# EMS performance evaluation with analytical stochastic models 

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The answer to Rob's question from yesterday ...

## CV Assumptions

- For an average call, travel time $T_{\text {avg }}$ has: variance $=b_{0}+b_{1} \times$ mean $\leftarrow$ Herman and Lam (1974)
- Random variable $B$ (mean 1 , variance $b_{2}$ ) captures call-to-call variability via $T=B \times T_{\text {avg }}$, where $T$ is travel time for a randomly chosen call
- $B$ and $T_{\text {avg }}$ are independent
- Functional form of CV vs. median relation is the same as for CV vs. mean


## Parametric CV function

- Then: $\mathrm{CV}(d)=\frac{\sqrt{b_{0}\left(b_{2}+1\right)+b_{1}\left(b_{2}+1\right) m(d)+b_{2} m(d)^{2}}}{m(d)}$
- Interpretation of parameters:
- $b_{0}$ : "fixed variability"-data recording errors, time spent finding an address, spatial aggregation, etc.
- $b_{1}$ : short-term variability in speed during a trip
- $b_{2}$ : long-term call-to-call variability, due to factors not included in the model
- CV approaches $\sqrt{ } b_{2}$ as distance goes to infinity
- CV has same breakpoint as median


## Parametric Functions



## Outline

- Performance Evaluation Models
- Using the Erlang B Performance Evaluation Model for Yellow and Red Alerts


## Performance Evaluation

## Decomposing Performance

- Performance estimates:
- $p_{i j}=$ estimated performance for calls from $j$ if station $i$ responds
- "performance:" could be coverage probability / survival probability / average response time / ...
- Dispatch probabilities:
$-f_{i j}=\operatorname{Pr}\{$ station $i$ responds | call from j\}
- This is where queueing / service syste these
- Call arrival rates:
- Neighborhood $j: \lambda_{j}$, system: $\lambda$
- System performance: $\sum_{j} \frac{\lambda_{j}}{\lambda} \sum_{i} f_{i j} p_{i j}$


## ("Simplest Interesting"?) Example



2 stations, each with 1 unit
2 neighborhoods
$1 / \mu=$ avg. service time $=1$ hour
$\lambda=$ call arrival rate $=1 /$ hour

Performance estimates:
$p_{11}=\operatorname{Pr}\{$ response time $\leq$ standard $\mid$ call from 1,1 responds $\}$

$$
=0.95
$$

$p_{12}=p_{21}=0.5$
$p_{22}=0.95$

## Model 1: "Always Available"



Assumes all stations have an available ambulance at all times
Provides upper bound on performance
Used in some station location optimization models

## Model 2: Binomial



Input:
$p=$ average busy fraction = $0.4=$ probability that an ambulance is busy, independent of status of all other ambulances

Used in some ambulance allocation optimization models

## Model 3: Erlang B



| Model | $f_{11}$ | $f_{21}$ | $f_{12}$ | $f_{22}$ | B | Performance |
| :--- | :---: | :---: | :---: | :---: | :---: | :---: |
| Always available | 1.00 | 0.00 | 0.00 | 1.00 | 0.00 | 0.95 |
| Binomial | 0.60 | 0.24 | 0.24 | 0.60 | 0.16 | 0.69 |
| Erlang B | 0.60 | 0.20 | 0.20 | 0.60 | 0.20 | 0.67 |


| $p_{11}$ | $p_{21}$ | $p_{12}$ | $p_{22}$ |
| :---: | :---: | :---: | :---: |
| 0.95 | 0.50 | 0.50 | 0.95 |

$\lambda$ Was chosen so that ambulance utilization $=p=0.4$


Probability that closest ambulance responds is the same as in binomial model

Probability that $2^{\text {nd }}$-closest ambulance responds is lower, because $\operatorname{Pr}\left\{2^{\text {nd }}\right.$-closest ambulance is busy $\mid$ closest ambulance is busy $\}>p$

## Model 4: Hypercube Queueing Model

|  | Model | $f_{11}$ | $t_{21}$ | $f_{12}$ | $f_{22}$ | B | Performance |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | Always available | 1.00 | 0.00 | 0.00 | 1.00 | 0.00 | 0.95 |
|  | Binomial | 0.60 | 0.24 | 0.24 | 0.60 | 0.16 | 0.69 |
|  | Erlang B | 0.60 | 0.20 | 0.20 | 0.60 | 0.20 | 0.67 |
| $\lambda_{1}=0.2 / \mathrm{hr}$. | HQM | 0.66 | 0.14 | 0.26 | 0.54 | 0.20 | 0.65 |
| - |  | $p_{11}$ | $p_{21}$ | $p_{12}$ | $p_{22}$ |  |  |
| $\lambda_{2}=0.8 / \mathrm{hr} .$ |  | 0.95 | 0.50 | 0.50 | 0.95 |  |  |



In this model, the two ambulances are distinguishable
$\rightarrow$ Ambulance 2 is busier
$\rightarrow$ Neighborhood 2 has a lower probability of closest station responding

## NoPenting

$\lambda_{1}=0.2 / \mathrm{hr} . /$|  |  |  |  |  |  |  |
| :--- | :---: | :---: | :---: | :---: | :---: | :---: |
| Model | $f_{11}$ | $f_{21}$ | $f_{12}$ | $f_{22}$ | B | Performance |
| Always available | 1.00 | 0.00 | 0.00 | 1.00 | 0.00 | 0.95 |
| Binomial | 0.60 | 0.24 | 0.24 | 0.60 | 0.16 | 0.69 |
| Erlang B | 0.60 | 0.20 | 0.20 | 0.60 | 0.20 | 0.67 |
| HQM | 0.66 | 0.14 | 0.26 | 0.54 | 0.20 | 0.65 |
| Repositioning | 0.45 | 0.35 | 0.09 | 0.72 | 0.20 | 0.70 |
|  | $p_{11}$ | $p_{21}$ | $p_{12}$ | $p_{22}$ |  |  |
|  | 0.95 | 0.50 | 0.50 | 0.95 |  |  |

Models 1-4 assume an ambulance always returns to its home station

Model 5: If only one ambulance is available and it is at Station 1, then move it to Station 2 (avg. move time $=6 \mathrm{~min}$.)

