Service Engineering: Data-Based Science & Teaching in support of Service Management

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

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Based on joint work with Sergey Zeltyn
Technion Statistics / SEE Lab: Paul Feigin, Valery Trofimov, RA's

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- 5. Scientifically-based design principles and tools (software), that support the balance of service quality, process efficiency and business profitability, from the (often-conflicting) views of customers, servers, managers: Service Engineering.

Background Material (Downloadable)

► Technion's "Service-Engineering" Course (≥ 1995): http://ie.technion.ac.il/serveng

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- Gans (U.S.A.), Koole (Europe), and M. (Israel):
 "Telephone Call Centers: Tutorial, Review and Research Prospects." MSOM, 2003.
- Brown, Gans, M., Sakov, Shen, Zeltyn, Zhao: "Statistical Analysis of a Telephone Call Center: A Queueing-Science Perspective." JASA, 2005.
- Trofimov, Feigin, M., Ishay, Nadjharov:
 "DataMOCCA: Models for Call/Contact Center Analysis."
 Technion Report, 2004-2006.
- ► M. "Call Centers: Research Bibliography with Abstracts." Version 7, December 2006.



The First Prerequisite: Data & Measurements

Empirical "Axiom": The data one needs is **never** there for one to use – always problems with historical data (eg. lacking, contaminated, averaged, . . .)

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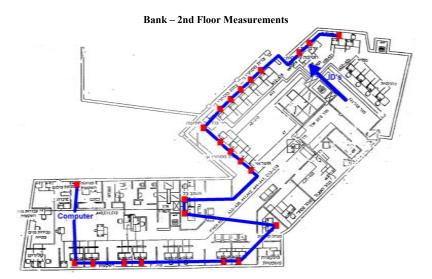
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Sources: "Service-floor" (vs. Industry-level, Surveys, ...)

- Administrative (Court, via "paper analysis")
- ► Face-to-Face (Bank, via bar-code readers)
- ► Telephone (Call Centers, via ACD)
- Future: Hospitals (via RFID)

Measurements: Face-to-Face Services

23 Bar-Code Readers at a Bank Branch



Measurements: Telephone Call-by-Call Data (Log-File)

