

Data-Based Service Networks:

A Research Framework for Asymptotic Inference, Analysis & Control of Service Systems

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

SAMSI Workshop, August 2012

- **Overheads available at SAMSI and my Technion websites**

Research Partners

▶ **Students:**

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

▶ **Theory:**

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

▶ **Empirical/Statistical Analysis:**

Brown, Gans, Shen, Zhao; Zeltyn; Ritov, Goldberg; Gurvich, Huang, Liberman; Armony, Marmor, Tseytlin, Yom-Tov; Nardi, Plonsky; Gorfine, Ghebali; Pang, ...

▶ **Industry:**

Mizrahi Bank (A. Cohen, U. Yonissi), Rambam Hospital (R. Beyar, S. Israelit, S. Tzafrir), IBM Research (OCR Project), Hapoalim Bank (G. Maklef, T. Shlasky), Pelephone Cellular, ...

▶ **Technion SEE Center / Laboratory:**

Feigin; Trofimov, Nadjharov, Gavako, Kutsy; Liberman, Koren, Plonsky, Senderovic; Research Assistants, ...

Contents

- ▶ Service Networks: Call Centers, Hospitals, Websites, . . .
- ▶ Redefine the paradigm of modeling/asymptotics via Data
- ▶ **ServNets: QNets, SimNets; FNets, DNets**
- ▶ **Ultimate Goal: Data-based creation and validation of ServNets, automatically in real-time**

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- ▶ **Ultimate Goal: Data-based creation and validation of ServNets, automatically in real-time**
- ▶ Why be Optimistic? Pilot at the Technion SEELab
 - ▶ Lacking but Feasible: Dynamics, Durations, Protocols
 - ▶ Simple Models at the Service of Complex Realities
 - ▶ State-Space Collapse (Queues, Waiting Times)
 - ▶ Congestion Laws (LN, Little, ImPatience, Staffing)
 - ▶ Universal Approximations: Simplifying the Asymptotic Landscape
 - ▶ Stabilizing Time-Varying Performance (Offered-Load)
- ▶ Successes: Palm/Erlang-R (ED Feedback = **FNet**),
Palm/Erlang-A (CC Abandonment = **DNet**)

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- ▶ **Elsewhere** : Process Mining (Petri Nets, BPM), Networks (Social, Biological, Complex, . . .), Simulation-based, . . .
- ▶ **Scenic Route** : Open Problems, New Directions, Uncharted Territories

On Asymptotic Research of Queueing Systems

Queueing asymptotics has grown to become a central research theme in Operations Research and Applied Probability, beyond just queueing theory. Its claim to fame has been the deep insights that it provides into the dynamics of Queueing Networks (**Q**Nets), and rightly so:

- ▶ Kingman's invariance principle in conventional heavy-traffic
- ▶ Whitt's sample-path (functional) framework
- ▶ Reiman's network analysis via oblique reflection
- ▶ Bramson-Williams' framework for state-space collapse
- ▶ Laws' resource pooling
- ▶ Harrison's paradigm for asymptotic control (Wein; van Mieghem's $Gc\mu$)
- ▶ Dai's fluid-based stability
- ▶ Halfin-Whitt's (QED regime) ($\sqrt{\cdot}$ -staffing for many-server queues)
- ▶ $P = NP$: Atar's equivalence of Preemptive and Non-Preemptive SBR; Stolyar, Gurvich
- ▶ Massey-Whitt's research of time-varying queues

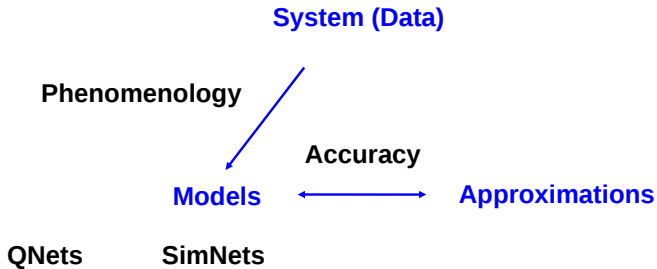
Applying Queueing Asymptotics

There are by now numerous insightful asymptotic queueing models at our disposal, and many arise from, and create, deep beautiful theory:

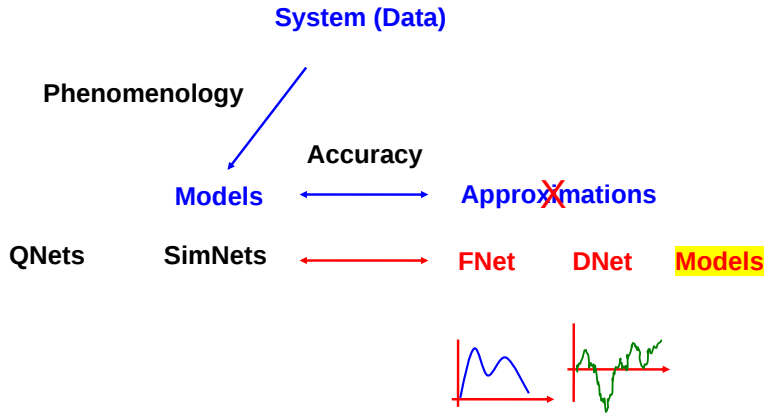
Has it helped one approximate or simulate a service system more efficiently, estimate its parameter more accurately, teach it to our students more effectively, perhaps even manage the system better?

I am of the opinion that the answers to such questions have been too often negative, that positive answers must have theory and applications nurture each other, which is good, and my approach to make this good happen is by marrying theory with data .

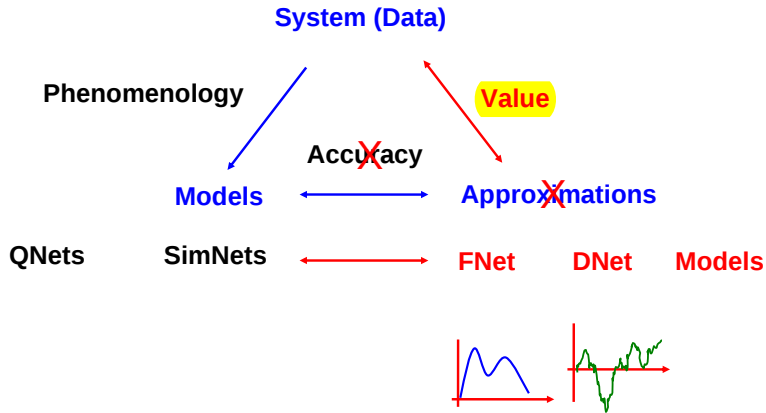
Prevalent Asymptotic Approximations



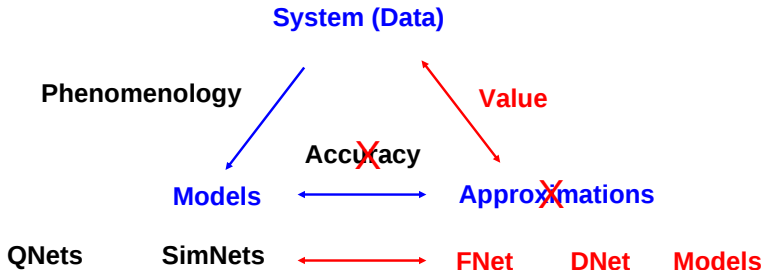
Data-Based ~~Prevalent~~ Asymptotic ~~Approximations~~ **Models**



Data-Based ~~Prevalent~~ Asymptotic ~~Approximations~~ Models



Data-Based ~~Prevalent~~ Asymptotic ~~Approximations~~ Models



System = Coin Tossing, **Model** = Binomial ; de Moivre 1738

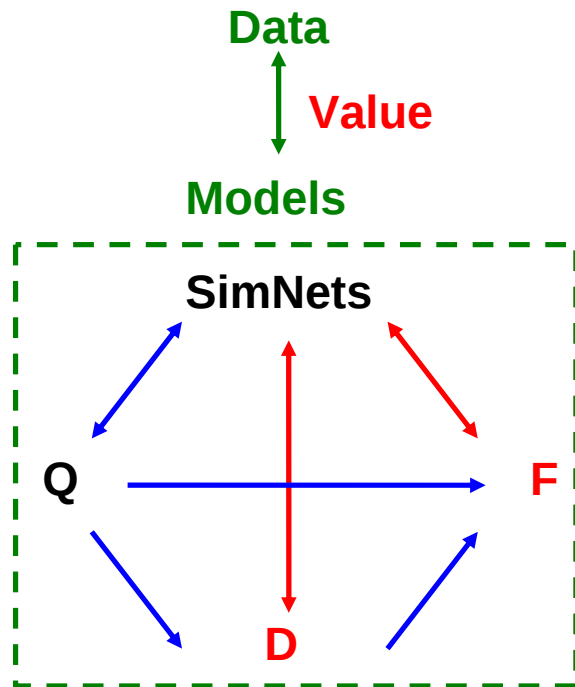
Approx. / Models: SLLN (**FNets**), CLT (**DNets**) ; Laplace 1810

Value: Exceeds Value of originating stylized model

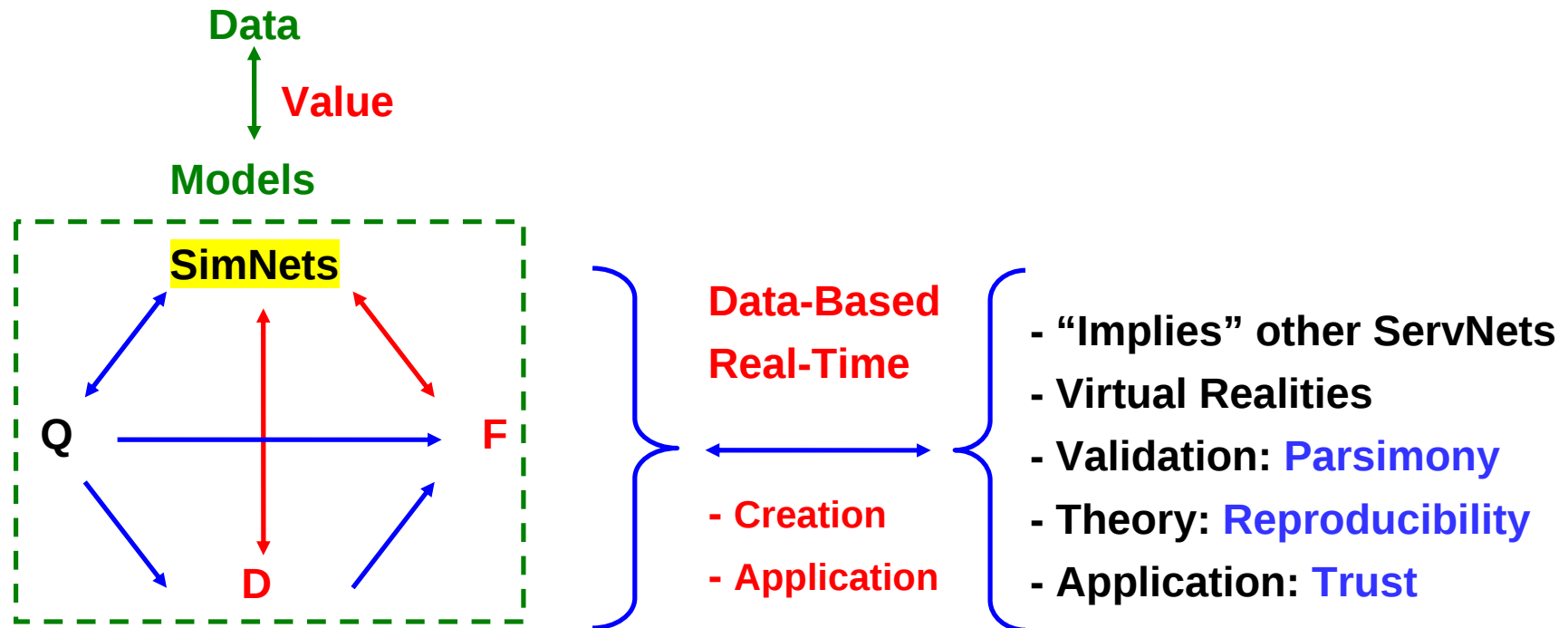
Normal, Brownian Motion ; Bachalier 1900

Poisson ; Poisson 1838

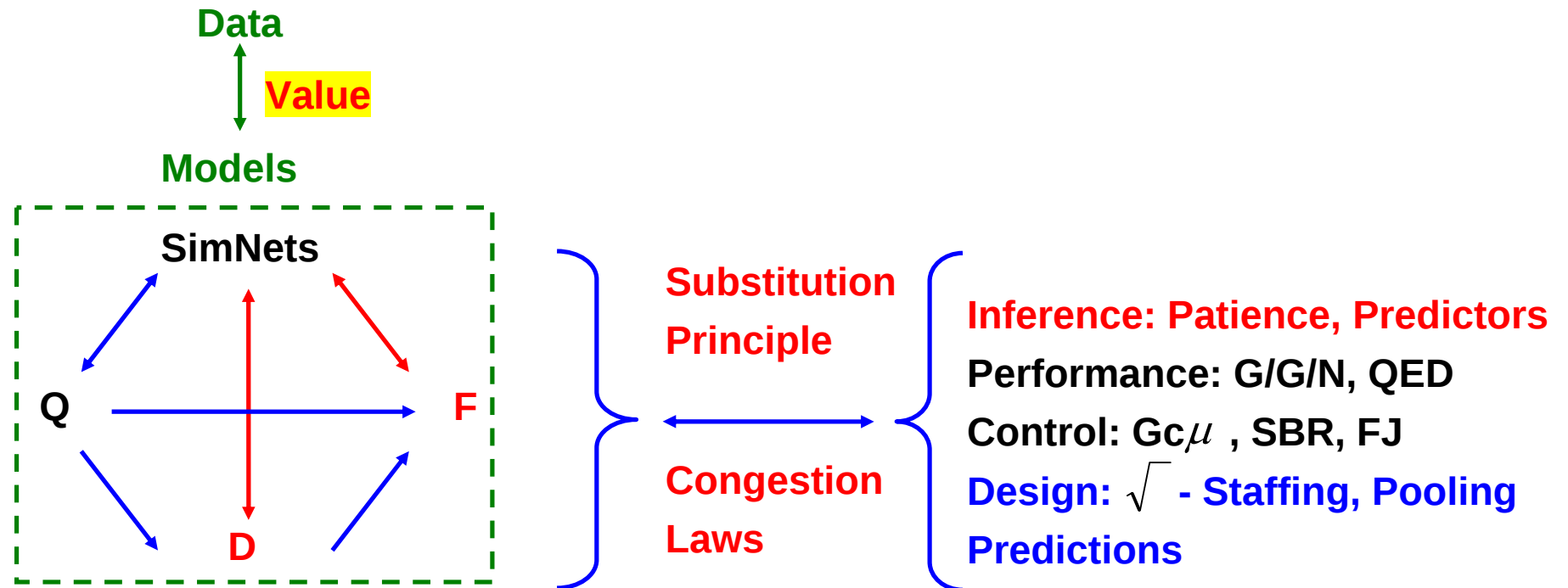
Data-Based Framework: (Almost) All Models Born Equal



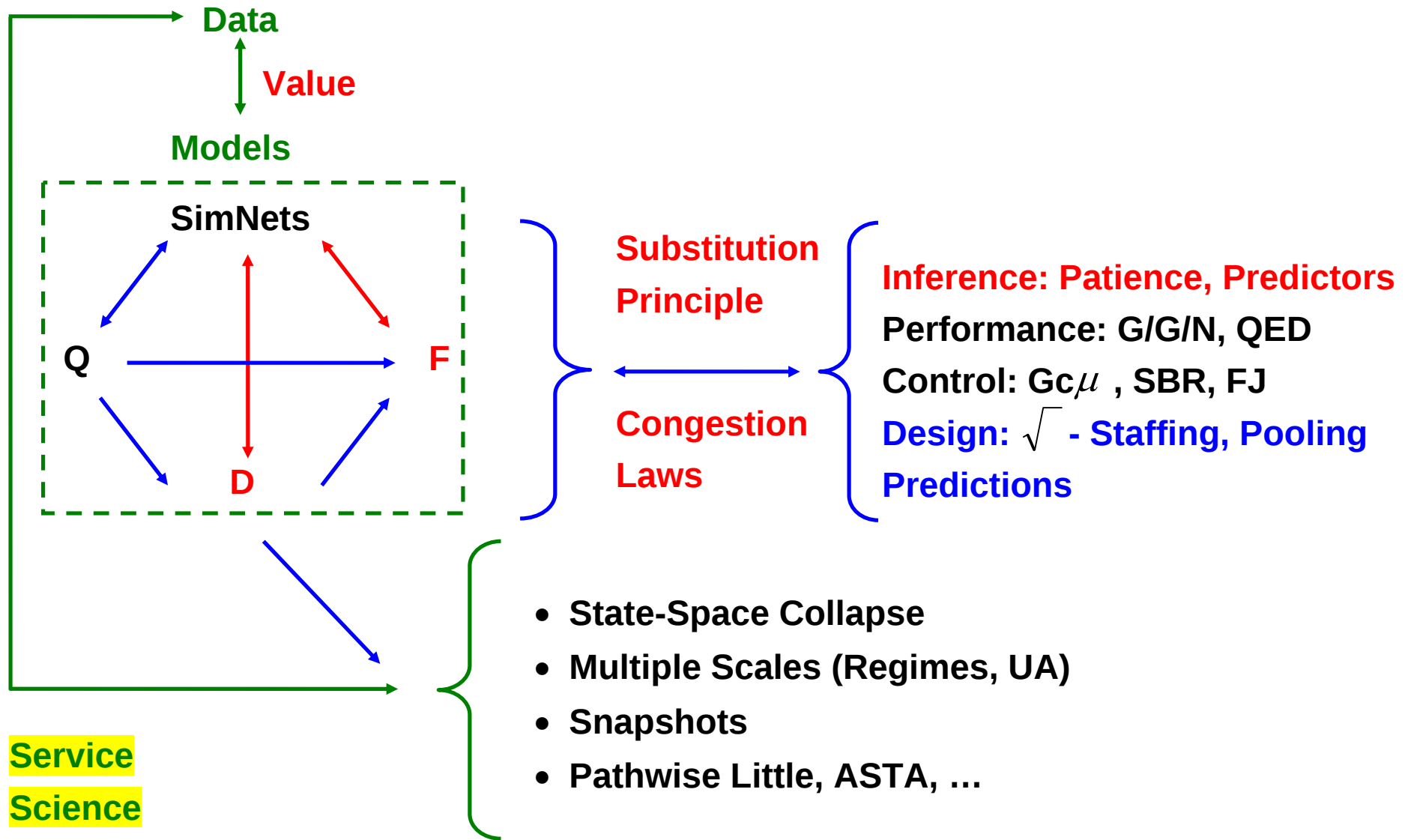
Data-Based (Asymptotic) Framework: Simulation Mining



Data-Based Asymptotic Framework: Added Value



Ultimately: Automatic “Discovery, Conformance, Enhancement”



Scope of the Service Industry

Guangzhou Railway Station, Southern China



Call Centers, Then Hospitals, Now Internet

Call Centers - U.S. Stat.

