Call-level Performance Analysis of Wired and Wireless Networks

TUTORIAL

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Preamble (cont.1)

Call Arrival Process



 Batch Poisson arrivals (*infinite number of traffic sources*).
 Calls from different service-classes arriving in batches, while batches arriving randomly.

time



Preamble (cont.3)

Call's behavior while in service





constant-bit-rate/stream traffic



bandwidth compression/expansion

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Preamble (cont.5)

Teletraffic (Loss) Models

- Importance of QoS assessment through teletraffic models:

 - **Avoidance of too costly over-dimensioning of the network.**
 - Prevention of excessive network throughput degradation, through traffic engineering mechanisms.
- Applicability:
 - Connection Oriented Communication Networks, in general.
 - IP based networks with resource reservation capabilities.
 - Cellular networks (e.g. UMTS).
 - All-optical core networks (MPλS/GMPLS).

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STRUCTURE

- **Teletraffic Models for:**
- (A) Random Traffic
- (B) Quasi-random Traffic
- (C) Batched Poisson Traffic

STRUCTURE (cont.1)

- (A) Random Traffic
 - (A1) Random arriving calls with either fixed (certain) or elastic bandwidth requirements upon arrival, and constant use of the assigned bandwidth (constant-bitrate/stream traffic) while in service.
 - (A2) Random arriving calls with either fixed or elastic bandwidth requirements upon arrival, and elastic bandwidth (compression/expansion) while in service.

STRUCTURE (cont.2)

• (B) Quasi-random Traffic

- (B1) Quasi-random arriving calls with either fixed or elastic bandwidth requirements upon arrival, and constant use of the assigned bandwidth (constant-bitrate/stream traffic) while in service.
- (B2) Quasi-random arriving calls with either fixed or elastic bandwidth requirements upon arrival, and elastic bandwidth (compression/expansion) while in service.

STRUCTURE (cont.3)

•• (C) Batched Poisson Traffic

- (C1) Batched Poisson arriving calls with fixed bandwidth requirements and continuous use of the assigned bandwidth (constant-bit-rate/stream traffic) while in service.
- (C2) Batched Poisson arriving calls with fixed bandwidth requirements upon arrival, and elastic bandwidth (compression/expansion) while in-service.

STRUCTURE – Where We Are

- (A) Random Traffic
 - (A1) Constant-bit-rate/stream traffic^o
 - (A2) Elastic/adaptive traffic while in service
- (B) Quasi-random Traffic
 - (B1) Constant-bit-rate/stream traffic
 - (B2) Elastic/adaptive traffic while in service
- (C) Batched Poisson Traffic
 - (C1) Constant-bit-rate/stream traffic
 - (C2) Elastic/adaptive traffic while in service

We

are

here!

(A) Random Traffic

(A1) Random arriving calls with either fixed (certain) or elastic bandwidth requirements upon arrival, and constant use of the assigned bandwidth (constant-bit-rate/stream traffic) while in service.

State of the art

- The Erlang Multi-rate Loss Model (EMLM) 1981
 - **The Retry Models** 1992

Furthermore

- The Connection Dependent Threshold Model (CDTM) 2002
- The CDTM under the Bandwidth Reservation Policy 2002



EMLM Analysis – Classical Method

State Space Ω Complete Sharing Policy – A coordinate convex policy Global Balance (rate_in=rate_out) - Statistical equilibrium



EMLM Analysis – Classical Method (cont.1)

Local Balance (Rate_up = rate_down)



EMLM Analysis – Classical Method (cont.2)

Product Form Solution

Product Form Solution of the State **Probabilities**

$$P(\boldsymbol{n}) = \boldsymbol{G}^{-1} \left(\prod_{k=1}^{K} \frac{a_k^{n_k}}{n_k!} \right)$$

where $n = (n_1, n_2, ..., n_k),$ $\alpha_k = \lambda_k / \mu_k$ (offered traffic load, in erl) $G = G(\Omega) = \sum_{k=1}^{K} \left(\prod_{k=1}^{K} \frac{a_k^{n_k}}{n_k!} \right)$ normalization constant

Product Form \longleftrightarrow Local Balance \iff Reversible Markov Chain High accuracy in Call Blocking Probability calculation

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EMLM Analysis – Classical Method (cont.3)

Call Blocking Probability Determination – Classical Method



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EMLM Analysis – Classical Method (cont.4)

Call Blocking Probability Determination – Classical Method



Necessity for recursive formulas

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EMLM Analysis – Recursive formula

Kaufman, IEEE Trans. on Commun. 1981

Macro-states – One-dimensional Markov chain

C = 8, K=2, $b_1 = 1$, $b_2 = 2$ Macro-state $j=n_1b_1+n_2b_2$ denotes the occupied link bandwidth



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EMLM Analysis – Recursive formula (cont.)

Call Blocking Probability – Recursive Calculation

Call Blocking Probability:
$$P_{b_k} = \sum_{j=C-b_k+1}^{C} G^{-1}q(j)$$
 where $G = \sum_{j=0}^{C} q(j)$

q(j)/G – Macro-state Probabilities





EMLM/BR Analysis

State Space Ω, Local-Global Balance? Product Form Solution?

$C = 8, K = 2, b_1 = 1, b_2 = 2, t_1 = 1 (t_2 = 0)$



EMLM/BR – Roberts' Method

Roberts, International Teletraffic Congress 1983

Macro-states – One-dimensional Markov chain

 $C = 8, K=2, b_1 = 1, b_2 = 2, t_1 = 1 (t_2 = 0)$



$$q(j) = \begin{cases} 1 & \text{for } j = 0 \\ \frac{1}{j} \sum_{k=1}^{K} a_k D_k (j - b_k) q(j - b_k) & \text{for } j = 1, \dots, C \\ 0 & \text{otherwise} \end{cases}$$
where
$$D_k (j - b_k) = \begin{cases} b_k & \text{when } j \le C - t_k \\ 0 & \text{when } j > C - t_k \end{cases}$$
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approximation

$$y_{k}(j) = \begin{cases} \frac{a_{k}q(j-b_{k})}{q(j)} & \text{for } j \le C - t_{k} \\ 0 & \text{for } j > C - t_{k} \end{cases}$$

EMLM/BR – Roberts' Method (cont.)

Call Blocking Probability – Recursive Calculation



The Retry Models



The Retry Models (cont.)

Kaufman, IEEE INFOCOM 1992, Performance Evaluation 1992 **Assumptions – Approximations**

- **Local Balance**
- When $j \leq C b_{kr_{s-1}} + b_{kr_s}$ (migration space) then $y_{kr_s}(j) = 0$ (Migration Approximation, M.A.)



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CDTM - The analytical model

Moscholios et al. Performance Evaluation 2002

Assumptions – Approximations

- 1) Local Balance
- 2) Migration Approximation, M.A $(\delta_{kc_s}(j))$ 3) Upward migration Approximation, U.A $(\delta_k(j))$

$$q(j) = \begin{cases} 1 & \text{for } j = 0 \\ \frac{1}{j} \left(\sum_{k=1}^{K} a_k b_k \delta_k(j) q(j - b_k) + \sum_{k=1}^{K} \sum_{s=1}^{S(k)} a_{kc_s} b_{kc_s} \delta_{kc_s}(j) q(j - b_{kc_s}) & \text{for } j = 1, ..., C \right) \\ 0 & \text{otherwise} \end{cases}$$

$$a_{kc_s} = \lambda_k \mu_{kc_s}^{-1} \quad (if \ 1 \le j \le J_{k0} + b_k \text{ and } b_{kc_s} > 0) \text{ or } (if \ 1 \le j \le C \text{ and } b_{kc_s} = 0) \\ 0 & \text{otherwise} \end{cases}$$

$$U.A$$

$$\delta_{kc_s}(j) = \begin{cases} 1 & \text{if } J_{ks} + b_{kc_s} \ge j > J_{ks-1} + b_{kc_s} \text{ and } b_{kc_s} > 0 \\ 0 & \text{otherwise} \end{cases}$$

$$M.A$$
Call Blocking Probability: $P_{b_k} = \sum_{j=C-b_{kc_{S(k)}}^{C} j^{-1}} G^{-1}q(j) \quad \text{where } G = \sum_{j=0}^{C} q(j)$
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Importance of the CDTM

