

Modeling Dynamics of Exponentially Averaged Queue

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Outline

- Introduction
- Background: Fluid queues
- Modeling approach for exponentially averaged queue
- Remarks about solving the model
- Examples with M/M/1/K queue
- Further work





- Instantaneous queue length $L(t), L(t) \in \{0, 1, ..., K\}$
- Exponentially averaged queue length S(t), $\frac{d}{dt}S(t) = -\alpha(S(t) L(t))$
- Poisson arrivals, packet drop probability depends on S(t)
- \bullet Exponential service times with parameter μ



Introduction (continued)

- Motivation for modeling dynamics of exponentially averaged queue
 - RED mechanism
 - DiffServ architecture: Assured Forwarding PHB
- The model provides information about
 - The stationary properties of exponentially averaged queue length (PDF)
 - The average packet drop probability
- Results may be utilized to
 - Determine the parameters of RED/AF buffers



Background: Fluid Queues



- Arriving work is regarded as continuous fluid flow
 - Arrival intensity R(t)
 - Drain rate \boldsymbol{c}
 - The amount of fluid X(t) in the container changes with rate R(t) c
- Arrival intensity is controlled by Markov process (MMRP)
- Interesting issue: Stationary distribution of X(t)

Background: Fluid Queues(continued)

- Assume underlying Markov process is birth-death process $Z(t) \in \{0, 1, ..., N\}$
- Set $R(t) = r_i$, when Z(t) is in state *i*,
- Define partial CDF $P_i(t, x) = P\{X(t) \le x, Z(t) = i\}$
- Consider first how the $P_i(t, x)$ evolves in time

 $P_{i}(t + \Delta t, x) = \lambda_{i-1} \Delta t P_{i-1}(t, x - (r_{i-1} - c)\Delta t) + \mu_{i+1} \Delta t P_{i+1}(t, x - (r_{i+1} - c)\Delta t) + [1 - (\lambda_{i} + \mu_{i})\Delta t] P_{i}(t, x - (r_{i} - c)\Delta t) + O(\Delta t^{2})$





Background: Fluid Queues(continued)

• Taking the limit $\Delta t \rightarrow 0$

$$\frac{\partial}{\partial t}P_i(t,x) + (r_i - c)\frac{\partial}{\partial x}P_i(t,x) = \lambda_{i-1}P_{i-1}(t,x) + \mu_{i+1}P_{i+1}(t,x) - (\lambda_i + \mu_i)P_i(t,x)$$

- We are interested in the time-independent properties of process X(t):
 - Setting $\frac{\partial}{\partial t}P_i(t,x) = 0$,
 - Denoting $F_i(x) = P\{X \le x, Z = i\}$ (equilibrium probabilities) $(r_i - c)\frac{\partial}{\partial x}F_i(x) = \lambda_{i-1}F_{i-1}(x) + \mu_{i+1}F_{i+1}(x) - (\lambda_i + \mu_i)F_i(x), \forall i \in \{1, 2, ...N\}$
- The equations can be expressed in matrix form $\mathbf{M} \frac{d}{dx} \mathbf{F}(x) = \mathbf{A} \mathbf{F}(x)$
- The ODE system is linear, homogenous, constant coefficient system and it is basically easy to solve.



Modelling approach for exponentially averaged queue

- Consider now exponentially averaged queue
 - Instantaneous queue length follows stochastic process $L(t) \in \{0, 1, 2, ..., K\}$
 - Exponentially averaged queue length $S(t), \frac{d}{dt}S(t) = -\alpha(S(t) L(t))$
 - -L(t) is similar to birth-death process
 - \ast Poisson arrivals, packet drop probability depends on S(t) , $\lambda(S(t))$
 - \ast Exponential service times with parameter μ



• Interesting issue: Stationary distribution of S(t) and L(t)



Comparison with fluid queue

Z(t) independent of $X(t) \sim L(t)$ depends on S(t)

- Z(t) controls the arrival rate $\sim L(t)$ describes the instantaneous queue length amount of work in queue $X(t) \sim$ exponentially averaged queue length S(t)X(t) changes with rate $R(t) - c \sim S(t)$ changes with rate $-\alpha(S(t) - L(t))$
- Let's take modelling approach similar to fluid queues
 - Consider a process $\{S(t), L(t)\}$
 - Define partial CDF $P_i(t,s) = P\{S(t) \le s, L(t) = i\},\$
 - and partial PDF $p_i(t,s) = \frac{\partial}{\partial s} P_i(t,s)$



• $P_i(t,s)$ evolves in time step Δt

$$P_{i}(t + \Delta t, s) = \int_{0}^{s + \alpha(s-i)\Delta t} [1 - (\lambda_{i}(x) + \mu_{i})\Delta t]p_{i}(t, x)dx + \int_{0}^{s + \alpha(s-(i-1))\Delta t} \lambda_{i-1}(x)\Delta tp_{i-1}(t, x)dx + \int_{0}^{s + \alpha(s-(i+1))\Delta t} \mu_{i+1}\Delta tp_{i+1}(t, x)dx + O(\Delta t^{2})$$

• Take the limit $\Delta t \rightarrow 0$

$$\frac{\partial}{\partial t}P_i(t,s) - \alpha(s-i)\frac{\partial}{\partial s}P_i(t,s) = -\int_0^s (\lambda_i(x) + \mu_i)p_i(t,x)dx + \int_0^s \lambda_{i-1}(x)p_{i-1}(t,x)dx + \int_0^s \mu_{i+1}p_{i+1}(t,x)dx$$



