# ANALYSIS OF READY QUEUE PROCESSING TIME UNDER PPS-LS AND SRS-LS SCHEME IN MULTIPROCESSING ENVIRONMENT 

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#### Abstract

In operating system, process sequencing is an open problem and solved by many scientists/authors suggesting different scheduling schemes. Every process needs a time span to be processed by the CPU. Lottery scheduling is one such scheme where the process selection is purely on random basis. The ready queue is used for processes to wait there until selected for processor. This paper considers the environment of many processors, a ready queue, lottery scheduling and presents an efficient method to predict about total time needed to process the entire ready queue if only few are processed in a specified time. Confidence internals are calculated based on PPS-LS and compared with SRS-LS. The PPS-LS found better over SRS-LS.


Keywords: Scheduling, Probability Proportional to Size Lottery Scheduling (PPS-LS), Bias, Variance, Confidence interval, Simple Random Sampling Lottery Scheduling (SRS-LS), Estimator, Sampling, Ready Queue.

## 1. INTRODUCTION

In multiprocessor and multi-user environment, in a fraction of second, many job requests arrive at the ready queue. Scheduler's role is to decide which of the process should be assigned to which processor at an instant of time. There are many CPU scheduling schemes exist in literature \{See [15], [17], [18]\} and Lottery scheduling is one of them whose algorithm is as under:
i. Allot token number to each of the process entering into the ready queue.
ii. Pickup a random number and search matching process token number to random number.
iii. The process which is matched to both numbers shall be assigned to processor by scheduler.
iv. Continue steps (i) (ii) and (iii) unless ready queue is vacated.

A modification in the Lottery scheduling procedure for $n$ processors environment is to choose $n$ random numbers at a time and perform $n$ matching of tokens of processes. After matching, select $n$ processes at a time randomly.

Carl et al. [1] discussed the proportional share resource management technique in lottery scheduling. David et al. [3] presented the specialization matching methodology in context to lottery scheduling. Raz et al. [4] presented procedure of deciding priorities among jobs by maintaining fairness in selection procedure.

Shukla et.al [7] picked up multiprocessor environment and lottery scheduling and discussed a procedure to obtain ready queue time estimate. The kind of prediction is important for backup management when sudden failure of power supply, machine disorder occur. Shukla et al [8] discussed similar problem but in grouped setup of ready queue using lottery scheduling. Shukla and Jain [5], [6] tackled Markov Chain based study of transition behaviors of scheduler in multilevel queue scheduling. Some other contributions are [9], [10], [11], [12], [13] and [14]. Sampling theory contains tools and techniques useful to estimate the unknown population parameter. The basis for this is information contained in a small sample which is a part of the whole \{See [2] and [16]\}.

One can think of utilizing additional information for ready queue processing time estimation. Suppose the size of each process is available in terms of bytes at the time of entering into the ready queue. This information could be a source of efficient estimation. Aim of present content is to
utilize this in the estimation of ready queue processing time under PPS-LS setup and multiprocessor environment by using size measure of the process.

## 2. PROBLEM DEFINITION

Let the size of coming $i^{\text {th }}$ process in the ready queue be $X_{i}(i=1,2, \ldots ., N)$ in terms of bytes, the total size being $X=\Sigma \AA$. We associate the random numbers 1 to $X$ with the first unit, the unit with which this number is associated is selected. This gives the surety of selection of $\mathrm{i}^{\text {th }}$ unit in the ready queue or pool of processes with probability proportional to its size $\mathrm{X}_{\mathrm{i}}$. This procedure is repeated $n$ times with replacement of the processes. The $X_{i}$ is a size measure of processes based on bytes.

## 3. ESTIMATION OF READY QUEUE PROCESSING TIME

## (A)PPS-LS scheme with replacement

Consider a pool of N processes and let $\mathrm{X}_{\mathrm{i}}$ be size value of the process Pi where ( $\mathrm{i}=1 \ldots \mathrm{~N}$ ). Suppose ( $\mathrm{c}_{\mathrm{i}}=\mathrm{X}_{\mathrm{i}} / \mathrm{X}$ ) be the chance that $\mathrm{i}^{\text {th }}$ process the process $\mathrm{c}_{\mathrm{i}}$ is selected in processor such that $\sum_{i}^{N} c_{i}=1$.

