

Proof of the Theorem in NSF Proposal “Multi-job Production Systems Engineering”

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Theorem: Consider a two-machine serial line defined by work-based model (i)-(v) and operating in the synchronous SJP regime. Then, the expressions for the performance metrics of this system are:

$$\begin{aligned} PR\left(\frac{w_i}{W_i}\right) &= e_2 \left(1 - Q\left(\lambda_1, \mu_1, \lambda_2, \mu_2, N \frac{w_i}{W_i}\right)\right) \\ &= e_1 \left(1 - Q\left(\lambda_2, \mu_2, \lambda_1, \mu_1, N \frac{w_i}{W_i}\right)\right), \end{aligned} \quad (1)$$

$$TP\left(\frac{w_i}{W_i}\right) = \frac{60W_i}{w_i} PR\left(\frac{w_i}{W_i}\right), \quad (2)$$

$$WIP\left(\frac{w_i}{W_i}\right) = \begin{cases} \frac{D_5\left(\frac{w_i}{W_i}\right)}{D_2\left(\frac{w_i}{W_i}\right) + D_3\left(\frac{w_i}{W_i}\right) + D_4\left(\frac{w_i}{W_i}\right)}, & \text{if } e_1 \neq e_2, \\ \frac{\left(\frac{D_2\left(\frac{w_i}{W_i}\right)}{2} + D_4\left(\frac{w_i}{W_i}\right)\right)N \frac{w_i}{W_i}}{D_2\left(\frac{w_i}{W_i}\right) + D_3\left(\frac{w_i}{W_i}\right) + D_4\left(\frac{w_i}{W_i}\right)}, & \text{if } e_1 = e_2, \end{cases} \quad (3)$$

$$ST_2\left(\frac{w_i}{W_i}\right) = e_2 Q\left(\lambda_1, \mu_1, \lambda_2, \mu_2, N \frac{w_i}{W_i}\right), \quad (4)$$

$$BL_1\left(\frac{w_i}{W_i}\right) = e_1 Q\left(\lambda_2, \mu_2, \lambda_1, \mu_1, N \frac{w_i}{W_i}\right), \quad (5)$$

where (λ_i, μ_i) , $i = 1, 2$ are the breakdown and repair rates of the machines, respectively, N is the buffer capacity, and

$$Q\left(\lambda_1, \mu_1, \lambda_2, \mu_2, N \frac{w_i}{W_i}\right) = \begin{cases} \frac{(1-e_1)(1-\phi)}{1-\phi e^{-\beta N \frac{w_i}{W_i}}}, & \text{if } \frac{\lambda_1}{\mu_1} \neq \frac{\lambda_2}{\mu_2}, \\ \frac{\lambda_1(\lambda_1+\lambda_2)(\mu_1+\mu_2)}{(\lambda_1+\mu_1)[(\lambda_1+\lambda_2)(\mu_1+\mu_2)+\lambda_2\mu_1(\lambda_1+\lambda_2+\mu_1+\mu_2)N \frac{w_i}{W_i}]}, & \text{if } \frac{\lambda_1}{\mu_1} = \frac{\lambda_2}{\mu_2}, \end{cases} \quad (6)$$

$$e_i = \frac{\lambda_i}{\lambda_i + \mu_i}, \quad \phi = \frac{e_1(1 - e_2)}{e_2(1 - e_1)}, \quad \beta = \frac{(\lambda_1 + \lambda_2 + \mu_1 + \mu_2)(\lambda_1\mu_2 - \lambda_2\mu_1)}{(\lambda_1 + \lambda_2)(\mu_1 + \mu_2)},$$