vru+line	coll id	customer id	priority	type	date	YOU CORY	vru exit	vru time	o start	a exit	a time	outcome	ser start	ser exit	ser time	server
AA0101	44749	27644400	,	PS	990901	11:45:33	11 45 39	6	11 45 39	11:46:58	79	AGENT	11 46 57	11:51:00	243	DORIT
A A0101	44750	12887816	-	PS	990905	14 49 00	14 49 06	-		14.53.00	234	AGENT	14 52 59	14:54:29	90	ROTH
AA0101	44967	58660291	2	PS	990905	14:58:42	14:58:48	6	14.49.06	15:02:31	223	AGENT	15.02.31		99	ROTH
AA0101	44968	0	0	NW	990905	15:10:17	15:10:26	9	15:10:26	15:13:19	173	HANG	00.00.00	00:00:00	0	NO SERVER
AA0101	44969	63193346	2	PS	990905	15:22:07	15:22:13	6	15:22:13	15:23:21	68	AGENT	15:23:20	15:25:25	125	STEREN
AA0101	44970	0	0	NW	990905	15:31:33	15:31:47	14	00:00:00	00.00.00	0	AGENT	15:31:45	15:34:16	151	STEREN
AA0101	44971	41630443	2	PS	990905	15:37:29	15:37:34	5	15:37:34	15:38:20	46	AGENT	15:38:18	15:40:56	158	TOVA
AA0101	44972	64185333	2	PS	990905	15:44:32	15:44:37	5	15:44:37	15:47:57	200	AGENT	15:47:56	15:49:02	66	TOVA
AA0101	44973	3.06E+08		PS	990905	15:53:05	15:53:11	6	15:53:11	15:56:39	208	AGENT	15:56:38	15:56:47	9	MORIAH
AA0101	44974	74780917	2	NE	990905	15:59:34	15:59:40	6	15:59:40	16:02:33	173	AGENT	16:02:33	16:26:04	1411	ELI
A A0101	44975	55979755	,	PS	SOURCES	1647.46	16-07-51	5	16:07:51	16:08:01	10	HANG	00 00 00	00:00:00	0	NO SERVER
AA0101	44976	0	0	NW	990905	16:11:38	16:11:48	10	16:11:48	16:11:50	2	HANG	00 00 00	00:00:00	0	NO SERVER
AA0101	44977	33689787	,	PS	990905	16:14:27	16:14:33	6	16:14:33	16:14:54	21	HANG	00.00.00		0	NO SERVER
A A0101	44978	23817067	,	PS	990905	16 19 11	16:19:17	6	16 19 17	16:19:39	22	MGENT	16:19:38	16:21:57	139	TOVA
A A0101	44764	0	0	PS	990901	15/03/26	15.03.36	10	00.00.00	00 00 00	0	AGENT	15 03 35	150636	181	ZOHARI
AA0101		25219700	,	PS	990901	15 14 46	15:14:51	5		15-15-10	19	MIENT	15:15:09		111	SHARON
A A0101	44766	0	0	PS	990901	15.25.48	15:26:00	12	00.00.00	00.00.00	0	AGENT	15.25.59	15:28:15	136	ANAT
AA0101	44767	58859752	2	PS	990901	15:34:57	15:35:03	6	15:35:03	15:35:14	11	AGENT	15:35:13	15:35:15	2	MORIAH
AA0101	44768	0	0	PS	990901	15:46:30	15:46:39	9	00.00.00	00.00.00	0	AGENT	15:46:38	15:51:51	313	ANAT
AA0101	44769	78191137	2	PS	990901	15:56:03	15:56:09	6	15:56:09	15:56:28	19	AGENT	15:56:28	15:59:02	154	MORIAH
AA0101	44770	0	0	PS	990901	16:14:31	16:14:46	15	00.00.00	00.00.00	0	AGENT	16:14:44	16:16:02	78	BENSION
AA0101	44771	0	0	PS	990901	16:38:59	16:39:12	13	00.00.00	00.00.00	0	AGENT	16:39:11	16:43:35	264	VICKY
AA0101	44772	0	0	PS	990901	16:51:40	16:51:50	10	00.00.00	00.00.00	0	AGENT	16:51:49	16:53:52	123	ANAT
AA0101	44773	0	0	PS	990901	17:02:19	17:02:28	9	00.00.00	00.00.00	0	AGENT	17:02:28	17:07:42	314	VICKY
AA0101	44774	32387482	1	PS	990901	17:18:18	17:18:24	6	17:18:24	17:19:01	37	AGENT	17:19:00	17:19:35	35	VICKY
AA0101	44775	0	0	PS	990901	17:38:53	17:39:05	12	00.00.00	00.00.00	0	AGENT	17:39:04	17:40:43	99	TOVA
AA0101	44776	0	0	PS	990901	17:52:59	17:53:09	10	00.00.00	00.00.00	0	AGENT	17:53:08	17:53:09		NO SERVER
AA0101	44777	37635950	2	PS	990901	18:15:47	18:15:52	5	18:15:52	18:16:57	65	AGENT	18:16:56	18:18:48	112	ANAT
AA0101	44778	0	0	NE	990901	18:30:43	18:30:52	9	00:00:00	00.00.00	0	AGENT	18:30:51	18:30:54	3	MORIAH
AA0101	44779	0	0	PS	990901	18:51:47	18:52:02	15	00.00.00	00.00.00	0	AGENT	18:52:02	18:55:30	208	TOVA
AA0101	44780	0	0	PS	990901	19:19:04	19:19:17	13	00.00.00	00.00.00	0	AGENT	19:19:15	19:20:20	65	MEIR
AA0101	44781	0	0	PS	990901	19.39.19	19:39:30	11	00.00.00	00.00.00	0	AGENT	19.39.29	19:41:42	133	BENSION
AA0101	44782	0	0	NW	990901	20:08:13	20:08:25	12	00.00.00	00.00.00	0	AGENT	20.08:28	20:08:41	13	NO_SERVER
AA0101	44783	0	0	PS	990901	20:23:51	20:24:05	14	00.00.00	00.00.00	0	AGENT	20:24:04	20:24:33	29	BENSION
AA0101	44784	o .	0	NW	990901	20:36:54	20:37:14	20	00.00.00	00.00.00	0	AGENT	20:37:13	20:38:07	54	BENSION
AA0101	44785	0	0	PS	990901	20:50:07	20:50:16	9	00:00:00	00.00.00	0	AGENT	20:50:15	20:51:32	77	BENSION
AA0101	44786	0	0	PS	990901	21:04:41	21:04:51	10	00:00:00	00.00.00	0	AGENT	21:04:50	21:05:59	69	TOVA
AA0101	44787	0	0	PS	990901	21:25:00	21:25:13	13	00:00:00	00.00.00	0	AGENT	21:25:13	21:28:03	170	AVI
AA0101	44788	0	0	PS	990901	21:50:40	21:50:54	14	00:00:00	00.00.00	0	AGENT	21.50.54	21:51:55	61	AVI
AA0101	44789	9103060	2	NE	990901	22:05:40	22:05:46	6	22:05:46	22:09:52	246	AGENT	22.09.51	22:13:41	230	AVI
AA0101	44790	14558621	2	PS	990901	22:24:11	22:24:17	6	22:24:17	22:26:16	119	AGENT	22:26:15	22:27:28	73	VICKY
AA0101	44791	0	0	PS	990901	22.46:27	22:46:37	10	00.00.00	00.00.00	0	AGENT	22.46:36	22:47:03	27	AVI
AA0101	44792	67158097	2	PS	990901	23:05:07	23:05:13	6	23:05:13	23:05:30	17	AGENT	23:05:29	23:06:49	80	VICKY
AA0101	44793	15317126	2	PS	990901	23:28:52	23:28:58	6	23:28:58	23:30:08	70	AGENT	23:30:07	23:35:03	296	DARMON
AA0101	44794	0	0	PS	990902	00:10:47	00:12:05	78	00.00.00	00.00.00	0	HANG	00.00.00	00:00:00	0	NO SERVER
AA0101	44795	0	0	PS	990902	07:16:52	07:17:01	9	00.00.00	00.00.00	0	AGENT	07:17:01	07:17:44	43	ANAT
AA0101	44796	0	0	PS	990902	07:50:05	07:50:16	11	00:00:00	00.00.00	0	AGENT	07:50:16	07:53:03	167	STEREN
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Averages Prevalent

