APPENDIX A

More Formulas Derivations for Job Scheduling

1) State steady equation: The stationary probability p_k for state k can be determined by solving the set of balance equations, which state that the flux into a state should be equal to the flux out of this state when the system is stationary [1]:

When state
$$k=0, \lambda p_0=\mu p_1,$$
 $p_1=\rho_1 p_0=n\rho p_0;$ When state $k=1, \lambda p_1=2\mu p_2,$ $p_2=\frac{\rho_1^2}{2!}p_0=\frac{n^2}{2!}\rho^2 p_0;$ When state $k=2, \lambda p_2=3\mu p_3,$ $p_3=\frac{\rho_1^3}{3!}p_0=\frac{n^3}{3!}\rho^3 p_0;$...

When state $k=n-1, \lambda p_{n-1}=n\mu p_n,$ $p_n=\frac{\rho_1^n}{n!}p_0=\frac{n^n}{n!}\rho^n p_0;$ When state $k=n, \lambda p_n=n\mu p_{n+1},$ $p_{n+1}=\frac{\rho_1^{n+1}}{n!n}p_0=\frac{n^n}{n!}\rho^{n+1}p_0;$

In general,

$$p_k = \begin{cases} \frac{\rho_1^k}{k!} p_0 = \frac{n^k}{k!} \rho^k p_0, & 0 \le k < n \\ \frac{\rho_1^k}{n! n^{k-n}} p_0 = \frac{n^n}{n!} \rho^k p_0; & k \ge n \end{cases}$$
 (2)

A. Mean queue length, sojourn time and waiting time:

The variance of mean number of jobs waiting in the queue shows as follow. Since,

When state k = n + r - 1, $\lambda p_{n+r-1} = n\mu p_{n+r}$,

 $p_{n+r} = \frac{\rho_1^{n+r}}{n!n^r} p_0 = \frac{n^n}{n!} \rho^{n+r} p_0.$

$$E(\bar{L}_{wai}^{2}) = \sum_{k=n}^{\infty} (k-n)^{2} p_{k} = \sum_{h=1}^{\infty} h^{2} p_{h+n}$$

$$= \sum_{h=1}^{\infty} \frac{h^{2}}{n!n^{h}} (n\rho)^{h+n} p_{0}$$

$$= \frac{(n\rho)^{n} \rho^{2} p_{0}}{n!} \sum_{h=2}^{\infty} h(h-1)\rho^{h-2} + \frac{(n\rho)^{n} \rho p_{0}}{n!} \sum_{h=1}^{\infty} h\rho^{h-1}$$

$$= \frac{2\rho^{2} \rho_{1}^{n} p_{0}}{n!(1-\rho)^{3}} + \bar{L}_{wai} = \frac{1+\rho}{1-\rho} \bar{L}_{wai}$$
(3)

Thus.

$$\sigma^{2}(\bar{L}_{wai}) = E(\bar{L}_{wai}^{2}) - [E(\bar{L}_{wai})]^{2} = \bar{L}_{wai}(\frac{1+\rho}{1-\rho} - \bar{L}_{wai})$$
 (4)

In addition,

$$E(L_{sys}) = E(L_{wai}) + E(L_{ser})$$
(5)

$$E(W_{soi}) = E(W_{wai}) + E(T_{ser}) \tag{6}$$

we can easily get

$$E(T_{ser}) = \frac{1}{\mu} \tag{7}$$

If we only consider servers of the system, without regarding as the waiting lines outside the servers, it is easy to observe that there are no losses, and therefore the arrival rate in this cloud system is λ , and the mean waiting time of each customer is $E(T_{ser}) = \frac{1}{u}$ [2] [3].

To obtain $E(L_{wai}|q \ge n)$, noted that the evolution of the M/M/n queue during the time when $q \ge n$ is equal to that of M/M/1 queue with the arrival rate λ and the service rate $n\mu$. Therefore, the mean queue length of this kind of M/M/1 queue is equivalent to $\frac{1}{1-\rho}$, where $\rho = \frac{\lambda}{n\mu}$. Therefore,

$$E(L_{wai}|q \ge n) = \frac{\rho/n}{1 - \rho/n} = \frac{\rho}{n - \rho}$$
(8)