$$D_1 = \frac{\mu_1 + \mu_2}{\lambda_1 + \lambda_2},$$

$$\begin{aligned}
D_2 \left(\frac{w_i}{W_i} \right) &= \begin{cases} \left(2 + D_1 + \frac{1}{D_1} \right) \frac{e^{KN \frac{w_i}{W_i}} - 1}{K \frac{w_i}{W_i}}, & \text{if } e_1 \neq e_2, \\ \left(2 + D_1 + \frac{1}{D_1} \right) N \frac{w_i}{W_i}, & \text{if } e_1 = e_2, \end{cases} \\
D_3 \left(\frac{w_i}{W_i} \right) &= \frac{(\lambda_1 + \mu_1 + \lambda_2 + \mu_2)(\lambda_2 + \mu_1) + \lambda_1 \mu_2 - \lambda_2 \mu_1}{\lambda_2 \mu_1 (\lambda_1 + \mu_1 + \lambda_2 + \mu_2) \frac{w_i}{W_i}}, \\
D_4 \left(\frac{w_i}{W_i} \right) &= \frac{(\lambda_1 + \mu_1 + \lambda_2 + \mu_2)(\lambda_2 + \mu_1) + \lambda_2 \mu_1 - \lambda_1 \mu_2}{\lambda_1 \mu_2 (\lambda_1 + \mu_1 + \lambda_2 + \mu_2) \frac{w_i}{W_i}} e^{KN \frac{w_i}{W_i}}, \\
D_5 \left(\frac{w_i}{W_i} \right) &= \frac{D_2 \left[1 + \left(KN \frac{w_i}{W_i} - 1 \right) e^{KN \frac{w_i}{W_i}} \right]}{K \frac{w_i}{W_i} \left(e^{KN \frac{w_i}{W_i}} - 1 \right)} + D_4 N \frac{w_i}{W_i}, \\
K &= \begin{cases} \frac{(\lambda_1 + \lambda_2 + \mu_1 + \mu_2)(\lambda_2 \mu_1 - \lambda_1 \mu_2)}{(\lambda_1 + \lambda_2)(\mu_1 + \mu_2)}, & \text{if } e_1 \neq e_2, \\ 0, & \text{if } e_1 = e_2. \end{cases}
\end{aligned}$$

Proof: System under consideration can be described by a continuous time, mixed space Markov process with the states defined as a triple (h, s_1, s_2) , where h is the state of the buffer and $s_i, i = 1, 2$ are the states of the first and the second machine, respectively. States in this problem are categorized into two groups: (1) boundary states $(0, s_1, s_2), (N, s_1, s_2)$, and (2) internal states $(h, s_1, s_2), 0 < h < N$. Boundary states are described by their probability mass functions whereas internal states are described by their density functions.

Transition rates of the process from state $(s_1 = i, s_2 = j)$ to state $(s_1 = k, s_2 = l)$, i.e., $\nu_{kl,ij}$, are as follows. Note that below, we use the notation τ as a shorthand to refer to $\tau := \frac{w_1}{W_1} (= \frac{w_2}{W_2})$.

$$\begin{aligned}
\nu_{11,00} &= 0, & \nu_{10,00} &= \mu_1, & \nu_{01,00} &= \mu_2, \\
\nu_{11,01} &= \mu_1, & \nu_{10,01} &= 0, & \nu_{00,01} &= \lambda_2, \\
\nu_{11,10} &= \mu_2, & \nu_{01,10} &= 0, & \nu_{00,10} &= \lambda_1, \\
\nu_{10,11} &= \lambda_2, & \nu_{01,11} &= \lambda_1, & \nu_{00,11} &= 0.
\end{aligned}$$

While the discrete part of the Markov process at hand is described by the above transition rates, the continuous part, i.e., buffer occupancy, is characterized by

$$\frac{dh(t)}{dt} \Big|_{0 < h(t) < N} = \begin{cases} \frac{1}{\tau}, & \text{if } s_1(t) = 1, s_2(t) = 0, \\ 0, & \text{if } s_1(t) = s_2(t), \\ -\frac{1}{\tau}, & \text{if } s_1(t) = 0, s_2(t) = 1, \end{cases} \quad (7)$$

$$\frac{dh(t)}{dt} \Big|_{h(t)=0} = \begin{cases} \frac{1}{\tau}, & \text{if } s_1(t) = 1, s_2(t) = 0, \\ 0, & \text{otherwise,} \end{cases} \quad (8)$$

$$\frac{dh(t)}{dt} \Big|_{h(t)=N} = \begin{cases} -\frac{1}{\tau}, & \text{if } s_1(t) = 0, s_2(t) = 1, \\ 0, & \text{otherwise,} \end{cases} \quad (9)$$