ACD Report: Health Insurance

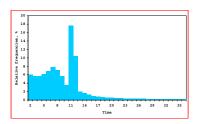
Time	Calls	Answered	Abandoned%	ASA	AHT	Occ%	# of agents
Total	20,577	19,860	3.5%	30	307	95.1%	
8:00	332	308	7.2%	27	302	87.1%	59.3
8:30	653	615	5.8%	58	293	96.1%	104.1
9:00	866	796	8.1%	63	308	97.1%	140.4
9:30	1,152	1,138	1.2%	28	303	90.8%	211.1
10:00	1,330	1,286	3.3%	22	307	98.4%	223.1
10:30	1,364	1,338	1.9%	33	296	99.0%	222.5
11:00	1,380	1,280	7.2%	34	306	98.2%	222.0
11:30	1,272	1,247	2.0%	44	298	94.6%	218.0
12:00	1,179	1,177	0.2%	1	306	91.6%	218.3
12:30	1,174	1,160	1.2%	10	302	95.5%	203.8
13:00	1,018	999	1.9%	9	314	95.4%	182.9
13:30	1,061	961	9.4%	67	306	100.0%	163.4
14:00	1,173	1,082	7.8%	78	313	99.5%	188.9
14:30	1,212	1,179	2.7%	23	304	96.6%	206.1
15:00	1,137	1,122	1.3%	15	320	96.9%	205.8
15:30	1,169	1,137	2.7%	17	311	97.1%	202.2
16:00	1,107	1,059	4.3%	46	315	99.2%	187.1
16:30	914	892	2.4%	22	307	95.2%	160.0
17:00	615	615	0.0%	2	328	83.0%	135.0
17:30	420	420	0.0%	0	328	73.8%	103.5
18:00	49	49	0.0%	14	180	84.2%	5.8

Beyond Averages: Waiting Times in a Call Center

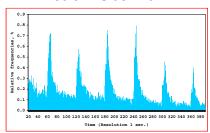
Small Israeli Bank

201 % Main = 58 80 = 105 20 % 124 % 3.0 % 3.1 % 2.3 % 1.7 %

Large U.S. Bank



Medium Israeli Bank



The Second Prerequisite: Models

Through **Examples** Only.