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

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Healthcare - similar, plus unique challenges:

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

Internet - ...

Call-Center Environment: Service Network



Operational Focus



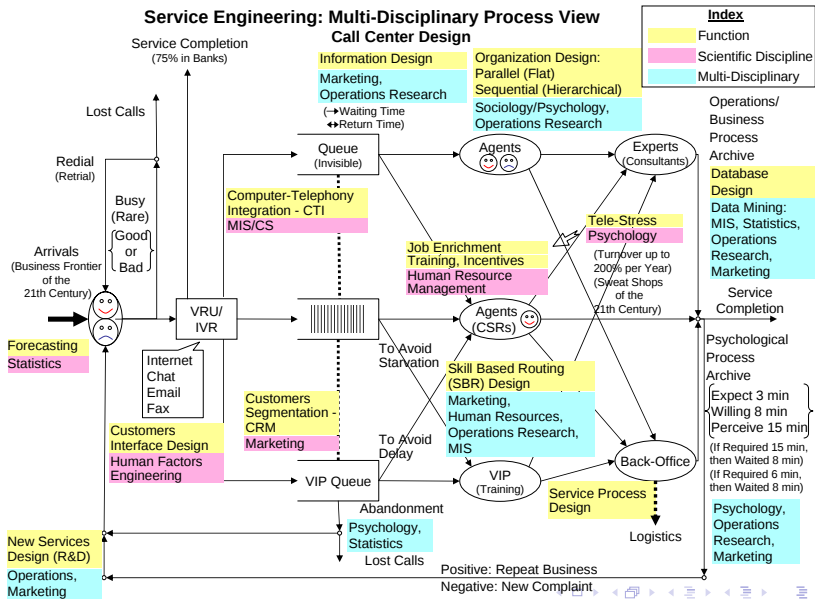
Operational Measures:

- ▶ Surrogates for overall performance: Financial, Psychological; Clinical
- ▶ Easiest to quantify, measure, track online, react upon / Research

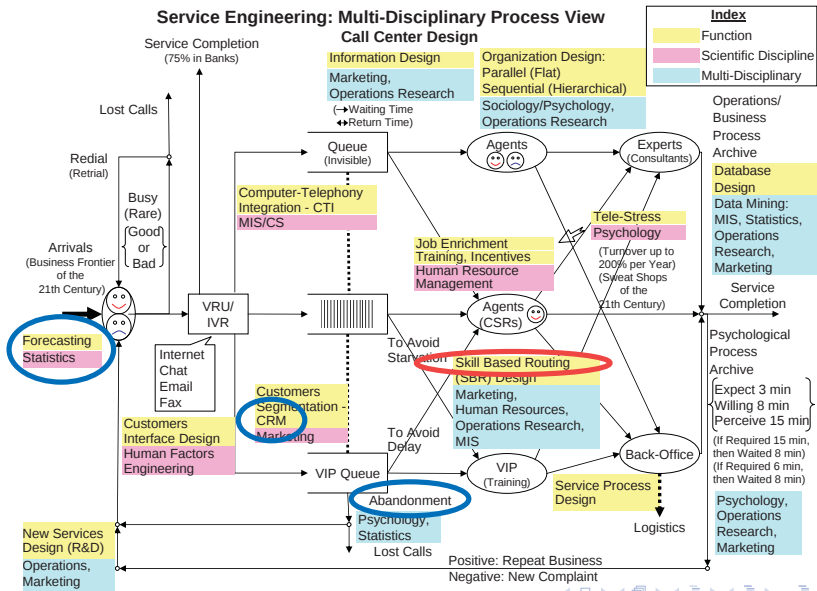
Call-Center Network: Gallery of Models

Service Engineering: Multi-Disciplinary Process View

Call Center Design



Call-Center Network: Gallery of Models



Skills-Based Routing in Call Centers

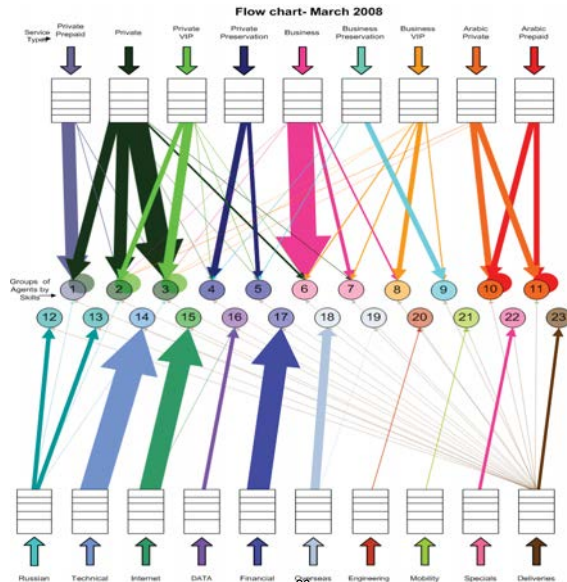
EDA and OR, with I. Gurvich and P. Liberman

Mktg. ⇒

OR ⇒

HRM ⇒

MIS ⇒



ER / ED Environment: Service Network

Acute (Internal, Trauma)



Walking



Multi-Trauma



ED-Environment in Israel

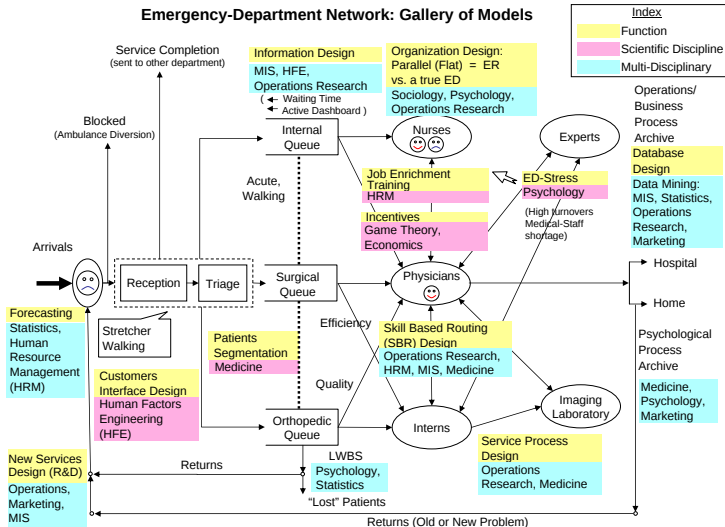


Queueing in a “Good” Hospital

Tong-ren Hospital at 6am, Beijing



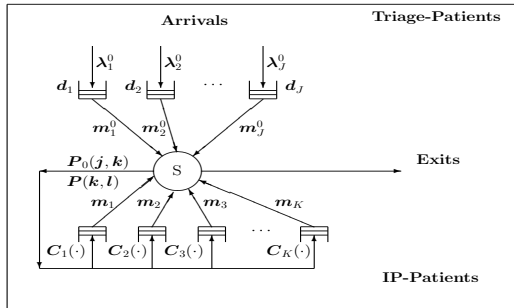
Emergency-Department Network: Gallery of Models



► **Forecasting**, Abandonment = **LWBS**, SBR $\approx$ **Flow Control**

ED Patient Flow: The Physicians View

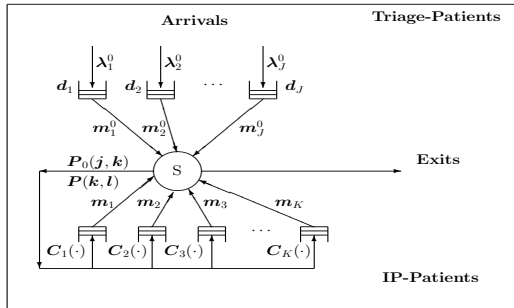
with J. Huang, B. Carmeli



- ▶ **Goal:** Adhere to **Triage-Constraints**, then **release In-Process** Patients
- ▶ **Model** = Multi-class Q with Feedback: Min. convex **congestion costs** of IP-Patients, s.t. **deadline constraints** on Triage-Patients.

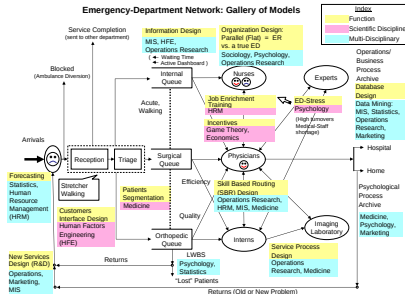
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- ▶ **Goal:** Adhere to **Triage-Constraints**, then **release In-Process** Patients
- ▶ **Model** = Multi-class Q with Feedback: Min. convex **congestion costs** of IP-Patients, s.t. **deadline constraints** on Triage-Patients.
- ▶ **Solution:** In conventional heavy-traffic, **asymptotic least-cost** s.t. **asymptotic compliance** (as in Plambeck, Harrison, Kumar, who applied admission control):
 - ▶ Triage or IP? former, if some deadline is “too” close
 - ▶ Triage Priorities: Chose the closest deadline
 - ▶ IP-Priorities: $Gc\mu$, modified (simply) to account for feedback

Emergency-Department Network: Flow Control



- ▶ ***Queueing-Science**, w/ **Armony, Marmor, Tseytlin, Yom-Tov**
- ▶ ***Fair ED-to-IW Routing** (Patients vs. Staff), w/ **Momcilovic, Tseytlin**
- ▶ ***Triage vs. InProcess/Release (Plambeck et al, van Mieghem)** in EDs, w/ **Carmeli, Huang; Shimkin**
- ▶ ***Staffing Time-Varying Q's with Re-Entrant Customers (de Vericourt & Jennings)**, w/ **Yom-Tov**
- ▶ **The Offered-Load in Fork-Join Nets (Adlakha & Kulkarni)**, w/ **Kaspi, Zaeid**
- ▶ **Synchronization Control** of Fork-Join Nets, w/ **Atar, Zviran**

Prerequisite I: Data

Averages Prevalent (and could be useful / interesting).

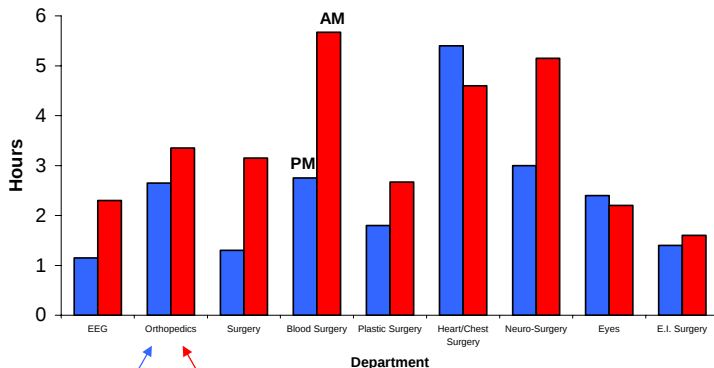
But I need data at the level of the **Individual Transaction**:

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

operational history = time-stamps of events (events-log files).

Interesting Averages: The Human Factor, or Even “Doctors” Can Manage

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



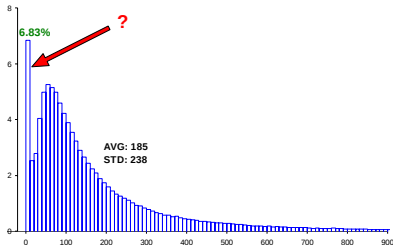
**Afternoon,
by Case**

**Morning,
by Hour**

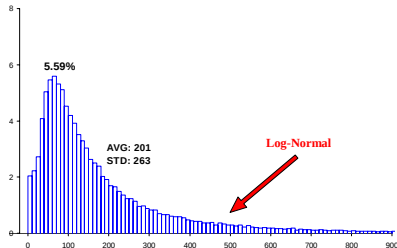
Beyond Averages: The Human Factor

Histogram of Service-Time in an Israeli Call Center, 1999

January-October



November-December

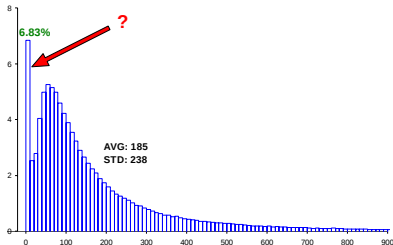


► 6.8% Short-Services:

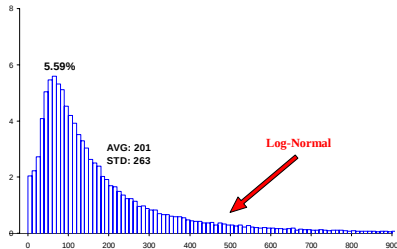
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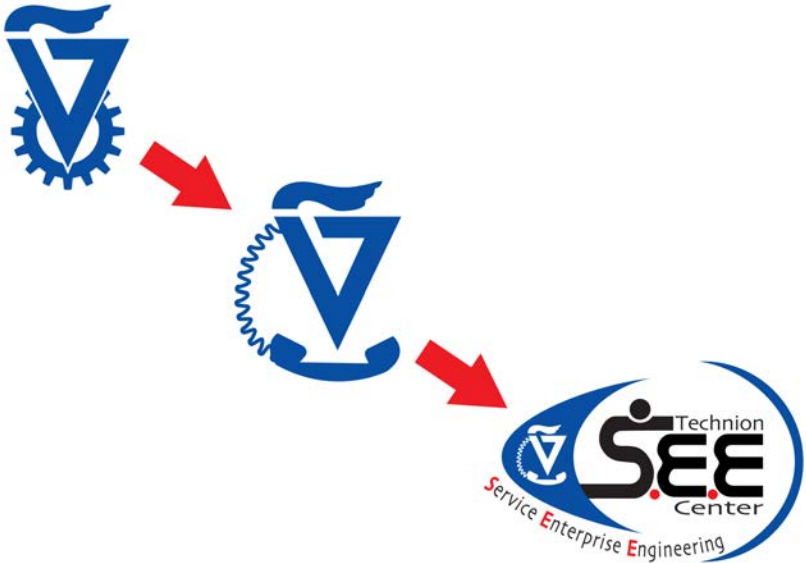
November-December



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

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 year, 350K calls by 15 agents** - in 2000.
Brown, Gans, Sakov, Shen, Zeltyn, Zhao (JASA),
paved the way to:
- ▶ U.S. Bank: **2.5 years, 220M calls, 40M by 1000 agents**
- ▶ Israeli Cellular: **2.5 years, 110M calls, 25M calls by 750 agents**
- ▶ Israeli Bank: **from January 2010, daily-deposit at a SEESafe**
- ▶ Home (Rambam) Hospital: **4 years, 1000 beds**, ward-level flow
- ▶ 5 EDs: **gathered by the late David Sinreich**, ED arrivals & LOS

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SEESat: Environment for graphical EDA in real-time

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

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SEEServer: Free for academic use

- ▶ Register
- ▶ Access **U.S. Bank, Bank Anonymous, Home Hospital**

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 (helps 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) !
- **"Psychology"** of Data Trust and Transfer (e.g. 2 years till transfer)