The following can be obtained for the steady state pdf for the internal states $h \in (0, N)$,

$$\begin{aligned}
f_{H,I,J}(h, 1, 1) &= f_{H,I,J}(h, 1, 0) \frac{\mu_2}{\lambda_1 + \lambda_2} + f_{H,I,J}(h, 0, 1) \frac{\mu_1}{\lambda_1 + \lambda_2}, \\
\frac{\partial f_{H,I,J}(h, 1, 0)}{\partial h} &= f_{H,I,J}(h, 1, 1) \lambda_2 \tau - f_{H,I,J}(h, 1, 0) (\lambda_1 + \mu_2) \tau + f_{H,I,J}(h, 0, 0) \mu_1 \tau, \\
\frac{\partial f_{H,I,J}(h, 0, 1)}{\partial h} &= f_{H,I,J}(h, 0, 1) (\lambda_2 + \mu_1) \tau - f_{H,I,J}(h, 1, 1) \lambda_1 \tau - f_{H,I,J}(h, 0, 0) \mu_2 \tau, \\
f_{H,I,J}(h, 0, 0) &= f_{H,I,J}(h, 1, 0) \frac{\lambda_1}{\mu_1 + \mu_2} + f_{H,I,J}(h, 0, 1) \frac{\lambda_2}{\mu_1 + \mu_2}.
\end{aligned} \tag{10}$$

Since $h(t)$ is continuous in t , events

$$A_1 = \{h(t) = h, h(t) \text{ is increasing in } t\}$$

and

$$A_2 = \{h(t) = h, h(t) \text{ is decreasing in } t\}$$

occur alternately with the equal frequencies. Therefore,

$$f_{H,I,J}(h, 1, 0) = f_{H,I,J}(h, 0, 1) \tag{11}$$

The stationary probabilities of the boundary states, $h = 0$ and $h = N$, are derived as follows. The evolution equations for these probabilities are:

$$\begin{aligned}
P_{0,11}(t + \delta t) &= P_{0,11}(t)[1 - (\lambda_1 + \lambda_2)\tau\delta t] + P_{0,01}(t)\mu_1\tau\delta t + o(\delta t), \\
P_{0,10}(t + \delta t) &= 0, \\
P_{0,01}(t + \delta t) &= P_{0,01}(t)[1 - (\mu_1 + \lambda_2)\tau\delta t] + P_{0,11}(t)\lambda_1\tau\delta t + P_{0,00}(t)\mu_2\tau\delta t + f_{H,I,J}(0, 0, 1, t)\delta t + o(\delta t), \\
P_{0,00}(t + \delta t) &= P_{0,11}(t)[1 - (\mu_1 + \mu_2)\tau\delta t] + P_{0,01}(t)\lambda_2\tau\delta t + o(\delta t), \\
P_{N,11}(t + \delta t) &= P_{N,11}(t)[1 - (\lambda_1 + \lambda_2)\tau\delta t] + P_{N,10}(t)\mu_2\tau\delta t + o(\delta t), \\
P_{N,10}(t + \delta t) &= P_{N,10}(t)[1 - (\lambda_1 + \mu_2)\tau\delta t] + P_{N,11}(t)\lambda_2\tau\delta t + P_{N,00}(t)\mu_1\tau\delta t + f_{H,I,J}(N, 1, 0, t)\delta t + o(\delta t), \\
P_{N,01}(t + \delta t) &= 0, \\
P_{N,00}(t + \delta t) &= P_{N,00}(t)[1 - (\mu_1 + \mu_2)\tau\delta t] + P_{N,10}(t)\lambda_1\tau\delta t + o(\delta t).
\end{aligned}$$

In the limit $\delta t \rightarrow 0$, these equations lead to the following steady state relationships:

$$\begin{aligned}
P_{0,11} &= \frac{\mu_1}{\lambda_1 + \lambda_2} P_{0,01}, \\
P_{0,10} &= 0, \\
P_{0,01} &= \frac{1}{\mu_1 + \lambda_2} \left[\frac{1}{\tau} f_{H,I,J}(0, 0, 1) + \lambda_1 P_{0,11} + \mu_2 P_{0,00} \right], \\
P_{0,00} &= \frac{\lambda_2}{\mu_1 + \mu_2} P_{0,01}, \\
P_{N,11} &= \frac{\mu_2}{\lambda_1 + \lambda_2} P_{N,10}, \\
P_{N,10} &= \frac{1}{\lambda_1 + \mu_2} \left[\frac{1}{\tau} f_{H,I,J}(N, 1, 0) + \lambda_2 P_{N,11} + \mu_1 P_{N,00} \right], \\
P_{N,01} &= 0, \\
P_{N,00} &= \frac{\lambda_1}{\mu_1 + \mu_2} P_{N,10}.
\end{aligned} \tag{12}$$