Each example starts with a **Complex Reality** and ends with a **useful** insight due to a **Simple Model**.

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"Theorem": A useful model must be simple (yet not too simple).

Models in decreasing simplicity-levels:

- Conceptual: Service Networks = Queueing Networks
- Descriptive: Averages, Histograms
- Explanatory: Comparative, Regression
- Analytical/Mathematical: Little's Law, Fluid Models, Queueing Models, Diffusion Refinements.

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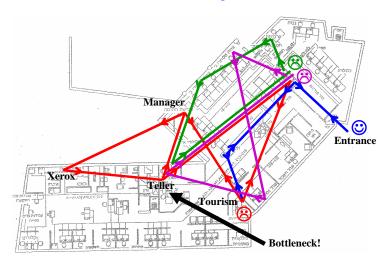
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"Corollary": To be useful, a simple model sometimes requires deep analysis.

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Conceptual Model: Face-to-Face Services

Bank Branch = Queueing Network



Descriptive Model: Transition Probabilities (Averages)

Transition Frequencies Between Units in The Private and Business Sections:

		Private Banking				Business				
	To Unit From Unit	Bankers	Authorized Personal	Compens - - ations	Tellers	Tellers	Overdrafts	Authorized Personal	Full Service	Exit
	Bankers		1%	1%	4%	4%	0%	0%	0%	90%
Private	Authorized Personal	12%		5%	4%	6%	0%	0%	0%	73%
Banking	Compensations	7%	4%		18%	6%	0%	0%	1%	64%
	Tellers	6%	0%	1%		1%	0%	0%	0%	90%
	Tellers	1%	0%	0%	0%		1%	0%	2%	94%
Services	Overdrafts	2%	0%	1%	1%	19%		5%	8%	64%
	Authorized Personal	2%	1%	0%	1%	11%	5%		11%	69%
	Full Service	1%	0%	0%	0%	8%	1%	2%		88%
	Entrance	13%	0%	3%	10%	58%	2%	0%	14%	0%

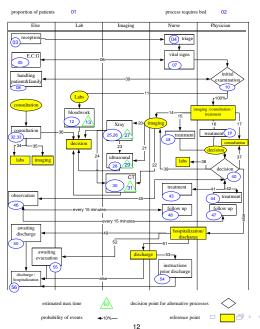
Legend:

0%-5% 5%-10% 10%-15% >15%

Dominant Paths - Business:

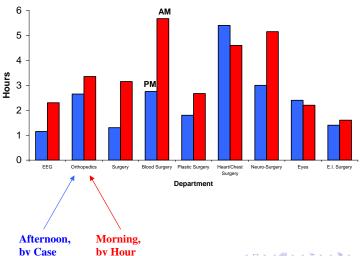
Unit Parameter	Station 1 Tourism	Station 2 Teller	Total Dominant Path		
Service Time	12.7	4.8	17.5		
Waiting Time	8.2	6.9	15.1		
Total Time	20.9	11.7	32.6		
Service Index	0.61	0.41	0.53		

Conceptual Model: Hospital (ED) Network (Sinreich)



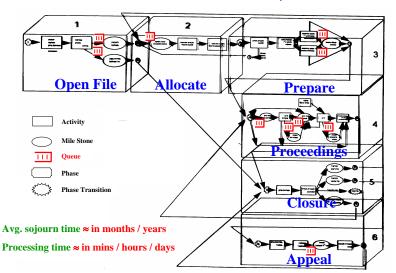
Descriptive Model: Service Times (Averages) or, Even "Doctors" Can Manage

Operations Time - Morning (by Hour) vs. Afternoon (by Case):



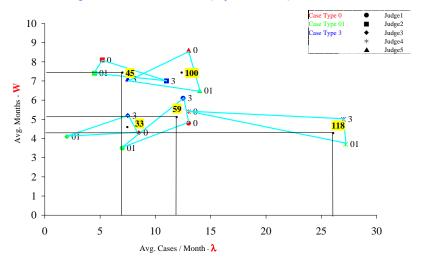
Conceptual Model: The "Production of Justice"

