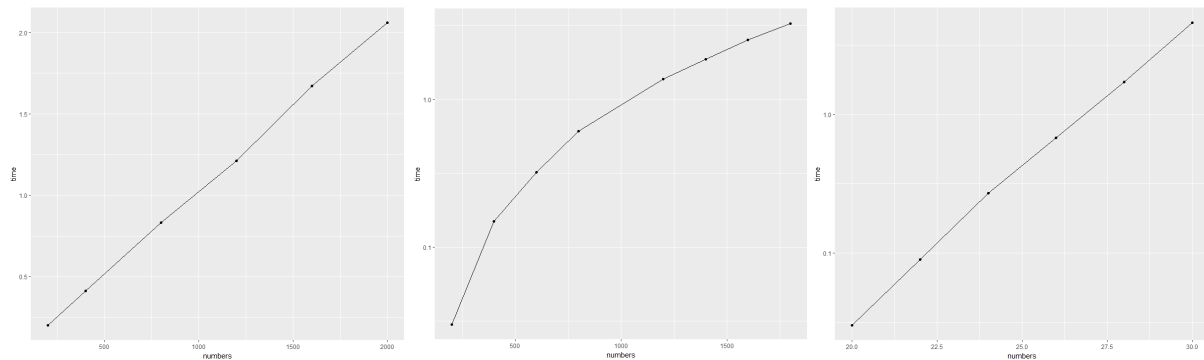


*log\_factorial*:  $O(N)$ . Linear on normal scale plot. (Figure 1 left).

*sum\_log\_factorial*:  $O(N^\alpha)$ . Non-linear on y-log10-scale plot (Figure 1 middle).

*fibonacci*:  $O(\alpha^N)$ . Linear on y-log10-scale plot (Figure 1 right).

**Figure 1:** Plots of running time versus numbers. **Left:** *log\_factorial*, x, y normal scale. **Middle:** *sum\_log\_factorial*, x normal scale, y log10 scale. **Right:** *Fibonacci*, x normal scale, y log10 scale.



- To plot Figure 1 left, I added a constant time in *log\_factorial* function. “return (constant.time() + log(n) + log\_factorial(n - 1))”.

**Figure 2:** Snip of my code.

```
15 return (sum)
16 }
17
18 fibonacci <- function(n) {
19   # Return nth Fibonacci number
20   if (n <= 1)
21     return (n)
22   return (fibonacci(n - 1) + fibonacci(n - 2))
23 }
24
25 constant.time <- function(){
26   count = 0
27   for(i in seq(1:50)){
28     for(j in seq(1:50)){
29       count = count + 1
30     }
31   }
32   return (0)
33 }
34
35 running.time <- function(function.name, num.seq){
36   nums = c()
37   time = c()
38   for(n in num.seq){
39     nums = append(nums, n)
40     time = append(time, system.time(a <- function.name(n))[[1]])
41   }
42   time.dataframe = data.frame(numbers = nums, time = time)
43   return (time.dataframe)
44 }
45
46 options(expressions=500000)
47 #rt = running.time(log_factorial, c( 200, 400, 800,1200,1600, 2000))
48 #rt = running.time(sum_log_factorial, c( 200, 400, 600, 800,1200, 1400, 1600,
49 rt = running.time(fibonacci, c(20, 22, 24, 26, 28, 30))
50
51 p <- ggplot(data = rt, aes(numbers, time)) +
52   geom_point() +
53   geom_line() +
54   scale_y_log10()
55
56 print (p)
57
```