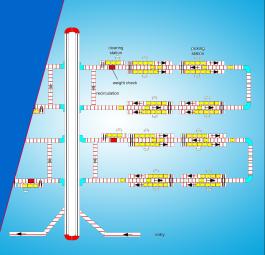
# Stochastic Models of Manufacturing Systems

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# **Fundamental relations**

#### Little's law

#### Consider system in equilibrium

- *E*(*L*) is mean number in system
- *E*(*S*) is mean time spent in system
- $\lambda$  is arrival rate (or departure rate)

Then:

 $E(L) = \lambda E(S)$ 

The definition of system is flexible (e.g. queue, server, queue+server)

**PASTA:** Poisson Arrivals See Time Averages

Poisson arrivals see the system in equilibrium: They see the same as random outside observer!



- Poisson arrivals with rate  $\lambda$
- Exponential service times with mean  $1/\mu$
- Stability:  $\lambda < \mu$  or  $\rho = \lambda/\mu < 1$
- Single server and FCFS service

Then:

- $p_k = P(k \text{ in system}) = (1 \rho)\rho^k$ , k = 0, 1, ... (Geometric)
- $E(L) = \sum_{k=0}^{\infty} kp_k = (1-\rho)\rho \sum_{k=0}^{\infty} k\rho^{k-1} = (1-\rho)\rho \frac{1}{(1-\rho)^2} = \frac{\rho}{1-\rho}$
- $E(S) = E(L)/\lambda = \frac{1/\mu}{1-\rho}$
- $E(Q) = \sum_{k=1}^{\infty} (k-1)p_k = \frac{\rho^2}{1-\rho}$

• 
$$E(W) = E(Q)/\lambda = \frac{\rho/\mu}{1-\mu}$$

or via PASTA+Little...



- Poisson arrivals with rate  $\lambda$
- Exponential service times with mean  $1/\mu$
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Via PASTA+Little...

$$E(S) = E(\boldsymbol{L}^{\boldsymbol{a}})\frac{1}{\mu} + \frac{1}{\mu},$$

where  $L^a$  is the number on arrival. By PASTA,  $E(L^a) = E(L)$ , so

$$E(S) = E(L)\frac{1}{\mu} + \frac{1}{\mu}$$

and thus by Little's law,  $E(L) = \lambda E(S)$ ,

$$E(S) = \frac{1/\mu}{1-\rho}.$$



- Poisson arrivals with rate  $\lambda$
- Exponential service times with mean  $1/\mu$
- Stability:  $\lambda < \mu$  or  $\rho = \lambda/\mu < 1$
- Single server and FCFS service

Then:

- Let  $a_k$  be arrival distribution, then by PASTA  $a_k = P(k \text{ in system just before arrival}) = p_k$
- Let  $d_k$  be departure distribution, then  $d_k = P(k \text{ in system just after departure}) = a_k (= p_k)$
- $P(S > t) = e^{-\mu(1-\rho)t}$ ,  $t \ge 0$  (Exponential)
- $P(W > t) = \rho e^{-\mu(1-\rho)t}$ ,  $t \ge 0$  (nearly Exponential)
- Output process is again Poisson!

- Poisson arrivals with rate  $\lambda$
- Exponential service times with mean  $1/\mu$
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Let  $f_{k+1}(t)$  be density of k + 1 independent exponential service times,

$$f_{k+1}(t) = \mu e^{-\mu t} \frac{(\mu t)^k}{k!}$$
 (Erlang- $k + 1$  distribution).