The Y be processing time as main variable .Let n independent choice be made with replacement method and the value of $y_{i}$ for each selected process is observed. Take $y_{i}$ (time), $c_{i}$ (chance) be the size and chance of selection of the $\mathrm{i}^{\text {th }}$ process in the sample. It can be seen that random variate $\left(\frac{y_{i}}{c_{i}}\right)(\mathrm{i}=1 \ldots \mathrm{~N})$ are independent and identically distributed. If $\mathrm{c}_{\mathrm{i}}=1 / \mathrm{N}$ it gives rise to a simple random sample. Consider estimator $\hat{Y}_{p p s}=\frac{1}{n} \sum_{i}^{n}\left(\frac{y_{i}}{c_{i}}\right)$ which is unbiased estimator of the ready queue processing time total $Y$.

## Sampling variance

We define $Z_{i}=Y_{i} / \mathrm{Nc}_{\mathrm{i}}$ and $\mathrm{z}_{\mathrm{i}}=\mathrm{y}_{\mathrm{i}} / \mathrm{Nc}_{\mathrm{i}}$. The $\mathrm{Z}_{\mathrm{i}}$ denotes values of processes in ready queue and $\mathrm{z}_{\mathrm{i}}$ relates to value in sample $n$ assigned to $n$ processors ( $n<N$ ).

$$
\begin{aligned}
= & \frac{1}{n} \sum_{i}^{N} \varphi_{i}\left(\frac{x}{q_{1}}-V\right)^{2} \\
& \operatorname{cov}\left(\frac{x}{\varepsilon_{i}} \frac{y_{i}}{c_{i}}\right)
\end{aligned}
$$

The variance of estimator is inversely proportional to the sample size $n$ as in simple random sampling, wr. An unbiased estimator of population mean is, $\bar{Y}$ is given by

$$
\hat{f}_{p p s}=\frac{1}{n_{N} \mathbb{N}} \sum_{i}^{2}\left(\frac{q}{\rho}\right) \text { with its sampliug vatiance. }
$$

In PPS-LS sampling wr,

$$
\begin{gathered}
V\left(\rho_{p g n}\right)=\frac{1}{n(n-1)} \sum_{i}^{n}\left(\frac{y_{1}}{q_{1}}-\rho_{p p z}\right)^{2} \\
z_{i}\left(=\frac{y_{i}}{c_{i}}\right), i=1,2 \ldots \ldots \ldots \ldots . \ldots
\end{gathered}
$$

independent unbiased estimator of $Y$ having the same variance. In PPS-LS sampling, wr an unbiased estimator of $v\left(\hat{\bar{Y}}_{p p s}\right)$ is given by

$$
s^{2}=\sum_{i}^{n}\left(\frac{z_{1}-\bar{z}}{n-1}\right)^{2}
$$

where is unbiased estimator of with the usual meaning.

$$
\begin{aligned}
& \nabla\left(f_{y N}\right)=\frac{1}{n} \sum_{1}^{y} c_{i}\left(\frac{y_{z}}{N c_{i}}-\bar{F}\right)^{2} \\
& v\left(f_{p p s}\right)=v(j)=\frac{1}{n^{2}} \sum^{n} v\left(\sigma_{p}\right) \\
& \because E\left(\gamma_{p p 2}\right)=E(\bar{z})=\sum_{i}^{n} \frac{1}{n} E\left(z_{i}\right)=Y \\
& v\left(\gamma_{p m}\right)=\frac{1}{n} \sum_{1}^{n} c_{1}\left(\frac{Y}{c_{1}}-V\right)^{2} \\
& =\frac{1}{n(n-1)}\left[\sum_{i}^{n}\left(\frac{y_{i}}{c_{i}}\right)^{2}-n \hat{Y}_{p p s}^{2}\right]
\end{aligned}
$$

So, an unbiased estimator of $V\left(\hat{\bar{Y}}_{p p s}\right)$ is given by

$$
\begin{aligned}
& v\left(\hat{f}_{p m}\right)=\frac{1}{n(n n-1)} \sum^{2}\left(\frac{n}{r q}-\hat{\gamma}_{p p s}\right)^{2} \\
& =\frac{1}{n(n-1) N}\left[\sum^{n}\left(\frac{x}{c_{2}}\right)^{2}-n \hat{p}_{p n}\right]
\end{aligned}
$$