Substitute the distribution for the density function of the waiting time, we get

$$f_{w}(x) = \frac{p_{0}(\frac{\lambda}{\mu})^{n}}{n!} n\mu e^{-n\mu x} \sum_{j=0}^{\infty} \frac{(\rho n\mu x)^{j}}{j!}$$

$$= \frac{(\frac{\lambda}{\mu})^{n}}{n!} p_{0} n\mu e^{-(n\mu - \lambda)x}$$

$$= \frac{(\frac{\lambda}{\mu})^{n}}{n!} p_{0} n\mu e^{-n\mu(1-\rho)x}$$

$$= \frac{(\frac{\lambda}{\mu})^{n}}{n!} p_{0} \frac{1}{1-\rho} n\mu (1-\rho) e^{-n\mu(1-\rho)x}$$

$$= P(Waiting) n\mu (1-\rho) e^{-n\mu(1-\rho)x}$$
(9)

Thus for the complement of the the distribution function, we have

$$P(W > x) = \int_{x}^{\infty} f_w(u) du = P(Waiting) e^{-n\mu(1-\rho)x}$$

= $C(n, \rho) \bullet e^{-\mu(n-\frac{\rho}{n})x}$ (10)

The distribution function of waiting time can be written as:

$$F_{w}(x) = 1 - P(Waiting) + P(Waiting)(1 - e^{-n\mu(1-\rho)x})$$

$$= 1 - P(Waiting)e^{-n\mu(1-\rho)x}$$

$$= 1 - C(n,\rho) \bullet e^{-\mu(n-\frac{\rho}{n})x}$$
(11)

By applying the law of total probability for the density function of the sojourn time $f_s(x)$ is given as follow:

$$f_s(x) = P(No\ waiting)\mu e^{-\mu x} + f_{w+ser}(x)$$
 (12)

Whereas, the density function of sojourn time for the job that need to wait first $f_{w+ser}(x)$:

$$f_{w+ser}(z) = \int_{0}^{z} f_{w}(x)\mu e^{-\mu(z-x)}dx$$

$$= P(Waiting)n\mu(1-\rho)\mu\int_{0}^{z} e^{-n\mu(1-\rho)x}e^{-\mu(z-x)}dx$$

$$= \frac{(n\rho)^{n}}{n!}p_{0}\frac{1}{(1-\rho)}n\mu(1-\rho)\mu e^{-z\mu}\int_{0}^{z} e^{-\mu(n-1-\frac{\lambda}{\mu})x}dx$$

$$= \frac{(n\rho)^{n}}{n!}p_{0}n\mu\frac{1}{(n-1-\frac{\lambda}{\mu})}e^{-z\mu}(1-e^{-\mu(n-1-\frac{\lambda}{\mu})z})$$
(13)

Therefore,

$$f_{s}(x) = \left(1 - \left(\frac{\lambda}{\mu}\right)^{n} \frac{p_{0}}{n!(1-\rho)}\right) \mu e^{-\mu x} + \frac{\left(\frac{\lambda}{\mu}\right)^{n}}{n!} n \mu p_{0} \frac{1}{(n-1-\frac{\lambda}{\mu})} e^{-\mu x} \left(1 - e^{-\mu(n-1-\frac{\lambda}{\mu})x}\right)$$

$$= \mu e^{-\mu x} \left(1 - \frac{\left(\frac{\lambda}{\mu}\right)^{n} p_{0}}{n!(1-\rho)} + \frac{\left(\frac{\lambda}{\mu}\right)^{n}}{n!} n p_{0} \frac{1}{(n-1-\frac{\lambda}{\mu})} \left(1 - e^{-\mu(n-1-\frac{\lambda}{\mu})x}\right)\right)$$

$$= \mu e^{-\mu x} \left(1 + \frac{\left(\frac{\lambda}{\mu}\right)^{n} p_{0}}{n!(1-\rho)} \frac{1 - \left(n - \frac{\lambda}{\mu}\right) e^{-\mu(n-1-\frac{\lambda}{\mu})x}}{(n-1-\frac{\lambda}{\mu})}\right)$$
(14)

For the complement of the distribution function of the response time, we get

$$P(S > x) = \int_{x}^{\infty} f_{s}(y)dy =$$

$$\int_{x}^{\infty} \mu e^{-\mu y} + \frac{(\frac{\lambda}{\mu})^{n} p_{0}}{n!(1-\rho)} \frac{1}{(n-1-\frac{\lambda}{\mu})} (\mu e^{-\mu y} - \mu (n-\frac{\lambda}{\mu})e^{-\mu (n-\frac{\lambda}{\mu})y})dy$$

$$= e^{-\mu x} + (\frac{\lambda}{\mu})^{n} p_{0} \frac{1}{n!(1-\rho)(n-1-\frac{\lambda}{\mu})} (e^{-\mu x} - e^{-\mu (n-\frac{\lambda}{\mu})x})$$

$$= e^{-\mu x} \left(1 + \frac{(\frac{\lambda}{\mu})^{n} p_{0}}{n!(1-\rho)} \frac{1 - e^{-\mu (n-1-\frac{\lambda}{\mu})x}}{(n-1-\frac{\lambda}{\mu})}\right)$$
(15)

