

A Data-Based Science for Service Engineering and Management, or “Empirical Adventures in Call-Centers and Hospitals”

Avi Mandelbaum

Technion, Haifa, Israel

<http://ie.technion.ac.il/serveng>

LOIS Lecture, Eindhoven, October 2010

Research Partners

- ▶ **Students:**

Aldor*, Baron*, Carmeli, Feldman, Garnett*, Gurvich*, Khudiakov*, Maman*, Marmor*, Reich, Rosenshmidt*, Shaikhet*, Senderovic, Tseytlin*, Yom-Tov*, Zaied, Zeltyn*, Zohar*, Zviran, ...

- ▶ **Empirical/Statistical Analysis:**

Feigin; Brown, Gans, Zhao; Shen; Ritov, Goldberg; Allon, Bassamboo, Gurvich; Armony, ...

- ▶ **Theory:**

Armony, Atar, Gurvich, Jelenkovic, Kaspi, Massey, Momcilovic, Reiman, Shimkin, Stolyar, Wasserkrug, Whitt, Zeltyn, ...

- ▶ **Industry:**

IBM Research (OCR: Carmeli, Vortman, Wasserkrug, Zeltyn), Rambam Hospital, Hapoalim Bank, Mizrahi Bank, Pelephone Cellular, ...

- ▶ **Technion SEE Center / Laboratory:**

Feigin; Trofimov, Nadjharov, Gavako, Kutsyy; Liberman, Koren, Rom, Plonsky; Research Assistants, ...

History, Resources (Downloadable)

- ▶ Math. + C.S. + Stat. + O.R. + Mgt. $\Rightarrow$ **IE** (≥ 1990)
- ▶ **“Service-Engineering” Course** (≥ 1995):
<http://ie.technion.ac.il/serveng> - website
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- ▶ Search Google-Scholar for **<Call Centers>**
- ▶ **SEELab** (≥ 2007), following StatLab (≥ 2000):
Data Repositories for Research & Teaching; Reports, Tutorials:
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- ▶ **OCR Project** (≥ 2008): **Hospitals**
IBM Research + Rambam Hospital + Technion IE&M
http://ie.technion.ac.il/Labs/Serveng/closed/OCR_Documents.php

The Case for Service Science / Engineering

- ▶ **Service Science / Engineering** (vs. Management) are emerging **Academic Disciplines**. For example, universities (world-wide), IBM (SSME, a la Computer-Science), USA NSF (SEE), Germany IAO (ServEng), ...

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 - **Call Centers**
 - **Hospitals**
 - **Transportation**
 - Justice, Fast Food, Police, Internet, . . .
- ▶ **Simple models** at the Service of **Complex Realities** (Human)
Note: Simple yet rooted in **deep analysis**.

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- ▶ Mostly **What Can Be Done** vs. **How To**

Title: Expands the Scientific Paradigm

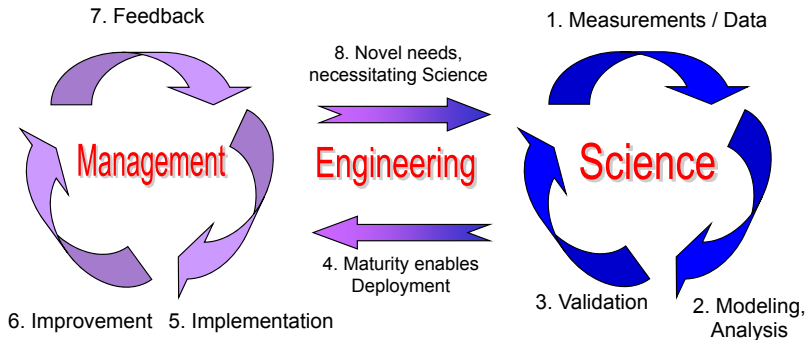
Physics, Biology, . . . : Measure, Model, Experiment, Validate, Refine.
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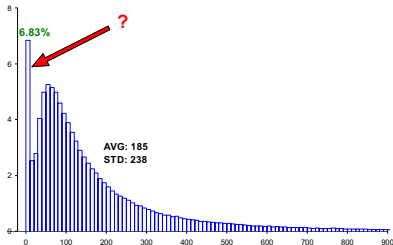
Starting with **Data**, expand to:



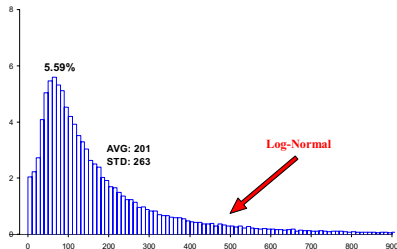
Beyond Averages: The Human Factor

Histogram of Service-Time in a (Small Israeli) Bank, 1999

January-October



November-December

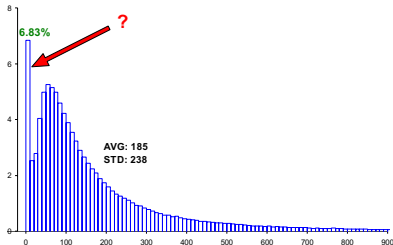


► 6.8% Short-Services:

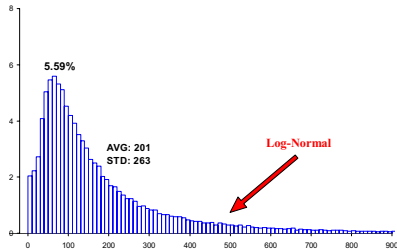
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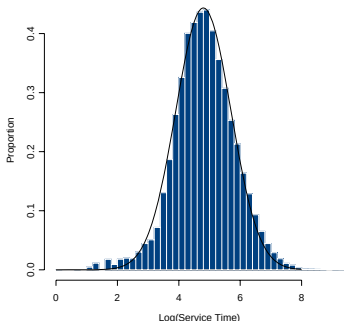


- ▶ **6.8% Short-Services:** Agents' "Abandon" (improve bonus, rest), (mis)lead by **incentives**
- ▶ **Distributions** must be measured (in **seconds** = **natural scale**)
- ▶ **LogNormal** service times common in call centers

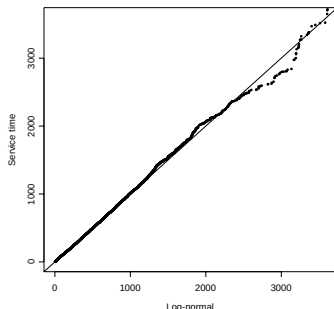
Validating LogNormality of Service-Times

Israeli Call Center, Nov-Dec, 1999

Log(Service Times)



LogNormal QQPlot

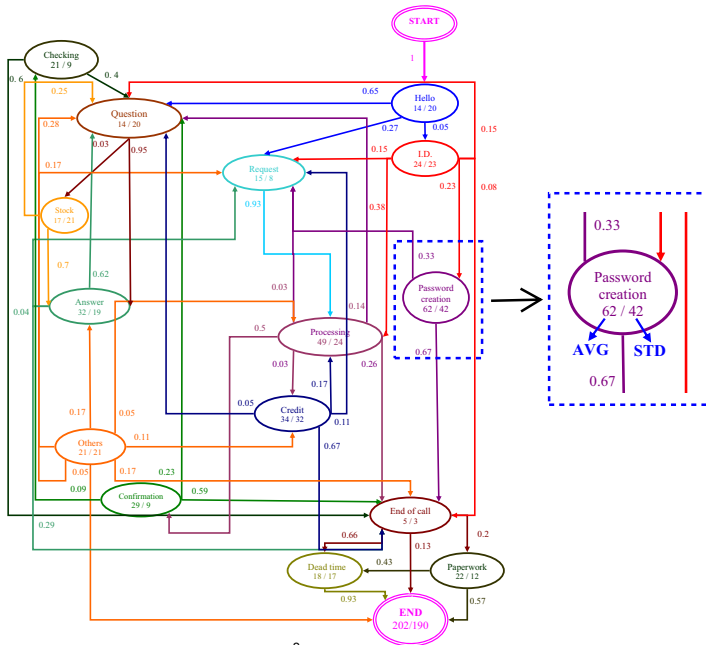


- ▶ **Practically Important:** (mean, std)(log) characterization
- ▶ **Theoretically Intriguing:** Why LogNormal ? Naturally multiplicative but, in fact, also **Infinitely-Divisible** (Generalized Gamma-Convolutions)
- ▶ Simple-model of a complex-reality? The **Service Process:**

(Telephone) Service-Process = "Phase-Type" Model

Retail
Service
(Israeli
Bank)

Statistics
OR
IE



Why Bother?

