

Bounding the partition function of BCMP multiclass queueing networks

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Outline

1 Motivation

- BCMP Queueing Networks
- Difficulties of BCMP Queueing Networks

2 Our Result

- Hölder's Inequality
- Bounding the Partition Function
- Load-Dependent Stations

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- Closed BCMP (**B**asket, **C**handy, **M**untz, **P**alacios, 1975) queueing networks have been widely adopted to analytically evaluate the performance of computer and communication systems,
- Extension of Gordon-Newell (1967) networks
 - 1 Jobs belong to multiple classes,
 - 2 more types of service disciplines,
 - 3 more general service times probability distributions.

Notation of a Closed BCMP Network

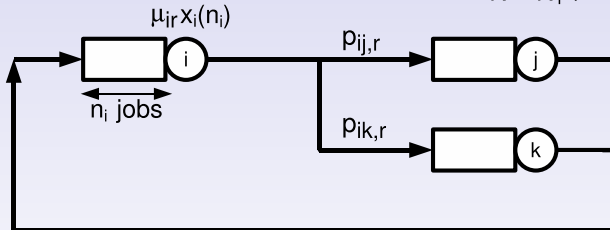
An Example

R : # job classes

M : # stations

N_r : # jobs of class r

$N = N_1 + \dots + N_R$



$\rho_{ir} = e_{ir} / \mu_{ir}$: mean loading (service demand) of class- r jobs in station i



Main assumptions of Closed BCMP Networks

	Per-class service time distribution
FCFS	Exponential (with mean μ_{ir}^{-1} , and $\mu_1 = \dots = \mu_R$)
LCFS,PS,IS	Coxian (with mean μ_{ir}^{-1})

Routing probabilities $p_{ij,r}$ are constants.

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Stationary Distribution of Closed BCMP Networks

The complexity of the model

BCMP models are CTMC, and the stationary distribution is

$$\pi(n_{11}, \dots, n_{MR}) = \frac{1}{G(\mathbf{N})} \prod_{i=1}^M \left(\sum_{r=1}^R n_{ir} \right)! \prod_{r=1}^R \frac{D_{ir}^{n_{ir}}}{n_{ir}!} \quad (1)$$

$$G(\mathbf{N}) = \sum_{\mathbf{S}(\mathbf{N})} \prod_{i=1}^M \left(\sum_{r=1}^R n_{ir} \right)! \prod_{r=1}^R \frac{D_{ir}^{n_{ir}}}{n_{ir}!} \quad (2)$$

$$\mathbf{S}(\mathbf{N}) = \left\{ n_{11}, \dots, n_{MR} : \sum_{i=1}^M n_{ir} = N_r, 1 \leq r \leq R \right\}. \quad (3)$$

Then, the main problem is the computation of $G(\mathbf{N})$. In fact,

$$|\mathbf{S}(\mathbf{N})| = \prod_{r=1}^R \binom{N_r + M - 1}{M - 1} \quad (4)$$

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What are the solutions?

A number of alternative analyses

Given that

- no (exact) polynomial algorithm is known for the solution of closed BCMP models,

alternative solutions are available:

- 1 Bounding Analysis
- 2 Asymptotic Analysis
- 3 Approximate Analysis
- 4 Bottleneck Analysis

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Hölder's Inequality in two dimensions

- Let p and q be positive real numbers such that $1/p + 1/q = 1$.
- If S is a measurable subset on \mathbb{R}^n , and f_1 and f_2 are measurable real-valued functions on S , then

$$\int_S |f_1(\mathbf{x})f_2(\mathbf{x})| \, d\mathbf{x} \leq \left(\int_S |f_1(\mathbf{x})|^p \, d\mathbf{x} \right)^{1/p} \left(\int_S |f_2(\mathbf{x})|^q \, d\mathbf{x} \right)^{1/q}. \quad (5)$$

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Our Objective

To apply the Hölder's Inequality (in the general case)

$$\int_S \left| \prod_{r=1}^R f_r(x) \right| dx \leq \prod_{r=1}^R \left(\int_S |f_r(x)|^{p_r} dx \right)^{1/p_r} \quad (6)$$

with $\sum_{r=1}^R 1/p_r = 1$, to the partition function of closed BCMP queueing networks

$$G(\mathbf{N}) = \sum_{\mathbf{S}(\mathbf{N})} \prod_{i=1}^M \left(\sum_{r=1}^R n_{ir} \right)! \prod_{r=1}^R \frac{D_{ir}^{n_{ir}}}{n_{ir}!} \quad (7)$$

to obtain a computationally efficient bound.

