>%
  filter(!is.na(RF) & InnOuts > 0 & !is.na(teamID))
sta_func <- function(data) {</pre>
data %>%
```

```
group_by(teamID) %>%
    summarise(mean_RF = mean(RF, na.rm = TRUE))
}
stratified_bootstrap <- function(data, group_col, sta_func, n_bootstrap = 1000) {
  replicate(n_bootstrap, {
    sampled_data <- data %>%
      group_by(!!sym(group_col)) %>%
      sample_frac(replace = TRUE) %>%
     ungroup()
    if (nrow(sampled_data) == 0) {
     return(data.frame(teamID = NA, mean_RF = NA))
    }
    sta_func(sampled_data)
  }, simplify = FALSE)
# Without any parallel processing
set.seed(123)
bootstrap <- stratified_bootstrap(fielding, "teamID", sta_func, n_bootstrap = 1000)</pre>
bootstrap_sum <- bind_rows(bootstrap) %>%
  group_by(teamID) %>%
  summarise(
   mean_RF = mean(mean_RF, na.rm = TRUE),
    se_RF = sd(mean_RF, na.rm = TRUE)
  ) %>%
  mutate(se_RF = ifelse(is.na(se_RF), 0, se_RF)) %>%
  arrange(desc(mean_RF)) %>%
  slice_head(n = 10)
print(bootstrap_sum)
# A tibble: 10 x 3
   teamID mean_RF se_RF
           <dbl> <dbl>
   <chr>
 1 RC1
           0.571
 2 LS1
          0.533
          0.525
 3 ELI
                      0
 4 MLU
          0.516
                      0
 5 KEO
            0.512
                      0
 6 RIC
          0.508
```

```
7 BLA
            0.495
                       0
 8 LS3
            0.489
                       0
 9 TRN
            0.481
                       0
10 PHU
            0.478
                       0
# Parallel
cl <- makeCluster(detectCores() - 1)</pre>
clusterExport(cl, varlist = c("fielding", "stratified_bootstrap", "sta_func"), envir = envir
clusterEvalQ(cl, library(dplyr))
[[1]]
[1] "dplyr"
                             "graphics" "grDevices" "utils"
                                                                   "datasets"
                 "stats"
[7] "methods"
                "base"
[[2]]
[1] "dplyr"
                "stats"
                             "graphics" "grDevices" "utils"
                                                                   "datasets"
[7] "methods"
                "base"
[[3]]
[1] "dplyr"
                             "graphics" "grDevices" "utils"
                                                                   "datasets"
                 "stats"
[7] "methods"
                 "base"
[[4]]
[1] "dplyr"
                 "stats"
                             "graphics" "grDevices" "utils"
                                                                   "datasets"
[7] "methods"
                "base"
[[5]]
[1] "dplyr"
                 "stats"
                             "graphics" "grDevices" "utils"
                                                                   "datasets"
[7] "methods"
                "base"
[[6]]
[1] "dplyr"
                             "graphics" "grDevices" "utils"
                                                                   "datasets"
                "stats"
[7] "methods"
                 "base"
[[7]]
[1] "dplyr"
                 "stats"
                             "graphics" "grDevices" "utils"
                                                                   "datasets"
[7] "methods"
                 "base"
set.seed(123)
bootstrap_parallel <- parLapply(cl, 1:1000, function(i) {</pre>
  stratified_bootstrap(fielding, "teamID", sta_func, n_bootstrap = 1)
})
```

```
bootstrap_sum_parallel <- bind_rows(bootstrap_parallel) %>%
  group_by(teamID) %>%
  summarise(
    mean_RF = mean(mean_RF, na.rm = TRUE),
    se_RF = ifelse(n_distinct(mean_RF) > 1, sd(mean_RF, na.rm = TRUE), 0)
) %>%
  mutate(se_RF = ifelse(is.na(se_RF), 0, se_RF)) %>%
  arrange(desc(mean_RF)) %>%
  slice_head(n = 10)

print("Results with Parallel Package:")
```

[1] "Results with Parallel Package:"

```
print(bootstrap_sum_parallel)
```

```
# A tibble: 10 \times 3
  teamID mean_RF se_RF
   <chr>
           <dbl> <dbl>
1 RC1
           0.571
                      0
2 LS1
           0.529
                      0
3 ELI
           0.526
                      0
4 MLU
          0.514
5 RIC
          0.510
                      0
6 KEO
          0.509
                      0
7 BLA
          0.496
                      0
8 LS3
          0.489
                      0
9 PHU
           0.482
                      0
10 TRN
           0.478
                      0
```

```
# Future
plan(multisession)

set.seed(123)
bootstrap_future <- future_map(1:1000, ~ {
    sampled_data <- fielding %>%
        group_by(teamID) %>%
        sample_frac(replace = TRUE) %>%
```

```
ungroup()
sta_func(sampled_data)
}, .options = furrr_options(seed = TRUE))

bootstrap_sum_future <- bind_rows(bind_rows(bootstrap_future)) %>%
group_by(teamID) %>%
summarise(
mean_RF = mean(mean_RF, na.rm = TRUE),
se_RF = sd(mean_RF, na.rm = TRUE)
) %>%
mutate(se_RF = ifelse(is.na(se_RF), 0, se_RF)) %>%
arrange(desc(mean_RF)) %>%
slice_head(n = 10)

print("Results with Future Package:")
```

[1] "Results with Future Package:"

```
print(bootstrap_sum_future)
```

```
# A tibble: 10 x 3
  teamID mean_RF se_RF
           <dbl> <dbl>
  <chr>
1 RC1
           0.574
2 LS1
           0.532
                      0
3 ELI
           0.523
                      0
4 RIC
           0.514
                      0
5 KEO
           0.506
                      0
6 MLU
           0.503
                      0
7 BLA
           0.495
                      0
8 LS3
           0.489
9 PHU
           0.481
                      0
10 TRN
           0.480
```

b

A tibble: 10×7

	teamID mea	an_RF_no_parallel	se_RF_no_parallel	$mean_RF_parallel$	se_RF_parallel
	<chr></chr>	<dbl></dbl>	<dbl></dbl>	<dbl></dbl>	<dbl></dbl>
1	RC1	0.571	0	0.571	0
2	LS1	0.533	0	0.529	0
3	ELI	0.525	0	0.526	0
4	MLU	0.516	0	0.514	0
5	KEO	0.512	0	0.509	0
6	RIC	0.508	0	0.510	0
7	BLA	0.495	0	0.496	0
8	LS3	0.489	0	0.489	0
9	TRN	0.481	0	0.478	0
10	PHU	0.478	0	0.482	0

i 2 more variables: mean_RF_future <dbl>, se_RF_future <dbl>

c

The mean_RF values differ slightly across the three versions, which is expected due to slight differences in the bootstrap.

Non-Parallel is the slowest since it performs all bootstrap iterations sequentially

Future is comparable to the Parallel, but slightly more flexible due to the future package.