CS 433 (2) HW3 Report

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1 Data Analysis

1.1 Data Import

First we load the simulation data, filter out processes that never finished, and give the data a once over.

```
library(readr)
library(dplyr)
library(magrittr)

col_spec <- cols(
    scheduler = col_character(),
    pid = col_integer(),
    arrival = col_integer(),
    finish = col_integer(),
    cpu = col_integer(),
    io = col_integer(),
    wait = col_integer()
)

data10 <- read_csv('./os_sim_10.csv', col_types = col_spec)
data20 <- read_csv('./os_sim_20.csv', col_types = col_spec)
data100 <- read_csv('./os_sim_100.csv', col_types = col_spec)
data1000 <- read_csv('./os_sim_1000.csv', col_types = col_spec)</pre>
```

```
data10$num_procs <- 10
data20$num_procs <- 20
data100$num_procs <- 100
data1000$num_procs <- 1000
glimpse(data10, width = 60)
## Observations: 20
## Variables: 8
## $ scheduler <chr> "FCFS", "FCFS", "FCFS", "FCFS", "FCFS...
## $ pid
          <int> 9, 8, 7, 6, 1, 0, 2, 3, 4, 5, 9, 8, 7...
## $ arrival <int> 290671, 274023, 271748, 250512, 40644...
## $ finish
            <int> 0, 0, 0, 0, 46512, 53631, 224614, 0, ...
## $ cpu
              <int> 1971, 9584, 7586, 17198, 3024, 6755, ...
## $ io
              <int> 1951, 6210, 7253, 16848, 2581, 7474, ...
## $ wait
              <int> 5326, 10167, 13339, 15441, 263, 1305,...
## $ num_procs <dbl> 10, 10, 10, 10, 10, 10, 10, 10, 10, 10.
glimpse(data20, width = 60)
## Observations: 40
## Variables: 8
## $ scheduler <chr> "FCFS", "FCFS", "FCFS", "FCFS", "FCFS...
            <int> 19, 18, 17, 16, 5, 4, 3, 2, 0, 1, 6, ...
## $ pid
## $ arrival <int> 298950, 297875, 290671, 290319, 83552...
## $ finish <int> 0, 0, 0, 0, 196672, 249049, 209215...
## $ cpu
              <int> 15, 102, 1263, 575, 11303, 24683, 493...
              <int> 96, 211, 1227, 1252, 61354, 29269, 48...
## $ io
## $ wait
              <int> 532, 1376, 6820, 7703, 143596, 76408,...
## $ num_procs <dbl> 20, 20, 20, 20, 20, 20, 20, 20, 20, 2...
glimpse(data100, width = 60)
## Observations: 200
## Variables: 8
## $ scheduler <chr> "FCFS", "FCFS", "FCFS", "FCFS", "FCFS...
## $ pid
           <int> 99, 98, 97, 96, 95, 94, 93, 92, 91, 9...
## $ arrival <int> 298950, 298232, 297875, 294344, 29118...
## $ finish
              <int> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0...
              <int> 0, 0, 0, 58, 42, 7, 42, 167, 162, 121...
## $ cpu
## $ io
              <int> 0, 0, 0, 31, 52, 42, 99, 95, 144, 127...
## $ wait
              <int> 0, 0, 0, 4791, 4688, 4605, 9516, 9371...
glimpse(data1000, width = 60)
## Observations: 2,000
## Variables: 8
```

```
## $ scheduler <chr> "FCFS", "FCFS", "FCFS", "FCFS", "FCFS...
              <int> 999, 998, 997, 996, 995, 994, 993, 99...
## $ arrival <int> 299735, 298954, 298950, 298852, 29836...
## $ finish
              <int> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0...
## $ cpu
              <int> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0...
## $ io
              <int> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0...
## $ wait
              <int> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0...
## $ num_procs <dbl> 1000, 1000, 1000, 1000, 1000, 1000, 1...
data <- rbind(data10, data20, data100, data1000) %>%
    # filter out processes that never finished
   filter(finish != 0) %>%
    # convert ms to s
   mutate at(.funs = list(~./1000), .vars = 3:7) %>%
    # col spec of col_factor() doesnt seem to work
   mutate(scheduler = as.factor(scheduler),
        turnaround = finish - arrival)
glimpse(data, width = 60)
## Observations: 34
## Variables: 9
## $ scheduler <fct> FCFS, FCFS, FCFS, FCFS, FCFS, SJF, S...
## $ pid
              <int> 1, 0, 2, 4, 5, 1, 0, 2, 4, 4, 3, 2, ...
               <dbl> 40.644, 38.097, 66.312, 189.715, 244...
## $ arrival
              <dbl> 46.512, 53.631, 224.614, 236.296, 26...
## $ finish
## $ cpu
               <dbl> 3.024, 6.755, 58.732, 24.784, 7.550,...
               <dbl> 2.581, 7.474, 89.009, 19.978, 11.004...
## $ io
               <dbl> 0.263, 1.305, 10.561, 1.819, 2.479, ...
## $ wait
## $ num_procs <dbl> 10, 10, 10, 10, 10, 10, 10, 10, 10, ...
## $ turnaround <dbl> 5.868, 15.534, 158.302, 46.581, 21.0...
```