The Labor-Court Process in Haifa, Israel

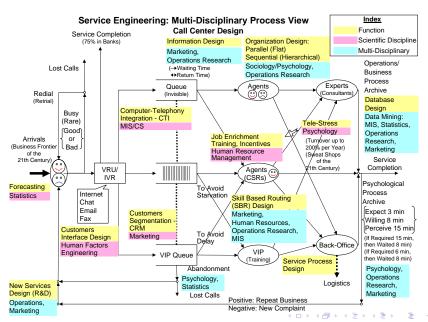


Analytical Model: Little's Law in Court (still Averages)

Judges: The Best/Worst (Operational) Performer



Call-Center Network: Gallery of Models



The "Phases of Waiting" for Service

Common Experience:

- Expected to wait 5 minutes, Required to 10
- ► Felt like 20, Actually waited 10 (hence Willing ≥ 10)

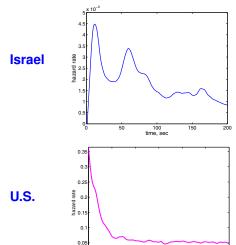
An attempt at "Modeling the Experience":

```
Experienced customers ⇒ Expected = Required  
"Rational" customers ⇒ Perceived = Actual.
```

Then left with (τ, V) .

Call Center Data: Hazard Rates (Un-Censored)





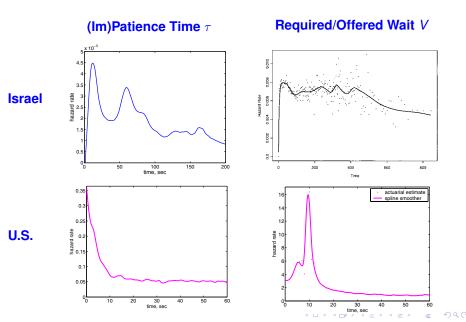
10



50

time, sec

Call Center Data: Hazard Rates (Un-Censored)



A Patience Index

Quantifying (Im)Patience: "Willing to wait 15 min" = Patient?

Theoretical Patience-Index
$$\stackrel{\triangle}{=} \frac{\text{Willing to wait}}{\text{Expected to wait}} = \frac{\text{E}[\tau]}{\text{E}[V]}$$

"assuming" Experienced;



19

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"assuming" Experienced; further "assuming" that τ and V are Exponential, the M-L estimate of Index is the easily-measurable:

Empirical Patience-Index
$$\stackrel{\triangle}{=} \frac{\% \text{ Served}}{\% \text{ Abandoning}}$$



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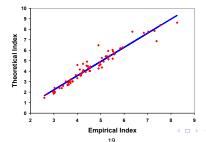
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Index Validation: Theoretical vs. Empirical



Predicting Performance

Model Primitives:

- Arrivals to service (random process)
- (Im)Patience while waiting τ (r.v.)
- Service times (r.v.)
- # Servers / Agents (parameter / r.v.)

Model Output: Offered-Wait V (r.v.)

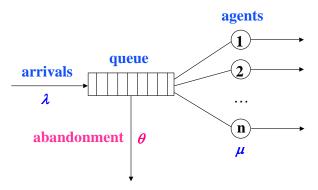
Operational Performance Measure calculable in terms of (τ, V) .

- ▶ eg. Average Wait = E[min{\tau, V}]
- eg. % Abandonment = $P\{\tau < V\}$

Application: Staffing - How Many Agents? (When? Who?)



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



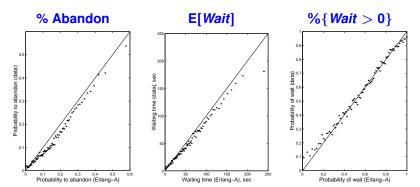
Erlang-A Parameters:

- λ **Arrival** rate (Poisson)
- μ **Service** rate (Exponential)
- \bullet θ Impatience rate (Exponential)
- ▶ n Number of Service-Agents.