Therefore the distribution function can be presented

$$F_s(x) = 1 - P(S > x)$$
 (16)

APPENDIX B

MORE EXPERIMENTS RESULTS

We evaluate our PDSonQueue vs YARN fair scheduler performance using 4 real world applications. Case 1 has been

presented in the main technical report. Here we will describe case 2, 3 and 4 and their experimental results.

A. Real world applications

Case 2: Number plate image recognition: This License Plate Recognition System (LPRS) recognises a vehicle plate license from images with edge detection used to identify points in digital images with discontinuities. A 40G image data file is loaded and read once from disk. This application is predominantly *CPU-intensive*.

Case 3: Hadoop log file text search: This application tracks Hadoop's logs to search for error information using a simple lambda expression based on the "error" string, identifying a cluster's health status and weakness. Its complexity is low and it consumes little CPU resource. A 40G log file is buffered in memory, read and searched. Hence, this application is *Memory-intensive*.

Case 4: Hadoop data migration: In a Hadoop cluster, the input file is split into one or more blocks stored in a set of DataNodes (running on commodity machines). When data volume is huge, tasks split from jobs are deployed on one node, however the needed data may be stored on different nodes and even different racks. Thus the system needs to copy other nodes' data to this destination node. A 40G telecommunications data file is copied and transmitted among nodes. It is *I/O-intensive*.

B. Real application evaluation results

Deadline-based QoS, completion time, completion rate and throughput when running use case 2, 3 and 4 together:

There are 6 different group combinations. Group 1 has 4,000 deadline jobs including 2,000 use case2 jobs, 1,000 use case3 jobs and 1,000 use case4 jobs and other group combinations are presented in horizontal axis in Fig. 1. Fig. 1 presents QoS rate, and our PDSonQueue's QoS rate is 85.37% and fair scheduler' QoS rate is 69.62%, which has 15.75% improvement. When deadline jobs occupy 33% of total jobs, for both schedulers, the QoS achievement rate is higher than that of the deadline jobs occupying 50% of the total jobs. When the number of deadline jobs is high, there are more preemptions from regular jobs and more failed deadline jobs.

In Fig. 2, PDSonQueue's average throughput is 655.37 job units/s and fair scheduler's is only 509.47 job units/s. PDSonQueue can improve 28.63% for throughput. When the rate of deadline jobs is 33%, PDSonQueue's throughput is 686.02 job units/s and fair scheduler's throughput is only 524.35 job units/s. Yet, when the rate of deadline jobs is 50%, PDSonQueue's throughput reduces to 624.72 job units/s, and fair scheduler's throughput also cuts down to 494.58 job units/s. When deadline jobs are less, then throughput is higher.

Fig. 3 shows completion times of different job combinations. The average completion time is 6112.83s using PDSon-Queue and average completion time is 6331.45s. PDSonQueue can reduce completion time of 218.62s, compared with YARN scheduler.

In Fig. 4, PDSonQueue's completion rate is 70.98% and fair scheduler's is 51.48%, which is a near 20% improvement.

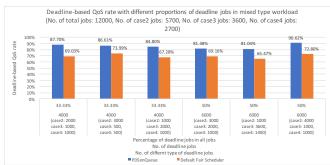


Fig. 1. Deadline-based QoS rate on a mixed use cases workload

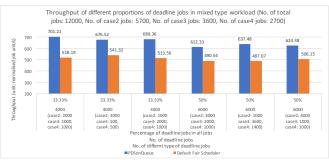


Fig. 2. Throughput on a mixed use cases workload

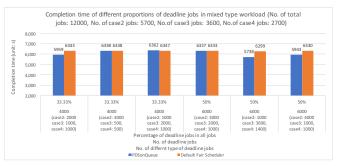


Fig. 3. Completion time on a mixed use cases workload

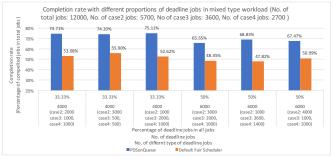


Fig. 4. Completion rate on a mixed use cases workload

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