Solving (10) and (11), we obtain

$$\begin{aligned}
f_{H,I,J}(h, 1, 0) &= f_{H,I,J}(h, 0, 1) = C_0 e^{Kh\tau}, \\
f_{H,I,J}(h, 1, 1) &= C_0 D_1 e^{Kh\tau}, \\
f_{H,I,J}(h, 0, 0) &= \frac{C_0}{D_1} e^{Kh\tau},
\end{aligned} \tag{13}$$

where D_1 is defined in the theorem and C_0 is a free constant. Substituting (13) into (12) results in

$$\begin{aligned}
P_{0,11} &= \frac{(\mu_1+\mu_2)C_0}{\lambda_2(\lambda_1+\lambda_2+\mu_1+\mu_2)\tau}, \\
P_{0,10} &= 0, \\
P_{0,01} &= \frac{(\lambda_1+\lambda_2)(\mu_1+\mu_2)C_0}{\lambda_2\mu_1(\lambda_1+\lambda_2+\mu_1+\mu_2)\tau}, \\
P_{0,00} &= \frac{(\lambda_1+\lambda_2)C_0}{\mu_1(\lambda_1+\lambda_2+\mu_1+\mu_2)\tau}, \\
P_{N,11} &= \frac{(\mu_1+\mu_2)C_0}{\lambda_2\lambda_1(\lambda_1+\lambda_2+\mu_1+\mu_2)\tau}e^{KN\tau}, \\
P_{N,10} &= \frac{(\lambda_1+\lambda_2)(\mu_1+\mu_2)C_0}{\lambda_1\mu_2(\lambda_1+\lambda_2+\mu_1+\mu_2)\tau}e^{KN\tau}, \\
P_{N,01} &= 0, \\
P_{N,00} &= \frac{(\lambda_1+\lambda_2)C_0}{\mu_2(\lambda_1+\lambda_2+\mu_1+\mu_2)\tau}e^{KN\tau}.
\end{aligned} \tag{14}$$

To evaluate the free constant, substitute (13) and (14) into the total probability formula,

$$\sum_{i=0}^1 \sum_{j=0}^1 \int_{0^+}^{N^-} f_{H,I,J}(h, i, j) dh + \sum_{i=0}^1 \sum_{j=0}^1 P_{0,ij} + \sum_{i=0}^1 \sum_{j=0}^1 P_{N,ij} = 1, \tag{15}$$

leading to

$$C_0 = \frac{1}{D_2 + D_3 + D_4}. \tag{16}$$

Thus the steady state distributions of buffer occupancy are as follows:

$$\begin{aligned}
f_{H,I,J}(h, 1, 1) &= \frac{D_1\tau}{D_2+D_3+D_4} e^{Kh\tau}, \\
f_{H,I,J}(h, 1, 0) &= \frac{\tau}{D_2+D_3+D_4} e^{Kh\tau}, \\
f_{H,I,J}(h, 0, 1) &= \frac{\tau}{D_2+D_3+D_4} e^{Kh\tau}, \\
f_{H,I,J}(h, 0, 0) &= \frac{\tau}{D_1(D_2+D_3+D_4)} e^{Kh\tau}, \\
P_{0,11} &= \frac{\mu_1+\mu_2}{\lambda_2(\lambda_1+\lambda_2+\mu_1+\mu_2)(D_2+D_3+D_4)\tau}, \\
P_{0,10} &= 0, \\
P_{0,01} &= \frac{(\lambda_1+\lambda_2)(\mu_1+\mu_2)}{\lambda_2\mu_1(\lambda_1+\lambda_2+\mu_1+\mu_2)(D_2+D_3+D_4)\tau}, \\
P_{0,00} &= \frac{\lambda_1+\lambda_2}{\mu_1(\lambda_1+\lambda_2+\mu_1+\mu_2)(D_2+D_3+D_4)\tau},
\end{aligned} \tag{17}$$

$$\begin{aligned}
P_{N,11} &= \frac{\mu_1 + \mu_2}{\lambda_1(\lambda_1 + \lambda_2 + \mu_1 + \mu_2)(D_2 + D_3 + D_4)\tau} e^{KN\tau}, \\
P_{N,01} &= 0, \\
P_{N,10} &= \frac{(\lambda_1 + \lambda_2)(\mu_1 + \mu_2)}{\lambda_1\mu_2(\lambda_1 + \lambda_2 + \mu_1 + \mu_2)(D_2 + D_3 + D_4)\tau} e^{KN\tau}, \\
P_{N,00} &= \frac{\lambda_1 + \lambda_2}{\mu_2(\lambda_1 + \lambda_2 + \mu_1 + \mu_2)(D_2 + D_3 + D_4)\tau} e^{KN\tau},
\end{aligned}$$