In large banking call centers:

+One Second to Service-Time implies **+Millions** in costs, annually

⇒ **Time and "Motion" Studies** (**Classical IE** with New-age IT)

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 - ▶ **Work Design** (w/ **Khudiakov**)
eg. **Cross-Selling**: higher profit vs. (costlier) longer services;
Analysis yields (congestion-dependent) cross-selling protocol
 - ▶ **"Worker" Design** (w/ **Gans & Shen**)
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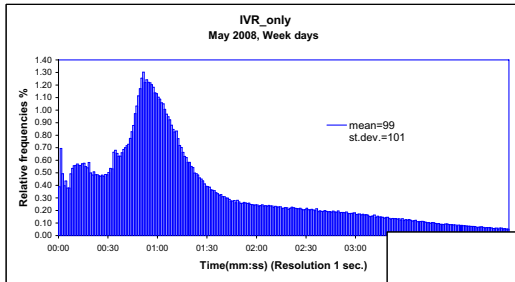
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- ▶ **IVR-Process Model**: Customer-Machine Interaction
75% services, poor design, yet scarce research;
Same approach, automatic (easier) data

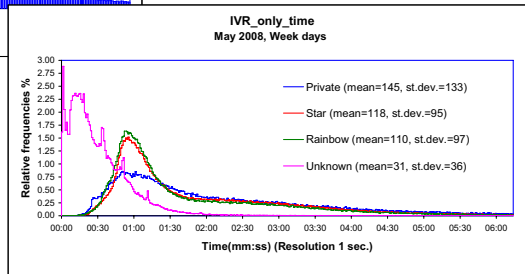
IVR-Time: Histograms

Israeli Bank: Served only by IVR, May 2008

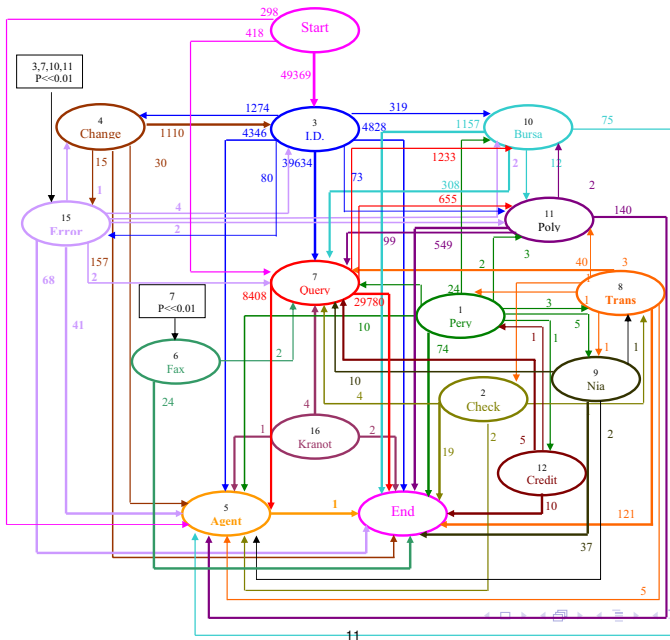
All Customers



By Service Type

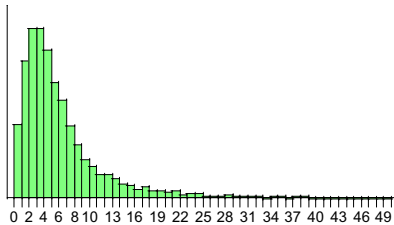


IVR-Process: "Phase-Type" Model



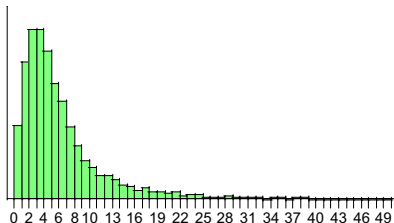
Beyond Averages: Length-of-Stay in a Hospital

Israeli Hospital, in Days: LN

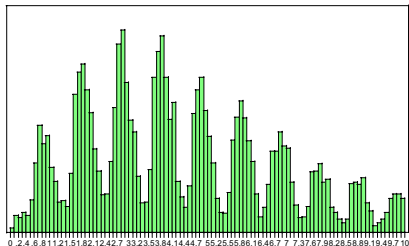


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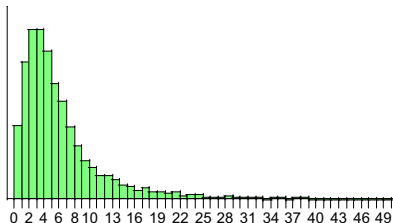


Israeli Hospital, in Hours

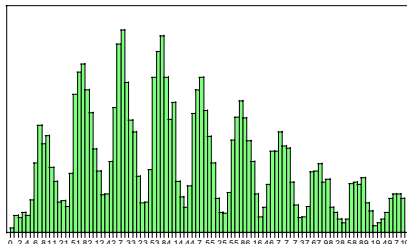


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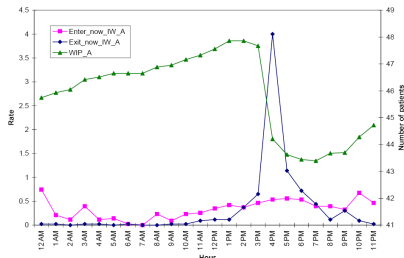
Israeli Hospital, in Hours



“Explanation”: Patients released around **3pm** (1pm in Singapore)

Why Bother ?

Staffing, Bed Management, ...



Started with Call Centers, Expanded to Hospitals

Call Centers - U.S. (Netherlands) Stat.

- ▶ \$200 – \$300 billion annual expenditures (0.5)
- ▶ 100,000 – 200,000 call centers (1500-2000)
- ▶ “Window” into the company, for better or worse
- ▶ Over 3 million agents = **2% – 4% workforce** (100K)

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Healthcare - similar and unique challenges:

- ▶ Cost-figures far more staggering
- ▶ Risks much higher
- ▶ ED (initial focus) = hospital-window
- ▶ Over 3 million nurses

Call-Center Environment: Service Network

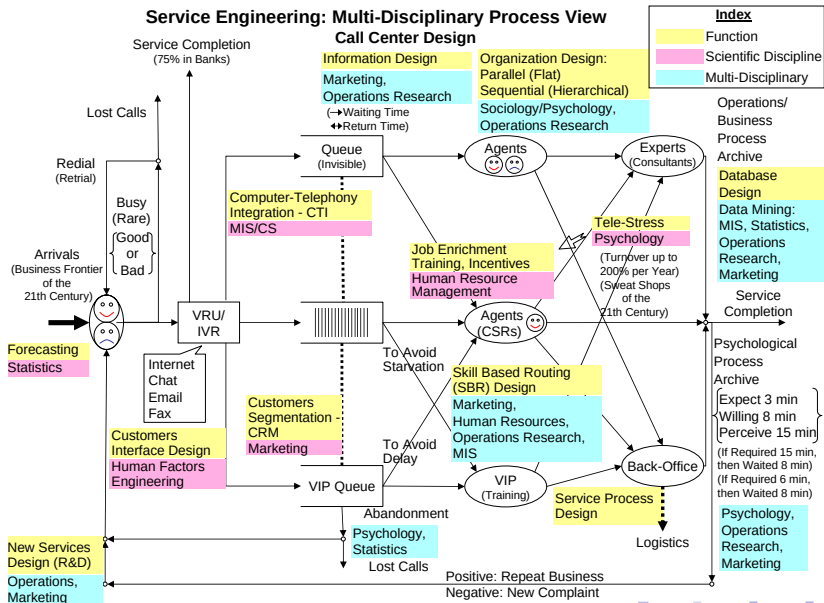


Call-Centers: “Sweat-Shops of the 21st Century”