The confidence interval for mean in PPS-LS is

$$
\frac{\mathrm{T}_{\mathrm{pz}}}{\mathrm{~N}}-\frac{1}{\mathrm{~N}} \sqrt{\mathrm{~V}\left(\mathcal{Y}_{\mathrm{p}}\right.}, \quad \frac{\mathrm{T}_{\mathrm{pm}}}{\mathrm{x}}+\frac{1}{\mathrm{~N}} \sqrt{\mathrm{~V} \rho_{\mathrm{p}}}
$$

## (B) SRS-LS scheme

Consider $\quad c_{i}=$ for all then we get SRS-LS set-up as $\quad z_{i}=$ and $\quad z_{i}=$ with process sample mean

$$
\begin{gathered}
\bar{y}=\frac{1}{n} \sum_{=1}^{n} x \text { Alse whole ready queue mean a square is } s^{2} \\
=\frac{1}{n-1} \sum_{=i}^{n}(\mathrm{n}-\bar{y})^{2}
\end{gathered}
$$

The $\bar{y}$ denotes mean time of sample processes in the n processors at a time.This generates confidence interval for SRS-LS for mean $\left[\bar{y}-3 \sqrt{p(\bar{y})}, \bar{y}+3 \sqrt{V(G)}\right.$ where $V(\bar{y})=\frac{N-n}{N n}$.

Note that when confidence interval length is high, the estimation procedure is less efficient. This comparative methodology is adopted here in this content.

## 4. NUMERICAL DATA

Consider 30 processes in ready queue at a time whose size measure X is also given in terms of bytes. If we assume that all the processes are processed completely in the ready queue, the CPU burst time Y is mentioned against them. The table 1 presents computation of size measure probability.

Table 1: Processes parameters in ready queue.

| Process Number | Process |  | $\mathrm{c}_{\mathrm{i}}=\mathrm{X}_{\mathrm{i}} / \mathbf{X}$ | PPS-LS | SRS-LS |
| :---: | :---: | :---: | :---: | :---: | :---: |
|  | $\operatorname{Size}\left(\mathrm{X}_{\mathrm{i}}\right)$ Parameter | $\begin{gathered} \text { CPU } \\ \text { Burst } \\ \text { Time }\left(\mathbf{Y}_{\mathbf{i}}\right) \end{gathered}$ |  | $\left(\mathbf{Y}_{i} / \mathrm{Nc}_{\mathrm{i}}{ }^{\text {) }}\right.$ | $\mathbf{Y i} /\left(\mathrm{Nc}_{\mathrm{i}}\right)$ |
| 1 | 210 | 30 | 0.01974 | 50.66 | 30.03 |
| 2 | 897 | 20 | 0.08434 | 7.9 | 20.02 |
| 3 | 312 | 112 | 0.02933 | 127.29 | 112.11 |
| 4 | 171 | 40 | 0.01608 | 82.92 | 40.04 |
| 5 | 461 | 59 | 0.04334 | 45.38 | 59.06 |
| 6 | 290 | 60 | 0.02727 | 73.34 | 60.06 |
| 7 | 379 | 30 | 0.03563 | 28.07 | 30.03 |
| 8 | 220 | 43 | 0.02068 | 69.31 | 43.04 |
| 9 | 470 | 101 | 0.04419 | 76.19 | 101.1 |
| 10 | 636 | 69 | 0.0598 | 38.46 | 69.07 |
| 11 | 455 | 138 | 0.04278 | 107.53 | 138.14 |
| 12 | 682 | 43 | 0.06412 | 22.35 | 43.04 |
| 13 | 952 | 109 | 0.08951 | 40.59 | 109.11 |
| 14 | 574 | 26 | 0.05397 | 16.06 | 26.03 |
| 15 | 536 | 74 | 0.05039 | 48.95 | 74.07 |
| 16 | 416 | 89 | 0.03911 | 75.85 | 89.09 |
| 17 | 788 | 123 | 0.07409 | 55.34 | 123.12 |
| 18 | 902 | 67 | 0.08481 | 26.33 | 67.07 |
| 19 | 623 | 58 | 0.05857 | 33.01 | 58.06 |
| 20 | 563 | 84 | 0.05293 | 52.9 | 84.08 |
| 21 | 111 | 143 | 0.01044 | 44.7 | 14.01 |
| 22 | 341 | 29 | 0.03206 | 30.15 | 29.03 |
| 23 | 775 | 147 | 0.07287 | 21.5 | 47.05 |
| 24 | 913 | 94 | 0.08584 | 36.5 | 94.09 |
| 25 | 745 | 131 | 0.07005 | 62.34 | 131.13 |
| 26 | 130 | 79 | 0.01222 | 215.49 | 79.08 |
| 27 | 877 | 46 | 0.08246 | 18.59 | 46.05 |
| 28 | 927 | 59 | 0.08716 | 22.56 | 59.06 |
| 29 | 424 | 72 | 0.03986 | 60.21 | 72.07 |
| 30 | 356 | 22 | 0.03347 | 21.91 | 22.02 |