The marginal steady state pdf of the buffer is given by

$$\begin{aligned}
f_H(h) &= \sum_{i=0}^1 \sum_{j=0}^1 f_{H,I,J}(h, i, j) + \sum_{i=0}^1 \sum_{j=0}^1 P_{0,ij}\delta(h) \\
&\quad + \sum_{i=0}^1 \sum_{j=0}^1 P_{N,ij}\delta(h - N).
\end{aligned} \tag{18}$$

Thus we have,

$$f_H(h) = A_1 e^{Kh\tau} + A_2 \delta(h) + A_3 \delta(h - N), \quad h \in [0, N], \tag{19}$$

where

$$\begin{aligned}
A_1 &= \frac{(2 + D_1 + \frac{1}{D_1})\tau}{D_2(\tau) + D_3(\tau) + D_4(\tau)}, \\
A_2 &= \frac{D_3(\tau)}{D_2(\tau) + D_3(\tau) + D_4(\tau)}, \\
A_3 &= \frac{D_4(\tau)}{D_2(\tau) + D_3(\tau) + D_4(\tau)},
\end{aligned}$$

and $D_i, i = 1, 2, 3, 4$, are introduced in the formulation of the theorem.

Using the above distribution, the system performance metrics can be expressed as follows:

$$\begin{aligned}
PR &= P[s_2 = 1](1 - P[h = 0, s_1 = 0 | s_2 = 1]) \\
&= P[s_1 = 1](1 - P[h = N, s_2 = 0 | s_1 = 1]) \\
&= e_2 \left(1 - \frac{P_{0,01}}{e_2} \right) = e_1 \left(1 - \frac{P_{N,10}}{e_1} \right) \\
WIP &= \int_0^N f_H(h) h dh = \sum_{i=0}^1 \sum_{j=0}^1 \int_{0^+}^{N^-} f_{H,I,J}(h, i, j) h dh + \sum_{i=0}^1 \sum_{j=0}^1 P_{N,ij} N, \\
BL_1 &= P[h = N, s_1 = 1, s_2 = 0] = P_{N,10}, \\
ST_2 &= P[h = 0, s_1 = 0, s_2 = 1] = P_{0,01}.
\end{aligned}$$

Using (17), and substituting τ with $\frac{w_i}{W_i}$, it can be shown that

$$\begin{aligned}
\frac{P_{0,01}}{e_2} &= Q\left(\lambda_1, \mu_1, \lambda_2, \mu_2, N \frac{w_i}{W_i}\right), \\
\frac{P_{N,10}}{e_1} &= Q\left(\lambda_2, \mu_2, \lambda_1, \mu_1, N \frac{w_i}{W_i}\right).
\end{aligned}$$

Thus, above equations can be rewritten as

$$\begin{aligned} PR\left(\frac{w_i}{W_i}\right) &= e_2 \left(1 - Q\left(\lambda_1, \mu_1, \lambda_2, \mu_2, N \frac{w_i}{W_i}\right)\right) \\ &= e_1 \left(1 - Q\left(\lambda_2, \mu_2, \lambda_1, \mu_1, N \frac{w_i}{W_i}\right)\right), \end{aligned}$$

$$TP\left(\frac{w_i}{W_i}\right) = \frac{60W_i}{w_i} PR\left(\frac{w_i}{W_i}\right),$$

$$WIP\left(\frac{w_i}{W_i}\right) = \begin{cases} \frac{D_5\left(\frac{w_i}{W_i}\right)}{D_2\left(\frac{w_i}{W_i}\right) + D_3\left(\frac{w_i}{W_i}\right) + D_4\left(\frac{w_i}{W_i}\right)}, & \text{if } e_1 \neq e_2, \\ \frac{\left(\frac{D_2\left(\frac{w_i}{W_i}\right)}{2} + D_4\left(\frac{w_i}{W_i}\right)\right)N \frac{w_i}{W_i}}{D_2\left(\frac{w_i}{W_i}\right) + D_3\left(\frac{w_i}{W_i}\right) + D_4\left(\frac{w_i}{W_i}\right)}, & \text{if } e_1 = e_2, \end{cases}$$

$$ST_2\left(\frac{w_i}{W_i}\right) = e_2 Q\left(\lambda_1, \mu_1, \lambda_2, \mu_2, N \frac{w_i}{W_i}\right),$$

$$BL_1\left(\frac{w_i}{W_i}\right) = e_1 Q\left(\lambda_2, \mu_2, \lambda_1, \mu_1, N \frac{w_i}{W_i}\right).$$

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