

# Week Ten Tasks

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*11/25/2018*

Need to compare timing of a few aspects within my function. The apply method below is much faster. I had hoped to use the map function, but I don't think that the purrr package can do what I would need it to do in this situation (a by column operation for matrices). I found this discussion board that points to this conclusion: <https://github.com/tidyverse/purrr/issues/341>

As a result I implemented apply in my code.

```
library(tidyverse)
library(here)
library(MASS)
library(broom)
library(ggbeeswarm)

whale <- read_csv(here("data", "sperm_whale_pop.csv"))

## Parsed with column specification:
## cols(
##   Year = col_integer(),
##   Population = col_integer()
## )

ntimes <- 100

# setup needed
df <- whale
fit.poi <- glm(df$Population ~ df$Year, data = df, family = "poisson")
best_model <- fit.poi
n <- max(df$Year) - min(df$Year) + 1
y.hat <- best_model$fitted.values
y.rep <- matrix(ncol = ntimes, data = c(rpois(n*ntimes, y.hat)))
years <- c(min(df$Year):max(df$Year))

# using apply
apply_method <- function(df, best_model, n, y.hat, y.rep){
  y.rep.sort <- -apply(-y.rep, 2, sort)
  y.coefs <- t(apply(y.rep.sort, 2, function(y.col) lm(y.col~years)$coef))
  end.dates <- -1* y.coefs[,1] / y.coefs[,2]
  return(end.dates)
}

apply_method(df, best_model, n, y.hat, y.rep)

## [1] 2558.934 2577.318 2569.007 2551.198 2556.999 2577.303 2540.984
## [8] 2562.693 2584.994 2589.143 2576.830 2590.650 2579.324 2574.598
## [15] 2556.175 2570.868 2566.836 2577.997 2570.868 2548.303 2575.871
## [22] 2567.968 2580.917 2565.659 2582.251 2559.069 2586.048 2562.023
## [29] 2571.221 2567.119 2581.335 2575.847 2548.440 2588.201 2557.332
## [36] 2557.611 2570.344 2554.550 2572.000 2589.329 2565.989 2581.891
## [43] 2589.150 2566.711 2606.226 2584.425 2578.575 2565.689 2552.717
```

```
## [50] 2583.258 2570.582 2571.976 2573.497 2576.274 2567.430 2592.127
## [57] 2560.634 2556.609 2559.147 2572.326 2571.744 2577.679 2581.375
## [64] 2585.476 2581.966 2596.716 2577.086 2579.970 2574.247 2552.709
## [71] 2564.523 2547.613 2556.384 2575.814 2563.710 2562.716 2592.009
## [78] 2576.618 2559.140 2560.477 2563.022 2571.770 2560.645 2566.718
## [85] 2579.679 2558.901 2559.191 2591.224 2585.122 2547.935 2564.124
## [92] 2571.617 2550.982 2582.263 2555.137 2561.210 2581.285 2546.668
## [99] 2569.502 2571.126
```

```
# using for loop
loop_method <- function(df, best_model, n, y.hat, y.rep){
  end_dates <- double(ntimes)
  for (i in 1:ntimes){
    sim_reg <- lm(sort(y.rep[,i], decreasing = TRUE) ~
                  years)
    end_dates[i] <- -1*summary(sim_reg)$coefficients[1,1] /
                  summary(sim_reg)$coefficients[2,1]
  }
  return(end_dates)
}

loop_method(df, best_model, n, y.hat, y.rep)
```

```
## [1] 2558.934 2577.318 2569.007 2551.198 2556.999 2577.303 2540.984
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## [99] 2569.502 2571.126
```

```
times <- bench::mark(
  apply_method(df, best_model, n, y.hat, y.rep),
  loop_method(df, best_model, n, y.hat, y.rep)
)
```

```
## Warning: Some expressions had a GC in every iteration; so filtering is
## disabled.
```

```
times
```

```
## # A tibble: 2 x 10
##   expression      min mean median      max `itr/sec` mem_alloc n_gc n_itr
##   <chr>          <bch:t> <bch> <bch:t> <bch:t>      <dbl> <bch:byt> <dbl> <int>
## 1 apply_met~ 80.1ms 83ms 81.3ms 92.6ms    12.1    563KB    14     7
## 2 loop_meth~ 130.7ms 134ms 134ms 136.8ms    7.47   470KB    12     4
## # ... with 1 more variable: total_time <bch:tm>
```

```
plot(times)
```