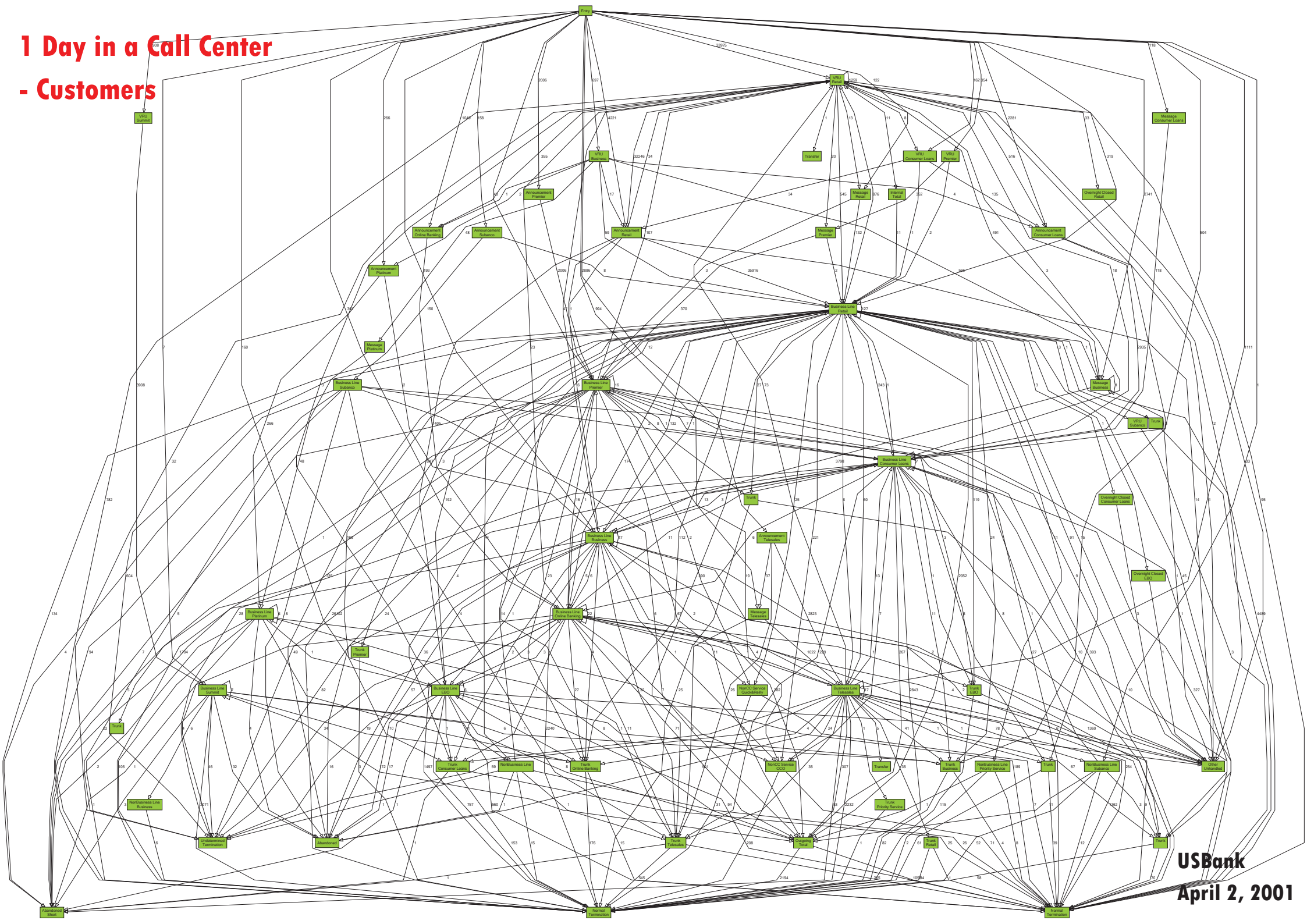
Event-Logs in a Call Center (Bank Anonymous)

A Data Sample (Excel worksheet)

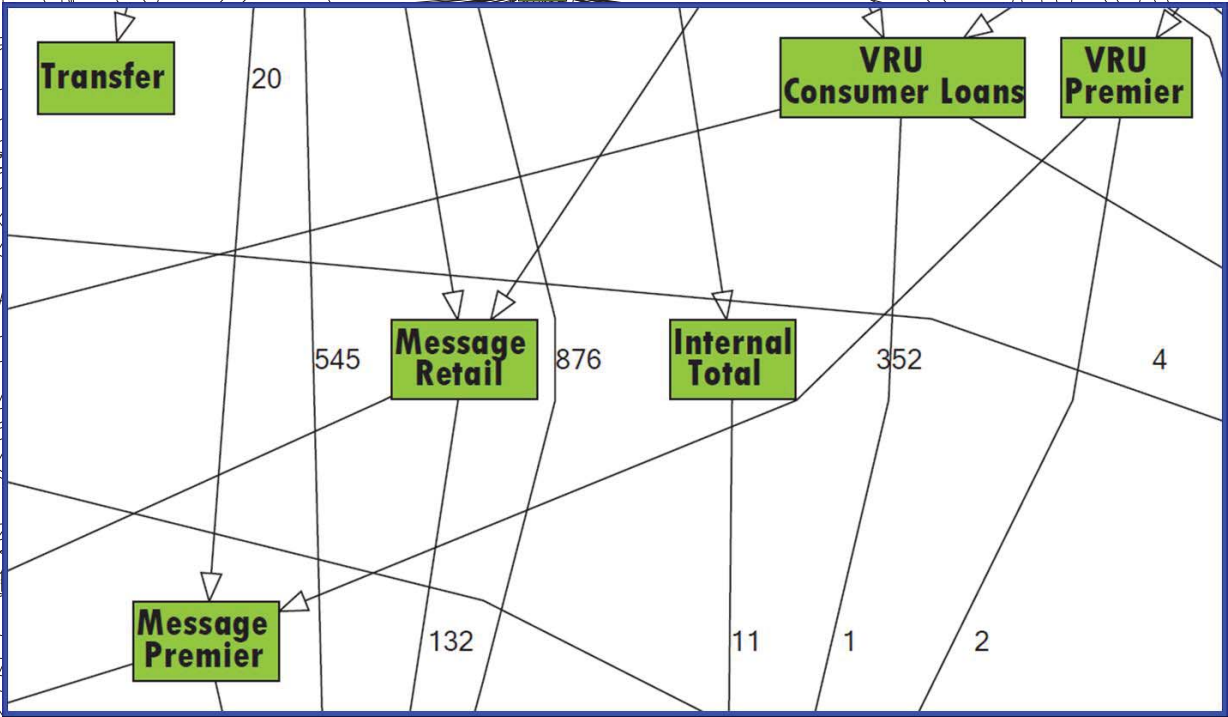
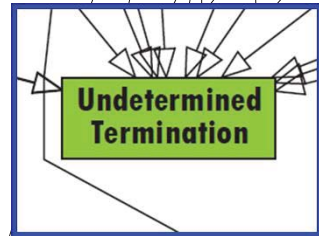
vru+line	call_id	customer_id	priority	type	date	vru_entry	vru_exit	vru_time	q_start	q_exit	q_time	outcome	ser_start	ser_exit	ser_time	server
AA0101	44749	27644400	2	PS	990901	11:45:33	11:45:39	6	11:45:39	11:46:58	79	AGENT	11:46:57	11:51:00	243	DORIT
AA0101	44750	12887816	1	PS	990905	14:49:00	14:49:06	6	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	14:58:48	15:02:31	223	AGENT	15:02:31	15:04:10	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	1	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
AA0101	44975	55920755	2	PS	990905	16:07:46	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	2	PS	990905	16:14:27	16:14:33	6	16:14:33	16:14:54	21	HANG	00:00:00	00:00:00	0	NO_SERVER
AA0101	44978	23817067	2	PS	990905	16:19:11	16:19:17	6	16:19:17	16:19:39	22	AGENT	16:19:38	16:21:57	139	TOVA
AA0101	44764	0	0	PS	990901	15:03:26	15:03:36	10	00:00:00	00:00:00	0	AGENT	15:03:35	15:06:36	181	ZOHARI
AA0101	44765	25219700	2	PS	990901	15:14:46	15:14:51	5	15:14:51	15:15:10	19	AGENT	15:15:09	15:17:00	111	SHARON
AA0101	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