Erlang-A: Fitting a Simple Model to a Complex Reality

Hourly Performance vs. Erlang-A Predictions (1 year)



- ► Empirically-Based & Theoretically-Supported Estimation of (Im)Patience: $\hat{\theta} = P\{Ab\}/E[W_a]$)
- Small Israeli Bank (more examples in progress)



Testing the Erlang-A Primitives

Arrivals: Poisson?

Service-durations: Exponential?

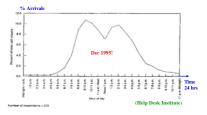
(Im)Patience: Exponential?

Validation: Support? Refute?

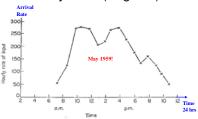
Arrivals to Service: only Poisson-Relatives

Arrival Rate to Three Call Centers

Dec. 1995 (U.S. 700 Helpdesks)



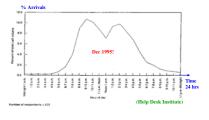
May 1959 (England)



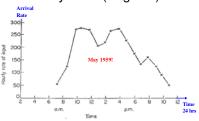
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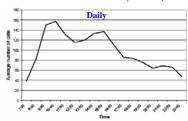
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May 1959 (England)



November 1999 (Israel)

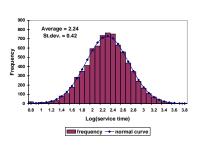


Observation:

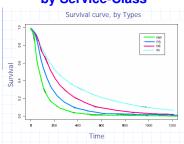
Peak Loads at 10:00 & 15:00

Service Durations: LogNormal Prevalent

Israeli Bank Log-Histogram



Survival-Functions by Service-Class



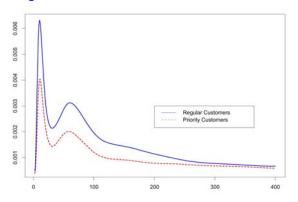
- New Customers: 2 min (NW);
- ► Regulars: 3 min (PS);

- ► Stock: 4.5 min (NE);
- Tech-Support: 6.5 min (IN).

Observation: VIP require longer service times.

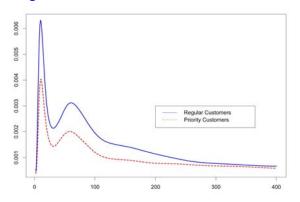
(Im)Patience while Waiting (Palm 1943-53)

Irritation ∝ Hazard Rate of (Im)Patience Distribution Regular over VIP Customers – Israeli Bank



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Irritation Hazard Rate of (Im)Patience Distribution Regular over VIP Customers − Israeli Bank



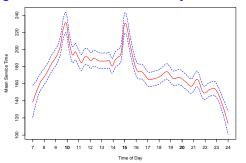
- Peaks of abandonment at times of Announcements
- ► Call-by-Call Data (DataMOCCA) required (& Un-Censoring).

Observation: VIP are more patient (Needy)



A "Service-Time" Puzzle at a Small Israeli Bank Inter-related Primitives

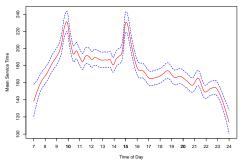
Average Service Time over the Day – Israeli Bank



Prevalent: Longest services at peak-loads (10:00, 15:00). Why?

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Average Service Time over the Day – Israeli Bank



Prevalent: Longest services at peak-loads (10:00, 15:00). Why? Explanations:

- Prevalent: Service protocol different (longer) during peak times.
- Operational: The needy abandon less during peak times; hence the VIP remain on line, with their longer service times.

Erlang-A: Simple, but Not Too Simple

Experience:

- ▶ Arrival process **not pure Poisson** (time-varying, σ^2 too large)
- Service times not exponential (typically close to lognormal)
- ▶ Patience times **not exponential** (various patterns observed).
- Customers and Servers not homogeneous (classes, skills)

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Questions naturally arise:

- 1. Why does Erlang-A practically work? justify robustness.
- 2. When does it fail? chart boundaries.
- 3. Generalize: time-variation, SBR, networks, uncertainty, ...

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Answers via **Asymptotic Analysis**, as load- and staffing-levels ↑:

- ► Efficiency-Driven (ED) regime: Fluid models (Whitt; Harrison, Zeevi; Bassamboo).
- Quality- and Efficiency-Driven (QED) regime: Diffusion refinements (Erlang, 1913; Halfin-Whitt, 1981; present surge).