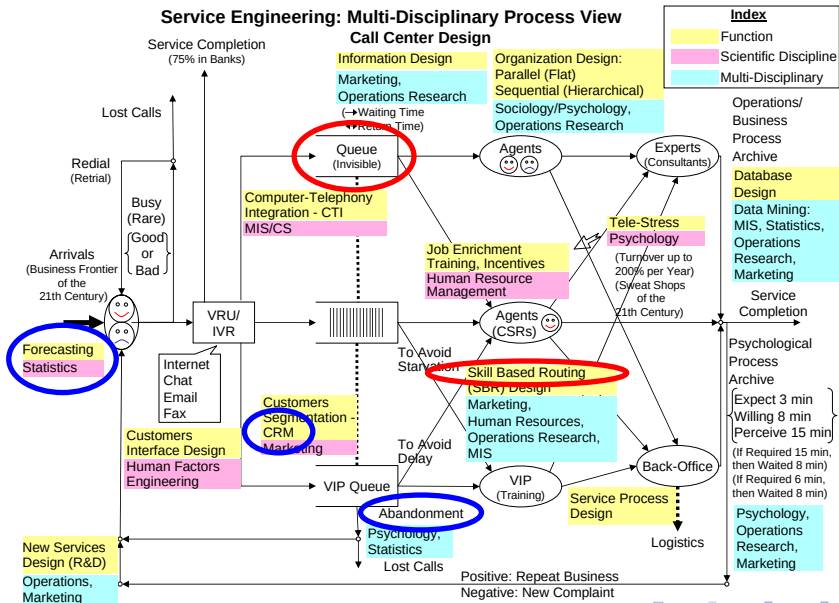


Call-Center Network: Gallery of Models

Service Engineering: Multi-Disciplinary Process View Call Center Design

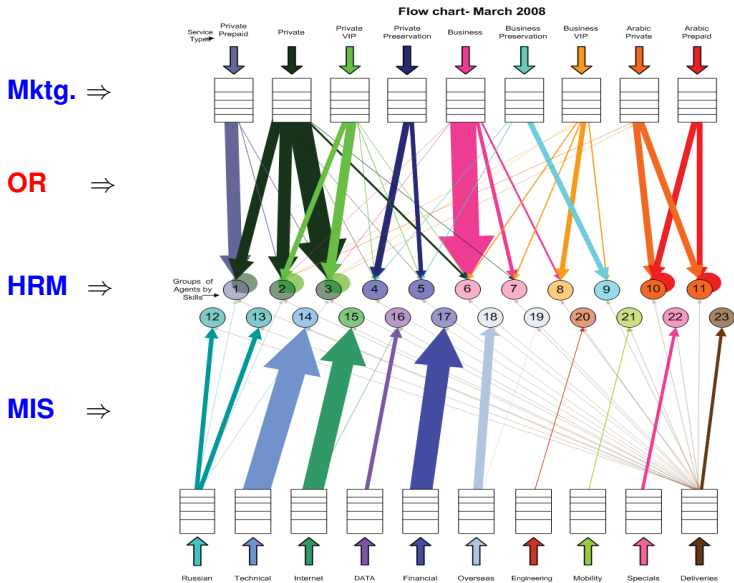


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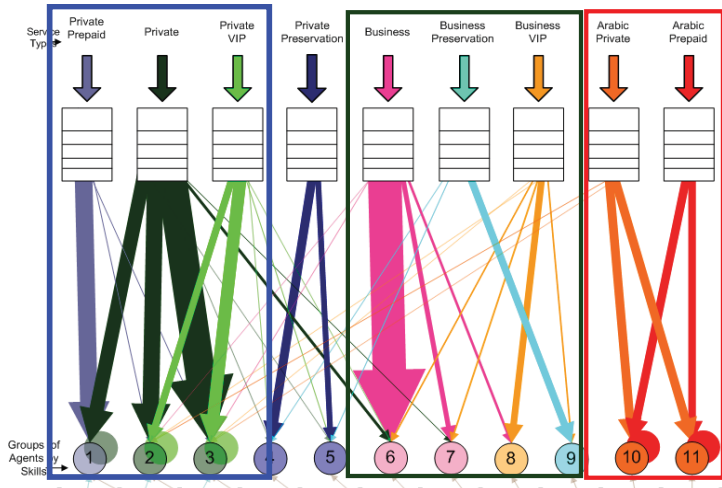
Skills-Based Routing in Call Centers

EDA and OR, with I. Gurvich and P. Liberman



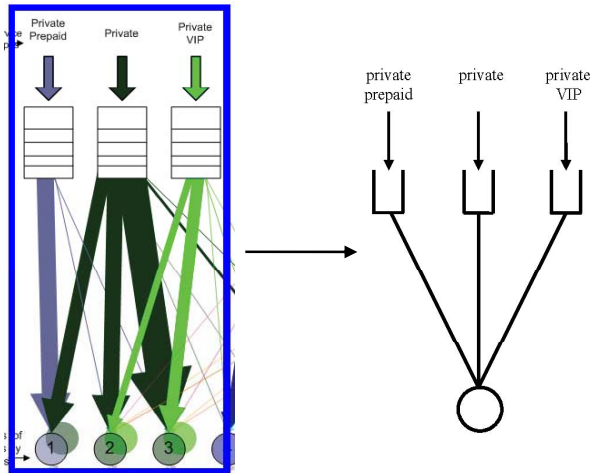
SBR Topologies: I; V, Reversed-V; N, X; W, M

Israeli Cellular, March 2008



SBR: Class-Dependent Services

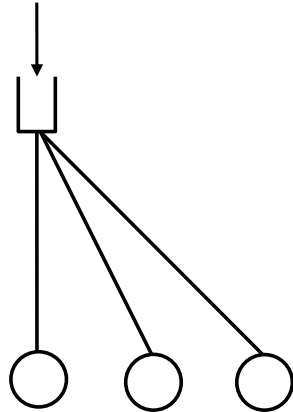
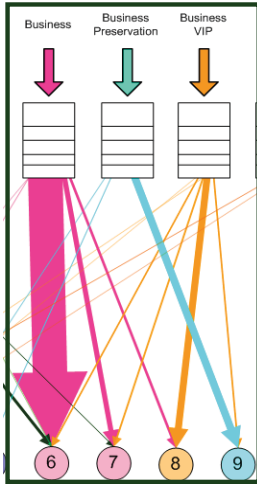
“Reduction” to V-Topology, with R. Atar and G. Shaikhet



Reduction in the sense of **equivalent Brownian Control Problems**

SBR: Pool-Dependent Services

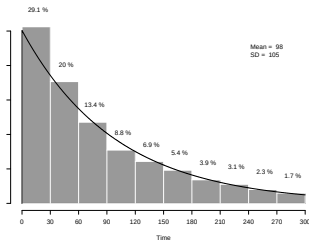
“Reduction” to Reversed-V and I, with R. Atar and G. Shaikhet



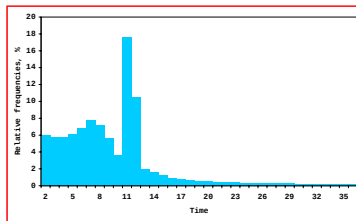
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Beyond Averages: Waiting Times in a Call Center

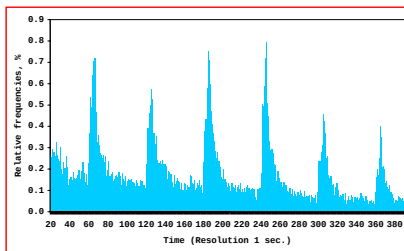
Small Israeli Bank



Large U.S. Bank



Medium Israeli Bank, in Seconds (Recall Hospital LOS, Hours)