- Generalizes the models of Thresholds, Retries and the EMLM
 - Incorporates the Thresholds models, by setting the same set of thresholds for all service-classes.
 - Incorporates the Retries models, when each service-class k has threshold: $J_{k_{s-1}} = C - b_{k_{cs-1}}$
 - Incorporates the EMLM by setting for each service-class k the threshold J_{ks-1} = C
- The CDTM models elastic traffic at the call setup phase

Elastic bandwidth requirements

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STRUCTURE – Where We Are

- (A) Random Traffic
 - (A1) Constant-bit-rate/stream traffic
 - (A2) Elastic/adaptive traffic while in service⁶
- (B) Quasi-random Traffic
 - (B1) Constant-bit-rate/stream traffic
 - (B2) Elastic/adaptive traffic while in service
- (C) Batched Poisson Traffic
 - (C1) Constant-bit-rate/stream traffic
 - (C2) Elastic/adaptive traffic while in service



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(A) Random Traffic

(A2) Random arriving calls with either fixed or elastic bandwidth requirements upon arrival, and elastic bandwidth (compression/expansion) while in service.



State of the art

 The Extended Erlang Multi-rate Loss Model (E-EMLM) 1997

Furthermore

- The E-EMLM for elastic and adaptive traffic 2002
- The Extended Connection Dependent Threshold Model (E-CDTM) 2007

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The Extended Erlang Multiple Rate Loss Model (E-EMLM)

Parameters

- C : link bandwidth capacity
- K: service-classes
- λ_k : arrival rate (Poisson)
- b_k: peak bandwidth requirement
- μ_k : service rate, μ_k^{-1} : service time (exponential)

If compression: "Bandwidth * Service-time" ⇒ constant ⇒ elastic traffic

- -j: total bandwidth demand ($0 \le j \le T$)
- T : maximum total bandwidth demand (T \ge C)
- s : real bandwidth allocation ($0 \le s \le C$)

Number of occupied b.u. if all in-service calls were receiving the requested bandwidth (without bandwidth compression)

The Extended Erlang Multiple Rate Loss Model (E-EMLM) (cont).

Transmission link: C=5, T=7In-service calls: $b_1=1$, $b_2=2$ Arriving call: $b_3=3$

j : system macro state, $0 \le j \le T$

s : real bandwidth allocation, $0 \le s \le C$

example



E-EMLM – The analytical model for elastic traffic

Stamatelos & Koukoulidis, IEEE/ACM Trans. Networking 1997

Total bandwidth demand:

Real bandwidth allocation:

$$j = \sum_{k=1}^{K} n_k b_k$$
$$s = \sum_{k=1}^{K} n_k b_k \Phi_k(\mathbf{n})$$

Where $b_k \Phi_k(\mathbf{n})$ is the actual allocated bandwidth to service-class k calls

$$\Phi_{k}(n) : \text{service-class } k \text{ and state } n \text{ dependent factor} \quad \Phi_{k}(n) = \begin{cases}
1 & \text{for } 0 \le j \le C \\
\frac{x(n_{k})}{x(n)} & \text{for } C < j \le T \\
0 & \text{otherwise}
\end{cases}$$

$$x(n) : \text{state multiplier or weight} \\
\text{associated with the state } n$$

$$x(n) = \begin{cases}
1 & \text{for } 0 \le j \le C \\
\frac{1}{C} \sum_{k=1}^{K} n_{k} b_{k} x(n_{k}) & \text{for } C < j \le T \\
0 & \text{otherwise}
\end{cases}$$

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E-EMLM – The analytical model for elastic traffic (cont.)

Link Occupancy Distribution

$$q(j) = \frac{1}{\min(C, j)} \sum_{k=1}^{K} \alpha_k b_k q(j - b_k), \quad j = 0, ..., T$$

$$q(x)=0 \text{ for } x < 0 \text{ and } \sum_{j=0}^{C} q(j) = 1$$

No product form solution

Call Blocking Probabilities (CBP)

CBP of service-class k:

$$P_{k:} P_{b_k} = \sum_{j=0}^{b_k - 1} q(T - j)$$

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E-EMLM – The analytical model for elastic and adaptive traffic

Racz, Gero and Fodor, Performance Evaluation 2002

$$q(j) = \frac{1}{\min(C,j)} \sum_{k \in K_e} a_k b_k q(j - b_k) + r(j) \sum_{k \in K_a} a_k b_k q(j - b_k), \quad j = 0, ..., T$$
$$q(x) = 0 \text{ for } x < 0, \qquad \sum_{j=0}^C q(j) = 1 \quad \text{and} \quad r(j) = \min(1, \frac{C}{j})$$

where K_e is the set of elastic service-classes and K_a is the set of adaptive service-classes No product form solution

CBP of service-class
$$k$$
: $B_k = \sum_{j=0}^{b_k - 1} q(T - j)$

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E-CDTM – The analytical model

Vassilakis et al., Int. Journal of Commun. Systems 2012

Link occupancy distribution

$$q(j) = \begin{cases} 1 & \text{for } j = 0 \\ \frac{1}{\min(C, j)} \sum_{k \in K_e} \sum_{l=0}^{S_k} a_{k_l} b_{k_l} \delta_{k_l}(j) q(j - b_{k_l}) + \\ + \frac{1}{j} \sum_{k \in K_a} \sum_{l=0}^{S_k} a_{k_l} b_{k_l} \delta_{k_l}(j) q(j - b_{k_l}) & \text{for } j = 1, ..., T \\ 0 & \text{otherwise} \end{cases} \quad G = \sum_{j=0}^{T} q(j)$$

Call Blocking Probability

$$P_{b_k} = \sum_{j=T-b_{k_{S_k}}}^{T} G^{-1}q(j)$$

Link Utilization

$$U = \sum_{j=1}^{C} j G^{-1} q(j) + \sum_{j=C+1}^{T} G^{-1} C q(j)$$

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E-CDTM versus E-EMLM



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STRUCTURE – Where We Are

- (A) Random Traffic
 - (A1) Constant-bit-rate/stream traffic
 - (A2) Elastic/adaptive Traffic while in service
- (B) Quasi-random Traffic
 - (B1) Constant-bit-rate/stream traffic $^{\circ}$
 - (B2) Elastic/adaptive Traffic while in service
 - (C) Batched Poisson Traffic
 - (C1) Constant-bit-rate/stream traffic
 - (C2) Elastic/adaptive Traffic while in service

We

are

here!

(B) Quasi-random Traffic

(B1) *Quasi-random arriving calls with either fixed or elastic bandwidth requirements upon arrival, and constant use of the* assigned bandwidth (constant-bit-rate/stream traffic) while in service.