Density  $f_S(t)$  of sojourn (waiting plus service) time S:

$$f_{S}(t) = \sum_{k=0}^{\infty} a_{k} f_{k+1}(t) = \sum_{k=0}^{\infty} (1-\rho) \rho^{k} \mu e^{-\mu t} \frac{(\mu t)^{k}}{k!}$$
$$= \mu (1-\rho) e^{-\mu t} \sum_{k=0}^{\infty} \frac{(\rho \mu t)^{k}}{k!} = \mu (1-\rho) e^{-\mu (1-\rho)t}$$

so *S* is exponential with parameter  $(1 - \rho)\mu!$ 



- Poisson arrivals with rate  $\lambda$
- Exponential service times with mean  $1/\mu$
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Density  $f_W(t)$  of waiting time *W*:

$$f_W(t) = \sum_{k=1}^{\infty} a_k f_k(t) = \sum_{k=1}^{\infty} (1-\rho) \rho^k \mu e^{-\mu t} \frac{(\mu t)^{k-1}}{(k-1)!}$$
$$= \rho \mu (1-\rho) e^{-\mu t} \sum_{k=1}^{\infty} \frac{(\rho \mu t)^{k-1}}{(k-1)!} = \rho \mu (1-\rho) e^{-\mu (1-\rho)t}.$$

Note that  $P(W > 0) = \rho$  by PASTA, so

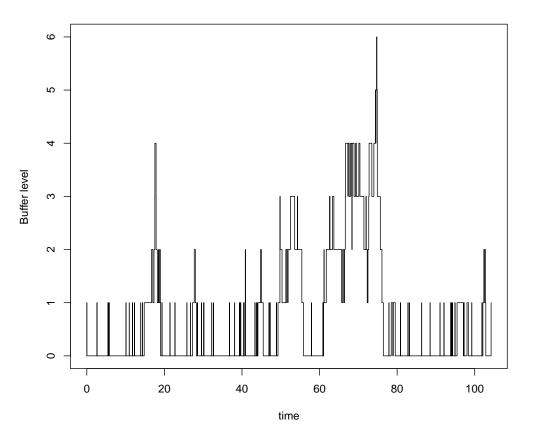
$$P(W > t | W > 0) = \frac{P(W > t)}{P(W > 0)} = e^{-\mu(1-\rho)t},$$

so conditional W|W > 0 is exponential with parameter  $(1 - \rho)\mu!$ 

**Tuesday May 26** 

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#### How does WIP behave over time?

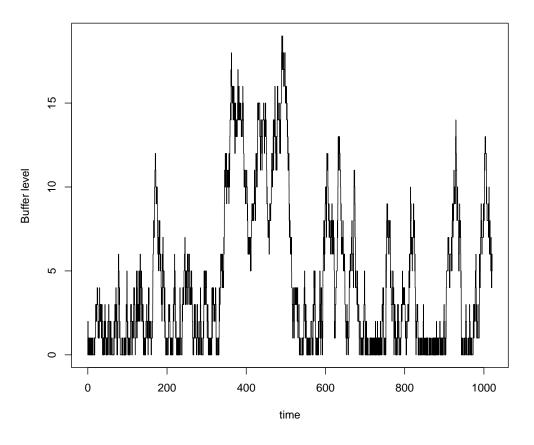


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Exponential model,  $\lambda = 1.0$ ,  $\rho = 0.5$ 

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#### How does WIP behave over time?

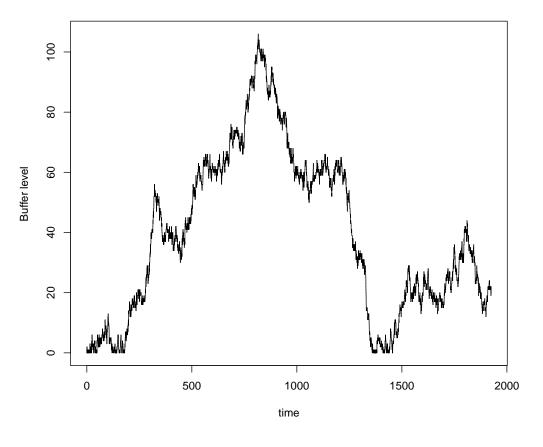


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Exponential model,  $\lambda = 1.0$ ,  $\rho = 0.9$ 

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#### How does WIP behave over time?