ER / ED Environment: Service Network

Acute (Internal, Trauma)



Walking



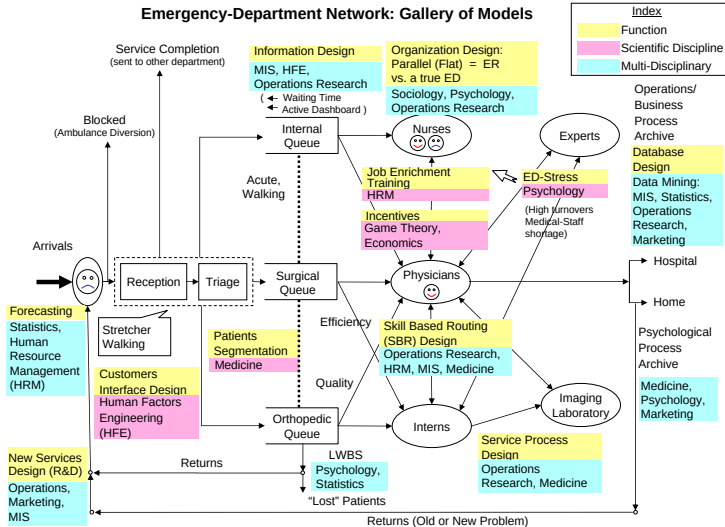
Multi-Trauma



Queueing in a “Good” Beijing Hospital, at 6am

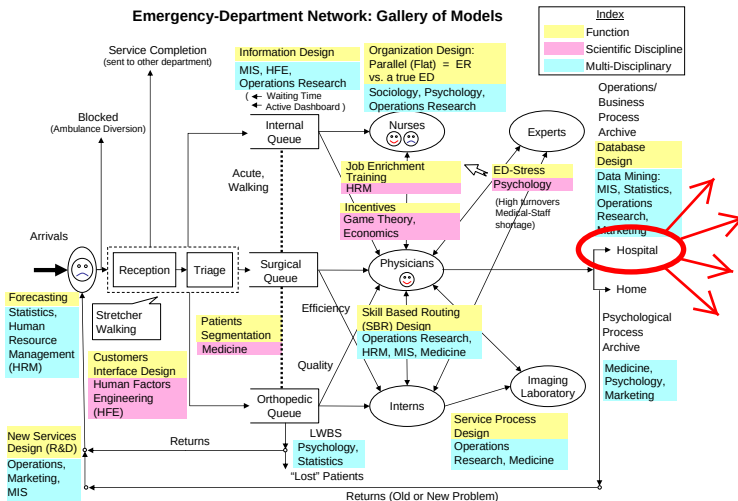


Emergency-Department Network: Gallery of Models



► **Forecasting**, $SBR \approx$ **Triage**, Abandonment = **LWBS**

Emergency-Department Network: Gallery of Models



- ▶ **Fork-Join** Q's, eg. After Physician: Nurse, Lab-Tests, X-Ray
- ▶ **Synchronization** Control, with **R. Atar** and **A. Zviran**
- ▶ **ED-to-IW** Routing

ED-to-IW Routing: A Hospital Bottleneck

Israeli Large Hospital (1/5/06 to 30/10/08, excluding 1-3/07)

	Ward A	Ward B	Ward C	Ward D
ALOS (days)	6.37	4.47	5.36	5.56
Avg Occupancy Rate	97%	95%	86%	92%
Avg # Patients per Month	206	187	210	210
Standard bed capacity	45	30	44	42
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- ▶ **Reversed-V**: **Queue** = ED, **Servers** in Pool = Beds in Ward (10's)
- ▶ **Information Analysis**: **QED/Sub-Diffusion Approx. (Natural)**

Prerequisite I: Data

Averages Prevalent (and could be useful / interesting).

But I need data at the level of the **Individual Transaction**:

For each service transaction (during a phone-service in a call center, or a patient's visit in a hospital, or browsing in a website, or ...), its

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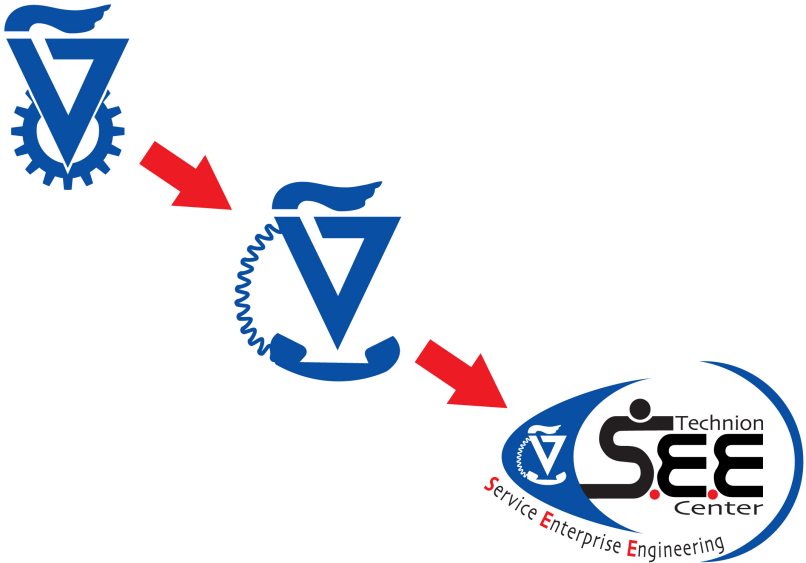
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- ▶ **Telephone** (Call Centers, via ACD / CTI, IVR/VRU)
- ▶ **Hospitals** (Emergency Departments, ...)
- ▶ Expanding:
 - ▶ Hospitals, via **RFID**, with **I. Cohen, S. Israelit (MD), Y. Marmor**
 - ▶ Operational + Financial + Contents (Marketing, Clinical)
 - ▶ Internet, Chat (multi-media)

Pause for a Commercial:

Pause for a Commercial: The Technion SEE Center



Technion SEE = Service Enterprise Engineering

SEELab: Data-repositories for research and teaching

- ▶ For example:
 - ▶ Bank Anonymous: **1 years, 350K calls by 15 agents** - in 2000.
 - ▶ U.S. Bank: **2.5 years, 220M calls, 40M by 1000 agents.**
 - ▶ Israeli Cellular: **2.5 years, 110M calls, 25M calls by 750 agents.**
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SEESat: Environment for graphical EDA in real-time

- ▶ **Universal Design, Internet Access, Real-Time Response.**

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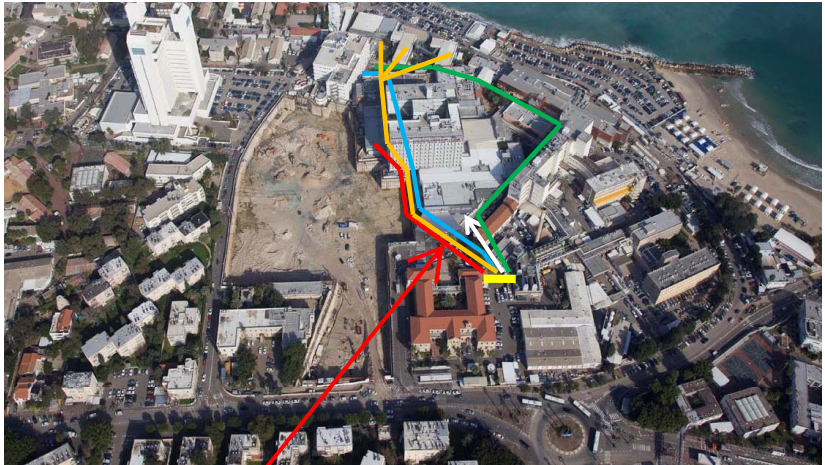
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SEEServer: Free for academic use

