

# Numerics for Queueing with Dispatcher Paper

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## Special Case: Exponential - Geometric - Exponential

Suppose the individual service times are exponential with parameter  $\xi$ , then:

```
In[1]:= xi[x_, ze_] := Assuming[ze > 0, ze / (ze + x)]
```

Suppose the batch sizes are geometric with parameter  $p$ , then:

```
In[2]:= a[z_, p_] := Assuming[0 <= p && p <= 1, p * z / (1 - (1 - p) * z)]
```

Suppose the interarrival times are Erlang distributed with parameters  $\lambda$  and  $r$ , then:

```
In[3]:= psi[x_, lambda_] := Assuming[lambda > 0, lambda / (lambda + x)]
```

The joint transform of the arrivals with position-independent marking and the above assumptions is as follows.

```
In[4]:= gamma[alpha_, theta_, z_, ze_, p_, lambda_] :=  
  a[z * xi[alpha, ze], p] * psi[alpha + theta, lambda]
```

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## The Joint Functional -- Theorem 2.2

The joint functional  $\Phi^d$

```
In[5]:= PhiD[alpha_, beta_, vartheta_, theta_, z_, ze_, p_, lambda_, T_] := InverseLaplaceTransform[  
  (1 / (x + beta + theta)) * (1 - gamma[0, x + beta + theta, 1, ze, p, lambda]) /  
  (1 - gamma[alpha + beta + x, vartheta + theta, z, ze, p, lambda]), x, T]
```

The joint functional  $\Phi^s$

```
In[6]:= PhiS[alpha_, beta_, vartheta_, theta_, z_, ze_, p_, lambda_, T_] := InverseLaplaceTransform[  
  (1 / x) * (gamma[beta, theta + x, z, ze, p, lambda] - gamma[beta + x, theta, z, ze, p, lambda]) /  
  (1 - gamma[alpha + beta + x, vartheta + theta, z, ze, p, lambda]), x, T]
```

Let's find the probability of the server being called back by the dispatcher for  $r=1$ .

```
In[7]:= Simplify[PhiS[0, 0, 0, 0, 1, ze, p, lambda, T]]
```

```
Out[7]= 
$$\frac{\lambda - e^{-T(\lambda + pze)} \lambda}{\lambda + pze}$$

```

```
In[8]:= Simplify[PhiD[0, 0, 0, 0, 1, ze, p, lambda, T]]
```

$$\text{Out[8]} = \frac{e^{-T(\lambda + pze)} \lambda + pze}{\lambda + pze}$$

The joint functional and primary result:

```
In[9]:= Phi[alpha_, beta_, vartheta_, theta_, z_, ze_, p_, lambda_, T_] :=
  PhiD[alpha, beta, vartheta, theta, z, ze, p, lambda, T] +
  PhiS[alpha, beta, vartheta, theta, z, ze, p, lambda, T]
```

## Special Case Results

Let's find the mean queue length.

```
In[10]:= Simplify[Assuming[ze > 0 && lambda > 0 && 0 <= p && p <= 1,
  Limit[D[Phi[0, 0, 0, 0, z, ze, p, lambda, T], {z, 1}], z -> 1]]]
```

$$\text{Out[10]} = \left( \lambda (\lambda - e^{-T(\lambda + pze)} \lambda + \lambda p T ze + p^2 T ze^2) \right) / \left( p (\lambda + pze)^2 \right)$$

```
In[11]:= Qmean[ze_, p_, lambda_, T_] :=
  (lambda (lambda - e^{-T (lambda+p ze)} lambda + lambda p T ze + p^2 T ze^2)) / (p (lambda + p ze)^2)
```

Next, we will look for the variance, finding first  $g''(1)$  and then  $\text{Var}(Q_{\text{nu}})$ .

```
In[12]:= Simplify[Assuming[ze > 0 && lambda > 0 && 0 <= p && p <= 1,
  Limit[D[Phi[0, 0, 0, 0, z, ze, p, lambda, T], {z, 2}], z -> 1]]]
```

$$\begin{aligned} \text{Out[12]} = & \frac{1}{p^2 (\lambda + pze)^4} e^{-T(\lambda + pze)} \lambda (\lambda - 2 e^{T(\lambda + pze)} (-1 + p) p^4 T ze^4 + \lambda p^2 ze^2 \\ & (-8 + 2 p^2 T ze + p (6 - 2 T ze) + e^{T(\lambda + pze)} (8 + p^2 T ze (-4 + T ze) + 2 p (-3 + T ze))) + \\ & \lambda \lambda^3 (2 (-1 + p + p^2 T ze) + e^{T(\lambda + pze)} (2 + p^2 T^2 ze^2 + 2 p (-1 + T ze))) + \\ & 2 \lambda \lambda^2 p ze ((-1 + 2 p) (2 + p T ze) + e^{T(\lambda + pze)} (2 + p (-4 + T ze) + p^2 T ze (-1 + T ze)))) \end{aligned}$$

```
In[13]:= QsecondDerivative[ze_, p_, lambda_, T_] :=
```

$$\begin{aligned} & \frac{1}{p^2 (\lambda + pze)^4} e^{-T(\lambda + pze)} \lambda (\lambda - 2 e^{T(\lambda + pze)} (-1 + p) p^4 T ze^4 + \lambda p^2 ze^2 \\ & (-8 + 2 p^2 T ze + p (6 - 2 T ze) + e^{T(\lambda + pze)} (8 + p^2 T ze (-4 + T ze) + 2 p (-3 + T ze))) + \\ & \lambda \lambda^3 (2 (-1 + p + p^2 T ze) + e^{T(\lambda + pze)} (2 + p^2 T^2 ze^2 + 2 p (-1 + T ze))) + \\ & 2 \lambda \lambda^2 p ze ((-1 + 2 p) (2 + p T ze) + e^{T(\lambda + pze)} (2 + p (-4 + T ze) + p^2 T ze (-1 + T ze)))) \end{aligned}$$

```
In[14]:= Qvariance[ze_, p_, lambda_, T_] :=
```