The table 2 presents the allotment of range of random numbers and table $3,4,5$ and 6 are for sample information in terms of size measure based probability. Note the true value of total is 1968 and entire mean is 65.6.

Table 2: Interval Random Number Association

| Process <br> Number | Size( $\mathbf{x}_{\mathbf{i}}$ ) <br> Parameter | Cumulative <br> Size Totals | Random <br> Numbers <br> Associated |
| :---: | :---: | :---: | :---: |
| $\mathbf{1}$ | 210 | 210 | $0-210$ |
| $\mathbf{2}$ | 897 | 1107 | $210-1107$ |
| $\mathbf{3}$ | 312 | 1419 | $1107-1419$ |
| $\mathbf{4}$ | 171 | 1590 | $1419-1590$ |
| $\mathbf{5}$ | 461 | 2051 | $1590-2051$ |
| $\mathbf{6}$ | 290 | 2341 | $2051-2341$ |
| $\mathbf{7}$ | 379 | 2720 | $2341-2720$ |
| $\mathbf{8}$ | 220 | 2940 | $2720-2940$ |
| $\mathbf{9}$ | 470 | 3410 | $2940-3410$ |
| $\mathbf{1 0}$ | 636 | 4046 | $3410-4046$ |
| $\mathbf{1 1}$ | 455 | 4501 | $4046-4501$ |
| $\mathbf{1 2}$ | 682 | 5183 | $4501-5183$ |
| $\mathbf{1 3}$ | 952 | 6135 | $5183-6135$ |
| $\mathbf{1 4}$ | 574 | 6709 | $6135-6709$ |
| $\mathbf{1 5}$ | 536 | 7245 | $6709-7245$ |
| $\mathbf{1 6}$ | 416 | 7661 | $7245-7661$ |
| $\mathbf{1 7}$ | 788 | 8449 | $7661-8449$ |
| $\mathbf{1 8}$ | 902 | 9351 | $8449-9351$ |
| $\mathbf{1 9}$ | 623 | 9974 | $9351-9974$ |
| $\mathbf{2 0}$ | 563 | 10537 | $9974-10537$ |
| $\mathbf{2 1}$ | 111 | 10648 | $10537-10648$ |
| $\mathbf{2 2}$ | 341 | 10989 | $10648-10989$ |
| $\mathbf{2 3}$ | 775 | 11764 | $10989-11764$ |
| $\mathbf{2 4}$ | 913 | 12677 | $11764-12677$ |
| $\mathbf{2 5}$ | 745 | 13422 | $12677-13422$ |
| $\mathbf{2 6}$ | 130 | 13552 | $13422-13552$ |
| $\mathbf{2 7}$ | 877 | 14429 | $13552-14429$ |
| $\mathbf{2 8}$ | 927 | 15356 | $14429-15356$ |
| $\mathbf{2 9}$ | 424 | 15780 | $15356-15780$ |
| $\mathbf{3 0}$ | 356 | 16136 | $15780-16136$ |
| $\mathbf{2 5}$ |  |  |  |