Register, then access (presently) U.S. Bank and Small Israeli Bank.

eg. RFID-Based Data: Mass Casualty Event (MCE)

Drill: Chemical MCE, Rambam Hospital, May 2010



Focus on **severely wounded** casualties (≈ 40 in drill)

Note: 20 observers support real-time control (will help validation)

Data Cleaning: MCE with RFID Support

Data-base				Company report		comment
Asset id	order	Entry date	Exit date	Entry date	Exit date	
4	1	1:14:07 PM		1:14:00 PM		
6	1	12:02:02 PM	12:33:10 PM	12:02:00 PM	12:33:00 PM	
8	1	11:37:15 AM	12:40:17 PM	11:37:00 AM		exit is missing
10	1	12:23:32 PM	12:38:23 PM	12:23:00 PM		
12	1	12:12:47 PM	12:35:33 PM		12:35:00 PM	entry is missing
15	1	1:07:15 PM		1:07:00 PM		
16	1	11:18:19 AM	11:31:04 AM	11:18:00 AM	11:31:00 AM	
17	1	1:03:31 PM		1:03:00 PM		
18	1	1:07:54 PM		1:07:00 PM		
19	1	12:01:58 PM		12:01:00 PM		
20	1	11:37:21 AM	12:57:02 PM	11:37:00 AM	12:57:00 PM	
21	1	12:01:16 PM	12:37:16 PM	12:01:00 PM		
22	1	12:04:31 PM	12:20:40 PM			first customer is missing
22	2	12:27:37 PM		12:27:00 PM		
25	1	12:27:35 PM	1:07:28 PM	12:27:00 PM	1:07:00 PM	
27	1	12:06:53 PM		12:06:00 PM		
28	1	11:21:34 AM	11:41:06 AM	11:41:00 AM	11:53:00 AM	exit time instead of entry time
29	1	12:21:06 PM	12:54:29 PM	12:21:00 PM	12:54:00 PM	
31	1	11:40:54 AM	12:30:16 PM	11:40:00 AM	12:30:00 PM	
31	2	12:37:57 PM	12:54:51 PM	12:37:00 PM	12:54:00 PM	
32	1	11:27:11 AM	12:15:17 PM	11:27:00 AM	12:15:00 PM	
33	1	12:05:50 PM	12:13:12 PM	12:05:00 PM	12:15:00 PM	wrong exit time
35	1	11:31:48 AM	11:40:50 AM	11:31:00 AM	11:40:00 AM	
36	1	12:06:23 PM	12:29:30 PM	12:06:00 PM	12:29:00 PM	
37	1	11:31:50 AM	11:48:18 AM	11:31:00 AM	11:48:00 AM	
37	2	12:59:21 PM		12:59:00 PM		

Imagine “Cleaning” 60,000+ customers per day (call centers) !

Prerequisite II: Models (Fluid Q's)

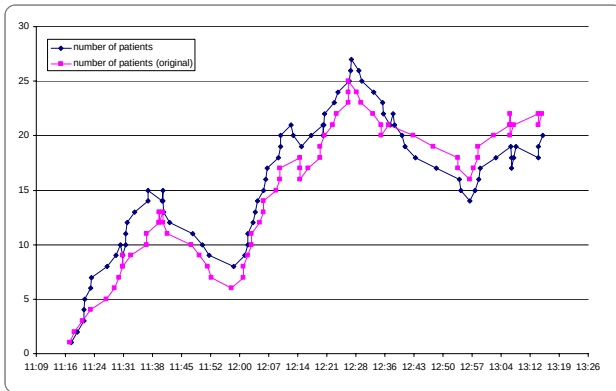
“Laws of Large Numbers” capture **Predictable** Variability

Deterministic Models: Scale Averages-out **Stochastic Individualism**

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Severely-Wounded Patients, 11:00-13:00



- ▶ Paths of doctors, nurses, patients (100+, **1 sec.** resolution)
eg. Help predict “What if **150+ casualties** severely wounded ?”
- ▶ **Transient** Q's, where **Service-Process** = **Needy-Content Cycles** (with G. Yom-Tov, PhD)

Prerequisite II: Models (Diffusion/QED's Q's)

Traditional Queueing Theory predicts that **Service-Quality** and **Servers' Efficiency** must be traded off against each other.

For example, **M/M/1** (single-server queue): **91%** server's utilization goes with

$$\text{Congestion Index} = \frac{E[\text{Wait}]}{E[\text{Service}]} = 10,$$

and only **9%** of the customers are served immediately upon arrival.

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Yet, heavily-loaded queueing systems with **Congestion Index = 0.1** (Waiting one order of magnitude less than Service) are prevalent:

- ▶ **Call Centers:** Wait **"seconds"** for **minutes** service;
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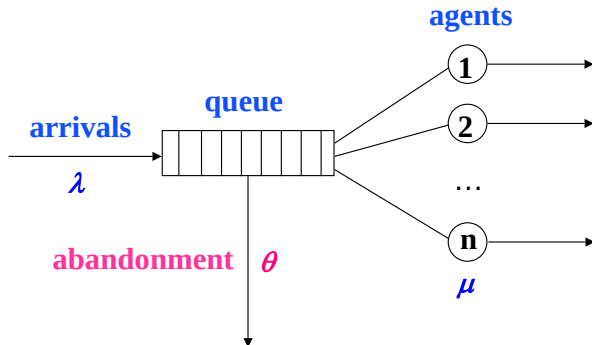
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and, moreover, a significant fraction are not delayed in queue. (For example, in well-run call-centers, **50%** served "immediately", along with over **90%** agents' utilization, is not uncommon) **?** **QED**

The Basic Staffing Model: Erlang-A (M/M/N + M)



Erlang-A (Palm 1940's) = **Birth & Death Q**, with parameters:

- ▶ λ – **Arrival** rate (Poisson)
- ▶ μ – **Service** rate (Exponential; $E[S] = \frac{1}{\mu}$)
- ▶ θ – **Patience** rate (Exponential, $E[\text{Patience}] = \frac{1}{\theta}$)
- ▶ n – Number of **Servers** (Agents).

Testing the Erlang-A Primitives

- ▶ **Arrivals:** Poisson?
- ▶ **Service-durations:** Exponential?
- ▶ **(Im)Patience:** Exponential?

Testing the Erlang-A Primitives

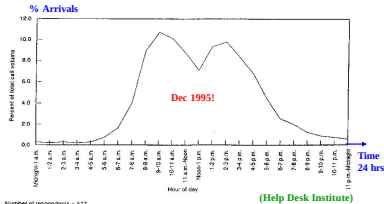
- ▶ **Arrivals:** Poisson?
- ▶ **Service-durations:** Exponential?
- ▶ **(Im)Patience:** Exponential?
- ▶ Primitives independent (eg. Impatience and Service-Durations)?
- ▶ Customers / Servers Homogeneous?
- ▶ Service discipline FCFS?
- ▶ ... ?

Validation: Support? Refute?

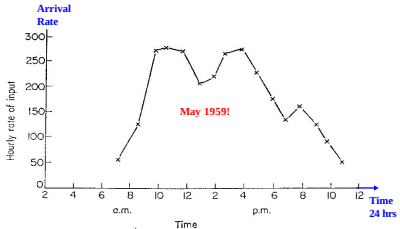
Arrivals to Service

Arrival-Rates to Three Call Centers

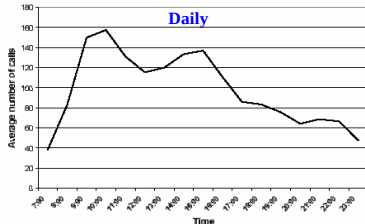
Dec. 1995 (U.S. 700 Helpdesks)



May 1959 (England)



November 1999 (Israel)

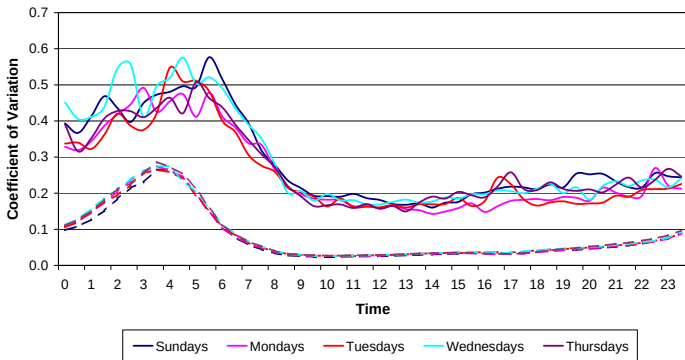


Random Arrivals “must be”
(Axiomatically)
Time-Inhomogeneous Poisson

Arrivals to Service: only Poisson-Relatives

Arrival-Counts: Coefficient-of-Variation (CV), per 30 min.