- Unsynchronized transition times, consistently

1 Day in a Call Center
- Customers



1 Day in a Call Center
- Customers



Server Network
- 1 Day
- All Agents
- Activities + Queues

Available

Inbound

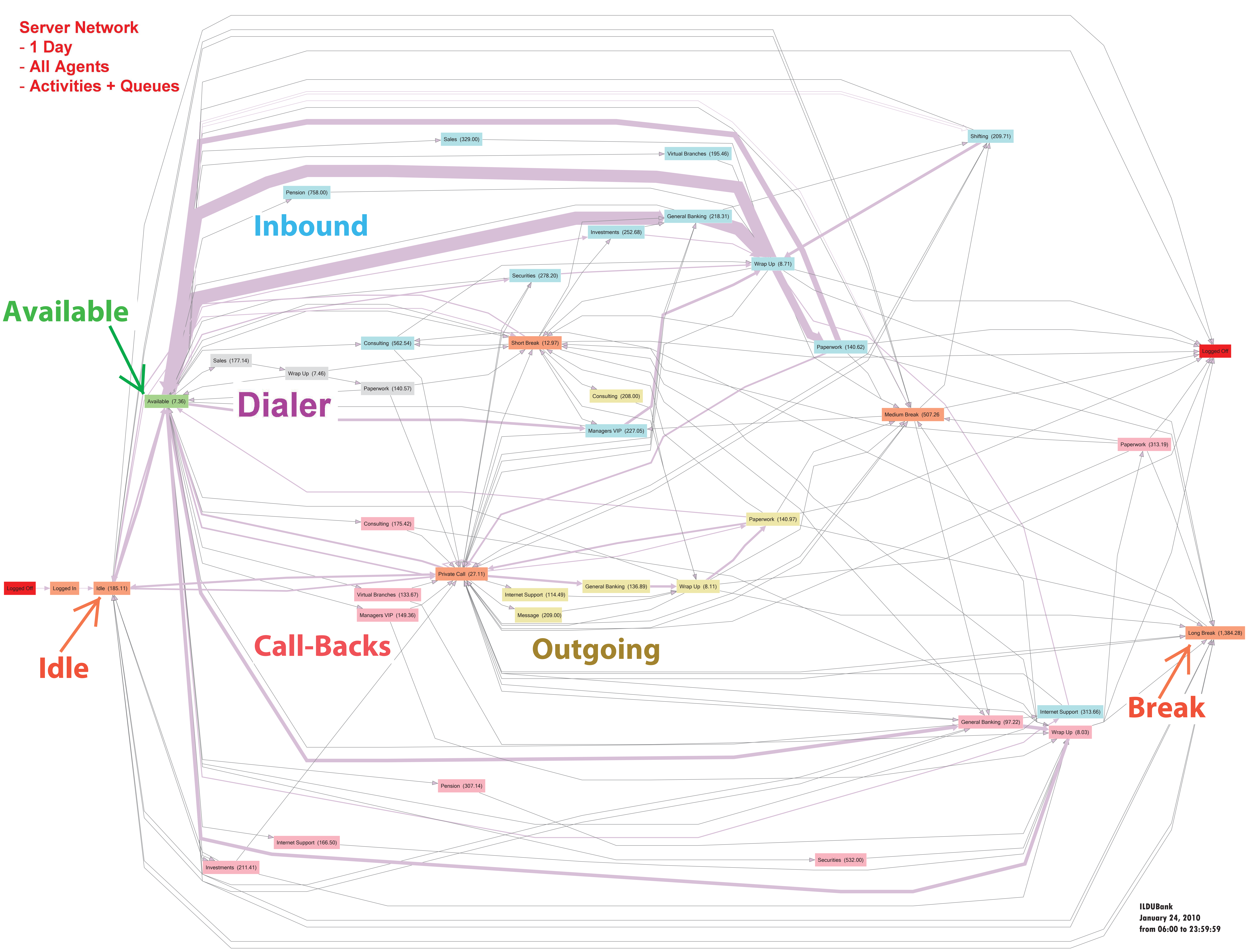
Dialer

Call-Backs

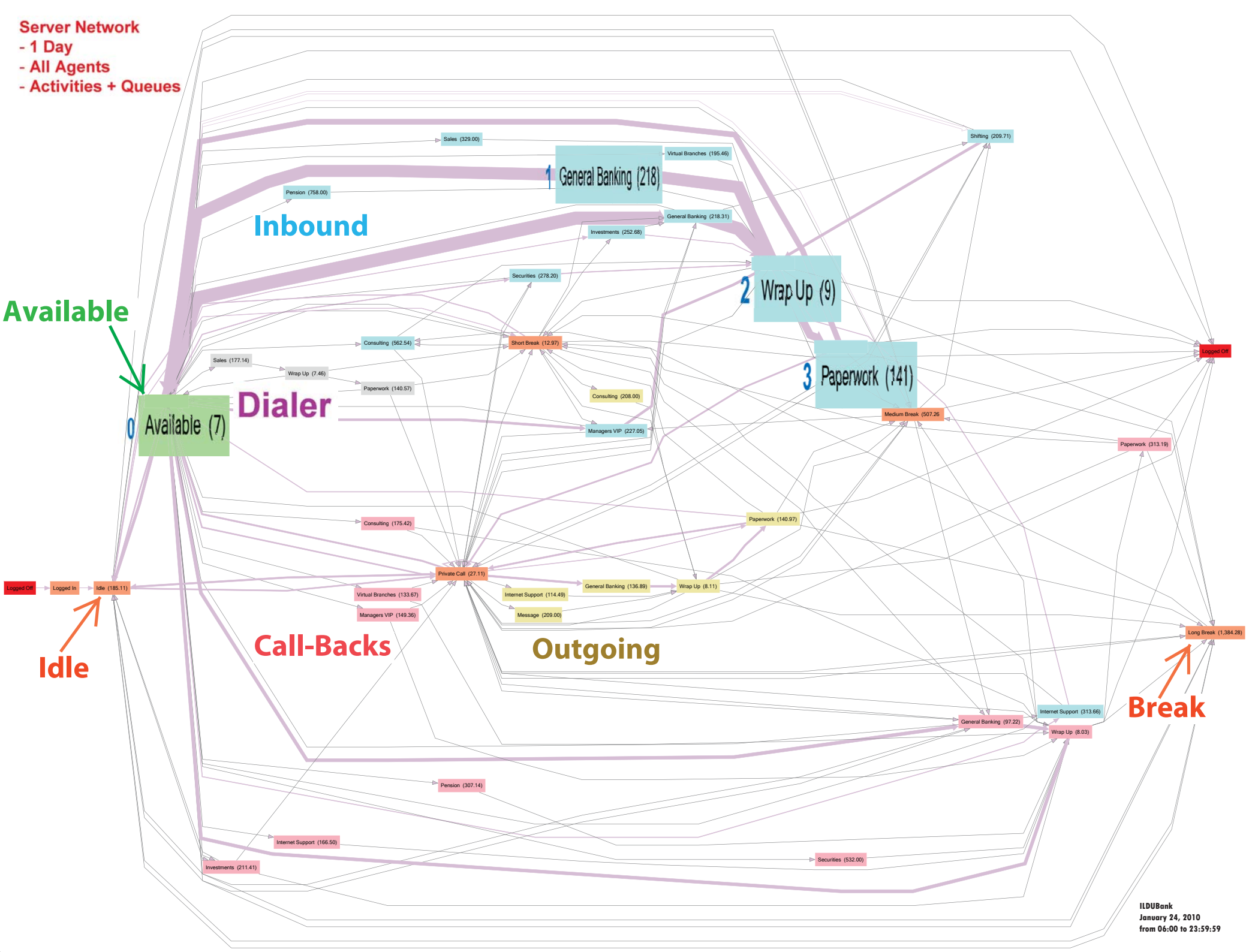
Outgoing

Idle

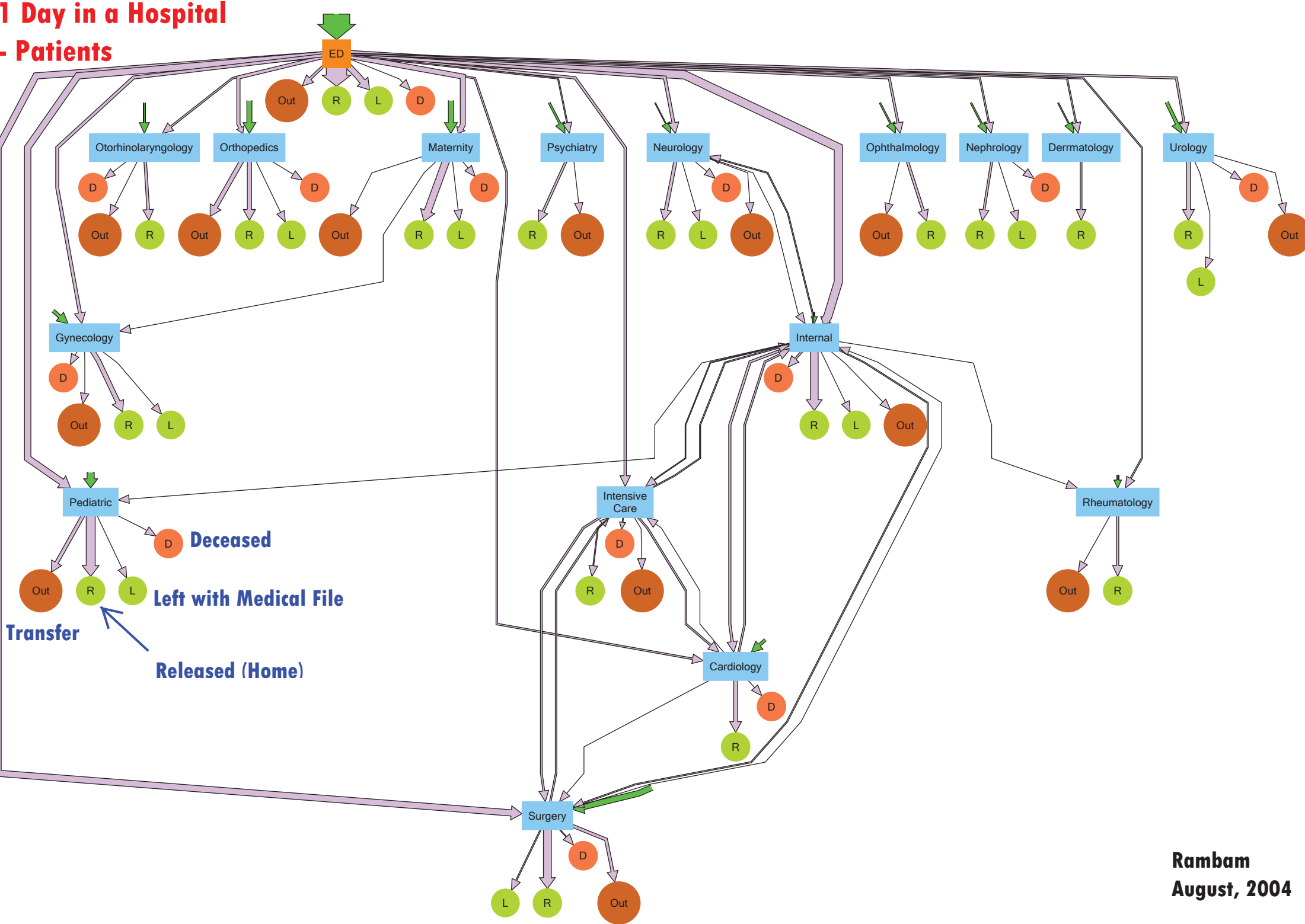
Break



Server Network
- 1 Day
- All Agents
- Activities + Queues



1 Day in a Hospital
- Patients



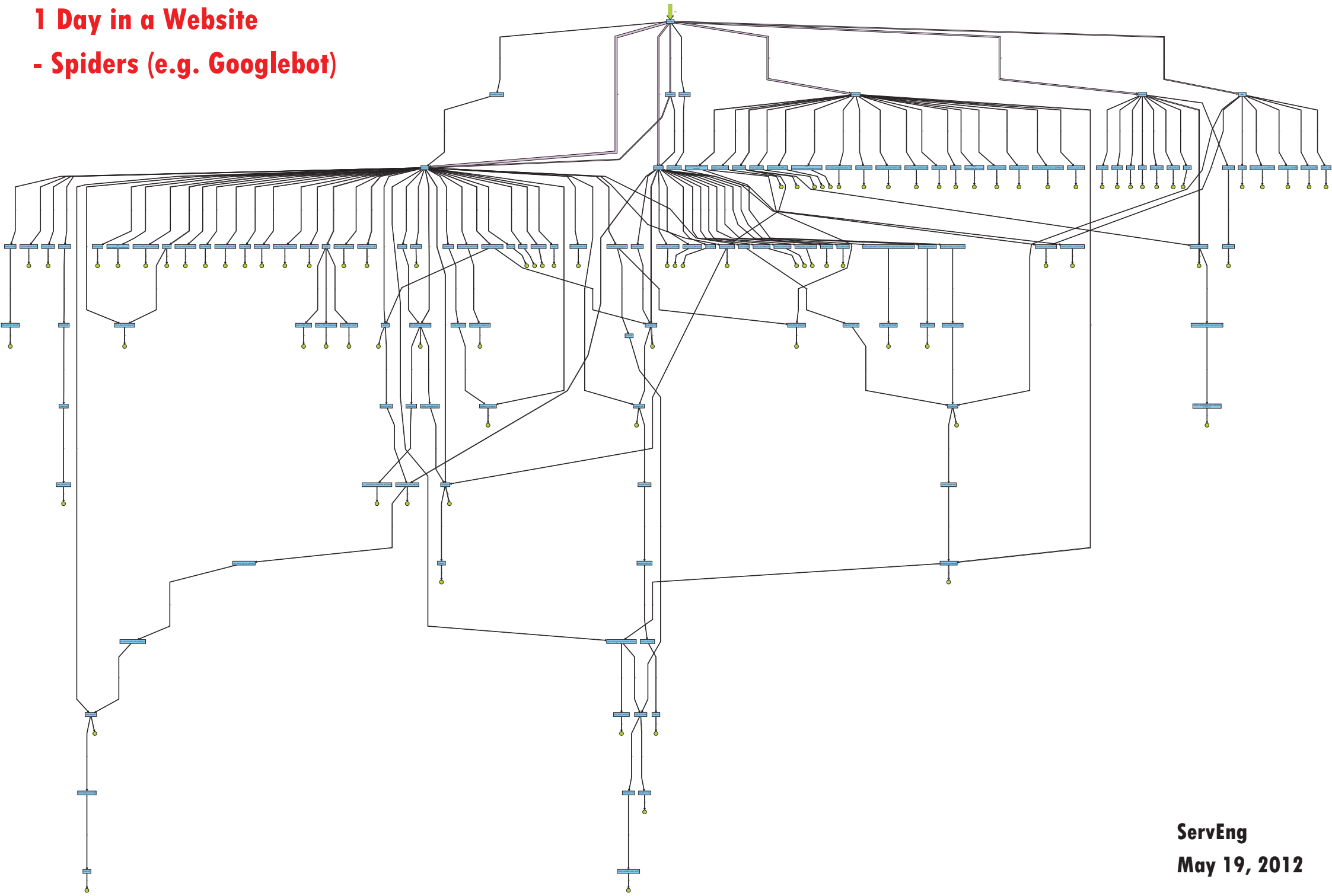
Deceased

Left with Medical File

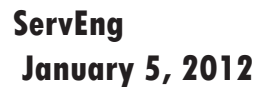
Transfer

Released (Home)