State of the art

- The Engset Multi-rate Loss Model (EnMLM) 1994
- The Single Retry Model for finite population (f-SRM) 1997

Furthermore

- The EnMLM for elastic and adaptive traffic
- The EnMLM under the Bandwidth Reservation Policy
- The f-SRM under the Bandwidth Reservation Policy
- The Multi Retry Model for finite population(f-MRM)
- The f-MRM under the Bandwidth Reservation Policy
- The CDTM for finite population (f-CDTM)
- The f-CDTM under the Bandwidth Reservation Policy
 - The Generalized f-CDTM when random and quasi-random traffic coexist

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The Engset Multi-rate Loss Model (EnMLM)



 n_k : number of service-class k calls (sources) which are in service

- v_k : fixed arrival rate per «free» source (not in service yet)
- λ_k : mean arrival rate of service-class k calls
- h_k : holding (service) time of service-class k calls

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EnMLM – The Analytical Model

A Product Form Solution model

$$P(\boldsymbol{n}) = G^{-1}\left(\prod_{k=1}^{K} \binom{N_k}{n_k} a_k^{n_k}\right) \quad Where \ G = G(\Omega) = \sum_{\boldsymbol{n} \in \Omega} \left(\prod_{k=1}^{K} \binom{N_k}{n_k} a_k^{n_k}\right)$$

Macro-states – One-dimensional Markov chain



EnMLM – The Analytical Model (cont.)

Stamatelos & Hayes, Computer Communications 1994

Link occupancy distribution – Recursive formula

$$q(j) = \begin{cases} 1 & \text{for } j = 0\\ \frac{1}{j} \sum_{k=1}^{K} (N_k - n_k + 1) \alpha_k b_k q(j - b_k) & \text{for } j = 1, ..., C\\ 0 & \text{otherwise} \end{cases}$$

Time congestion probability:

$$P_{b_k} = \sum_{j=C-b_k+1}^{C} G^{-1}q(j)$$

For
$$K = 1 \rightarrow P_{b_1} = \frac{\binom{N}{C}(\alpha_1)^C}{\sum_{i=0}^C \binom{N}{i}(\alpha_1)^i}$$
 Engset formula (1918)

For $N_k \rightarrow \infty$,
September 23, 2012q(j) results in Kaufman/Roberts recursion (EMLM)
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EnMLM – State Space Determination

The problem
In calculating the q(j)'s
The link occupancy j (macro-state)
\Leftrightarrow single state (not valid in many cases)
Example:
C = 5 b.u.
K = 3 service-classes
$N_1 = N_2 = N_3 = 10$ sources
$b_1 = 3$ b.u. (per call)
$b_2 = 2$ b.u. (per call)
$b_3 = 1$ b.u. (per call)
$a_1 = a_2 = a_3 = 0.1$ erl (per idle source)

$$q(4) = \frac{1}{4} \sum_{k=1}^{K} (N_k - n_k + 1) a_k b_k q (4 - b_k)$$

	macro state		
n1	n ₂	n ₃	j
0	0	0	0
0	0	1	1
0	0	2	2
0	0	3	3
0			4
0	0	5	5
0	1	0	2
0	1	1	3
0			4
0	1	3	5
0			4
0	2	1	5
1	0	0	3
1			4
1	0	2	5
1	1	0	5

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EnMLM – State Space Determination (cont.1)

The solution

State space

Blocking states

Theorem: Two stochastic

Two stochastic systems with the same state space and the same parameters K, N_k , a_k are equivalent – they have the same Blocking States

Lemma:

Modify only the b_k 's so that the resultant link occupancy per state is unique.

Example

By choosing $b_1=16$, $b_2=12$ and $b_3=5$ an equivalent system results with unique link occupancy per state, j_{eq} and capacity C=29.

<u> </u>						0	
n ₁	n ₂	n ₃	j	B ₁	B ₂	B ₃	j _{eq}
0	0	0	0				0
0	0	1	1				5
0	0	2	2				10
0	0	3	3	Þ			15
				Þ	Þ		20
0	0	5	5	Þ	Þ	Þ	25
0	1	0	2				12
0	1	1	3	P			17
				Þ	Þ		22
0	1	3	5	Þ	Þ	Þ	27
				Þ	Þ		24
0	2	1	5	P	Þ	Þ	29
1	0	0	3	Þ			16
				Þ	Þ		21
1	0	2	5	Þ	Þ	Þ	26
1	1	0	5	Þ	Þ	Þ	28

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The Single Retry Model for finite population (f-SRM)

Stamatelos & Koukoulidis, IEEE/ACM Trans. on Networking 1997

Local Palance \square Product Form Solution $\square > \approx P_{bk}$

Assumptions – Approximations

- Local Balance
- When $j \le C b_k + b_{kr}$ (migration space) then $y_{kr}(j) = 0$ (Migration approximation, M.A.)

$$q(j) = \begin{cases} 1 \text{ for } j = 0 \\ \frac{1}{j} \left(\sum_{k=1}^{K} (N_k - n_k + 1) a_k b_k q(j - b_k) + \sum_{k=1}^{K} (N_k - (n_k + n_{kr}) + 1) a_{kr} b_{kr} \gamma_k(j) q(j - b_{kr}) \right) \text{ for } j = 1, ..., C \\ 0 \text{ otherwise EnMLM} & \text{ calls with } b_{kr} \end{cases}$$

$$a_{kr} = v_{kr} \mu_{kr}^{-1}, \quad \gamma_k(j) = 1 \quad \text{when } j > C - b_k + b_{kr} \quad \text{otherwise } \gamma_k(j) = 0$$
For $N_k \to \infty$ \Longrightarrow the Single Retry Model (for random traffic)
Time Congestion Probability: $P_{b_k} = \sum_{j=C-b_{kr}+1}^{C} G^{-1}q(j) \quad \text{where } G = \sum_{j=0}^{C} q(j)$
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The Connection Dependent Threshold Model for finite population (f-CDTM)



f-CDTM – The Analytical Model

Moscholios et al., Performance Evaluation 2005

Assumptions - Approximations

- 1) Local Balance
- 2) Migration approximation, M.A. $(\delta_{kcs}(j))$
- 3) Upward approximation, U.A. $(\delta_{\kappa}(j))$

1 for j = 0 $q(j) = \begin{cases} \frac{1}{j} (\sum_{k=1}^{K} (N_{k} - n_{k} + 1)\alpha_{k} b_{k} \delta_{k}(j) q(j - b_{k}) + \sum_{k=1}^{K} \sum_{s=1}^{S(k)} (N_{k} - (n_{k} + n_{kc_{1}} + ... + n_{kc_{s}} + ... + n_{kc_{s}(k)}) + 1)\alpha_{kc_{s}} b_{kc_{s}} \delta_{kc_{s}}(j) q(j - b_{kc_{s}})) \text{ for } j = 1, ..., C \end{cases}$ 0 otherwise $\delta_{k}(j) = \begin{cases} 1 & (\text{if } 1 \le j \le J_{k0} + b_{k} \text{ and } b_{kc_{s}} > 0) \text{ or } (\text{if } 1 \le j \le C \text{ and } b_{kc_{s}} = 0) \\ 0 & \text{otherwise} \end{cases} \textbf{U.A}$ $a_{kc} = v_{kc} \mu_{kc}^{-1}$ $\delta_{kc_{s}}(j) = \begin{cases} 1 & \text{if } J_{ks} + b_{kc_{s}} \ge j > J_{ks-1} + b_{kc_{s}} \text{ and } b_{kc_{s}} > 0 \\ 0 & \text{otherwise} \end{cases}$ M.A **Time Congestion Probability**: $P_{b_k} = \sum_{\substack{j=C-b_{kc} \\ \text{Emerging}(2012)}}^{C} G^{-1}q(j)$ where $G = \sum_{\substack{j=0 \\ j=0}}^{C} q(j)$ 50

f-CDTM – State Space Determination

A Good Approximation - Without equivalent system!