The times are in seconds.

```
library(xtable)

data_summary <- data %>%
    group_by(scheduler, num_procs) %>%
    summarise(
        mean_arrival = mean(arrival),
        sd_arrival = sd(arrival),
        mean_finish = mean(finish),
        sd_finish = sd(finish),
        mean_cpu = mean(cpu),
        sd_cpu = sd(cpu),
        mean_io = mean(io),
        sd_io = sd(io),
```

```
mean_wait = mean(wait),
    sd_wait = sd(wait),
    mean_turnaround = mean(turnaround),
    sd_turnaround = sd(turnaround),
    throughput = n() / (max(finish) - min(arrival))
)

data_summary %>%
    select(scheduler, num_procs, mean_arrival:sd_arrival) %>%
    xtable()
```

	scheduler	num_procs	mean_arrival	sd_arrival
1	FCFS	10.00	115.84	95.06
2	FCFS	20.00	84.63	60.60
3	FCFS	100.00	31.97	44.92
4	SJF	10.00	83.69	71.82
5	SJF	20.00	60.49	27.10
6	SJF	100.00	87.34	52.02
7	SJF	1000.00	71.57	49.53

```
data_summary %>%
    select(scheduler, num_procs, mean_finish:sd_finish) %>%
    xtable()
```

	scheduler	num_procs	mean_finish	sd_finish
1	FCFS	10.00	165.30	106.27
2	FCFS	20.00	178.39	63.24
3	FCFS	100.00	238.52	40.24
4	SJF	10.00	140.31	104.36
5	SJF	20.00	132.29	65.43
6	SJF	100.00	156.94	48.93
7	SJF	1000.00	161.99	49.81

```
data_summary %>%
    select(scheduler, num_procs, mean_cpu:sd_io) %>%
    xtable()

data_summary %>%
    select(scheduler, num_procs, mean_wait:sd_wait) %>%
    xtable()

data_summary %>%
    select(scheduler, num_procs, mean_turnaround:throughput) %>%
    xtable()
```