Israeli-Bank Call-Center, 263 regular days (4/2007 - 3/2008)

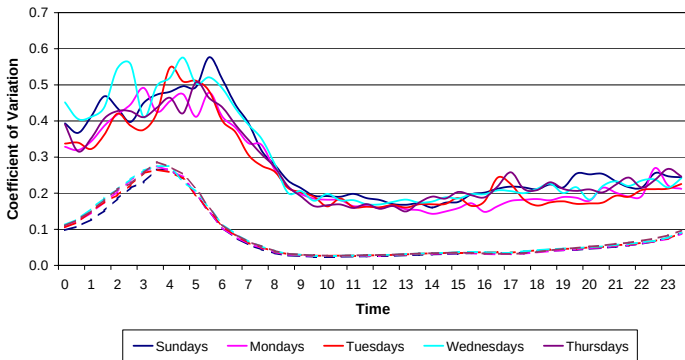


- ▶ **Poisson CV** (Dashed Line) = $1/\sqrt{\text{mean arrival-rate}}$
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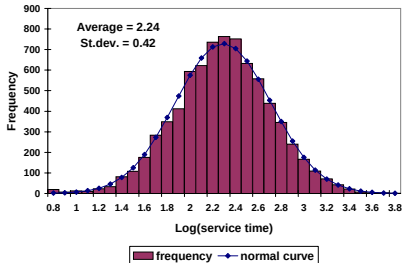
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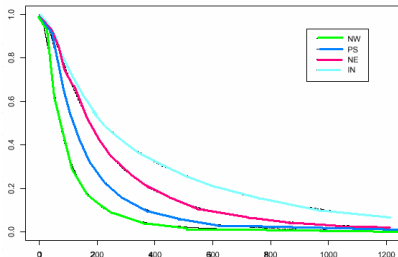
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- $\Rightarrow$ **Modeling** (Poisson-Mixture) of and **Staffing** ($> \sqrt{\cdot}$) against **Time-Varying Over-Dispersed** Arrivals (with **S. Maman** and **S. Zeltyn**)

Service Durations: LogNormal Prevalent

Israeli Bank Log-Histogram



Service-Classes Survival-Functions

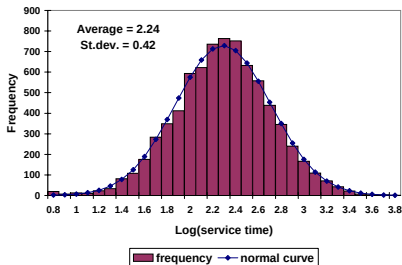


- **New Customers:** 2 min (NW);
- **Regulars:** 3 min (PS);

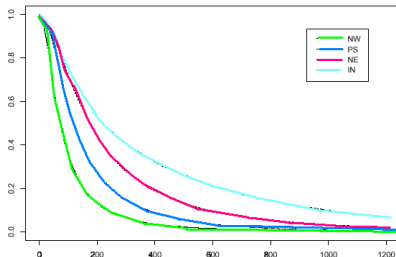
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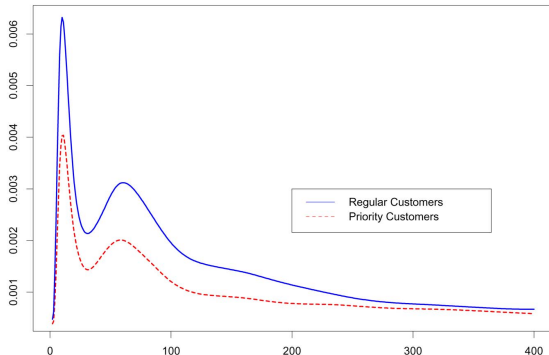
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- Service Durations are **LogNormal (LN)** and **Heterogeneous**

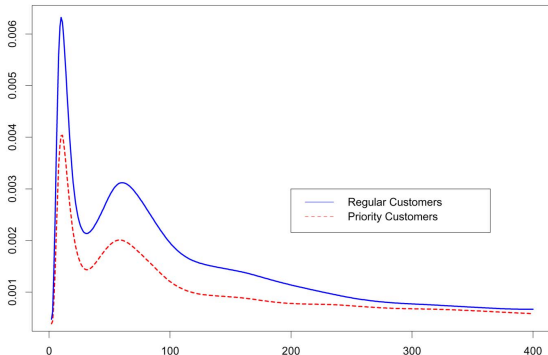
(Im)Patience while Waiting (Palm 1943-53)

Hazard Rate of (Im)Patience Distribution $\propto$ Irritation
Regular over **VIP** Customers – Israeli Bank



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- ▶ **VIP** Customers are **more Patient** (Needy)
- ▶ **Peaks** of abandonment at times of **Announcements**
- ▶ Stat. Challenge: **Un-Censoring** - requires **Call-by-Call Data**

Erlang-A: Practical Relevance?

Experience:

- ▶ Arrival process **not pure Poisson** (time-varying, σ^2 too large)
- ▶ Service times **not Exponential** (typically close to LogNormal)
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Question: **Is Erlang-A Practically Relevant?**

Answer, via **Fitting a Simple Model to a Complex Reality**

Erlang-A: Simple, but Not Too Simple

Natural Questions:

1. Fitting Erlang-A (with **O. Plonsky** and **S. Zeltyn**).
2. Why does it practically work? justify **robustness**.
3. When does it fail? chart **boundaries**.
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Motivation: Moderate-to-large service systems (**100's - 1000's** servers), notably **Call-Centers**.

Results turn out **accurate** enough to also cover **<10** servers:

- ▶ Practically Important: Relevant to **Healthcare** (F. de Véricourt and O. Jennings; with **G. Yom-Tov**; with **Y. Marmor**, **S. Zeltyn**)
- ▶ Theoretically Justifiable: Gap-Analysis by **B. Zhang**, **J. van Leeuwen**, **B. Zwart**.

Operational Regimes: Conceptual Framework

R: Offered Load

Def. **R** = Arrival-rate $\times$ Average-Service-Time = $\frac{\lambda}{\mu}$

eg. **R** = 25 calls/min. $\times$ 4 min./call = **100**

N = #Agents **?**

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- ▶ Framework developed in **O. Garnett's** MSc thesis
- ▶ Rigorously: $(N - R)/R \rightarrow \delta$, as **N**, $\lambda \uparrow \infty$, with μ fixed.
- ▶ Performance: Delays are rare events

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QED Regime: $N \approx R + \beta\sqrt{R}$, $-1 < \beta < +1$ (eg. $N = 100$)

- ▶ Erlang 1913-24, **Halfin & Whitt** 1981 (for Erlang-C)
- ▶ %Delayed between 25% and 75%
- ▶ $E[\text{Wait}] \propto \frac{1}{\sqrt{N}} \times E[\text{Service}]$ (**sec vs. min**); 1-5% Abandon.

Operational Regimes: Rules-of-Thumb, with S. Zeltyn

Constraint	P{Ab}		E[W]		P{W > T}	
Offered Load	Tight	Loose	Tight	Loose	Tight	Loose
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WFM: How to determine specific staffing level N ? e.g. β .