$n_k(j) \approx y_k(j)$

The parameters $n_k(j)$ can be approximated by the average number of service-class k calls in state j, $y_k(j)$, assuming infinite population for each service-class (i.e. from the corresponding CDTM)

Glabowski & Stasiak, Proc. MMB&PGTS 2004 Moscholios et al., MEDJCN 2007

Numerical example: f-CDTM versus CDTM

V	$N_1 = N_2 = 12$	(f-CDTM)	$N_1 = N_2 = \infty$ (CDTM)		
Σ	$P_{b1c2}(\%)$	P_{b2c1} (%)	$P_{b1c2}(\%)$	P_{b2c1} (%)	
1	1.96	1.07	4.49	2.48	
2	2.78	1.52	6.70	3.65	
3	3.76	2.05	9.39	5.10	
4	4.90	2.66	12.55	6.74	
5	6.19	3.34	16.06	8.62	
6	7.63	4.09	19.84	10.65	

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The Generalized f-CDTM where random and quasi-random traffic coexist

Moscholios et al., Performance Evaluation 2005

 K_{fin} service-classes of finite sources (quasi-random input).

K_{inf} service-classes of infinite sources (random – Poisson input).

Link occupancy distribution

$$q(j) = \begin{pmatrix} 1 & \text{for } j = 0 \\ \frac{1}{j} \sum_{k \in K_{fin}} (N_k - n_k + 1) \alpha_k b_k \delta_k(j) G(j - b_k) + \frac{1}{j} \sum_{k \in K_{fin}} \sum_{t=1}^{T} (N_k - (n_k + n_{kc_1} + \dots + n_{kc_t} + \dots + n_{kc_t}) + 1) a_{kc_t} b_{kc_t} \delta_{kc_t}(j) q(j - b_{kc_t}) \\ + \frac{1}{j} \sum_{k \in K_{inf}} \alpha_k b_k \delta_k(j) G(j - b_k) + \frac{1}{j} \sum_{k \in K_{inf}} \sum_{t=1}^{T} a_{kc_t} b_{kc_t} \delta_{kc_t}(j) q(j - b_{kc_t}) \text{ for } j = 1, \dots, C \\ 0 & \text{otherwise} \end{pmatrix}$$

Where:

$$\delta_k(j) = 1 \text{ when } 1 \le j \le C \text{ and } b_{kc} = 0, \text{ or, when } j \le J_{kt} + b_k \text{ and } b_{kc} > 0, \text{ otherwise } \delta_k(j) = 0.$$

$$\delta_{kct}(j) = 1 \text{ when } J_{kt} + b_{kct} \ge j > J_{kt} - l + b_{kct} \text{ otherwise } \delta_{kct}(j) = 0.$$

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STRUCTURE – Where We Are

- (A) Random Traffic
 - (A1) Constant-bit-rate/stream traffic
 - (A2) Elastic Traffic while in service
- (B) Quasi-random Traffic
 - (B1) Constant-bit-rate/stream traffic
 - (B2) Elastic Traffic while in service $^{\circ}$
- (C) Batched Poisson Traffic
- (D) ON-OFF Traffic
 - (D1) Poisson arrivals
 - (D2) Quasi-random arrivals
 - (D3) Batched Poisson arrivals



(B) Quasi-random Traffic

(B2) *Quasi-random arriving calls with either fixed or elastic bandwidth requirements upon arrival, and elastic bandwidth while in service.*



State of the art

• The Extended Engset Multi-rate Loss Model (E-EnMLM) 1997

Furthermore

 The Extended Connection Dependent Threshold Model for finite population (Ef-CDTM) 2007

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 h_k : holding (service) time of service-class k calls

If compression: "Bandwidth * Service-time" ⇒ constant ⇒ elastic traffic

j : total bandwidth demand $(0 \le j \le T)$

- **T** : maximum total bandwidth demand ($T \ge C$)
- s : real bandwidth allocation ($0 \le s \le C$)

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E-EnMLM – The analytical model

Stamatelos & Koukoulidis, IEEE/ACM Trans. Networking 1997

Link occupancy distribution

$$q(j) = \begin{cases} 1 & \text{for } j = 0 \\ \frac{1}{\min(C, j)} \sum_{k \in K_e} (N_k - n_k + 1) \alpha_k b_k q(j - b_k) + \\ + \frac{1}{j} \sum_{k \in K_a} (N_k - n_k + 1) \alpha_k b_k q(j - b_k) & \text{for } j = 1, \dots, T \\ 0 & \text{otherwise} \end{cases} \quad G = \sum_{j=0}^T q(j)$$

Time Congestion Probability Link Utilization

$$P_{b_k} = \sum_{j=T-b_k+1}^T G^{-l}q(j)$$

$$U = \sum_{j=1}^{C} j G^{-1} q(j) + \sum_{j=C+1}^{T} G^{-1} C q(j)$$

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The Extended Connection Dependent Threshold Model for finite population (Ef-CDTM)



Ef-CDTM – The analytical model

Vassilakis et al., IEICE Trans. Commun. 2008

Link occupancy distribution

$$q(j) = \begin{cases} 1 & \text{for } j = 0 \\ \frac{1}{\min(C,j)} \sum_{k \in K_e} \sum_{l=0}^{S_k} (N_k - \sum_{l=0}^{S_k} n_{k_l} + 1) a_{k_l} b_{k_l} \delta_{k_l}(j) q(j - b_{k_l}) + \\ + \frac{1}{j} \sum_{k \in K_a} \sum_{l=0}^{S_k} (N_k - \sum_{l=0}^{S_k} n_{k_l} + 1) a_{k_l} b_{k_l} \delta_{k_l}(j) q(j - b_{k_l}) & \text{for } j = 1, ..., T \\ 0 & \text{otherwise} \end{cases} \quad G = \sum_{j=0}^{T} q(j)$$

Time Congestion Probability Link Utilization

$$U = \sum_{j=1}^{C} j G^{-1} q(j) + \sum_{j=C+1}^{T} G^{-1} C q(j)$$

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 $P_{b_k} = \sum_{j=T-b_{k_{S_k}}+1}^T G^{-l}q(j)$



Ef-CDTM accuracy (cont.)



Ef-CDTM comparison with other models: EMLM, CDTM, E-CDTM



Service-class 1: elastic

Offered Traffic-Load per idle source = 0.025 erl Consequently, it increases by 0.025 erl

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Ef-CDTM comparison with other models: EMLM, CDTM, E-CDTM (cont.)



T=C

STRUCTURE – Where We Are

- (A) Random Traffic
 - (A1) Constant-bit-rate/stream traffic
 - (A2) Elastic Traffic while in service
- (B) Quasi-random Traffic
 - (B1) Constant-bit-rate/stream traffic
 - (B2) Elastic Traffic while in service
- (C) Batched Poisson Traffic
 - (C1) Constant-bit-rate/stream traffic⁶
 - (C2) Elastic Traffic while in service



(C) Batched Poisson Traffic

(C1) Batched Poisson arriving calls with fixed bandwidth requirements and continuous use of the assigned bandwidth (constant-bit-rate/stream traffic) while in service.



time



State of the art

The Batched Poisson Erlang Multirate Loss Model (BP-EMLM) 1996

Furthermore

• The Batched Poisson Erlang Multirate Loss Model under the Bandwidth Reservation Policy 2010

Batched Poisson arrival process



- batch arrival rate
- λ_k^{-1} batch interarrival time (exponentially distributed).
- $\mathbf{B}_{\mathbf{r}}^{\mathbf{k}}$ probability that there are *r* calls in an arriving batch of service-class *k*

 $\lambda_{\mathbf{k}}$

The Batched Poisson Erlang Multirate Loss Model (BP-EMLM)



The proportion of arriving calls The proportion of time that the that find the system congested. system is congested.

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The level L_n^k separates the state-vector $\mathbf{n} = (n_1, n_2, \dots, n_{k-1}, n_k, n_{k+1}, \dots, n_K)$ from the state-vector $(n_1, n_2, \dots, n_{k-1}, n_k + 1, n_{k+1}, \dots, n_K)$, for service-class k.