- The time-independent properties of process x(t):
 - Set $\frac{\partial}{\partial t}P_i(t,s) = 0$
 - Define CDF $F_i(s) = P\{S \le s, L = i\}$ (equilibrium probabilities),

- and PDF
$$f_i(s) = \frac{d}{ds} F_i(s)$$

 $\alpha(s-i)f_i(s) = \int_0^s (\lambda_i(x) + \mu_i)f_i(x)dx$
 $-\int_0^s \lambda_{i-1}(x)f_{i-1}(x)dx - \int_0^s \mu_{i+1}f_{i+1}(x)dx, \forall i$

• Assuming that $F_i(s)$ has continuous second derivative $\frac{\partial}{\partial s}f_i(s)$, we get

$$\alpha(s-i)\frac{\partial}{\partial s}f_i(s) = (\lambda_i(s) + \mu_i - \alpha)f_i(s) - \lambda_{i-1}(s)f_{i-1}(s) - \mu_{i+1}f_{i+1}(s), \forall i$$



• Thus, we get a system of differential equations

$$\mathbf{M}(s)\frac{d}{ds}\mathbf{f}(s) = \mathbf{A}(s)\mathbf{f}(s),$$

in which

$$\mathbf{f}(s) = \left(f_0(s) \ f_1(s) \ \dots \ f_K(s) \right)^T$$

$$\mathbf{M}(s) = \alpha \begin{pmatrix} s & 0 \\ s-1 & \\ & \ddots & \\ 0 & s-K \end{pmatrix}, \mathbf{A}(s) = \begin{pmatrix} \lambda(s) - \alpha & -\mu & 0 \\ -\lambda(s) & \lambda(s) + \mu - \alpha & -\mu & \\ & \ddots & \\ 0 & & -\lambda(s) & \mu - \alpha \end{pmatrix}$$

• Process S(t) will not reach boundaries in finite time \rightarrow boundary conditions: $f_i(0) = 0$, $f_i(K) = 0$, $\forall i$

Remarks about solving the model

- How to solve the system $\mathbf{M}(s)\frac{d}{dx}\mathbf{f}(s) = \mathbf{A}(s)\mathbf{f}(s)$?
- No direct way to solve the DE system analytically
- Numerical solution with Euler methods is difficult
 - M(s) pointwise singular in points $s = \{0, 1, ..., K\}$
 - \rightarrow Boundary conditions do not define the solution uniquely
 - \rightarrow System "infinitely stiff", rounding errors dominate the solution
- Other approaches
 - Solve the PDE equations in discretized time
 - * Any initial PDF will approach infinitely close to the stationary PDF
 - * Embedded chain approach
 - Solve the DE system with base function approximations

Examples with M/M/1/K queue

- Consider M/M/1/K queue
 - Arrival intensity λ (independent of S(t))
 - Service intensity μ
- Arrival intensity is now constant and the DE system takes form

$$\mathbf{M}(s)\frac{d}{ds}\mathbf{f}(s) = \mathbf{A}\mathbf{f}(s)$$

• In this case we can derive similar DE system for CDFs $\mathbf{F}(s)$

$$\mathbf{M}(s)\frac{d}{ds}\mathbf{F}(s) = \mathbf{A'F}(s),$$

in which $A' = A - \alpha I$

Examples with M/M/1/K queue (continued)

0.6

0.4

0.8

1

Analytical solution is found in case K=1

$$\begin{cases} \alpha s \frac{d}{ds} f_0(x) = (\lambda - \alpha) f_0(s) - \mu f_1(s) \\ \alpha(s-1) \frac{d}{ds} f_1(s) = (\mu - \alpha) f_1(s) - \lambda f_0(s) \\ \Rightarrow \begin{cases} f_0(s) = s^{\lambda/\alpha - 1} (1 - s)^{\mu/\alpha}, s \in [0, 1] \\ f_1(s) = s^{\lambda/\alpha} (1 - s)^{\mu/\alpha - 1}, s \in [0, 1] \end{cases}$$

0.2



Examples with M/M/1/K queue (continued)

• K = 2, solutions found in the special case $\lambda/\alpha = \mu/\alpha = 2N + 1, N \in \mathbb{N}$

$$\begin{cases} \alpha s \frac{d}{ds} f_0(s) = (\lambda - \alpha) f_0(s) - \mu f_1(s) \\ \alpha(s-1) \frac{d}{ds} f_1(s) = (\mu + \lambda - \alpha) f_1(s) - \lambda f_0(s) - \mu f_2(s) \\ \alpha(s-2) \frac{d}{ds} f_2(s) = (\mu - \alpha) f_2(s) - \lambda f_1(s) \end{cases}$$

• Solutions are of the form (The $P_i(s)$ are polynomials of degree 2N)

$$\Rightarrow \begin{cases} f_0(s) = s^{\lambda/\alpha - 1}(2 - s)^{\mu/\alpha + 1}P_0(s), s \in [0, 2] \\ f_1(s) = s^{\lambda/\alpha - 0}(2 - s)^{\mu/\alpha - 0}P_1(s), s \in [0, 2] \\ f_2(s) = s^{\lambda/\alpha + 1}(2 - s)^{\mu/\alpha - 1}P_2(s), s \in [0, 2] \end{cases}$$

- In the special case $\lambda/\alpha = \mu/\alpha = 2$ solution is also found
 - $-f_i(s)$ are picewise polynomials

Further work

- Develop more efficient methods for solving PDE/DE system
 - Speeding up the convergence of time discretized PDE system
 - Embedded chain approach
 - Base function approximations for the DE system
- Verify model results with simulation
- Further development of the model
 - More complex traffic model (e.g. MMRP)
 - Model for two connected queues (AF buffer)