$$QsecondDerivative[ze, p, lambda, T] + Qmean[ze, p, lambda, T] - (Qmean[ze, p, lambda, T])^2$$

```
In[15]:= Simplify[Qvariance[ze, p, lambda, T]]
```

$$\begin{aligned} \text{Out[15]} = & \frac{1}{p^2 (\lambda + pze)^4} e^{-2T(\lambda + pze)} \lambda (\lambda - e^{2T(\lambda + pze)} (-2 + p) p^4 T ze^4 + \\ & e^{T(\lambda + pze)} \lambda \lambda^2 p ze (-4 + 6 p + 4 p^2 T ze + e^{T(\lambda + pze)} (4 - 6 p + p^2 T ze)) + \\ & \lambda \lambda^3 (-1 + e^{2T(\lambda + pze)} (1 - p + p^2 T ze) + e^{T(\lambda + pze)} p (1 + 2 (1 + p) T ze)) - e^{T(\lambda + pze)} \\ & \lambda \lambda p^2 ze^2 (8 - 2 p^2 T ze + p (-5 + 2 T ze) + e^{T(\lambda + pze)} (-8 + p^2 T ze + p (5 - 2 T ze)))) \end{aligned}$$

Next, we will find the variance of the return time tau\_nu.

```
In[16]:= (-1) * Simplify[Assuming[ze > 0 && lambda > 0 && 0 <= p && p <= 1,
  Limit[D[Phi[0, 0, 0, theta, 1, ze, p, lambda, T], {theta, 1}], theta -> 0]]]
```

$$\text{Out[16]} = - \frac{-p^2 T ze^2 + \lambda (-1 + e^{-T(\lambda + p ze)} - p T ze)}{(\lambda + p ze)^2}$$

```
In[17]:= TauMean[ze_, p_, lambda_, T_] :=
  - ((-p^2 T ze^2 + lambda (-1 + e^{-T(\lambda + p ze)} - p T ze)) / (\lambda + p ze)^2)
```

```
In[18]:= TauMean[ze, p, lambda, T]
```

$$\text{Out[18]} = - \frac{-p^2 T ze^2 + \lambda (-1 + e^{-T(\lambda + p ze)} - p T ze)}{(\lambda + p ze)^2}$$

```
In[19]:= Simplify[Assuming[ze > 0 && lambda > 0 && 0 <= p && p <= 1,
  Limit[D[Phi[0, 0, 0, theta, 1, ze, p, lambda, T], {theta, 2}], theta -> 0]]]
```

$$\begin{aligned} \text{Out[19]} = & \frac{1}{(\lambda + p ze)^4} \\ & (-2 e^{-T(\lambda + p ze)} \lambda^3 T + p^4 T^2 ze^4 + 2 \lambda p ze (-2 + 2 e^{-T(\lambda + p ze)} + 2 p T ze + p^2 T^2 ze^2) + \\ & \lambda^2 (2 + 4 p T ze + p^2 T^2 ze^2 - 2 e^{-T(\lambda + p ze)} (1 + p T ze))) \end{aligned}$$

Next, we will find the variance of the return time tau\_nu.

```
In[20]:= TauSecondMean[ze_, p_, lambda_, T_] := \frac{1}{(\lambda + p ze)^4}
  (-2 e^{-T(\lambda + p ze)} \lambda^3 T + p^4 T^2 ze^4 + 2 \lambda p ze (-2 + 2 e^{-T(\lambda + p ze)} + 2 p T ze + p^2 T^2 ze^2) +
  \lambda^2 (2 + 4 p T ze + p^2 T^2 ze^2 - 2 e^{-T(\lambda + p ze)} (1 + p T ze)))
```

```
In[21]:= TauSecondMean[ze, p, lambda, T]
```

$$\begin{aligned} \text{Out[21]} = & \frac{1}{(\lambda + p ze)^4} \\ & (-2 e^{-T(\lambda + p ze)} \lambda^3 T + p^4 T^2 ze^4 + 2 \lambda p ze (-2 + 2 e^{-T(\lambda + p ze)} + 2 p T ze + p^2 T^2 ze^2) + \\ & \lambda^2 (2 + 4 p T ze + p^2 T^2 ze^2 - 2 e^{-T(\lambda + p ze)} (1 + p T ze))) \end{aligned}$$

```
In[22]:= TauVariance[ze_, p_, lambda_, T_] :=
  TauSecondMean[ze, p, lambda, T] - TauMean[ze, p, lambda, T]^2
```

```
In[23]:= Simplify[TauVariance[ze, p, lambda, T]]
```

$$\begin{aligned} \text{Out[23]} = & \frac{1}{(\lambda + p ze)^4} e^{-2 T (\lambda + p ze)} \lambda^2 \\ & (-2 e^{T(\lambda + p ze)} \lambda^2 T + 2 e^{T(\lambda + p ze)} p ze (2 + p T ze + e^{T(\lambda + p ze)} (-2 + p T ze)) + \\ & \lambda (-1 + e^{2 T (\lambda + p ze)} (1 + 2 p T ze))) \end{aligned}$$