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BP-EMLM – The analytical Model

Kaufman, Rege, Performance Evaluation 1996

- C link capacity
- K service classes
- **b**_k bandwidth requirements (k=1,...,K)
- λ_k batch arrival rate

 μ_k service rate

h_k

- $= \mu_k^{-1}$ service time (exponentially distributed).
- $\begin{array}{c} \mathbf{B}_{\mathbf{r}}^{\mathbf{k}} & \text{probability that there are } r \text{ calls in an arriving batch of service-class } k \\ \mathbf{j} & \text{occupied link bandwidth} \end{array}$

q(j) probability that j out of C bandwidth units are occupied

Link occupancy distribution

$$\mathbf{q}(j) = \frac{1}{j} \sum_{k=1}^{K} \alpha_k b_k \sum_{l=1}^{\lfloor j/b_k \rfloor} \hat{B}_{l-1}^k \mathbf{q}(j-lb_k)$$

where $\alpha_k = \lambda_k / \mu_k$ and $\hat{B}_l^k = \sum_{r=l+1}^{\infty} B_r^k$ (the complementary batch size distribution) September 23, 2012 Emerging 2012 Barcelona

BP-EMLM – The analytical Model (cont.)

Performance measures

 $E(n_k|j) = \frac{\alpha_k \sum_{l=1}^{\lfloor j/b_k \rfloor} \hat{B}_{l-1}^k q(j-lb_k)}{q(j)}$ Average number of service-class k calls in state j

 $\overline{n_k} = \sum_{j=1}^{C} E(n_k | j) q(j)$ Average number of service-class k calls in the system

$$C_{b_k} = \frac{\alpha_k \hat{B}_k - \overline{n}_k}{\alpha_k \hat{B}_k}$$
 Call congestion probability of service-class k

$$P_{b_k} = \sum_{j=C-b_k+1}^{C} G^{-1}q(j)$$
 Time congestion probability of service-class k

 $U = \sum_{j=1}^{C} jq(j)$ Link utilization
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The BP-EMLM under Bandwidth Reservation Policy (BP-EMLM/BR)



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BP-EMLM/BR – Roberts' Method



The reservation space of a service-class k includes the blocking states: $C-b_k-t_k+1,...,C$ e.g. for the 1st service-class, *j*=3 and 4.

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BP-EMLM/BR – Roberts' Method (cont.)

Moscholios and Logothetis, Computer Communications, 2010

Link Occupancy Distribution

$$q(j) = \frac{1}{j} \sum_{k=1}^{K} \alpha_k D_k (j - b_k) \sum_{l=1}^{\lfloor j/b_k \rfloor} \hat{B}_{l-1}^k q(j - lb_k) \qquad D_k (j - b_k) = 0$$

$$D_k(j-b_k) = \begin{cases} b_k & \text{when } j \le C - t_k \\ 0 & \text{when } j > C - t_k \end{cases}$$



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BP-EMLM/BR-Method of Stasiak & Glabowski (cont.)

$$E^{*}(n_{k}|j) = \begin{cases} a_{k} \sum_{l=1}^{\lfloor j/b_{k} \rfloor} \hat{B}_{l-1}^{k}q(j-lb_{k}) \\ q(j) \\ \vdots \\ \sum_{i=1, i \neq k}^{K} E^{*}(n_{k}|j-b_{i})w_{k,i}(j) \\ \vdots \\ p_{i} \\ i \neq k \end{cases} \text{ where } w_{k,i}(j) = \frac{a_{i}b_{i}}{\sum_{j=1, j \neq k}^{K} a_{j}b_{j}} \hat{B}_{l}^{k} = \sum_{r=l+1}^{\infty} B_{r}^{k} \\ \hat{B}_{r}^{k} = \sum_{r=l+1}^{\infty} B_{r}^{k} \\ \vdots \\ \dots, C \end{cases}$$

Link Occupancy Distribution

$$q(j) = \frac{1}{j^{*}} \sum_{k=1}^{K} \alpha_{k} b_{k} \sum_{l=1}^{\lfloor j/b_{k} \rfloor} \hat{B}_{l-1}^{k} q(j-lb_{k})$$

$$j^{*} = \sum_{k=1}^{K} b_{k} E^{*} (n_{k} | j)$$

$$j^{*} = \sum_{k=1}^{K} b_{k} E^{*} (n_{k} | j)$$

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Numerical example: BP-EMLM – BP-EMLM/BR

C = 100 b.u. $\mathbf{K} = \mathbf{3}$ $b_1 = 1$ b.u., $t_1 = 15$ b.u. $b_2 = 4$ b.u., $t_2 = 12$ b.u. $b_3=16$ b.u., $t_3=0$ b.u. $P_{r}(s_{k}=r) = (1 - \beta_{k})\beta_{k}^{r-1}$ (geometric distribution of batch size s_k) $\beta_1 = 0.75, \ \beta_2 = 0.5, \ \beta_3 = 0.2$ (note: average batch size is $1/(1-\beta_k)$) $\mu^{-1}_{1} = \mu^{-1}_{2} = \mu^{-1}_{3} = 1$ (exponentially distributed call service time) $\alpha_1 = 6 \text{ erl}, \alpha_2 = 4 \text{ erl}, \alpha_3 = 2 \text{ erl (offered traffic)}$

Numerical example: BP-EMLM – BP-EMLM/BR (cont.1)



Time Congestion Probabilities

Numerical example: BP-EMLM – BP-EMLM/BR (cont.2)

Call Congestion Probabilities (higher than time congestion probabilities)



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Numerical example: BP-EMLM – BP-EMLM/BR (cont.3)

Call congestion probabilities

		Roberts' method (%)			Method of S&G (%)			Simulation results (%)			
	α_1	1 st	2^{nd}	3 rd	1^{st}	2^{nd}	3 rd	1^{st}	2 nd	3 rd	
	1	class	class	class	class	class	class	class	class	class	
	6.0	26.28	28.45	27.67	26.83	28.98	28.21	27.38	29.32	28.23	
								±0.33	±0.40	±0.46	
	5.5	25.44	27.57	26.91	25.96	28.08	27.42	26.57	28.40	27.46	
								±0.17	±0.22	±0.33	
	5.0	24.61	26.69	26.15	25.10	27.17	26.63	25.59	27.28	26.67	
								±0.19	±0.16	±0.22	
	4.5	23.78	25.81	25.37	24.24	26.26	25.83	24.77	26.63	25.88	
								±0.30	±0.15	0.16	
	4.0	22.94	24.93	24.60	23.38	25.36	25.02	23.84	25.65	25.07	
								±0.14	±0.21	±0.17	
	3.5	22.12	24.06	23.81	22.52	24.45	24.21	23.03	24.62	24.37	
								±0.26	±0.29	±0.25	
	3.0	21.29	23.18	23.03	21.67	23.55	23.40	22.08	23.70	23.47	
								±0.13	±0.07	±0.08	
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Numerical example: BP-EMLM – BP-EMLM/BR (cont.5)

Equalizing Call Congestion Probabilities



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STRUCTURE – Where We Are

- (A) Random Traffic
 - (A1) Constant-bit-rate/stream traffic
 - (A2) Elastic Traffic while in service
- (B) Quasi-random Traffic
 - (B1) Constant-bit-rate/stream traffic
 - (B2) Elastic Traffic while in service

• (C) Batched Poisson Traffic

- (C1) Constant-bit-rate/stream traffic
 - (C2) Elastic Traffic while in service



(C) Batched Poisson Traffic

(C2) Batched Poisson arriving calls with fixed bandwidth requirements upon arrival, and elastic bandwidth while in service.