Tuesday May 26

Exponential model,  $\lambda = 1.0$ ,  $\rho = 0.95$ 

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# Single server: general service

- Poisson arrivals with rate  $\lambda$
- General service times *B* with distribution  $F_B(\cdot)$
- Stability:  $\rho = \lambda E(B) < 1$
- Single server and FCFS service

Then:

$$E(W) = E(Q^a)E(B) + \rho E(R) = E(Q)E(B) + \rho E(R)$$

where  $Q(Q^a)$  is number in queue (on arrival) and R is residual service time,

$$E(R) = \frac{E(B^2)}{2E(B)} = \frac{1}{2} E(B) (1 + c_B^2).$$

So with Little's law  $E(Q) = \lambda E(W)$ , we get...



# Single server: general service

- Poisson arrivals with rate  $\lambda$
- General service times *B* with distribution  $F_B(\cdot)$
- Stability:  $\rho = \lambda E(B) < 1$
- Single server and FCFS service

#### Then:

$$E(W) = \frac{\rho E(R)}{1 - \rho}$$

where *R* is residual service time,

$$E(R) = \frac{E(B^2)}{2E(B)} = \frac{1}{2} E(B) (1 + c_B^2)$$

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$$E(W) = \frac{\rho}{1-\rho} \frac{1}{2} E(B) (1+c_B^2)$$

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# Single server: residual service

Probability of randomly selected service X of size x is proportional to its size x and  $f_B(x)dx$ , which is number of services of size x:

 $P(x < X < x + dx) = f_X(x)dx = Cxf_B(x)dx$ 

where C is normalizing constant, so  $C = 1 / \int_{x=0}^{\infty} x f_B(x) dx = 1 / E(B)$  and

$$f_X(x) = \frac{xf_B(x)}{E(B)}$$

The mean of randomly selected service is

$$E(X) = \int_{x=0}^{\infty} x f_X(x) dx = \frac{1}{E(B)} \int_{x=0}^{\infty} x^2 f_B(x) dx = \frac{E(B^2)}{E(B)}$$

On average, in the middle of randomly selected service, so

$$E(R) = \frac{E(X)}{2} = \frac{E(B^2)}{2E(B)}$$



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# Single server: general

- General inter-arrival times A with distribution  $F_A(\cdot)$ , mean E(A), sd  $\sigma(A)$
- General service times *B* with distribution  $F_B(\cdot)$ , mean E(B), sd  $\sigma(B)$
- Stability:  $\rho = E(B)/E(A) < 1$
- Single server and FCFS service

Then:

$$E(W) \approx \frac{\rho}{1-\rho} \frac{1}{2} E(B) \left(c_A^2 + c_B^2\right)$$

where  $c_A$  and  $c_B$  are coefficients of variation of A and B:

$$c_A = \frac{\sigma(A)}{E(A)}, \quad c_B = \frac{\sigma(B)}{E(B)}.$$



$$E(W) \approx \frac{\rho}{1-\rho} \frac{1}{2} E(B) (c_A^2 + c_B^2)$$

Lessons:

- As  $\rho$  tends to 1 then E(W) tends to  $\infty$
- As  $\rho$  tends to 1 then approximation (1) becomes *exact*: the relative error tends to 0
- As  $\rho$  tends to 1 then waiting time distribution becomes exponential
- Summary: As system operates close to its maximum capacity, waiting times are long and exponential with mean (1)
- Insensitivity: Mean waiting time only depends on mean and standard deviation of inter-arrival times and service times!