1 Day in a Website
- Spiders (e.g. Googlebot)

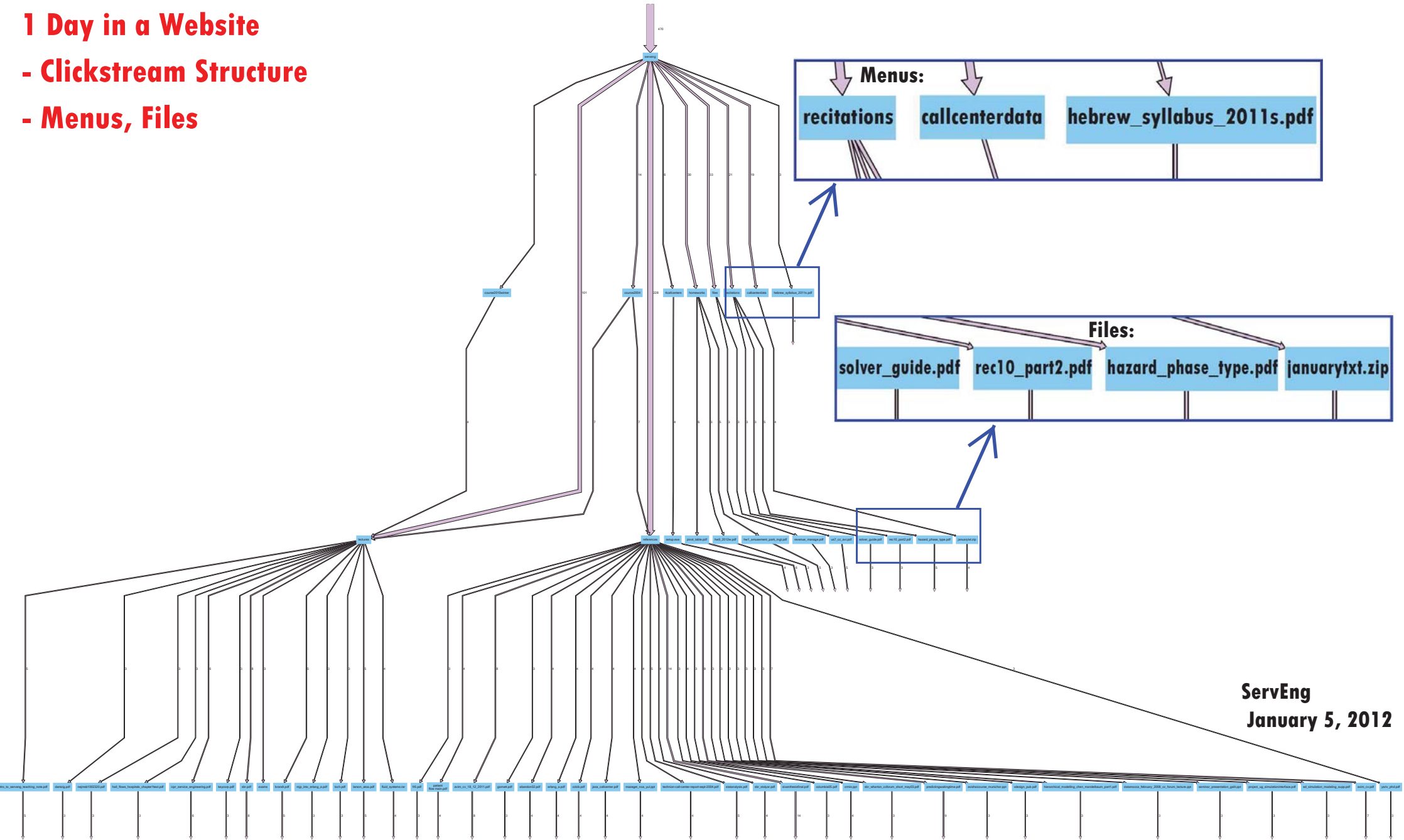


- Menus, Files



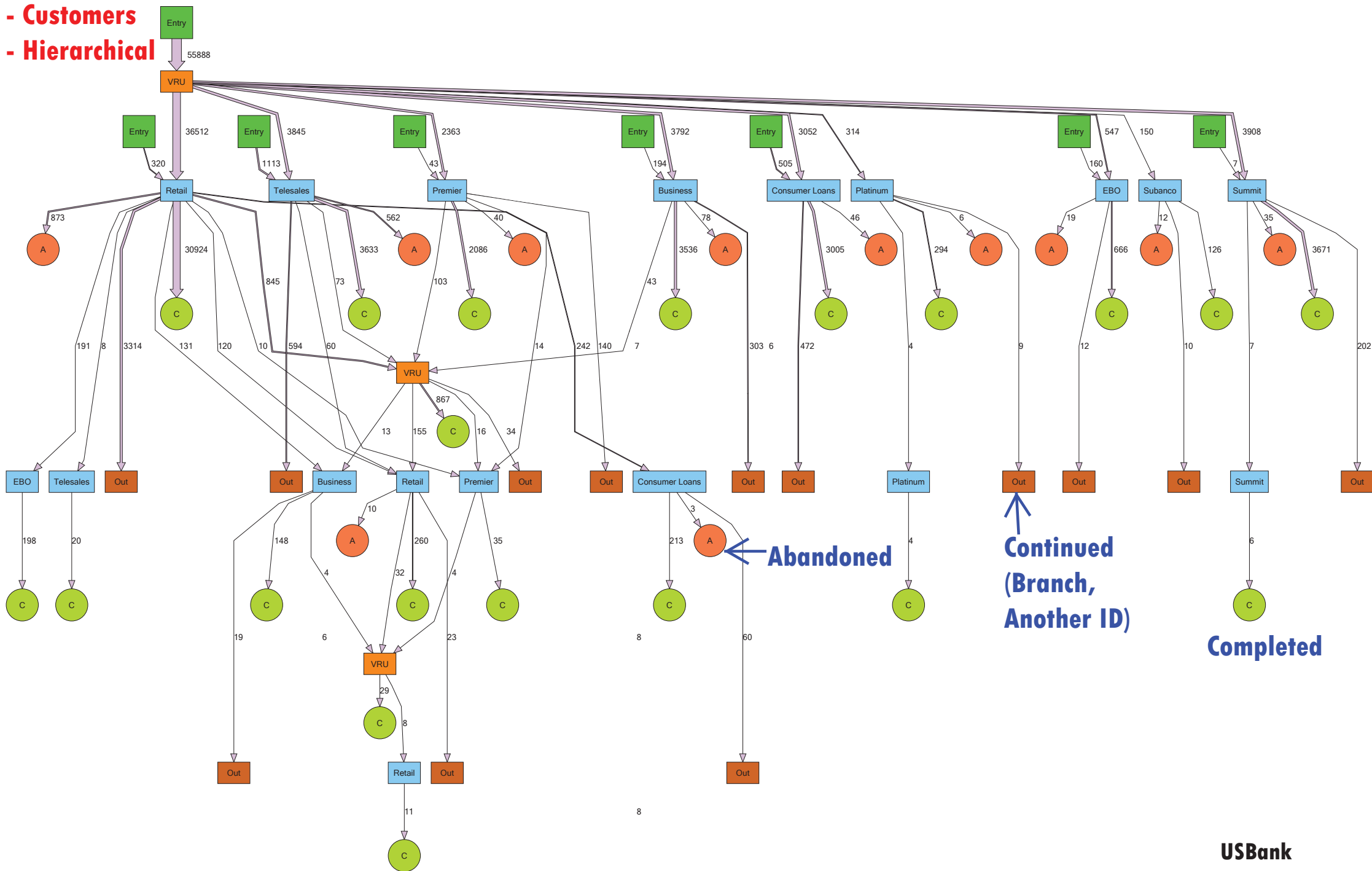
1 Day in a Website

- Clickstream Structure
- Menus, Files



1 Hour in a Call Center

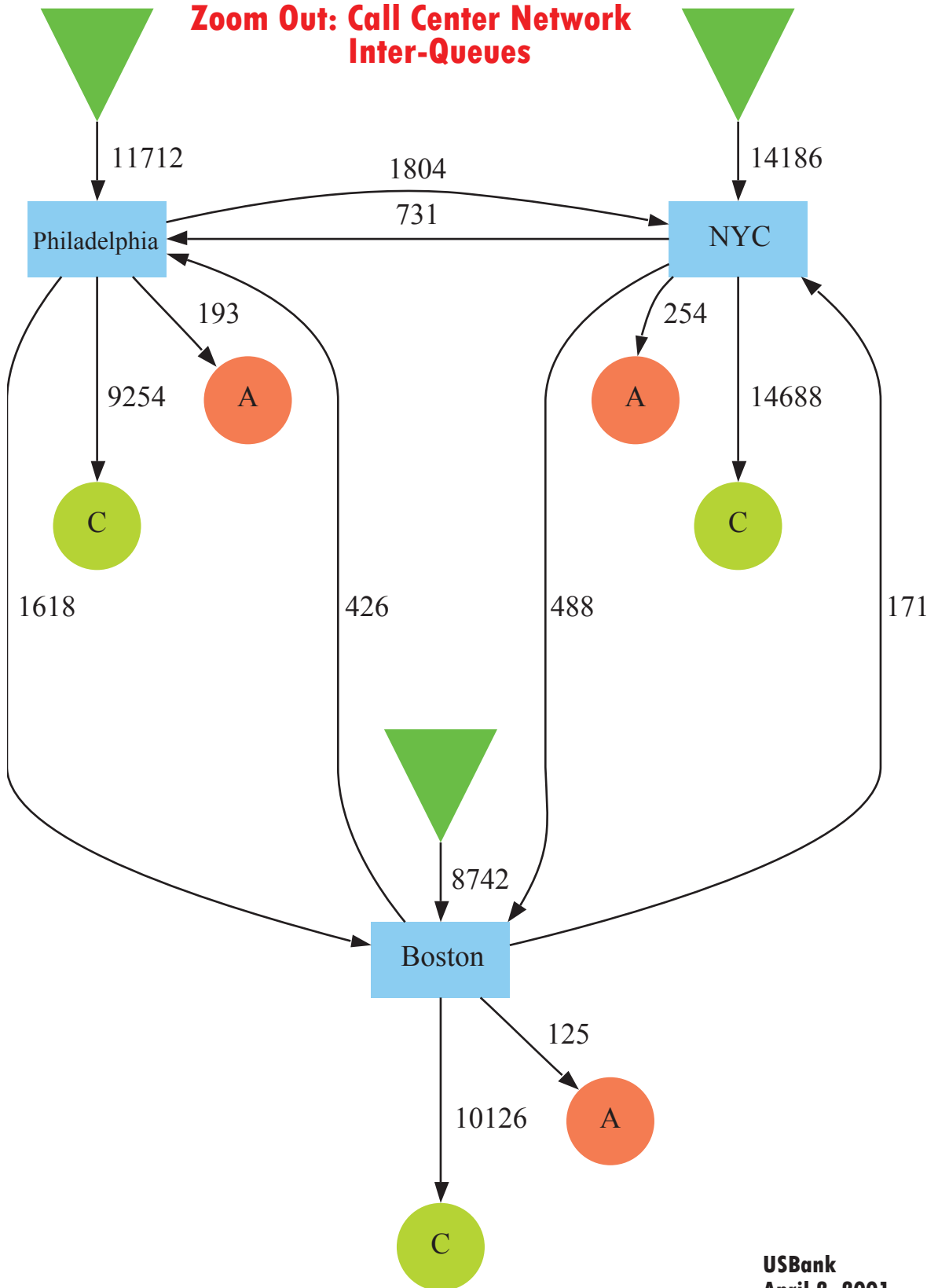
- Customers
- Hierarchical



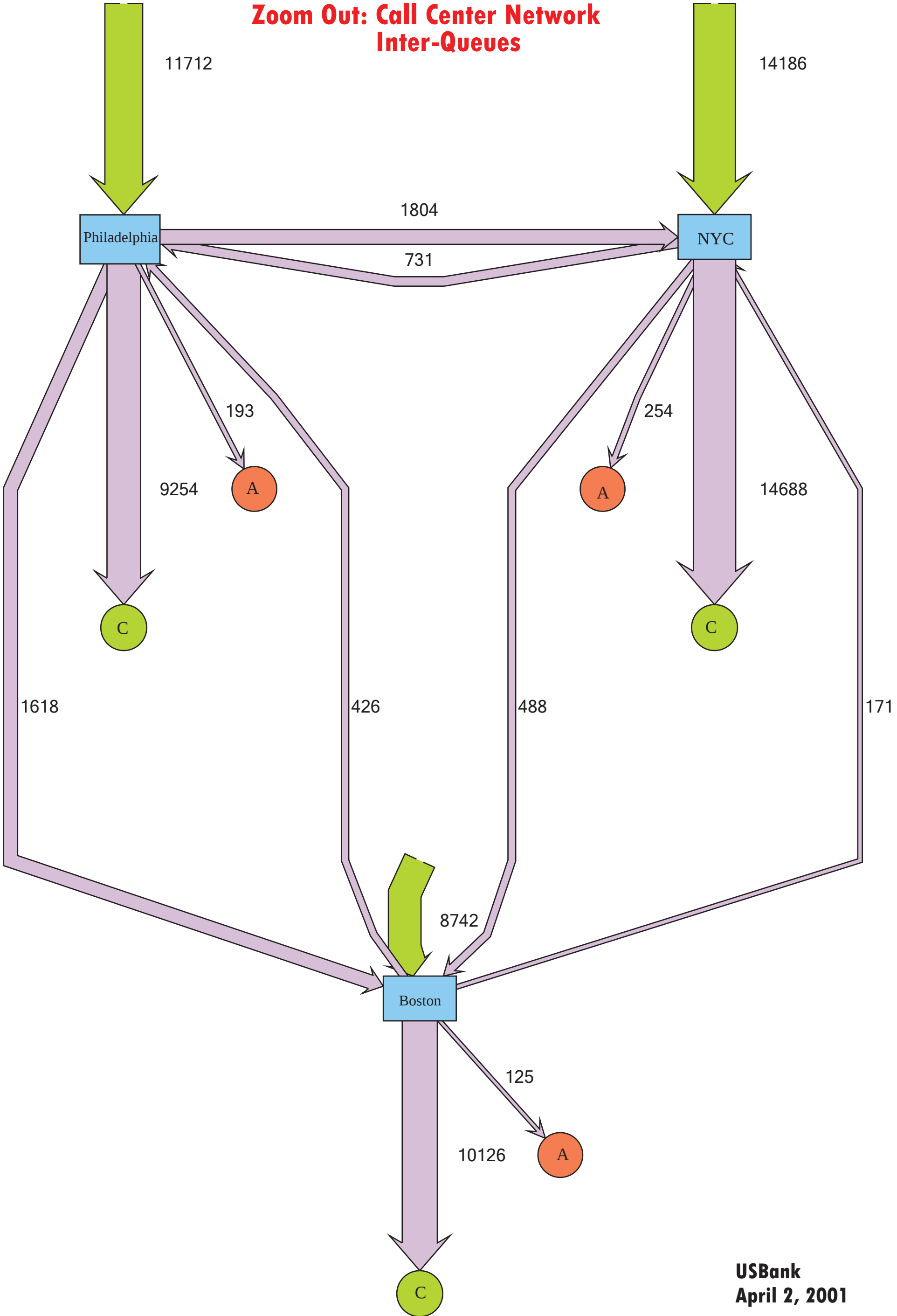
8 AM - 9 AM

USBank
April 2, 2001

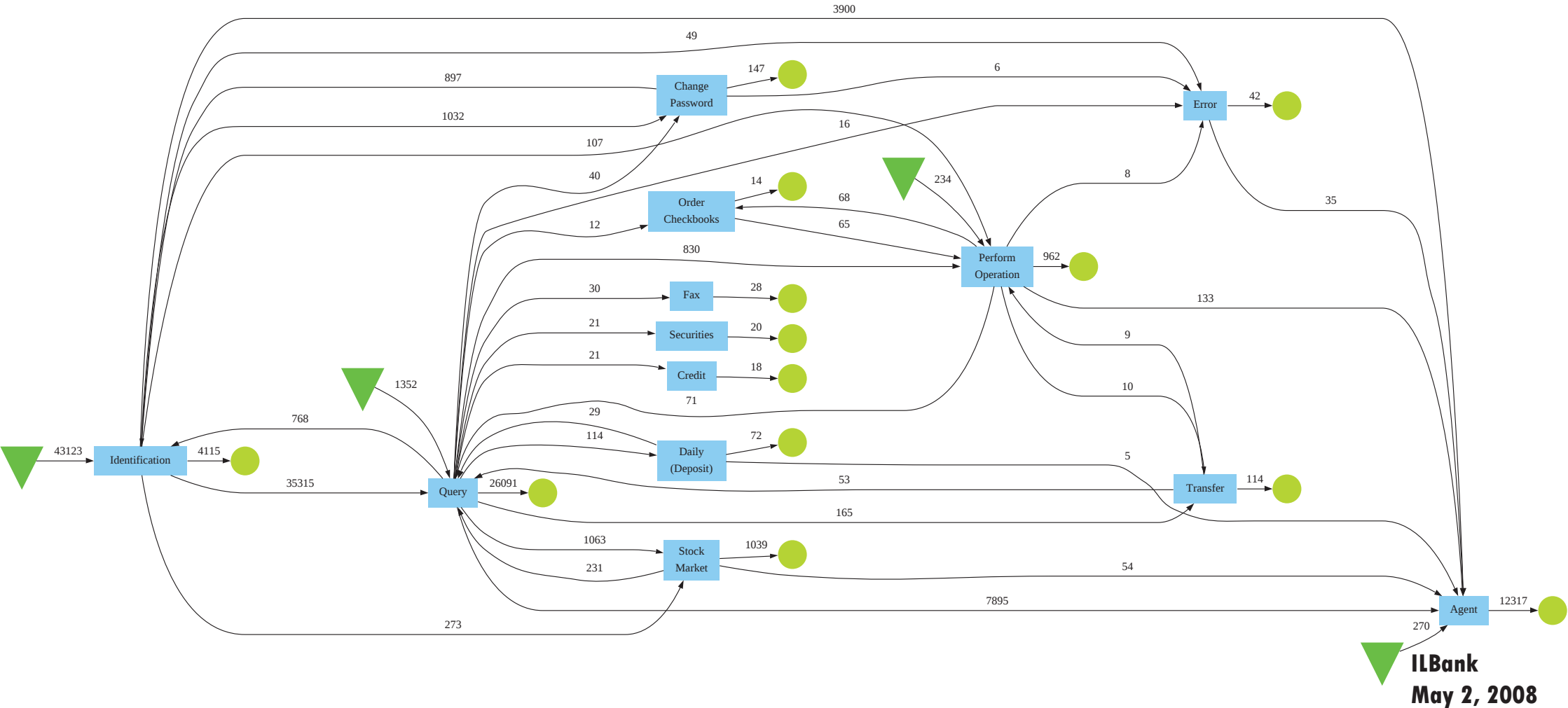
Zoom Out: Call Center Network Inter-Queues



Zoom Out: Call Center Network
Inter-Queues



Zoom In: Interactive Voice Response (IVR)

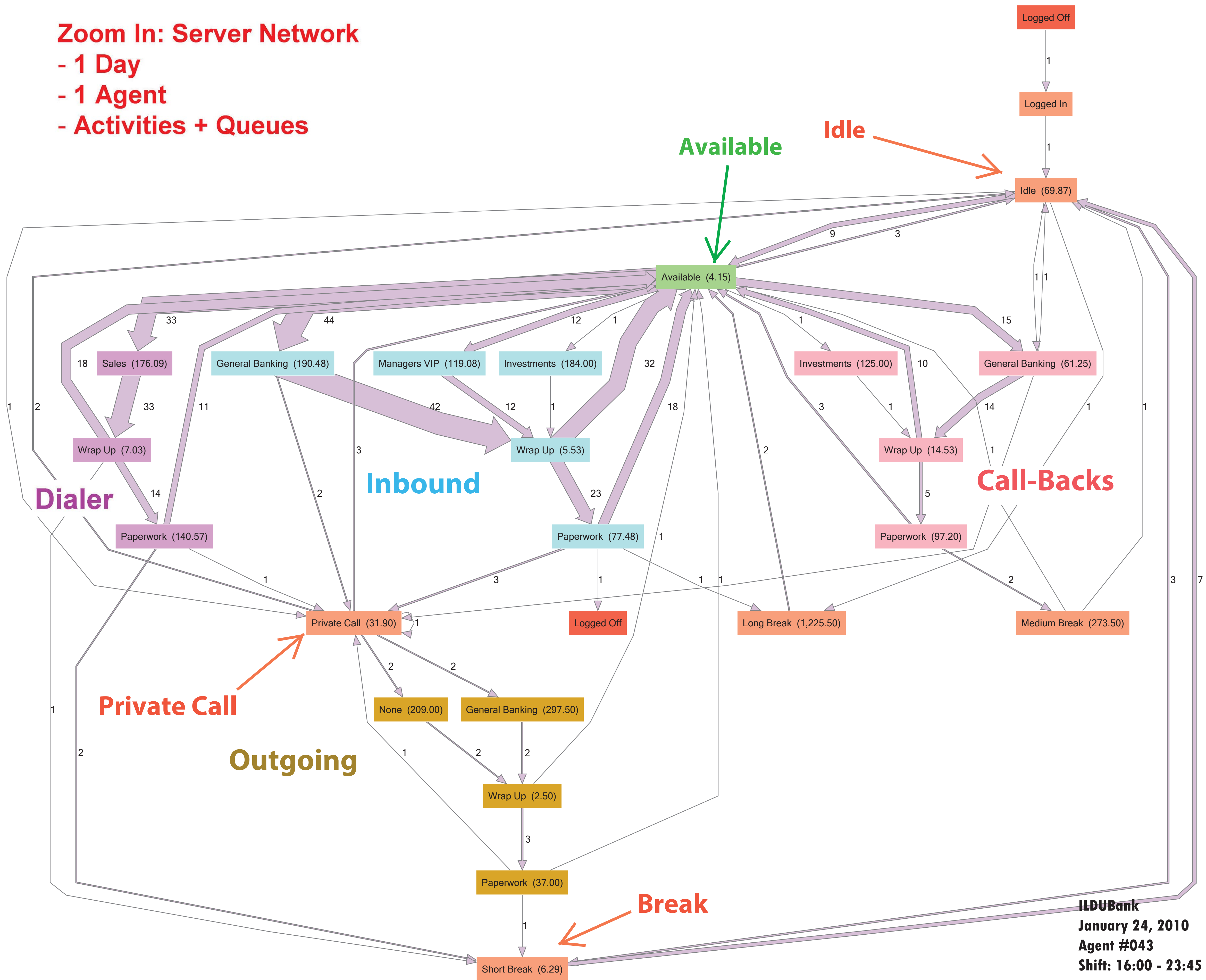


Zoom In: Server Network

- 1 Day

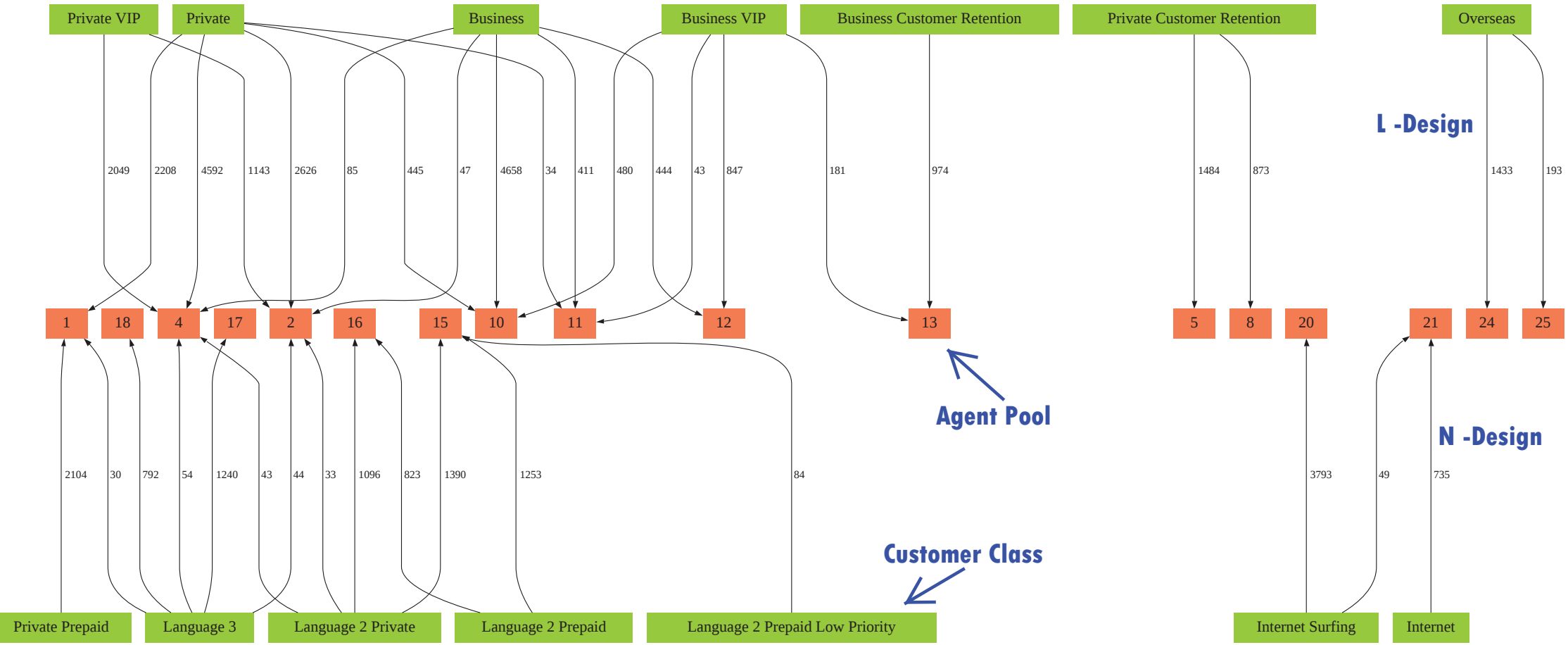
- 1 Agent

- Activities + Queues



Zoom In: Skills - Based Routing (SBR)

Network Structure (Protocol)