Table 3: Sampled processes ( $\mathrm{n}=8$ )
Sample No. Sampled Processes

| $\mathbf{1}$ | $\mathrm{P}_{30}$ | $\mathrm{P}_{17}$ | $\mathrm{P}_{6}$ | $\mathrm{P}_{25}$ | $\mathrm{P}_{15}$ |
| :--- | :--- | :--- | :--- | :--- | :--- |
| $\mathbf{2}$ | $\mathrm{P}_{8}$ | $\mathrm{P}_{4}$ | $\mathrm{P}_{18}$ | $\mathrm{P}_{25}$ | $\mathrm{P}_{28}$ |
| $\mathbf{3}$ | $\mathrm{P}_{20}$ | $\mathrm{P}_{4}$ | $\mathrm{P}_{11}$ | $\mathrm{P}_{1}$ | $\mathrm{P}_{22}$ |
| $\mathbf{4}$ | $\mathrm{P}_{12}$ | $\mathrm{P}_{22}$ | $\mathrm{P}_{4}$ | $\mathrm{P}_{16}$ | $\mathrm{P}_{24}$ |
| $\mathbf{5}$ | $\mathrm{P}_{3}$ | $\mathrm{P}_{6}$ | $\mathrm{P}_{7}$ | $\mathrm{P}_{18}$ | $\mathrm{P}_{12}$ |
| $\mathbf{6}$ | $\mathrm{P}_{27}$ | $\mathrm{P}_{19}$ | $\mathrm{P}_{13}$ | $\mathrm{P}_{2}$ | $\mathrm{P}_{6}$ |
| $\mathbf{7}$ | $\mathrm{P}_{24}$ | $\mathrm{P}_{16}$ | $\mathrm{P}_{8}$ | $\mathrm{P}_{5}$ | $\mathrm{P}_{19}$ |
| $\mathbf{8}$ | $\mathrm{P}_{4}$ | $\mathrm{P}_{12}$ | $\mathrm{P}_{15}$ | $\mathrm{P}_{16}$ | $\mathrm{P}_{29}$ |
| $\mathbf{Y}_{\mathbf{i}}$ | Sampled Processes CPU Burst Time |  |  |  |  |

burst time for

|  | Factor |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
| $\mathbf{1}$ | 22 | 123 | 60 | 131 | 74 |
| $\mathbf{2}$ | 43 | 40 | 67 | 131 | 59 |
| $\mathbf{3}$ | 84 | 40 | 138 | 30 | 29 |
| $\mathbf{4}$ | 43 | 29 | 40 | 89 | 94 |
| $\mathbf{5}$ | 112 | 60 | 33 | 67 | 43 |
| $\mathbf{6}$ | 46 | 58 | 109 | 20 | 60 |
| $\mathbf{7}$ | 94 | 89 | 43 | 59 | 58 |
| $\mathbf{8}$ | 40 | 43 | 74 | 89 | 72 |

Table 5: Size measure of sampled process ( $\mathrm{n}=\mathrm{8}$ )

| $\mathbf{X}_{\mathbf{i}}$ | Sampled Processes Weight Factor |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
| $\mathbf{1}$ | 356 | 788 | 290 | 745 | 536 |
| $\mathbf{2}$ | 220 | 902 | 745 | 171 | 927 |
| $\mathbf{3}$ | 563 | 171 | 455 | 210 | 341 |
| $\mathbf{4}$ | 682 | 171 | 341 | 913 | 416 |
| $\mathbf{5}$ | 312 | 290 | 379 | 902 | 682 |
| $\mathbf{6}$ | 877 | 623 | 952 | 879 | 290 |
| $\mathbf{7}$ | 913 | 416 | 220 | 561 | 623 |
| $\mathbf{8}$ | 171 | 682 | 536 | 416 | 424 |