scheduler	num_procs	$mean_cpu$	sd _cpu	$mean_io$	sd _io
FCFS	10.00	20.17	23.14	26.01	35.79
FCFS	20.00	22.79	16.09	22.54	15.81
FCFS	100.00	12.11	5.30	10.18	5.49
SJF	10.00	23.32	25.45	29.66	40.00
SJF	20.00	21.44	19.55	26.69	19.34
SJF	100.00	8.64	4.90	40.26	26.67
SJF	1000.00	6.45	4.44	71.21	49.11
	FCFS FCFS FCFS SJF SJF SJF	FCFS 10.00 FCFS 20.00 FCFS 100.00 SJF 10.00 SJF 20.00 SJF 20.00	FCFS 10.00 20.17 FCFS 20.00 22.79 FCFS 100.00 12.11 SJF 10.00 23.32 SJF 20.00 21.44 SJF 100.00 8.64	FCFS 10.00 20.17 23.14 FCFS 20.00 22.79 16.09 FCFS 100.00 12.11 5.30 SJF 10.00 23.32 25.45 SJF 20.00 21.44 19.55 SJF 100.00 8.64 4.90	FCFS 10.00 20.17 23.14 26.01 FCFS 20.00 22.79 16.09 22.54 FCFS 100.00 12.11 5.30 10.18 SJF 10.00 23.32 25.45 29.66 SJF 20.00 21.44 19.55 26.69 SJF 100.00 8.64 4.90 40.26

	scheduler	num_procs	$mean_wait$	sd_wait
1	FCFS	10.00	3.29	4.15
2	FCFS	20.00	48.43	35.67
3	FCFS	100.00	184.26	29.23
4	SJF	10.00	3.63	4.71
5	SJF	20.00	23.67	20.33
6	SJF	100.00	20.70	11.80
7	SJF	1000.00	12.76	8.37

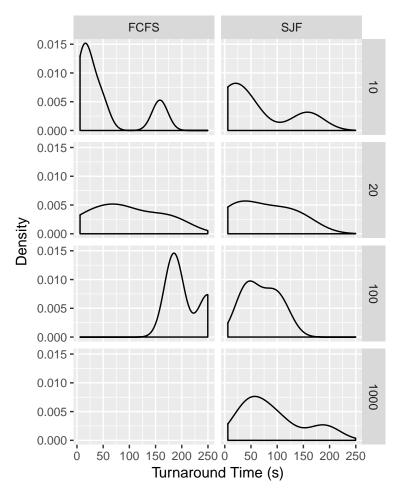
	scheduler	num_procs	$mean_turnaround$	$sd_turnaround$	throughput
1	FCFS	10.00	49.46	62.68	0.02
2	FCFS	20.00	93.76	65.91	0.04
3	FCFS	100.00	206.54	37.41	0.01
4	SJF	10.00	56.61	69.82	0.02
5	SJF	20.00	71.81	57.08	0.02
6	SJF	100.00	69.60	33.02	0.03
7	SJF	1000.00	90.43	61.40	0.03

1.2 Data Visualization

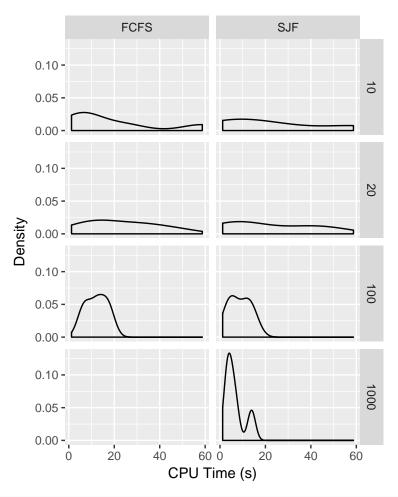
Next we plot the data using density plots and line plots.

```
library(ggplot2)

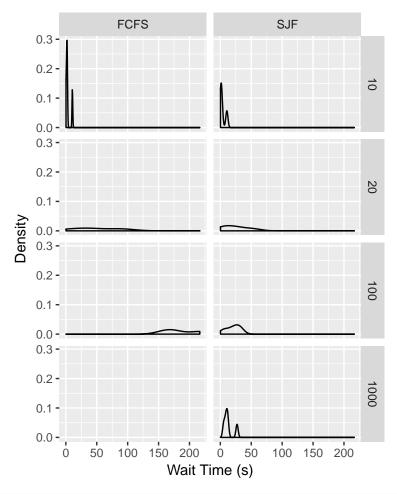
ggplot(data, aes(turnaround)) +
    geom_density() +
    labs(x = 'Turnaround Time (s)', y = 'Density') +
    facet_grid(num_procs~scheduler)
```