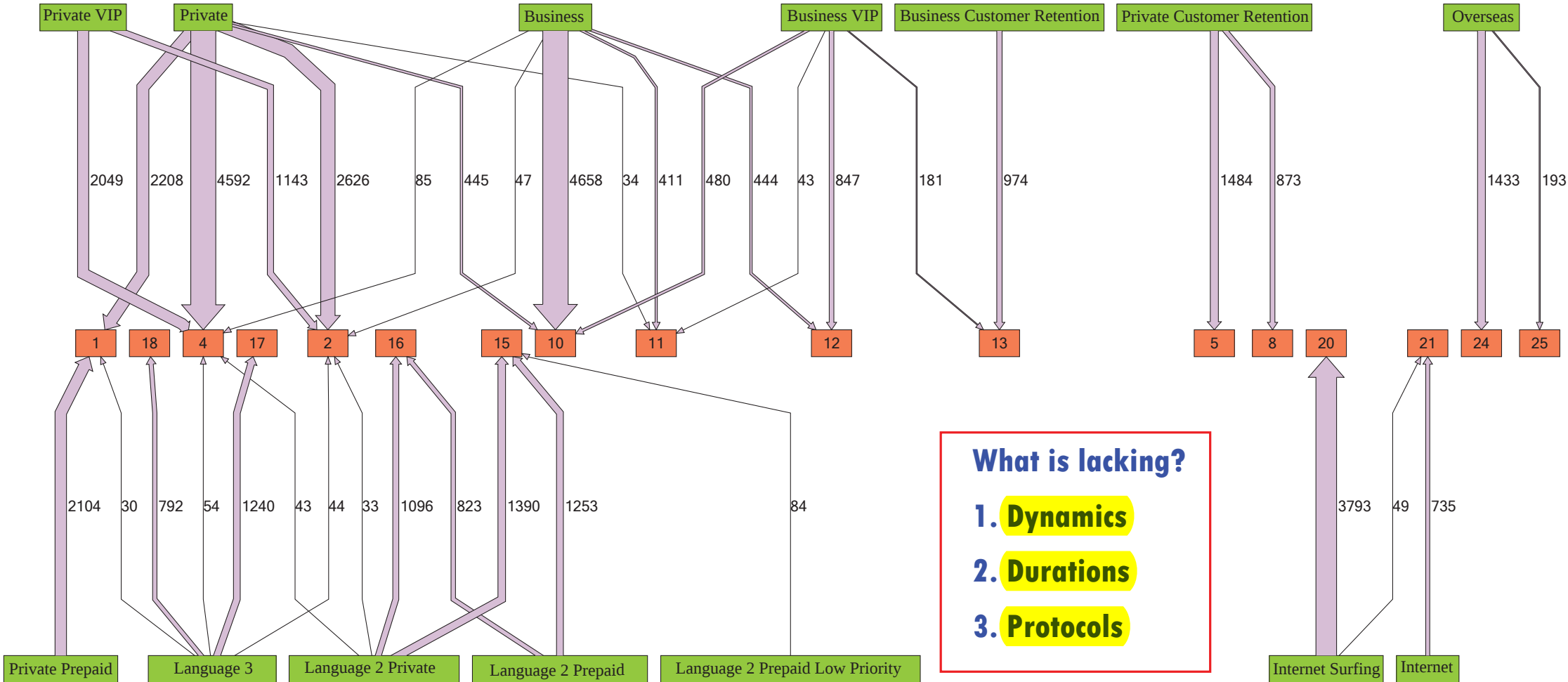


Connected Component

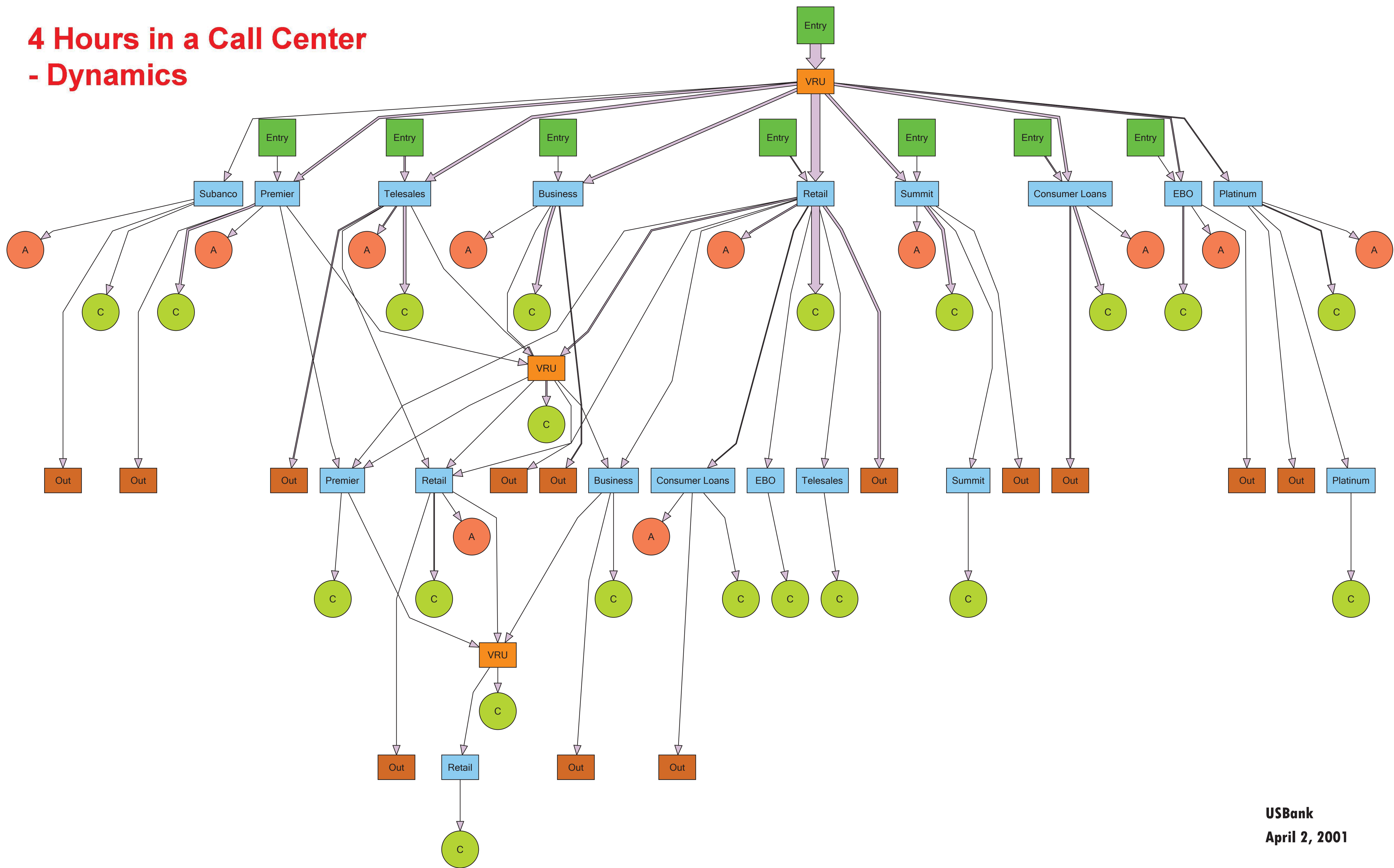
L -Design

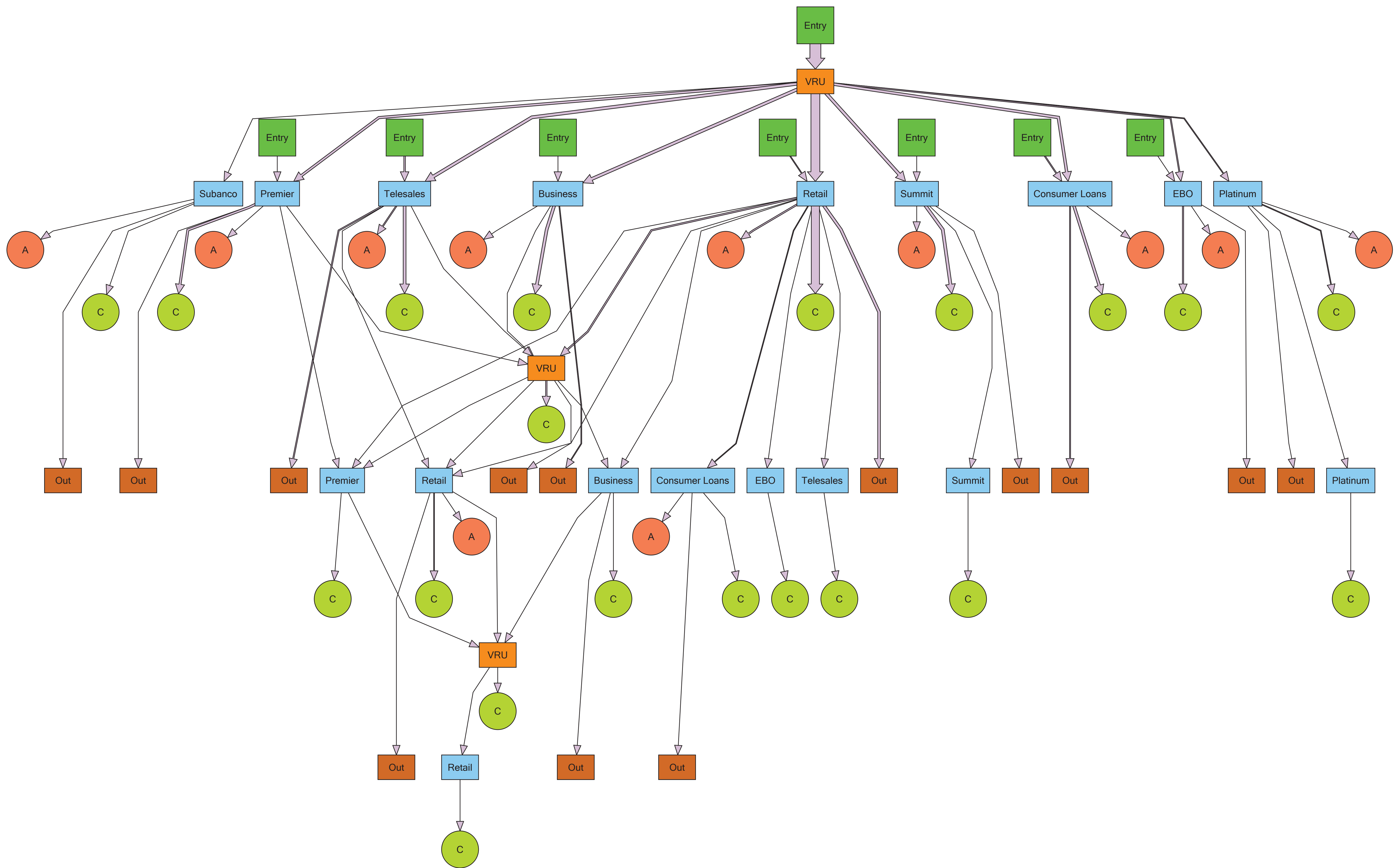
N -Design

Goal: Data-Based Real-Time Simulation (SimNet)

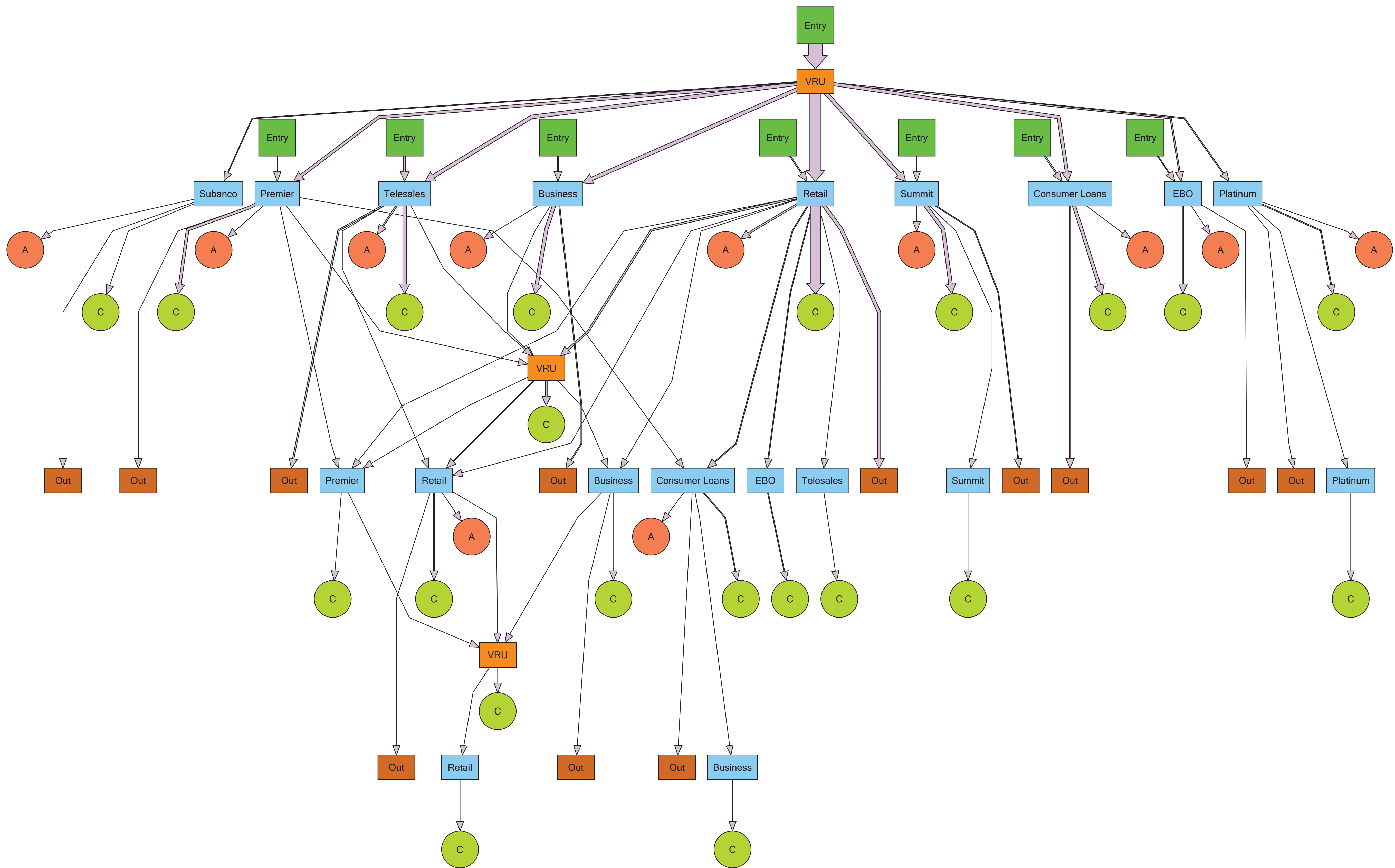


4 Hours in a Call Center
- Dynamics

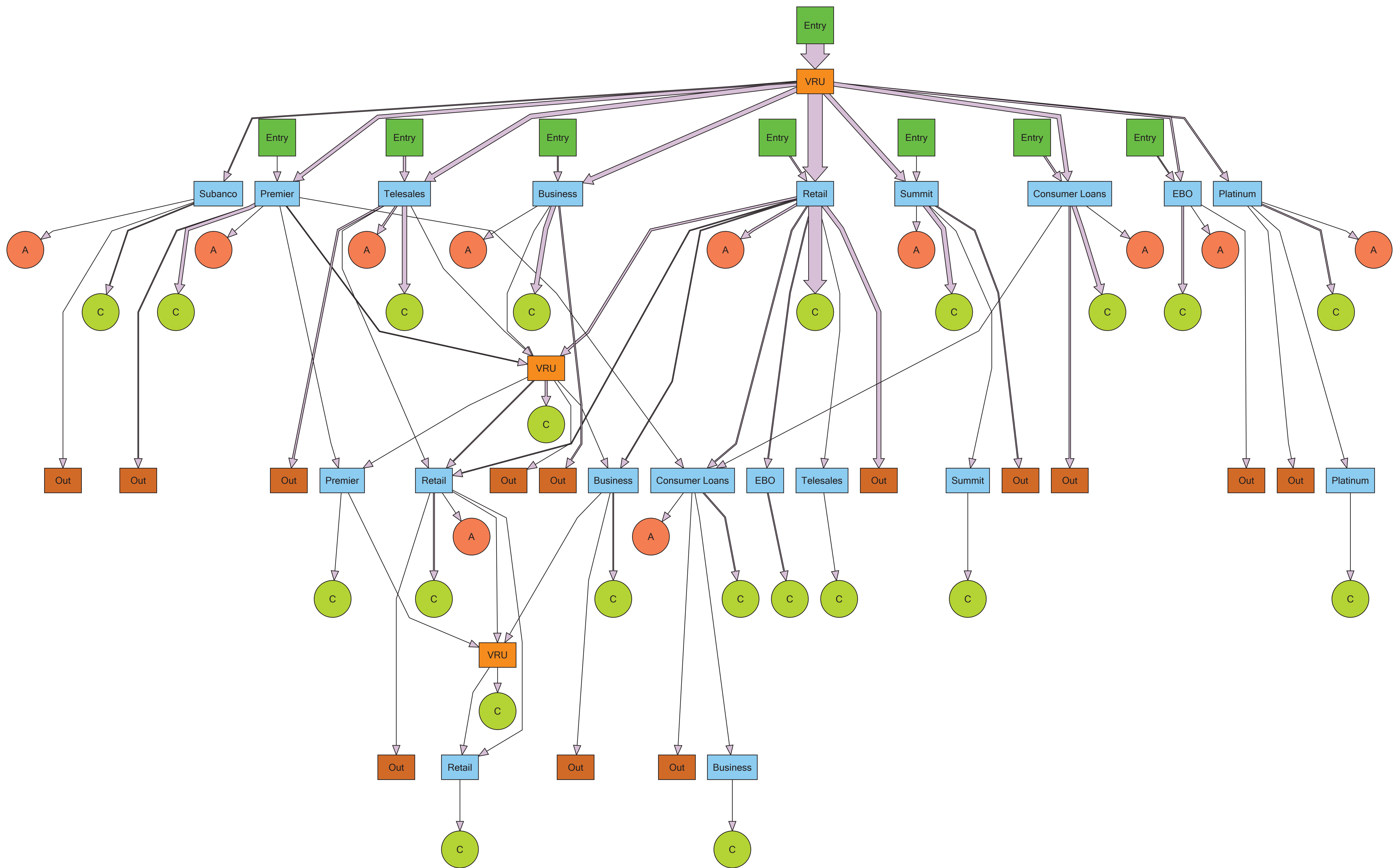




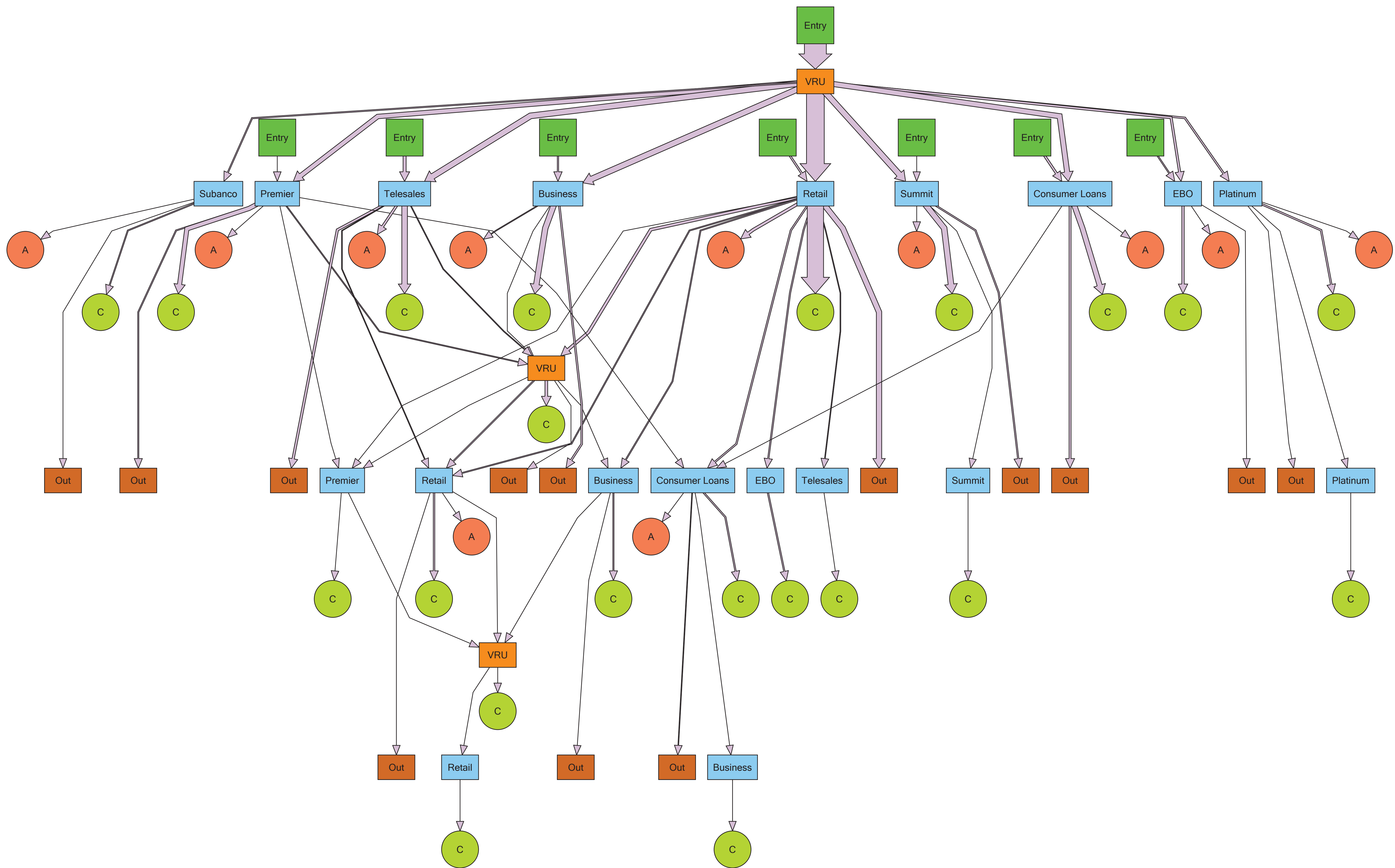
8 AM - 9 AM



9 AM - 10 AM



10 AM - 11 AM

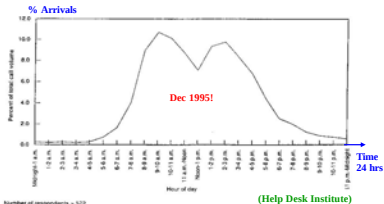


11 AM - 12 PM

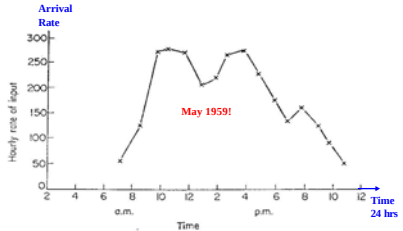
Dynamics: Time-Varying Arrival-Rates

2 Daily Peaks

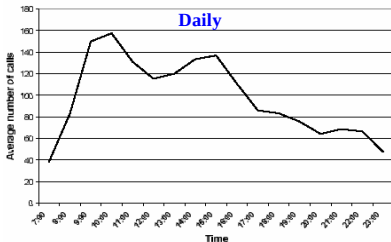
CC: Dec. 1995, (USA, 700 Helpdesks)