- ► **Technion**: P. Feigin, V. Trofimov, Statistics / SEE Laboratory.
- ▶ Wharton: L. Brown, N. Gans, H. Shen (UNC).
- ▶ industry:
 - U.S. Bank: 2.5 years, 220M calls, 40M by 1000 agents.
 - Israeli Cellular: 2.5 years, 110M calls, 25M calls by 750 agents; ongoing.

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- ► Clean Databases: operational-data of individual calls / agents.
- ► **Graphical Online Interface**: easily generates graphs and tables, at varying resolutions (seconds, minutes, hours, days, months).

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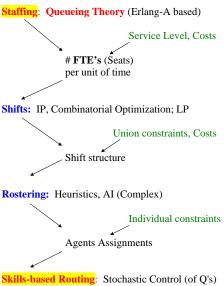
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Free for academic adoption: 7GB tables, or 20GB raw zipped, for each call center – ask for my mini-HD.

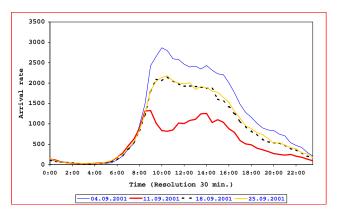


Call Centers: Hierarchical Operational View

Forecasting Customers (Statistics), Agents (HRM)



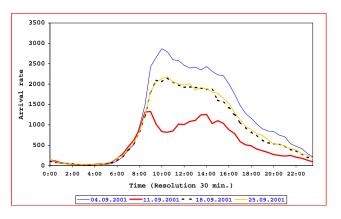
Arrivals to Service: Predictable vs. Random Arrival Rates on Tuesdays in a September – U.S. Bank



- ► Tuesday, September 4th: Heavy, following Labor Day.
- ► Tuesdays, September 18 & 25: Normal.



Arrivals to Service: Predictable vs. Random Arrival Rates on Tuesdays in a September – U.S. Bank



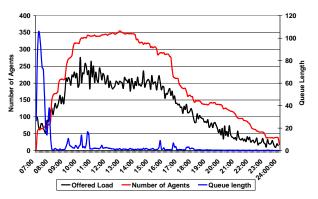
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- ► Tuesday, September 11th, 2001.



When: Shift Scheduling

Integer Programming, given piecewise-constant Staffing Levels.

U.S. Bank: Queue-length and Staffing on May 3, 2002



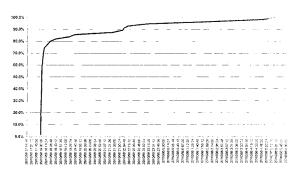
Shift scheduling matters: consistent under-staffing at 7:00am.

Who: Rostering

Assigning individual agents to shifts.

Typically, heuristics (AI) to accommodate individual constraints.

Israeli Technical Support Call Center: Online Shift Bidding

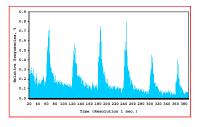


Shift-bidding starts at 18:00.

- ▶ 60% of "successful" agents are registered till 18:00.
- ▶ 80% till 18:24; 90% till 22:00; registration closed at 5:23am.



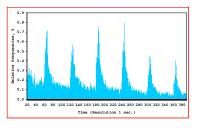
A "Waiting-Times" Puzzle at a Medium Israeli Bank



Peaks Every 60 Seconds. Why?

- ► Human: Voice-announcement every 60 seconds.
- System: Priority-upgrade (unrevealed) every 60 seconds.

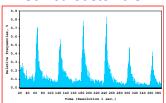
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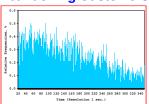
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Served Customers

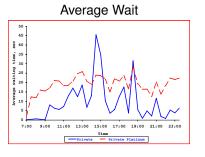


Abandoning Customers



Priorities, Economies-of-Scale, SBR

Regular vs. VIP Customers: Cellular – March 23, 2004

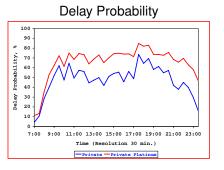


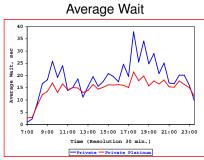


- Design: VIP-dedicated agents, Regular-dedicated Agents.
- VIP's are not served better than Regular's
- ▶ **Solutions:** Add VIP agents (costly), or Re-Design.