QED Theory (Erlang '13; Halfin-Whitt '81; Garnett MSc; Zeltyn PhD)

Consider a sequence of **steady-state** M/M/**N** + G queues, **N** = 1, 2, 3, ...
Then the following points of view are **equivalent**, as $N \uparrow \infty$:

- **QED** $\% \{\text{Wait} > 0\} \approx \alpha, \quad 0 < \alpha < 1 ;$
- **Customers** $\% \{\text{Abandon}\} \approx \frac{\gamma}{\sqrt{N}}, \quad 0 < \gamma ;$
- **Agents** $\text{OCC} \approx 1 - \frac{\beta + \gamma}{\sqrt{N}} \quad -\infty < \beta < \infty ;$
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- ▶ **QED performance**: Laplace Method (asymptotics of integrals).
- ▶ **Parameters**: Arrivals and Staffing - β , Services - μ ,
(Im)Patience - $g(0)$ = **patience density at the origin**.

QED Approximations: Some Examples

G – patience distribution,

g_0 – patience density at origin ($g_0 = \theta$, if $\exp(\theta)$).

$$N = \frac{\lambda}{\mu} + \beta \sqrt{\frac{\lambda}{\mu}} + o(\sqrt{\lambda}), \quad -\infty < \beta < \infty.$$

$$P\{\text{Ab}\} \approx \frac{1}{\sqrt{N}} \cdot [h(\hat{\beta}) - \hat{\beta}] \cdot \left[\sqrt{\frac{\mu}{g_0}} + \frac{h(\hat{\beta})}{h(-\beta)} \right]^{-1},$$

$$P\left\{W > \frac{T}{\sqrt{N}}\right\} \approx \left[1 + \sqrt{\frac{g_0}{\mu}} \cdot \frac{h(\hat{\beta})}{h(-\beta)} \right]^{-1} \cdot \frac{\bar{\Phi}(\hat{\beta} + \sqrt{g_0 \mu} \cdot T)}{\bar{\Phi}(\hat{\beta})},$$

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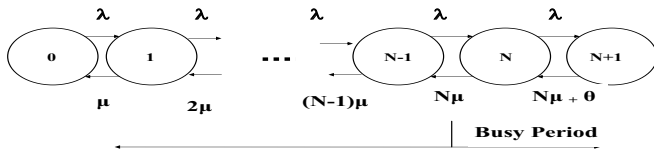
Here

$$\hat{\beta} = \beta \sqrt{\frac{\mu}{g_0}}$$

$$\bar{\Phi}(x) = 1 - \Phi(x),$$

$$h(x) = \phi(x)/\bar{\Phi}(x), \text{ hazard rate of } N(0, 1).$$

QED Intuition via Excursions: Busy-Idle Cycles



$Q(0) = N$: all servers busy, no queue.

Let $T_{N,N-1} = E[\text{Busy Period}]$ down-crossing $N \downarrow N-1$

$T_{N-1,N} = E[\text{Idle Period}]$ up-crossing $N-1 \uparrow N$

$$\text{Then } P(\text{Wait} > 0) = \frac{T_{N,N-1}}{T_{N,N-1} + T_{N-1,N}} = \left[1 + \frac{T_{N-1,N}}{T_{N,N-1}} \right]^{-1}.$$

QED Intuition via Excursions: Asymptotics

Calculate $T_{N-1,N} = \frac{1}{\lambda_N E_{1,N-1}} \sim \frac{1}{N\mu \times h(-\beta)/\sqrt{N}} \sim \frac{1}{\sqrt{N}} \cdot \frac{1/\mu}{h(-\beta)}$

$$T_{N,N-1} = \frac{1}{N\mu\pi_+(0)} \sim \frac{1}{\sqrt{N}} \cdot \frac{\beta/\mu}{h(\delta)/\delta}, \quad \delta = \beta\sqrt{\mu/\theta}$$

Both apply as $\sqrt{N}(1 - \rho_N) \rightarrow \beta, \quad -\infty < \beta < \infty.$

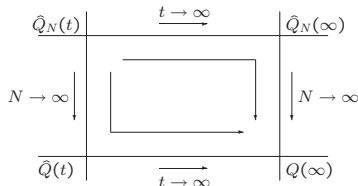
Hence, $P(Wait > 0) \sim \left[1 + \frac{h(\delta)/\delta}{h(-\beta)/\beta}\right]^{-1}.$

Process Limits (Queueing, Waiting)

- $\hat{Q}_N = \{\hat{Q}_N(t), t \geq 0\}$: **stochastic process** obtained by centering and rescaling:

$$\hat{Q}_N = \frac{Q_N - N}{\sqrt{N}}$$

- $\hat{Q}_N(\infty)$: stationary distribution of $\hat{Q}_N$
- $\hat{Q} = \{\hat{Q}(t), t \geq 0\}$: process defined by: $\hat{Q}_N(t) \xrightarrow{d} \hat{Q}(t)$.



Approximating (Virtual) **Waiting Time**

$$\hat{V}_N = \sqrt{N} V_N \Rightarrow \hat{V} = \left[\frac{1}{\mu} \hat{Q} \right]^+$$

Back to “Why does Erlang-A Work?”

Theoretical Answer:

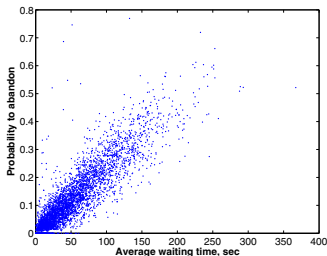
$$M_t^{?,J} / G / N_t + G \stackrel{d}{\approx} (M / M / N + M)_t, \quad t \geq 0.$$

- ▶ **General Patience**: Behavior at the origin is all that matters.
- ▶ **General Services**: Empirical insensitivity beyond the mean.
- ▶ **Time-Varying Arrivals**: Modified Offered-Load approximations.
- ▶ **Over-Dispersed Arrivals**: c -Staffing ($c > 1/2$).
- ▶ **Heterogeneous Customers**: 1-D state-collapse.

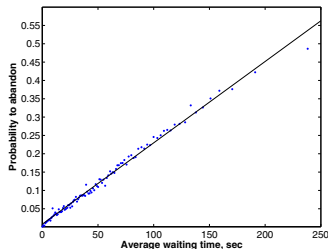
“Why does Erlang-A Work?” General Patience

Israeli Bank: Yearly Data

Hourly Data



Aggregated



Theory:

Erlang-A: $P\{Ab\} = \theta \cdot E[W_q]$;

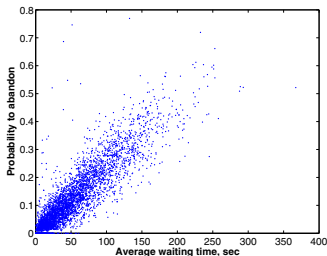
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$g(0)$ = Patience-density at origin

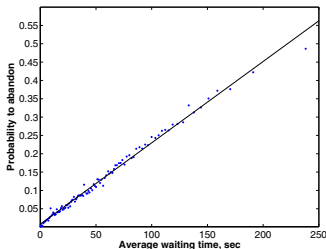
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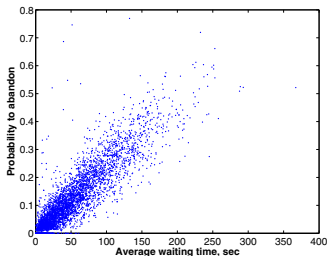
Recipe:

In both cases, use Erlang-A, with $\hat{\theta} = \widehat{P\{Ab\}} / \widehat{E[W_q]}$ (slope above).