State of the art

The Batched Poisson Erlang Multirate Loss Model (BP-EMLM) 1996

Furthermore

The BP-EMLM supporting elastic and adaptive traffic under the BR policy 2011, 2012

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The BP EMLM for elastic & adaptive traffic under the BR policy

Moscholios et. al (IEEE ICC 2012, Annals of Telecommunications 2012)

Link Occupancy Distribution

 $q(j) = \begin{cases} 1 & \text{for } j = 0 \\ \frac{1}{\min(j,C)} \sum_{k=1}^{K_e} \alpha_k D_k (j-b_k) \sum_{l=1}^{\lfloor j/b_k \rfloor} \widehat{B}_{l-1}^{(k)} G(j-lb_k) \\ + \frac{1}{j} \sum_{k=1}^{K_e} \alpha_k D_k (j-b_k) \sum_{l=1}^{\lfloor j/b_k \rfloor} \widehat{B}_{l-1}^{(k)} G(j-lb_k) \text{ for } j = 1, ..., T \\ 0 & \text{for } j < 0 \end{cases}$

where:
$$D_k(j-b_k) = \begin{cases} b_k & \text{for } j \le T-t_k \\ 0 & \text{for } j > T-t_k \end{cases}$$

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The BP EMLM for elastic & adaptive traffic under the BR policy (cont.)

TC probability of service-class k

CC probability of service-class k

$$P_{b_{k}} = \sum_{j=C-b_{k}-t_{k}+1}^{C} G^{-1}q(j)$$

Performance Metrics

$$C_{b_{k}} = \sum_{j=0}^{C} G^{-1}q(j) \sum_{m=\left|\frac{C-j}{b_{k}}\right|+1}^{\infty} B_{m}^{(k)}$$

Link Utilization

$$U = \sum_{j=1}^{C} jG^{1}q(j) + \sum_{j=C+1}^{T} CG^{1}q(j)$$

 No Product Form Solution
 Approx. calculation of link occupancy distribution and all performance measures.

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Numerical Results – Evaluation



Batch size, s_k : Geometrically distributed, $Pr(s_k=r)=(1-\beta_k)\beta_k^{r-1}$ $\beta_1=0.75, \beta_2=0.5, \beta_3=\beta_4=0.2.$

One set of BR parameters: $t_1 = 15, t_2 = 12, t_3 = 6, t_4 = 0$ (TC equalization among calls of all service-classes).

Three different values of *T*:

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- a) T = C = 200 b.u. (no bandwidth compression results coincide with BP-EMLM/BR)
- b) T = 220 b.u. (max compression factor C/T = 200/220) $b_1 = 1 \rightarrow b_{1min} = 0.91$
- c) T = 240 b.u. (max compression factor C/T = 200/240) $b_1 = 1 \rightarrow b_{1min} = 0.83$



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Offered traffic-load points

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Offered traffic-load points

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Introduction to W-CDMA

User Activity

Uplink: calls from the Mobile Users (MUs) to the Base Station (BS)



K service-classes (k=1,...,K)

- N_k : Number of traffic sources (MUs)
- R_k : Transmission bit rate

• $(E_b/N_0)_k$: Signal energy per bit divided by noise spectral density, required to meet a predefined Bit Error Rate (BER) parameter

 v_k : Activity factor

User Activity: users alternate between transmitting and silent periods

- Active users: have a call in progress (occupy system resources)
- **Passive users**: are silent (do not occupy any system resources)

Introduction to W-CDMA Interference & Call Admission Control

Interference



Wireless Erlang Multi-rate Loss Model (Wireless EMLM)

The EMLM is not suitable for W-CDMA Networks, since it does not take into account:
1) User activity (active and silent periods)
2) Blocking due to inter-cell interference (soft blocking)



D. Staehle and A. Mäder, "*An analytic approximation of the uplink capacity in a UMTS network with heterogeneous traffic,*" in proc. 18th International Teletraffic Congress (ITC18), Sept. 2003.

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Wireless EMLM

Cell Load, Load Factor and Local Blocking Probability

n =Cell Load: The ratio of the received power from all active users to the total received power

$$n = \frac{I_{intra} + I_{inter}}{I_{intra} + I_{inter} + P_{N}} = n_{intra} + n_{inter}$$

$$n_{intra}: cell load from users of the reference cell$$

$$n_{inter}: cell load from users of the neighboring cells$$

$$NR = \frac{I_{intra} + I_{inter} + P_{N}}{P_{N}}$$

$$n = \frac{NR - 1}{NR}$$

$$n = \frac{NR - 1}{NR}$$

$$n = \frac{NR - 1}{NR}$$

$$Typical value, n_{max} = 0.8$$
(can be considered as the shared system resource)

 $L_k = Load Factor$: can be seen as the bandwidth requirement of service-class k calls

$$L_{k} = \frac{(E_{b} / N_{0})_{k} * R_{k}}{W + (E_{b} / N_{0})_{k} * R_{k}}$$

 R_k : Transmission bit rate

 $(Eb/No)_k$: Bit error rate (BER) parameter

W = 3.84 Mcps: Chip rate of the W-CDMA carrier

 β_k = Local Blocking Probability: The prob. that a new call is blocked when arriving at an instant with intra-cell load n_{intra}. It depends on the system occupied bandwidth as well as on the calls requirement

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$$\beta_k(n_{intra}) = P(n_{intra} + n_{inter} + L_k > n_{max})$$

Wireless EMLM Intra-cell load and Inter-cell load

n_{intra}: Intra-cell load (cell load from users of the reference cell)

$$n_{intra} = \sum_{k=1}^{K} m_k L_k$$

where m_k is the number of active service-class k calls and

 L_k is the load factor of service-class k calls

*n*_{inter}: Inter-cell load (cell load from users of the neighboring cells)

$$n_{inter} = (1 - n_{\max}) \frac{I_{inter}}{P_N}$$

where I_{inter} is modeled as a lognormal random variable, that is independent of the intra-cell interference, with mean $E[I_{inter}]$ and variance $Var[I_{inter}]$

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Wireless EMLM

Bandwidth Discretization & Bandwidth Occupancy

g: basic cell load unit used for Banwidth Discretization

Bandwidth discretization is needed since the EMLM considers discrete state space

$$n \rightarrow j = \frac{n}{g}, \quad n_{\max} \rightarrow C = \frac{n_{\max}}{g}$$

 $L_k \rightarrow b_k = \operatorname{round}(\frac{L_k}{g})$

Due to the existence of passive users a state *j* does not represent the total number of occupied b.u.

A(c|j) = **Bandwidth Occupancy:** conditional probability that *c* b.u.are occupied in state *j*

Note that: c=0 *all users are passive,* c=j *all users active while in the EMLM,* c=j *always*

$$\Lambda(c \mid j) = \sum_{k=1}^{K} P_k(j) [v_k \Lambda(c - b_k \mid j - b_k) + (1 - v_k) \Lambda(c \mid j - b_k)]$$

for $j = 1, ..., j_{\text{max}}$ and $c \le j$
where $\Lambda(0 \mid 0) = 1$ and $\Lambda(c \mid j) = 0$ for $c > j$

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Wireless EMLM **Local Blocking Factor**

Local Blocking Factor: due to the inter-cell interference blocking may occur in every state *j* with probability $LB_k(j)$

- λ_k : arrival rate (Poisson)
- μ_k : service rate
- $n_k(j)$: number of in-service calls in state j
- $\lambda_k(1-LB_k(j))$: effective arrival rate in state j





Wireless EMLM Call Blocking Probabilities Calculation

State Probabilities

$$\hat{q}(j) = \begin{cases} 1 & \text{for } j = 0 \\ \sum_{k=1}^{K} \alpha_k (1 - LB_k (j - b_k) b_k \hat{q}(j - b_k)) & \text{for } j = 1, ..., j_{\max} \\ 0 & \text{otherwind} \end{cases}$$

$$q(j) = \frac{\hat{q}(j)}{\sum_{j=0}^{j_{\text{max}}} \hat{q}(j)}$$

Bandwidth Share

$$P_k(j) = \frac{a_k(1 - LB_k(j - b_k)b_kq(j - b_k))}{jq(j)}$$

Call Blocking Probabilities



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Wireless Engset Multirate Loss Model Vassilakis et. al (IEEE PIMRC 2007)

Due to the limited coverage area of a cell, it is certainly more realistic to consider that the number of mobile users, in a cell, is finite. This consideration is especially true in the case of microcells (small size cells).