Table 6: The selection probability based on size measure

| $\mathbf{c}_{\mathbf{i}}$ | Selection Chances of Sampled Processes |  |  |  |  |
| :--- | :--- | :--- | :--- | :--- | :--- |
| $\mathbf{1}$ | 0.03347 | 0.07409 | 0.02727 | 0.07005 | 0.05039 |
| $\mathbf{2}$ | 0.02068 | 0.01608 | 0.08481 | 0.07005 | 0.08716 |
| $\mathbf{3}$ | 0.05293 | 0.01608 | 0.04278 | 0.01974 | 0.03206 |
| $\mathbf{4}$ | 0.06412 | 0.03206 | 0.01608 | 0.03911 | 0.08584 |
| $\mathbf{5}$ | 0.02933 | 0.02727 | 0.03563 | 0.08481 | 0.06412 |
| $\mathbf{6}$ | 0.08246 | 0.05857 | 0.08951 | 0.08434 | 0.02727 |
| $\mathbf{7}$ | 0.08584 | 0.03911 | 0.02068 | 0.04334 | 0.05857 |
| $\mathbf{8}$ | 0.01608 | 0.06412 | 0.05039 | 0.03911 | 0.03986 |

## 5. RESULT AND DISCUSSION

In light of table 7 the mean, variance, confidence interval show that PPS-LS means are closer to true value than SRS-LS. The confidence intervals generated by PPS-LS are smaller than generated by SRS-LS. The estimator efficiently estimates the ready queue processing time.

Table 7: Computation of Sample Mean and Confidence Interval for ULS and PPS Scheduling Scheme

| Random <br> Samples | Sampled Process <br> $(\mathbf{k}=\mathbf{5})$ | Mean |  | Variance | Confidence Intervals |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| $\mathbf{1}$ | $\mathrm{P}_{30}, \mathrm{P}_{17}, \mathrm{P}_{6}, \mathrm{P}_{25}, \mathrm{P}_{15}$ | 52.37 | 82.08 | $(30.44-74.31)$ | $(21.16-143)$ |  |
| $\mathbf{2}$ | $\mathrm{P}_{8}, \mathrm{P}_{4}, \mathrm{P}_{18}, \mathrm{P}_{25}, \mathrm{P}_{28}$ | 61.02 | 65.26 | $(49.16-72.88)$ | $(115.53-109.98)$ |  |
| $\mathbf{3}$ | $\mathrm{P}_{20}, \mathrm{P}_{4}, P_{11}, \mathrm{P}_{1}, \mathrm{P}_{22}$ | 61.15 | 64.26 | $(47.43-74.87)$ | $(1.16-127.36)$ |  |
| $\mathbf{4}$ | $P_{12}, \mathrm{P}_{22}, \mathrm{P}_{4}, \mathrm{P}_{16}, \mathrm{P}_{24}$ | 85.34 | 56.05 | $(50.68-120.0)$ | $(20.7-91.4)$ |  |
| $\mathbf{5}$ | $\mathrm{P}_{3}, \mathrm{P}_{6}, \mathrm{P}_{7}, \mathrm{P}_{18}, \mathrm{P}_{12}$ | 55.47 | 62.46 | $(35.25-75.68)$ | $(20.42-104.42)$ |  |
| $\mathbf{6}$ | $\mathrm{P}_{27}, \mathrm{P}_{19}, \mathrm{P}_{13}, \mathrm{P}_{2}, \mathrm{P}_{6}$ | 34.69 | 58.6 | $(1.09-68.29)$ | $(15.13-102.07)$ |  |
| $\mathbf{7}$ | $\mathrm{P}_{24}, \mathrm{P}_{16}, \mathrm{P}_{8}, \mathrm{P}_{5}, \mathrm{P}_{19}$ | 52 | 68.66 | $(25.91-78.09)$ | $(39.23-98.09)$ |  |
| $\mathbf{8}$ | $\mathrm{P}_{4}, \mathrm{P}_{12}, \mathrm{P}_{15}, \mathrm{P}_{16}, \mathrm{P}_{29}$ | 58.05 | 63 | $(59.51-66.48)$ | $(53.53-73.46)$ |  |

## 6. CONCLUSION

The content of this paper suggests an estimation technique for obtaining sampled based estimate of total time required for processing for the ready queue. When we consider size measure of processes as additional information we get better estimate of processing time. This estimate is useful when sudden breakdown of system occurs and system manager needs time valuation to vacate the entire ready holding the ready processes. The estimate helps for disaster and backup management of CPU system. The proposed methodology PPS-LS is efficient than SRS-LS due to incorporation of additional information in the form of size measure of processes in ready queue.

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