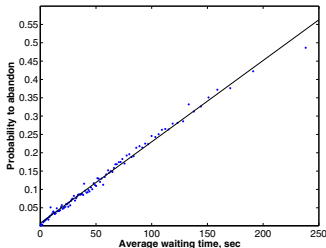
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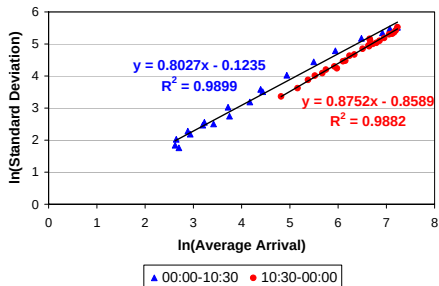
References on $g(0)$:

- Stationary M/M/N+GI, with **S. Zeltyn**
- Process G/GI/N+GI, with **P. Momcilovic**

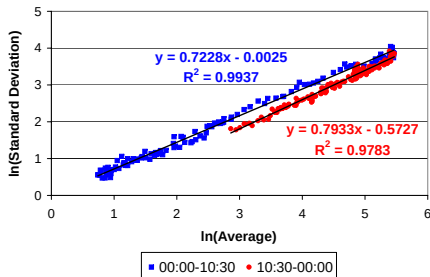
“Why does Erlang-A Work?” Over-Dispersion

$\ln(\text{STD})$ vs. $\ln(\text{AVG})$ (Israeli Bank, 4/2007-3/2008)

Tue-Wed, 30 min resolution



Tue-Wed, 5 min resolution



Significant linear relations (with **S. Aldor & P. Feigin**):

$$\ln(\text{STD}) = c \cdot \ln(\text{AVG}) + a$$

(Poisson: $\text{STD} = \text{AVG}^{1/2}$, hence $c = 1/2$, $a = 0$.)

Over-Dispersion: Random Arrival-Rates

Linear relation between $\ln(\text{STD})$ and $\ln(\text{AVG})$ gives rise to:

Poisson-Mixture (Doubly-Poisson, Cox) model for Arrivals:

Poisson(Λ) with **Random-Rate** of the form

$$\Lambda = \lambda + \lambda^c \cdot X, \quad c \leq 1;$$

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 $c = 1$, proportional to λ ; $c \leq 1/2$, Poisson-level;
 - In **Call Centers**: $c \approx 0.75 - 0.85$ (significant over-dispersion).
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QED-c Regime: Erlang-A, with Poisson(Λ) arrivals, amenable to asymptotic analysis (with **S. Maman & S. Zeltyn**)

Over-Dispersion: The QED-c Regime

QED-c Staffing: Under offered-load $R = \lambda \cdot E[S]$,

$$N = R + \beta \cdot R^c, \quad 0.5 < c < 1$$

Performance measures (M/M/N + G):

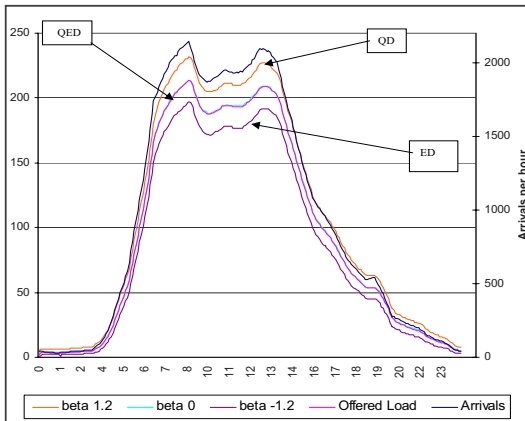
- Delay probability: $P\{W_q > 0\} \sim 1 - G(\beta)$
- Abandonment probability: $P\{Ab\} \sim \frac{E[X - \beta]_+}{n^{1-c}}$
- Average offered wait: $E[V] \sim \frac{E[X - \beta]_+}{n^{1-c} \cdot g_0}$
- Average actual wait: $E_{\Lambda, N}[W] \sim E_{\Lambda, N}[V]$

Why Does Erlang-A Work? Time-Varying Arrival Rates

Square-Root Staffing: $N_t = R_t + \beta\sqrt{R_t}$, $-\infty < \beta < \infty$

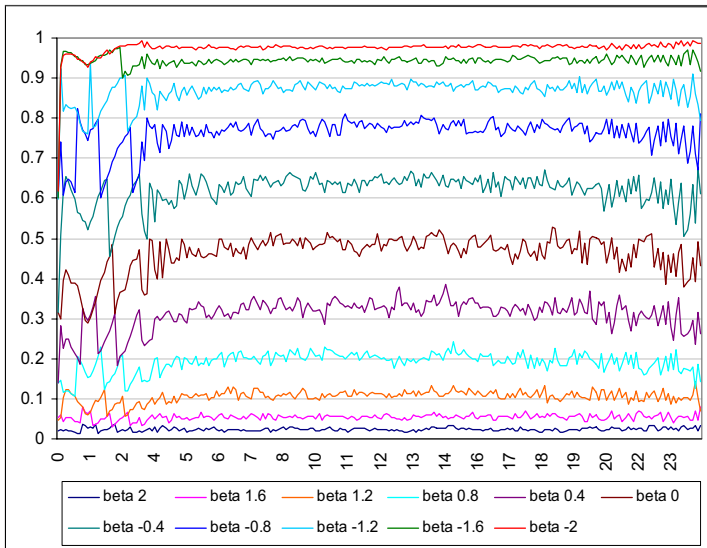
What is R_t , the **Offered-Load** at time t ? ($R_t \neq \lambda_t \times E[S]$)

Arrivals, Offered-Load and Staffing



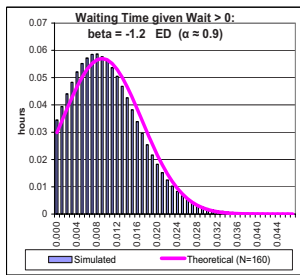
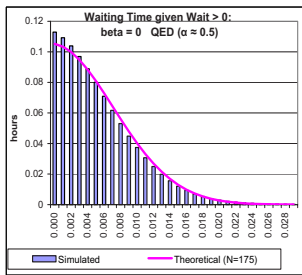
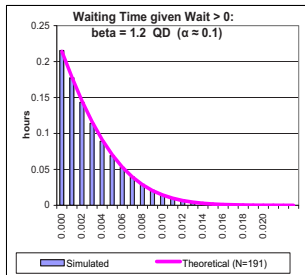
Time-Stable Performance of Time-Varying Systems

Delay Probability = As in the **Stationary Erlang-A** (Garnett)



Time-Stable Performance of Time-Varying Systems

Waiting Time, Given Waiting: Empirical vs. Theoretical Distribution



- **Empirical:** Simulate **time-varying** $M_t/M/N_t + M$ ($\lambda_t, N_t = R_t + \beta\sqrt{R_t}$)
- **Theoretical:** Naturally-corresponding **stationary** Erlang-A, with QED β -staffing
- **Generalizes** up to a station within a complex network (eg. Doctors in an Emergency Department).

What is the Offered-Load $R(t)$? Time-Varying Little

For $M_t/GI/N_t + GI$, the **Offered-Load function**, $\{R(t), t \geq 0\}$, is the **average number of customers (= busy servers)**, in a naturally corresponding $M_t/GI/\infty$ queue (MOL = Modified Offered Load).

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Four (all useful) representations, capturing “**work before t**”:

$$\begin{aligned} R(t) &= E[L(t)] = \int_{-\infty}^t \lambda(u) \cdot P(S > t - u) du = E[A(t) - A(t - S)] = \\ &= E\left[\int_{t-S}^t \lambda(u) du\right] = E[\lambda(t - S_e)] \cdot E[S]. \end{aligned}$$

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- **Stationary models:** $\lambda(t) \equiv \lambda$ then $R(t) \equiv \lambda/\mu$.
- **QED-c:** $N_t = R_t + \beta R_t^c$, $1/2 < c < 1$; ($c = 1$ separate analysis).

The Technion SEE Center / Laboratory

Data-Based Service Science / Engineering



Technion SEE = Service Enterprise Engineering

SEELab: Hub for data-based research and teaching

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SEEServer: Free for academic use

Register, then access (presently) U.S. Bank and Small Israeli Bank.