In that case the Wireless EMLM should be replaced by the Wireless Engset Multirate Loss Model (Wireless EnMLM).

Wireless Engset Multirate Loss Model Local Blocking Factor

 $LB_k(j) = \sum_{j=1}^{j} \beta_k(c) \Lambda(c \mid j)$

Local Blocking Factor: due to the inter-cell interference blocking may occur in every state *j* with probability $LB_k(j)$

- λ_k : arrival rate from an idle source
- µ_k: service rate
- *N_k*: number of traffic sources (MUs)
- **n**_k (j): number of in-service calls in state j
- $(N_k n_k(j))\lambda_k (1-LB_k(j))$: effective arrival rate in state j



Wireless EnMLM Call Blocking Probabilities Calculation

State Probabilities

$$\hat{q}(j) = \begin{cases} 1 & \text{for } j = 0\\ \sum_{k=1}^{K} (N_k - n_k + 1)\alpha_k (1 - LB_k (j - b_k)b_k \hat{q}(j - b_k)) & \text{for } j = 1, ..., j_{\max} \\ 0 & \text{otherwise} \end{cases}$$

$$q(j) = \frac{\hat{q}(j)}{\sum_{j=0}^{j_{\text{max}}} \hat{q}(j)}$$

Bandwidth Share

$$P_k(j) = \frac{(N_k - n_k + 1)a_k(1 - LB_k(j - b_k)b_kq(j - b_k))}{jq(j)}$$

Call Blocking Probabilities

$$B_k = \sum_{j=0}^{j_{max}} q(j) LB_k(j)$$

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Evaluation – Application Example

We compare:

- a) Analytical to Simulation CBP results of the Wireless-EnMLM
- b) The Wireless-EnMLM to the Wireless-EMLM (infinite source model)

	Data	Video				
Transmission rates (Kbps)	$R_1 = 64$	$R_2 = 144$				
Activity factor	v ₁ =1.0	v ₂ =0.3				
BER parameter (dB)	$(E_b/N_0)_1 = 4$	$(E_b/N_0)_2 = 3$				
Inter-cell Interference	$E[I_{inter}] = 2*10^{-18} \text{ mW and } CV[I_{inter}] = 1$					

Traffic load point	1	2	3	4	5	6	7	8	9	10
Number of sources $(N_1 = N_2)$	10	20	30	40	50	60	70	80	90	100
Offered traffic for Data (erl)	1.0	2.0	3.0	4.0	5.0	6.0	7.0	8.0	9.0	10.0
Offered traffic for Video (erl)	0.1	0.2	0.3	0.4	0.5	0.6	0.7	0.8	0.9	1.0

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Evaluation – Application Example (cont.)



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The Wireless EMLM including Handoff traffic (WH-EMLM)

Vassilakis et. al (IARIA AICT 2008)



User Activity: users alternate between transmitting and silent periods
Active users: have a call in progress (occupy system resources)
Passive users: are silent (do not occupy any system resources)

The WH-EMLM Interference & Call Admission Control



The WH-EMLM

Cell Load, Load Factor and Local Blocking Probability

n =Cell Load: The ratio of the received power from all active users to the total received power



The WH-EMLM

Bandwidth Discretization & Bandwidth Occupancy

In order to describe the system by a Markov Chain we express all parameters with integer values.

g: basic cell load unit used for Resource Discretization

$$n \rightarrow j = \frac{n}{g}, n_{\max} \rightarrow C = \frac{n_{\max}}{g}$$

 $L \rightarrow b = \operatorname{round}(\frac{L}{g})$

 $\Lambda(c|j) =$ **Resource Occupancy:** conditional probability that *c* resources are occupied in state *j*

$$\begin{array}{l} A(c \mid j) = P(j)[vA(c-b \mid j-b) + (1-v)A(c \mid j-b)], \\ \text{for } j = 1, \dots, j_{\max} \text{ and } c \leq j \\ \text{where } A(0 \mid 0) = 1 \text{ and } A(c \mid j) = 0 \text{ for } c > j \\ \text{where } A(0 \mid 0) = 1 \text{ and } A(c \mid j) = 0 \text{ for } c > j \\ \text{(active user arrived)} \\ \end{array}$$

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The WH-EMLM Local Blocking Factor

Local Blocking Factor: due to the inter-cell interference blocking may occur in every state *j* with probability LB(j)

New Calls

- λ_N : mean arrival rate of new calls (Poisson process)
- μ_N : mean service rate of a new call
- **Y_N (j):** number of in-service calls in state j
- $\lambda_N(j) = \lambda_N(1-LB_N(j))$: effective arrival rate in j

Handoff Calls

- λ_H : mean arrival rate of handoff calls (Poisson)
- μ_H : mean service rate of handoff calls
- Y_H (j): number of in-service handoff calls in state j
- $\lambda_H(\mathbf{j}) = \lambda_H(\mathbf{1}-LB_H(\mathbf{j}))$: effective arrival rate in j

$\mu_{\rm H} > \mu_{\rm N}$

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$$LB_N(j) = \sum_{c=0}^{j} \beta_N(c) \Lambda(c \mid j)$$

$$LB_H(j) = \sum_{c=0}^{j} \beta_H(c) \Lambda(c \mid j)$$

The WH-EMLM State Transition Diagram

- **s_N**: Number of New Calls
- **s_H**: Number of Handoff Calls
- $j = (s_H + s_N) b$: occupied bandwidth (system state)



The WH-EMLM Call Blocking Probabilities Calculation

State Probabilities

$$\hat{q}(j) = \begin{cases} 1 & \text{for } j = 0\\ \frac{1}{j} \alpha_N (1 - LB_N(j - b)) b \hat{q}(j - b) + \\ \frac{1}{j} \alpha_H (1 - LB_H(j - b)) b \hat{q}(j - b) & \text{for } j = 1, ..., j_{\text{max}} \\ 0 & \text{otherwise} \end{cases}$$

$$q(j) = \frac{\hat{q}(j)}{\sum_{j=0}^{j_{\text{max}}} \hat{q}(j)}$$

Call Blocking Probabilities

$$B_N = \sum_{j=0}^{j_{max}} q(j) LB_N(j)$$

$$B_H = \sum_{j=0}^{j_{max}} q(j) L B_H(j)$$

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The WH-EMLM Generalization to K Service-Classes State Probabilities

$$\hat{q}(j) = \begin{cases} 1 & \text{for } j = 0 \\ \frac{1}{j} \sum_{k=1}^{K} \alpha_{N,k} (1 - LB_{N,k}(j - b_k)) b_k \hat{q}(j - b) + \\ \frac{1}{j} \sum_{k=1}^{K} \alpha_{H,k} (1 - LB_{H,k}(j - b_k)) b_k \hat{q}(j - b_k) & \text{for } j = 1, ..., j_{\text{max}} \\ 0 & \text{otherwise} \end{cases}$$

$$q(j) = \frac{\hat{q}(j)}{\sum_{j=0}^{j_{\text{max}}} \hat{q}(j)}$$

Bandwidth Share



Evaluation – Application Example

We compare Analytical to Simulation CBP results

	Data	Video			
Transmission rates (Kbps)	$R_1 = 144$	$R_2 = 384$			
Activity factor	v ₁ =0.7	v ₂ =0.6			
BER parameter (dB)	$(E_b/N_0)_1 = 3$	$(E_b/N_0)_2 = 4$			
Inter-cell Interference	$E[I_{inter}] = 2*10^{-18} \text{ mW and } CV[I_{inter}] = 1$				