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(1)

# Single server: output process

Inter-departure times *D*:

- Conservation of flow gives E(D) = E(A) or output rate = input rate
- Squared coefficient of variation  $c_D^2 \approx (1 \rho^2)c_A^2 + \rho^2 c_B^2$
- This makes sense since:
  - If ho pprox 1 , then server nearly always busy, so

 $c_D \approx c_B$ 

– If  $\rho \approx 0$ , then E(B) is very small compared to E(A), so

 $c_D \approx c_A$ 



#### Multi server: exponential

- Poisson arrivals with rate  $\lambda$
- Exponential service times with mean  $1/\mu$
- Stability:  $\lambda < c\mu$  or  $\rho = \lambda/(c\mu) < 1$
- c parallel servers and FCFS service

Then:

$$p_{k} = P(k \text{ in system}) = \begin{cases} \frac{(c\rho)^{k}}{k!} p_{0} & k = 0, 1, \dots c - 1, \\ \rho^{k-c} \frac{(c\rho)^{c}}{c!} p_{0} & k = c, c + 1, \dots \end{cases}$$

where

$$\frac{1}{p_0} = \sum_{k=0}^{c-1} \frac{(c\rho)^k}{k!} + \frac{(c\rho)^c}{c!} \frac{1}{1-\rho}.$$



#### Multi server: exponential

- Poisson arrivals with rate  $\lambda$
- Exponential service times with mean  $1/\mu$
- Stability:  $\lambda < c\mu$  or  $\rho = \lambda/(c\mu) < 1$
- c parallel servers and FCFS service

Then:

- $E(W) = E(Q)\frac{1}{c\mu} + \Pi_W \frac{1}{c\mu}$ , so with Little's law,  $E(W) = \frac{\Pi_W}{1-\rho} \frac{1}{c\mu}$
- $\Pi_W$  is probability of waiting,

$$\Pi_W = P(W > 0)$$
  
=  $p_c + p_{c+1} + \dots = \frac{(c\rho)^c}{c!} \left(\frac{(c\rho)^c}{c!} + (1-\rho)\sum_{n=0}^{c-1} \frac{(c\rho)^n}{n!}\right)^{-1}$ 

•  $P(W > t) = \prod_{W} e^{-c\mu(1-\rho)t}$ ,  $t \ge 0$  (nearly Exponential)



### Multi server: exponential

- Let  $a_k$  be arrival distribution, then  $a_k = p_k$  by PASTA
- Let  $d_k$  be departure distribution, then  $d_k = a_k$  (=  $p_k$ )
- Output process is again Poisson!



# Multi server: general service

- Poisson arrivals with rate  $\lambda$
- General service times *B* with distribution  $F_B(\cdot)$
- Stability:  $\rho = \lambda E(B)/c < 1$
- c parallel servers and FCFS service

Then:

$$E(W) \approx \frac{\Pi_W}{1-\rho} \frac{E(R)}{c}$$

where  $\Pi_W$  is probability of waiting in corresponding exponential system, so

$$E(W) \approx \frac{\Pi_W}{1-\rho} \frac{1}{2} \frac{E(B)}{c} (1+c_B^2).$$

- $\Pi_W$  is fairly insensitive to service time distribution;
- Corresponding system means with same mean service times.



# Multi server: general

- General inter-arrival times A with distribution  $F_A(\cdot)$ , mean E(A), sd  $\sigma(A)$
- General service times *B* with distribution  $F_B(\cdot)$ , mean E(B), sd  $\sigma(B)$
- Stability:  $\rho = E(B)/(cE(A)) < 1$
- c parallel servers and FCFS service

Then:

$$E(W) \approx \frac{\Pi_W}{1-\rho} \frac{1}{2} \frac{E(B)}{c} (c_A^2 + c_B^2)$$

where  $c_A$  and  $c_B$  are coefficients of variation of A and B



# Multi server: output process

Inter-departure times *D*:

- Conservation of flow gives E(D) = E(A) or output rate = input rate
- Coefficient of variation  $c_D^2 \approx 1 + (1-\rho^2)(c_A^2-1) + \rho^2(c_B^2-1)/\sqrt{c}$