Integral representation of $G(\mathbf{N})$

- It is possible to show that

$$G(\mathbf{N}) = \frac{1}{\prod_{r=1}^R N_r!} \int_{\mathfrak{R}^{+M}} \prod_{r=1}^R H(r, \mathbf{u})^{N_r} e^{-(u_1 + \dots + u_M)} d\mathbf{u} \quad (8)$$

where $H(r, \mathbf{u}) = \rho_{0r} + \rho_{1r}u_1 + \dots + \rho_{Mr}u_M$.

- (8) can be rewritten as

$$G(\mathbf{N}) = \frac{1}{\prod_{r=1}^R N_r!} \int_{\mathfrak{R}^{+M}} \prod_{r=1}^R \left[H(r, \mathbf{u})^{N_r} e^{-(u_1 + \dots + u_M)} \right]^{\beta_r} d\mathbf{u} \quad (9)$$

where $\beta_r = N_r/N$.

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Applying Hölder's result

- Therefore, Hölder's inequality yields

$$G(\mathbf{N}) \leq \prod_{r=1}^R \frac{1}{N_r!} \left[\int_{\mathfrak{R}^{+M}} H(r, \mathbf{u})^N e^{-(u_1 + \dots + u_M)} d\mathbf{u} \right]^{\beta_r} \quad (10)$$

- Now, multiplying and dividing by $N!$, we obtain

$$G(\mathbf{N}) \leq N! \prod_{r=1}^R \frac{1}{N_r!} \left[\frac{1}{N!} \int_{\mathfrak{R}^{+M}} H(r, \mathbf{u})^N e^{-(u_1 + \dots + u_M)} d\mathbf{u} \right]^{\beta_r} \quad (11)$$

- \Rightarrow the expression in the brackets can be interpreted as the integral representation of the partition function of a singleclass network populated by N class- r jobs only.

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Main Result

- Hence,

$$G(\mathbf{N}) \leq \prod_{r=1}^R \frac{[N! G(N\mathbf{e}_r)]^{\beta_r}}{N_r!} = \binom{N}{N_1, \dots, N_R} \prod_{r=1}^R G(N\mathbf{e}_r)^{\beta_r} \quad (12)$$

where \mathbf{e}_r is the unit vector in direction r .

- \Rightarrow the upper bound on the partition function of a closed, multiclass BCMP queueing network is provided.

Notes on $G(Ne_r)$

$G(Ne_r)$ refers to a **single-class** queueing network. Therefore,

- it can be computed efficiently
 - 1 The computational complexity of the MVA is $O(MN)$
 - 2 The computational complexity of Koenigsberg's formula is $O(M^2)$
- Koenigsberg's formula provides the asymptotic behavior

$$G(Ne_r) \approx \frac{(\max_{i \geq 1} \rho_{ir})^{N+M-1}}{\prod_{j=1, j \neq \arg \max_{i \geq 1} \rho_{ir}}^M (\max_{i \geq 1} \rho_{ir} - \rho_{jr})} \quad (13)$$

i.e., the order of magnitude of $G(Ne_r)$.

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Extension to LD Stations

- Load-Dependent (LS) station: service rates μ_{ir} are multiplied by queue-length functions, i.e., $x_i(n)$ or $y_{ir}(n)$.
- Making the following replacements

$$\rho_{ir} \leftarrow \begin{cases} \rho_{ir} / \inf_n x_i(n) \\ \rho_{ir} / \inf_n y_{ir}(n) \end{cases} \quad (14)$$

the resulting queueing network is composed of only LI stations and the new value of the partition function bounds from above the original one, and

- Bound (12) can be applied again.

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Summary

- We proposed **a new inequality** upper bounding the partition function of multiclass, closed BCMP queueing networks in terms of R partition functions related to single-class, closed BCMP queueing networks,
- The upper bound can be computed efficiently,
- Beyond the theoretical interest, it can provide an estimate of the minimum amount of memory that exact solution algorithms implementations should allocate to avoid numerical instabilities.

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For Further Reading I



Jonatha Anselmi.

New Analyses in BCMP Queueing Networks Theory.

PhD Thesis, March 2009.

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