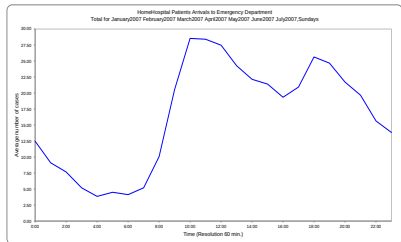
CC: May 1959 (England)



CC: Nov. 1999 (Israel)

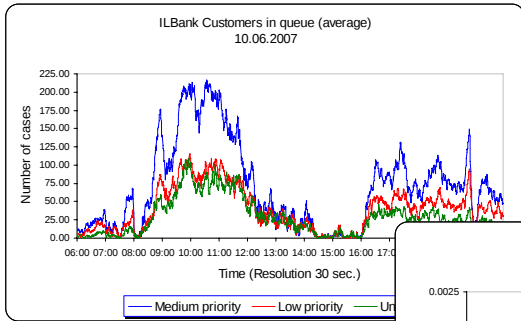


ED: Jan.—July 2007 (Israel)

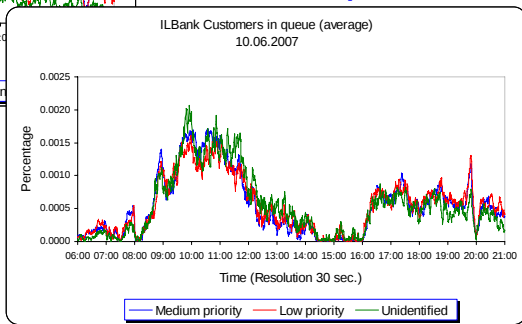


Dynamics: Parsimonious Models (Congestion Laws)

3 Queue-Lengths at 30 sec. resolution (ILBank, 10/6/2007)



Queue "Shape"

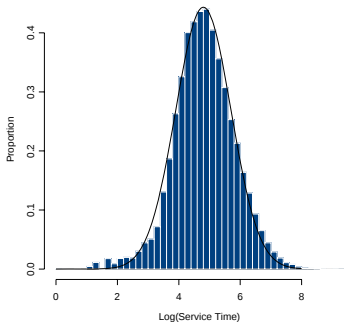


- ▶ Area normalized to 100%
- ▶ **State-Space Collapse**

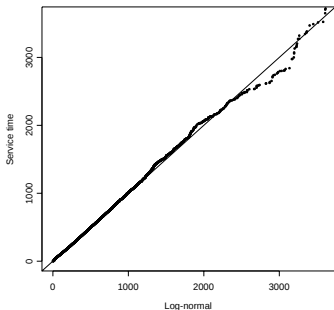
Durations: Phone Calls (2 Surprises)

Israeli Call Center, Nov–Dec, 1999

Log(Service Times)



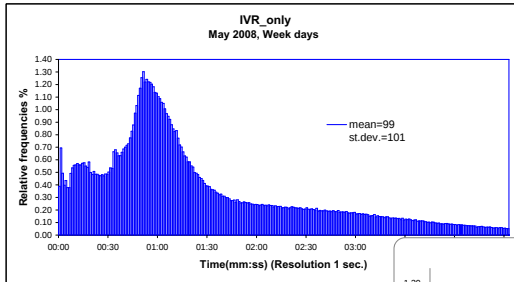
LogNormal QQPlot



- ▶ **Practically Important:** (mean, std)(log) characterization
- ▶ **Theoretically Intriguing:** Why LogNormal ? Naturally multiplicative but, in fact, also **Infinitely-Divisible** (Generalized Gamma-Convolutions)

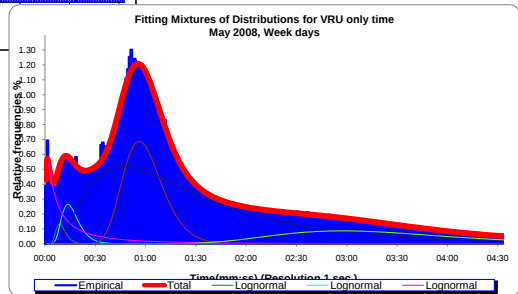
Durations: Answering Machine

Israeli Bank: IVR/VRU Only, May 2008



Mixture: 7 LogNormals

Tree-Network Tomography:
Reconstruct topology
from root and leaves

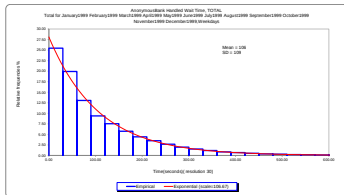


Durations: Waiting Times in a Call Center

⇒ Protocols

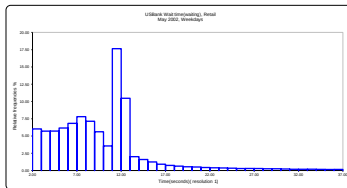
Exponential in Heavy-Traffic (min.)

Small Israeli Bank



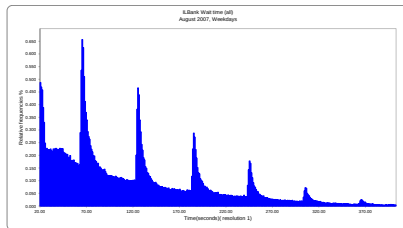
Routing via Thresholds (sec.)

Large U.S. Bank



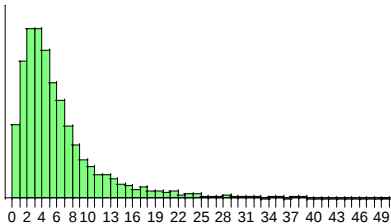
Scheduling Priorities (sec.) [compare Hospital LOS (hours)]

Medium Israeli Bank



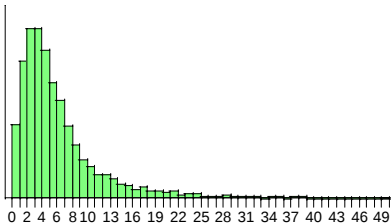
LogNormal & Beyond: Length-of-Stay in a Hospital

Israeli Hospital, in Days: LN

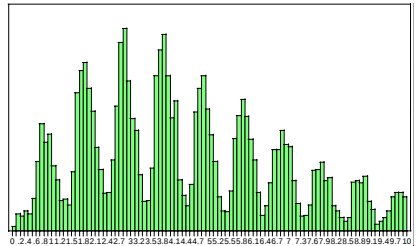


LogNormal & Beyond: Length-of-Stay in a Hospital

Israeli Hospital, in Days: LN

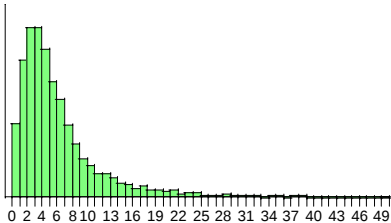


Israeli Hospital, in Hours: Mixture

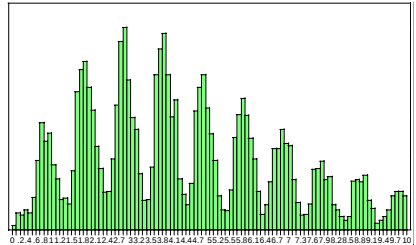


LogNormal & Beyond: Length-of-Stay in a Hospital

Israeli Hospital, in Days: LN



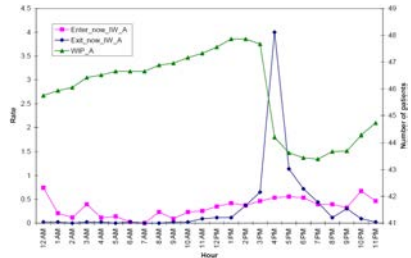
Israeli Hospital, in Hours: Mixture



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

Why Bother ?

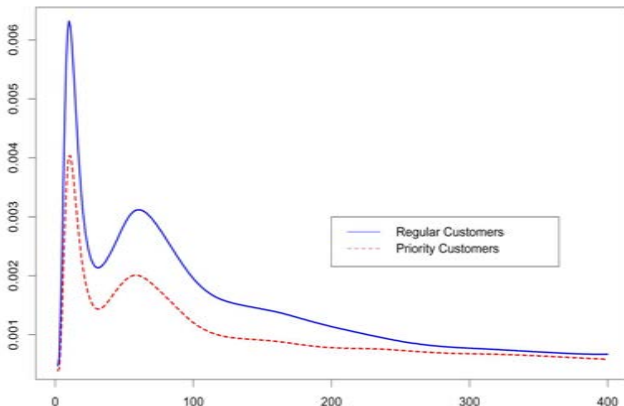
- ▶ Hourly Scale: Staffing,...
- ▶ Daily: Flow / Bed Control,...



Durations: (Im)Patience while Waiting (Psychology)

Palm: (1943–53): Irritation $\propto$ Hazard Rate

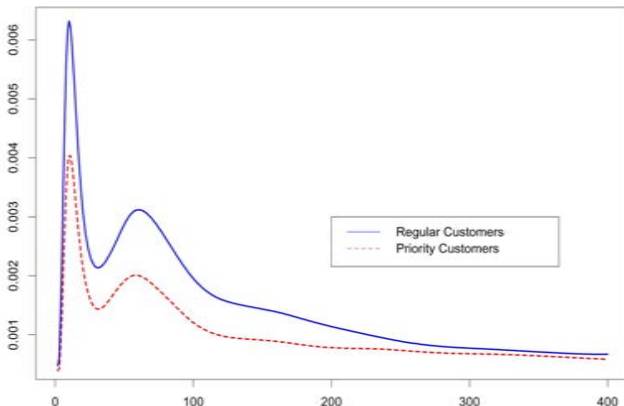
Regular over VIP Customers – Israeli Bank



Durations: (Im)Patience while Waiting (Psychology)

Palm: (1943–53): Irritation $\propto$ Hazard Rate

Regular over VIP Customers – Israeli Bank

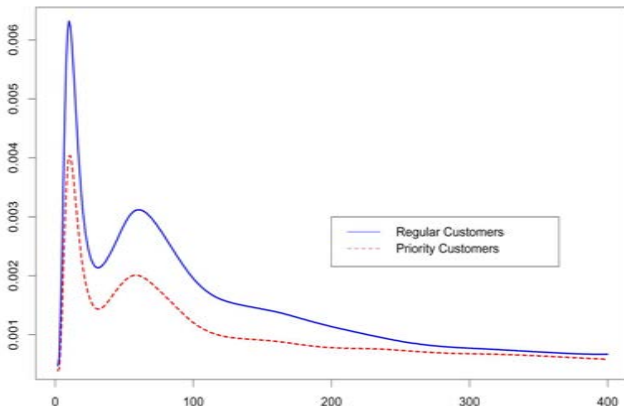


- Challenges: **Un-Censoring**, **Dependence**, **Smoothing**
- requires **Call-by-Call Data**

Durations: (Im)Patience while Waiting (Psychology)

Palm: (1943–53): Irritation $\propto$ Hazard Rate

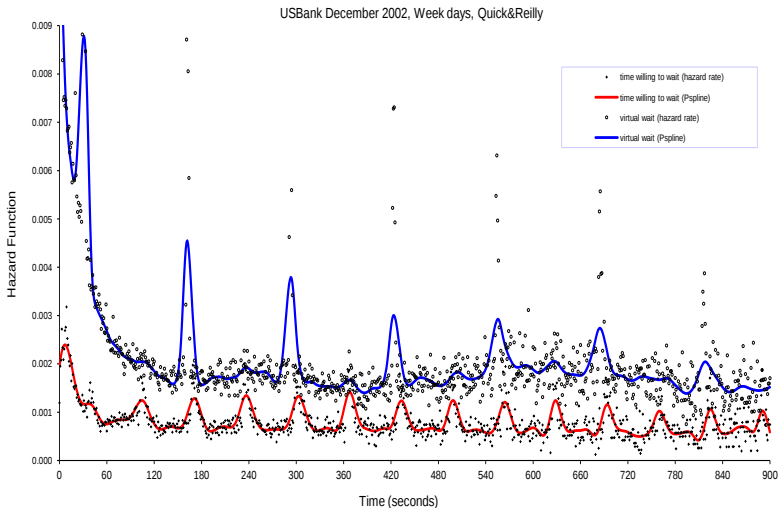
Regular over VIP Customers – Israeli Bank



- ▶ Challenges: **Un-Censoring**, **Dependence**, **Smoothing**
- requires **Call-by-Call Data**
- ▶ Here: **VIP** Customers are **more Patient** (Needy)
- ▶ **Peaks** of abandonment at times of **Announcements**

Protocols + Psychology

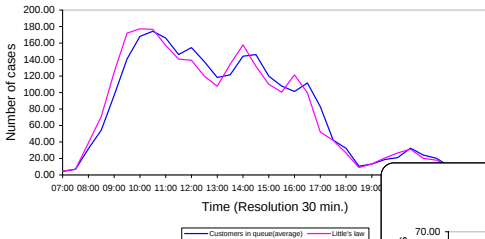
Patient Customers, Announcements, Priority Upgrades



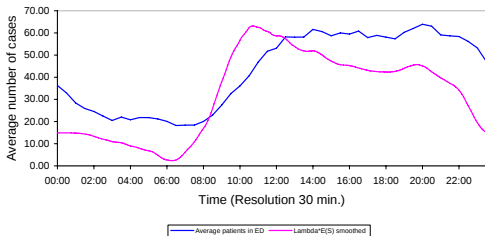
Little's Law: Call Center & Emergency Department

Time-Gap: # in **System** lags behind **Piecewise-Little** ($L = \lambda \times W$)

USBank Customers in queue(average), Telesales
10.10.2001



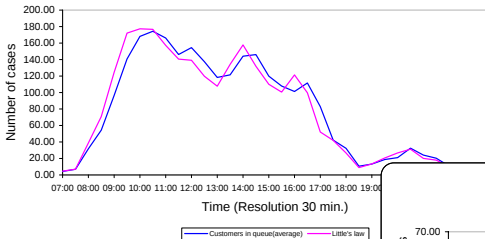
HomeHospital Average patients in ED
February 2004, Wednesdays



Little's Law: Call Center & Emergency Department

Time-Gap: # in **System** lags behind **Piecewise-Little** ($L = \lambda \times W$)

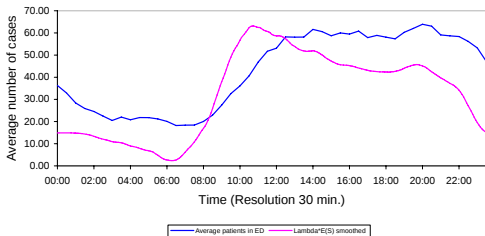
USBank Customers in queue(average), Telesales
10.10.2001



⇒ **Time-Varying Little's Law**

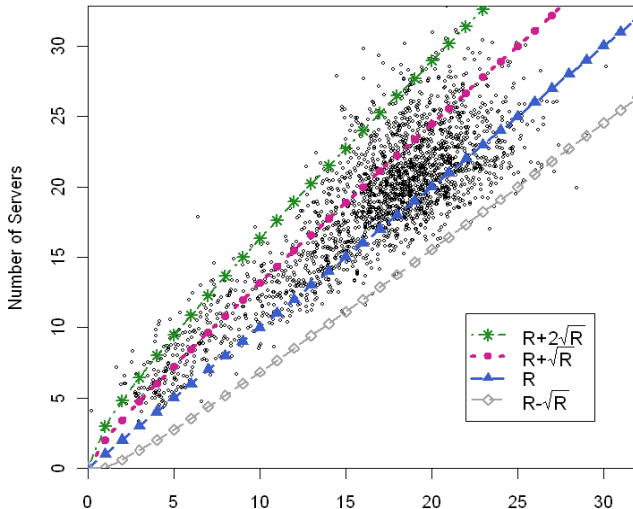
- ▶ Berstemas & Mourtzinou;
- ▶ Fralix, Riano, Serfozo; ...