Neighborhood 1 is better off, Neigbhorhood 2 is worse off

## Comparison of Models

| Model | Performance | Increased <br> realism | Repositioning | Incorporated in <br> math programs | Scaling <br> issues |
| :--- | :---: | :---: | :---: | :---: | :---: |
| Always available | 0.95 | $\square$ |  | $\checkmark$ |  |
| Binomial | 0.69 |  |  | $\checkmark$ |  |
| Erlang B | 0.67 |  |  |  | $?$ |
| HQM | 0.65 |  |  | $?$ |  |
| Repositioning | 0.70 |  | $\checkmark$ | $?$ |  |

Managing red and yellow alerts and the consequences of calling in additional units or expediting hospital turnaround

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Managing red and yellow alerts and the consequences of calling in additional units or expediting hospital furnaround


## Ambulance shortage periods

'Busy’ ambulance system causes concern for paramedics
During the first nine months of 2010, the city of Edmonton had no ambulances to cover medical emergencies for almost 10 hours in total.

- Edmonton Journal, Jan. 20, 2012

Too few paramedics to answer call: Union official

- Toronto Sun, May 13, 2012

Opposition demands EMS wait time review

- Calgary Sun, Feb. 24, 2012


## Alert periods

Periods during which:

- Most ambulances are busy

Yellow Alert
Available ambulances below a threshold.
Calgary EMS threshold $=12$ ambulances

- All ambulances are busy

Red Alert

## Yellow alert example



Time

## Descriptive Statistics

| Table 1 | EMS configuration in Edmonton in 2008 and in Calgary in 2009. |  |
| :--- | :--- | :--- | :--- |
| Parameter | Edmonton | Calgary |
| Yellow Alert threshold $(\theta)$ | 8 | 12 |
| Minimum number of scheduled ambulances | 19 | 28 |
| Maximum number of scheduled ambulances | 36 | 54 |
| Average number of scheduled ambulances | 25 | 41 |

Table 2 Descriptive statistics for the duration of alert periods in Edmonton in 2008 and in Calgary in 2009.

|  | Yellow Alert |  |  | Red Alert |  |
| :--- | :--- | :--- | :--- | :--- | :--- |
| Statistic | Edmonton | Calgary |  | Edmonton | Calgary |
| Sample Size | 1349 | 703 |  | 587 | 9 |
| Mean (min.) | 106.41 | 7.09 |  | 7.20 | 1.37 |
| Standard Deviation (min.) | 120.26 | 11.53 |  | 11.32 | 1.32 |
| Maximum (min.) | 1012.02 | 127.28 |  | 138.93 | 4.53 |
| Squared Coefficient of Variation | 1.28 | 2.64 |  | 2.47 | 0.94 |

## Decision faced by dispatchers

- Ride out the alert... or act?
- Possible actions:
- Reposition ambulances*
- Call in additional ambulances
- Free up busy ambulances in EDs
- ?
*Alanis et al. 2012, Maxwell et al. 2010, Schmid 2011

Mathematical Model: "Erlang B loss model"
Analogy:
phone lines = ambulances
busy signal = red alert
k-partial busy period:
$k$ or more of $c$ servers are busy


Yellow Alert $=(c-$ threshold +1$)$-partial busy period
Red Alert = c-partial busy period

## Relationship between alert periods and partial busy periods

Calgary: 41 servers
Yellow Alert $=30$-partial busy period


## Main Result

Equations to calculate average busy period durations:
$\mathrm{E}\left(B_{c}\right)=\frac{1}{c \mu}, \mathrm{E}\left(B_{k}\right)=\frac{\lambda \mathrm{E}\left(B_{k+1}\right)}{k \mu}+\frac{1}{k \mu}, k=c-1, \ldots, 1$.

Also have equations for variance and other quantities

## Validation-the whole year



## Reasons for poor fit

- Number of units varies with time
- Call rates vary with time
- "Service speed" varies with number of busy units
- Check how much fit improves after controlling for these factors


## Validation for weekday 9 am - 1 pm



## Aggregation over 16 time segments



## Actions and Performance Measures

- Actions
- Call in additional units
- Free up units in EDs Modeled as "increase service rate"
- Performance measures
- Average remaining Yellow Alert duration
- Average number of "missed" calls (because of red alert)


## Expediting Hospital Turnaround




## Calling in Additional Units



## "Optimal" Combination of Actions



Called in ambulances

(b) Expected number of lost calls.
(a) Expected residual Yellow Alert duration.

## The Optimal Combination can Depend on the Performance Measure



Called in ambulances

| sooutinquit pasearan | 0.32 |  |  |  |
| :---: | :---: | :---: | :---: | :---: |
|  | 0.36 | 0.31 |  |  |
|  | 0.40 | 0.34 | 0.31 |  |
|  | 0.45 | 0.38 | 0.34 | 0.32 |
|  | 0 | 1 | 2 | 3 |

(b) Expected number of lost calls.
(a) Expected residual Yellow Alert duration.