Priorities and Routing Protocols I

Regular vs. VIP Customers: Cellular – October 2004





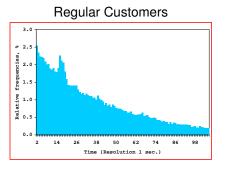
More VIPs delayed than Regulars, yet their average wait is shorter.

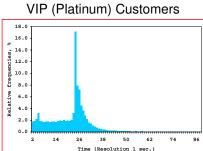
What changed since last March?



Priorities and Routing Protocols II

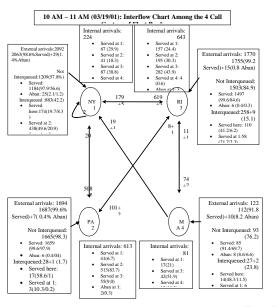
Waiting-Time Histograms: Cellular – October 2004





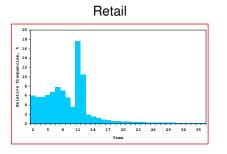
After **25 seconds** of wait, **VIP** customers are **routed** with **high priority** to Regular agents. Hence, almost **no long waiting times** for VIP's.

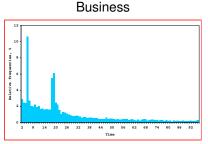
Network Balancing via Inter-Queues at a U.S. Bank



Balancing Protocols and Performance Level

U.S. Bank: Histograms of Waiting Times





Peak for **Retail** service at **10 seconds** – **Why?**After 10 seconds of wait, **Retail** customers sent into the **inter-queue**.

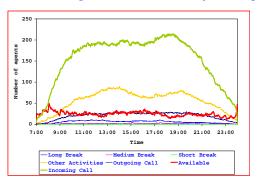
Business customers – peak at **5 seconds**, for the same reason. Second peak – unclear, maybe priority-upgrade.

The Planning-Reality Gap: Agent Status

Erlang-A Model \Rightarrow optimal **Staffing Level** n.

n = number-of-agents that show up? serve? No!

Israeli Bank, Agent Status: Monthly Averages



n (FTE) = Busy with "Incoming Calls" + "Available" for service.



Data-Based Service-Research (with DataMOCCA, even before tenure)

- Contrast with EmpOM: Industry / Company / Survey Data (Social Sciences)
- ► Converge to: Measure, Model, Validate, Experiment, Refine (**Physics, Biology,** ...).

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- Outcomes: Relevance, Credibility; Interest, Fun;
 Call Centers as a Pilot (eg. for Healthcare). Moreover,
 - ► Teaching: Class, Homework (Experimental Data Analysis); Cases.
 - Research: Validate Existing (Queueing) Theory/Laws and Suggest New Models/Research.
 - Practice: OM Tools (Scenario Analysis), Mktg. (Trends, Benchmarking).



Live Demonstration of DataMOCCA

5-7 minutes, to emphasize "online" capabilities.

U.S. Bank

- ▶ Daily Reports: October 2003, weekdays; typically takes 10-20 sec till a first output, but this is because of PowerPoint/Windows. Then do few additional Daily Reports, say Monday, Tuesday,... (starting with STATCCA, as opposed to by minimizing the powerpoint screnn) this will be now happening very fast.
- Time-Series: Number of agents, for ALL classes, all months, weekdays. (Including total). Shows scale, trends. Then do Service Durations, indicating that 1 second of 1000 agents could cost \$500M per year. Could also do Unhandled (lower middle entry in list), for only Retail and Premium Premium is worse, and deteriorating,
- Daily Summaries:
 - Tuesdays in September 2001; September 11th; shown during lecture under the heading "Predictable or Random";
 - ▶ 30 sec scale stoch. variability, 1 hour scale the "right" scale;
 - ▶ % to mean, to show very similar shape over 3 Tuesdays. Suggests the model $\lambda(t) = \lambda_0(t) \cdot Z$, for $0 \le t \le T$.