Traffic load point	1	2	3	4	5	6	7	8	9	
New call Offered traffic for Data, (erl)	1.0	1.5	2.0	2.5	3.0	3.5	4.0	4.5	5.0	
Handoff Call Offered traffic for Data (erl)	0.2	0.3	0.4	0.5	0.6	0.7	0.8	0.9	1.0	
New call Offered traffic for Video (erl)	0.2	0.3	0.4	0.5	0.6	0.7	0.8	0.9	1.0	
Handoff Call Offered traffic for Video (erl)	0.1	0.15	0.2	0.25	0.3	0.35	0.4	0.45	0.5	

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The Wireless finite CDTM

Vassilakis et. al (IEEE ICC 2008)



User Activity: users alternate between transmitting and silent periods

- Active users: have a call in progress (occupy system resources)
- Passive users: are silent (do not occupy any system resources)

The Wireless finite CDTM **Interference & Call Admission Control**



Interference

Intra-cell Interference: *I*_{intra}

Inter-cell Interference: *I*_{inter}

Thermal Noise: P_N

Need to preserve the QoS of in-service calls

Call Admission Control

Cell Load, Load Factor and Local Blocking Probability

 $n \equiv$ **Cell Load**: Shared system bandwidth/resource

$$n = \frac{I_{intra} + I_{inter}}{I_{intra} + I_{inter} + P_N} = n_{intra} + n_{inter} \qquad n = \frac{NR - 1}{NR} \implies n_{max} = \frac{NR_{max} - 1}{NR_{max}}$$

$$NR = \frac{I_{intra} + I_{inter} + P_N}{P_N} \qquad We \text{ use } Cell \text{ Load (instead of Noise Rise) for the CAC}$$

 $L_{k,l}$ = Load Factor: call resource requirement

$$L_{k,l} = \frac{(E_b / N_0)_{k,l} * R_{k,l}}{W + (E_b / N_0)_{k,l} * R_{k,l}}$$

R_{k,l} : Transmission bit rate

 $(Eb/No)_{k,l}$: Bit error rate (BER) parameter

W = 3.84 Mcps: Chip rate (bit rate of the spreading signal)

(NEW CAC CRITERION)

 $\beta_{k,l}$ = Local Blocking Probability: depends on the system occupied resources as well as on the calls requirement

$$P(n_{intra}) = P(n_{intra} + n_{inter} + L_{k,l} > n_{max})$$

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 β_{k}

The Wireless finite CDTM

Resource Discretization & Resource Occupancy

g: basic cell load unit used for Resource Discretization $n \rightarrow j = \frac{n}{g}$ $n_{\max} \rightarrow C = \frac{n_{\max}}{g}$ $L_{k,l} \rightarrow b_{k,l} = \text{round}(\frac{L_{k,l}}{\sigma})$ $c-b_k/j-b_k$ $c/j-b_k$ $\Lambda(c \mid j) =$ Resource Occupancy: conditional probability that *c* 1-V_k resources are occupied in state *j* (active user c/j (passive user arrived) arrived) $\Lambda(c \mid j) = \sum_{k=1}^{K} \sum_{k=1}^{S_k} P_{k,l}(j) [v_k \Lambda(c - b_{k,l} \mid j - b_{k,l}) + (1 - v_k) \Lambda(c \mid j - b_{k,l})],$ for $j = 1, ..., j_{\max}$ and $c \le j$ where $\Lambda(0 \mid 0) = 1$ and $\Lambda(c \mid j) = 0$ for c > jSeptember 23, 2012 Emerging 2012 Barcelona

The Wireless finite CDTM Local blocking factor

Local Blocking Factor: due to the inter-cell interference. Blocking may occur in every state *j* with probability $LB_{k,l}(j)$ $LB_{k,l}(j) = \sum_{c=0}^{j} \beta_{k,l}(c)A(c \mid j)$

- $\lambda_{k,l}$: arrival rate from an idle source
- $\mu_{k,l}$: service rate
- $n_{k,l}(j)$: number of in-service calls in state j
- $(N_k n_{k,l}(j)) \lambda_{k,l} (1-LB_{k,l}(j))$: effective arrival rate in state j





The Wireless finite CDTM Call blocking probabilities calculation

Un-normalized State Probabilities

$$\hat{q}(j) = \begin{cases} 1 & \text{for } j = 0 \\ \sum_{k=1}^{K} \sum_{l=0}^{S_k} (N_k - \sum_{l=0}^{S_k} n_{k,l}(j) + 1) A_{k,l}(j) \hat{q}(j - b_{k,l}) & \text{for } j = 1, ..., j_{\max} \\ 0 & \text{otherwise} \end{cases}$$

$$A_{k,l}(j) = \alpha_{k,l}(1 - LB_{k,l}(j - b_{k,l})b_{k,l}\delta_{k,l}(j)$$

$$n_{k,l}(j) \approx \frac{a_{k,l}(j)q(j-b_{k,l})(1-LB_{k,l}(j-b_{k,l}))}{q(j)}$$

Normalization

$$q(j) = \frac{\hat{q}(j)}{\sum_{i=0}^{j_{\text{max}}} \hat{q}(j)}$$

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The Wireless finite CDTM Call blocking probabilities

Performance Metrics

Bandwidth Share

$$P_{k,l}(j) = \frac{(N_k - \sum_{l=0}^{S_k} n_{k,l}(j) + 1) A_{k,l}(j)q(j - b_{k,l})}{jq(j)}$$

Call Blocking Probabilities

$$B_{k} = \sum_{j=0}^{j_{max}} q(j) \sum_{l=1}^{S_{k}} \omega_{k,l}(j) LB_{k}(j)$$

$$\omega_{k,1}(j) = \begin{cases} 1 & \text{when } j \leq J_{k,1} \\ 0 & \text{otherwise} \end{cases}$$
$$\omega_{k,l}(j) = \begin{cases} 1 & \text{when } J_{k,l} < j \leq J_{k,l+1} \\ 0 & \text{otherwise} \end{cases}, \text{ for } l > 1$$

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Evaluation – 1st Application Example

Service-class Data		Video			
Туре	Elastic	Elastic			
Transmission rate (Kbps)	$R_{1,1}$ =64 and $R_{1,2}$ =32	$R_{2,1}=144$, $R_{2,2}=128$ and $R_{2,3}=112$			
Thresholds	$J_{1,1} = 0.6$	$J_{2,1}=0.4$ and $J_{2,2}=0.6$			
Activity factor	v ₁ =1.0	v ₂ =0.7			
BER parameter (dB)	$(E_b/N_0)_1 = 4$	$(E_b/N_0)_2 = 3$			

Characteristics of the Service-classes

Evaluation – 1st Application Example (cont.)





Evaluation – 2nd Application Example

Characteristics of the Service-classes

Service-class	Voice	Data	Video			
Туре	Stream	Elastic	Elastic			
Transmission rate (Kbps)	$R_{1,1}=12.2$	$R_{2,1}=128$ and $R_{2,2}=64$	$R_{3,1}=384, R_{3,2}=144 \text{ and } R_{3,3}=128$			
Thresholds -		$J_{2,1}=0.6$	$J_{3,1}=0.4$ and $J_{3,2}=0.6$			
Activity factor	v ₁ =0.5	v ₂ =1.0	v ₃ =0.7			
BLER parameter (dB)	LER parameter (dB) $(E_b/N_0)_1=5$		$(E_b/N_0)_3 = 3$			
Number of sources N ₁ =100		N ₂ =50	N ₃ =10			

		Offered	I traffic	-load			
Traffic load point:		1	2	3	4	5	6
Offered traffic-load (erl)	Voice	4.0	6.0	8.0	10.0	12.0	14.0
	Data	1.0	1.4	1.8	2.2	2.6	3.0
	Video	0.1	0.2	0.3	0.4	0.5	0.6

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Evaluation – 2nd Application Example (cont.)

We compare Analytical to Simulation results



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