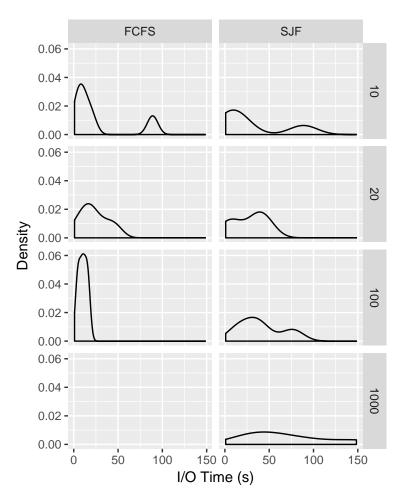
```
ggplot(data, aes(cpu)) +
    geom_density() +
    labs(x = 'CPU Time (s)', y = 'Density') +
    facet_grid(num_procs~scheduler)
```



```
ggplot(data, aes(wait)) +
    geom_density() +
    labs(x = 'Wait Time (s)', y = 'Density') +
    facet_grid(num_procs~scheduler)
```

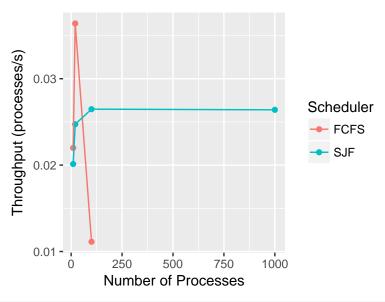


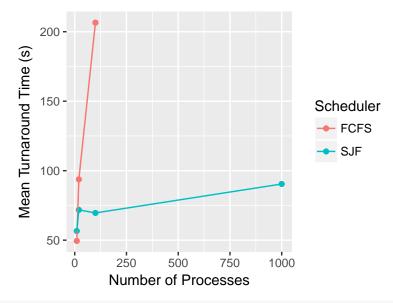
```
ggplot(data, aes(io)) +
   geom_density() +
   labs(x = 'I/O Time (s)', y = 'Density') +
   facet_grid(num_procs~scheduler)
```

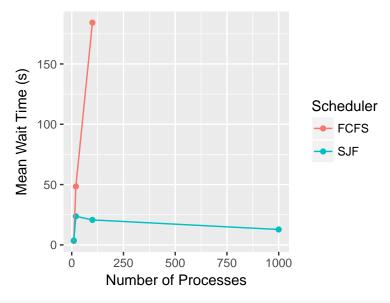


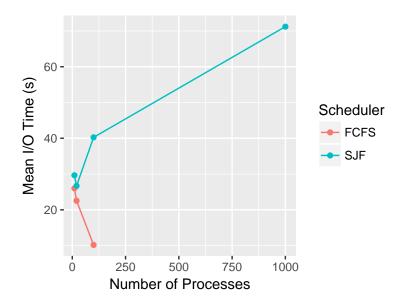
The density plots for FCFS at 1000 processes does not contain anything because when the small CPU burst finishes, the processes has to wait for IO and all the processes already in the ready queue which is a staggering number. Thus, every processes blocks each other for a very long time — longer than the simulation duration of $5\,\mathrm{min}$ — and no process can complete.

```
y = 'Throughput (processes/s)',
color = 'Scheduler')
```









2 Scheduling Algorithm Analysis

From the throughput vs. number of processes graph, one can see that at a sufficiently small number of processes, FCFS does a better job at maximizing throughput. However, when the number of processes is too high, the time to traverse the ready queue in FCFS is incredibly high. This leads to FCFS failing drastically. Conversely, SJF doesn't beat FCFS on throughput, but is stable under a large amount of processes.

From the mean wait time vs. number of processes, one can see that SJF is clearly better at minimizing wait times. Also, as noted before, SJF does not choke on a large amount of processes.

In conclusion, FCFS maximizes throughput and SJF minimizes wait times. However, with limited resources and a sufficiently large amount of processes, FCFS grinds to a halt.