HomeHospital Average patients in ED
February 2004, Wednesdays



Protocols: Staffing (N) vs. Offered-Load ($R = \lambda \times E(S)$)

IL Telecom; June-September, 2004; w/ Nardi, Plonski, Zeltyn



2205 half-hour intervals (13 summer weeks, week-days)

Technion - Israel Institute of Technology
The William Davidson Faculty of Industrial Engineering and Management
Center for Service Enterprise Engineering (SEE)
<http://ie.technion.ac.il/Labs/Serveng/>

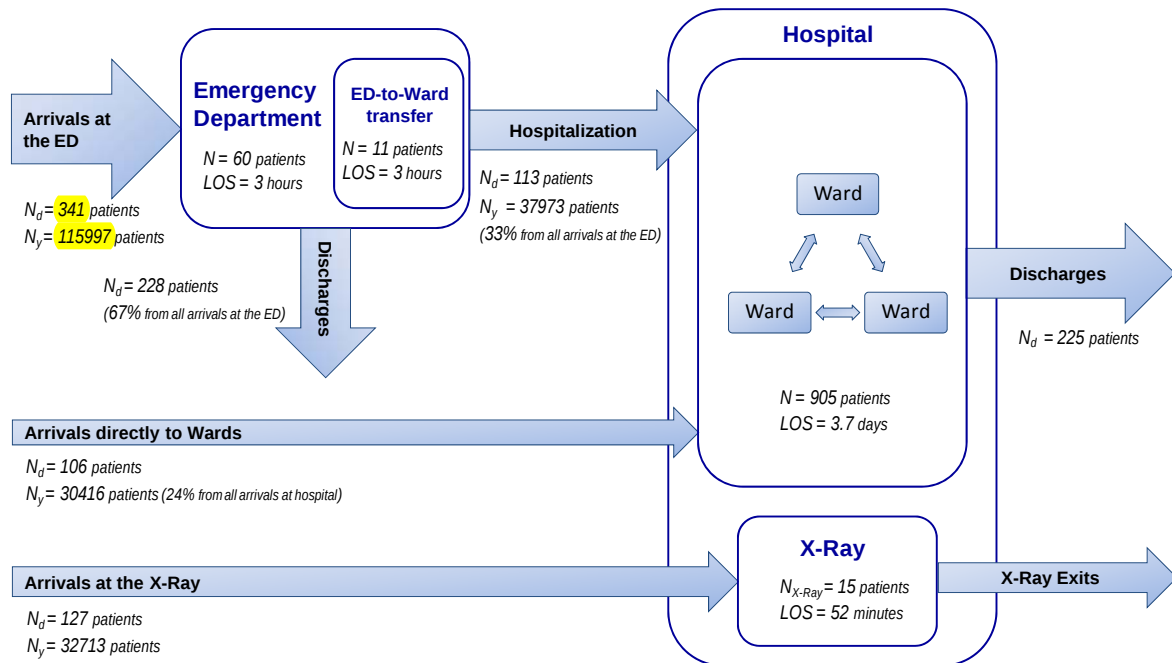


SEESat 3.0 Tutorial

August 23, 2012

HomeHospital Data

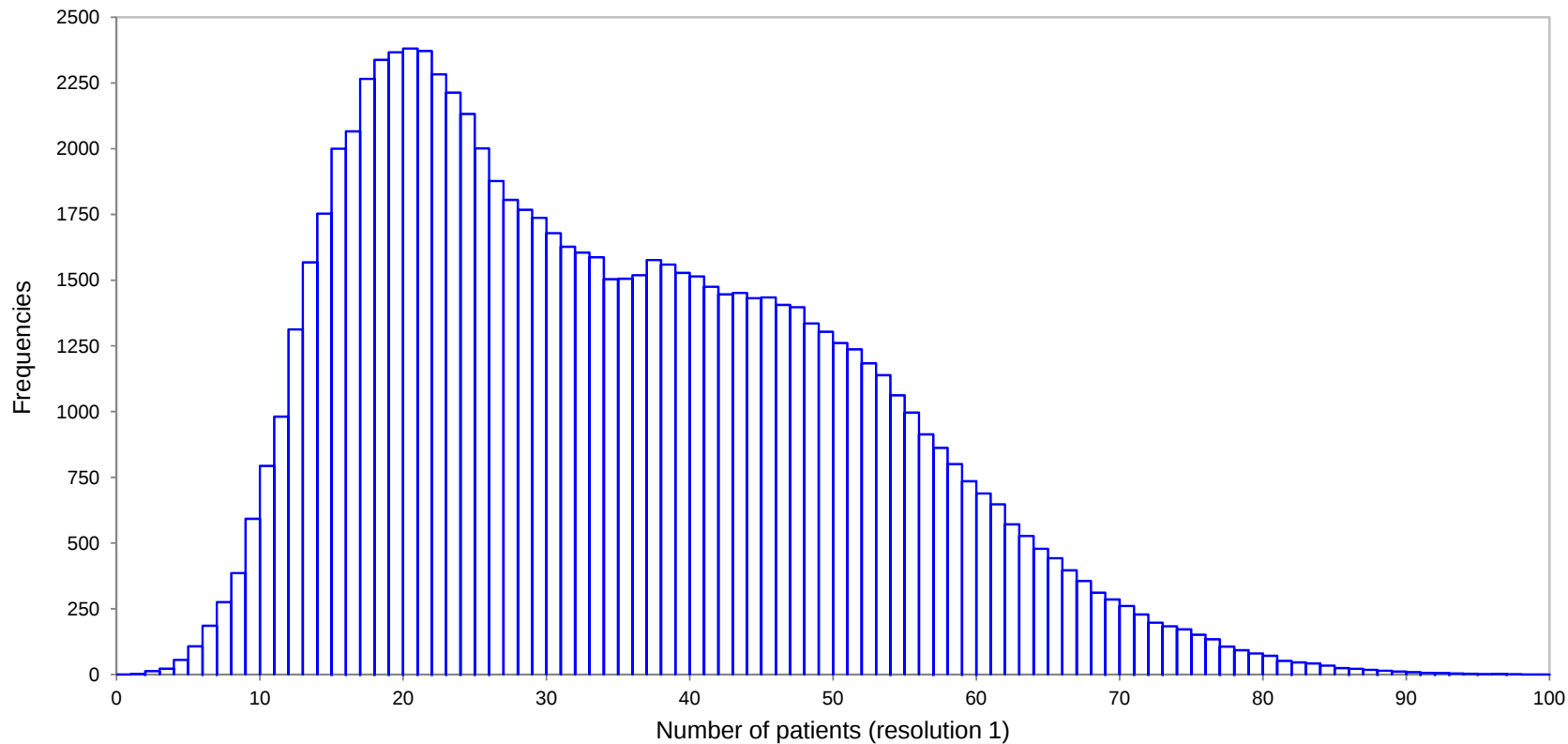
Background: The data we rely on was collected at a large Israeli hospital. This hospital consists of about 1000 beds and 45 medical units. The data includes detailed information on patient flow throughout the hospital, over a period of several years (January 2004–October 2007). In particular, the data allows one to follow the paths of individual patients throughout their stay at the hospital, including admission, discharge, and transfers between hospital units. The data does not acknowledge resolutions within the ED or within wards.



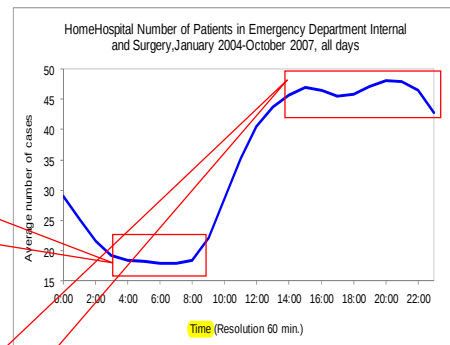
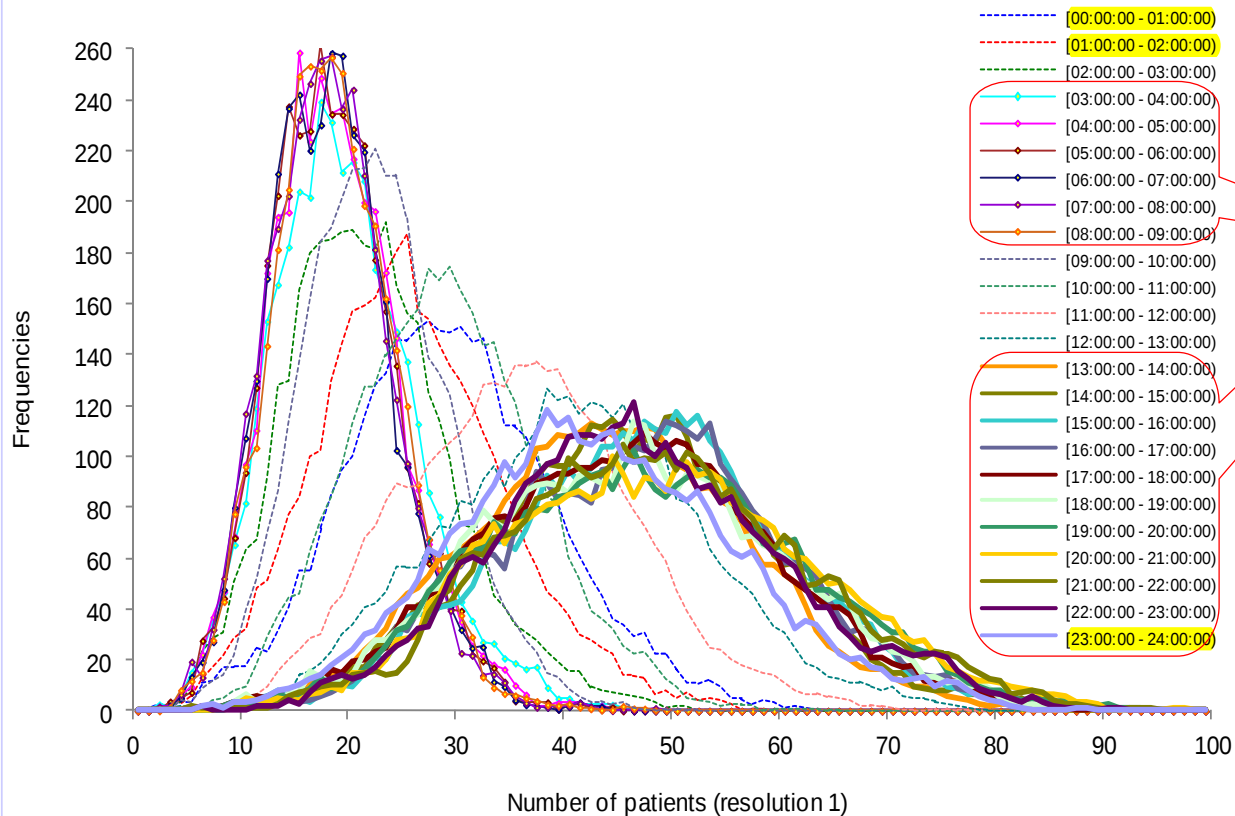
Number of patients in ED

HomeHospital Time by ED Internal and Surgery state (sec.)

January 2004-October 2007, all days



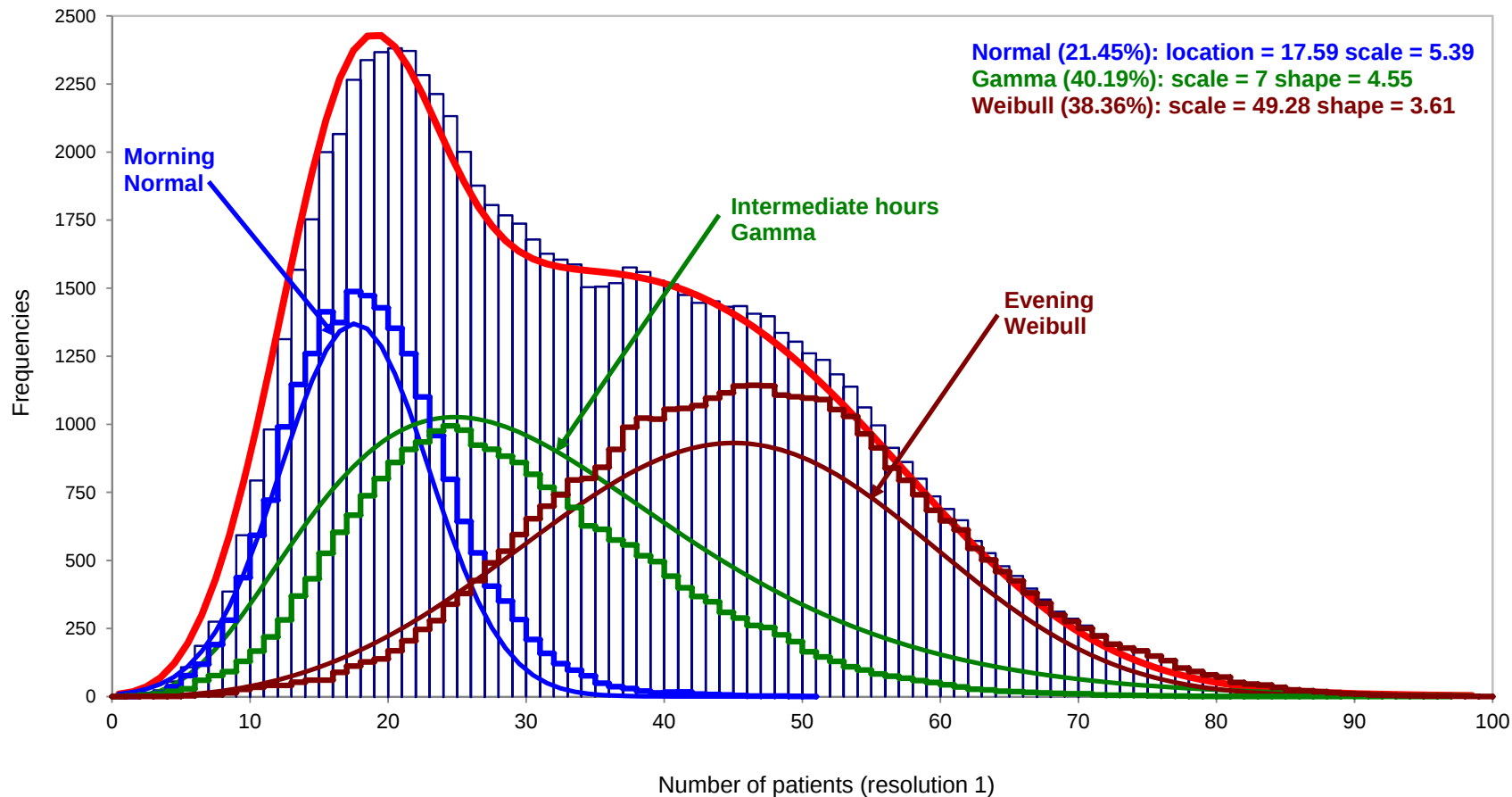
HomeHospital Time by ED Internal and Surgery state (sec.) January 2004-October 2007, all days



HomeHospital Time by ED Internal and Surgery state (sec.)

January 2004-October 2007, all days

Fitting Mixtures of Distributions



<i>Time by ED Internal and Surgery state (sec.) Statistics</i>	
N	120960000
N(average per day)	86400
Mean	34.1
Standard Deviation	16.34
Variance	266.87
Median	32
Minimum	0
Maximum	99
Skewness	0.524
Kurtosis	-0.44452
Standard Error Mean	0.00149
Interquartile Range	25
Mean Absolute Deviation	13.7
Median Absolute Deviation(MAD)	12
Coefficient of Variation (CV) (%)	47.9
L-moment 2 (half of Gini's Mean Difference)	9.25
L-Skewness	0.121
L-Kurtosis	0.0561
Coefficient of L-variation (L-CV)(%) (Gini's Coefficient)	27.12

<i>Parameter Estimates</i>						
<i>Components</i>	<i>Mixing Proportions (%)</i>	<i>Location</i>	<i>Scale</i>	<i>Shape</i>	<i>Mean</i>	<i>Standard Deviation</i>
1. Normal	21.45	17.59	5.39		17.59	5.39
2. Gamma	40.19		7.00	4.55	31.83	14.99
3. Weibull	38.36		49.28	3.61	44.42	13.679

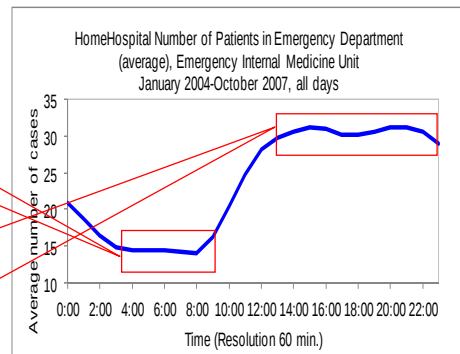
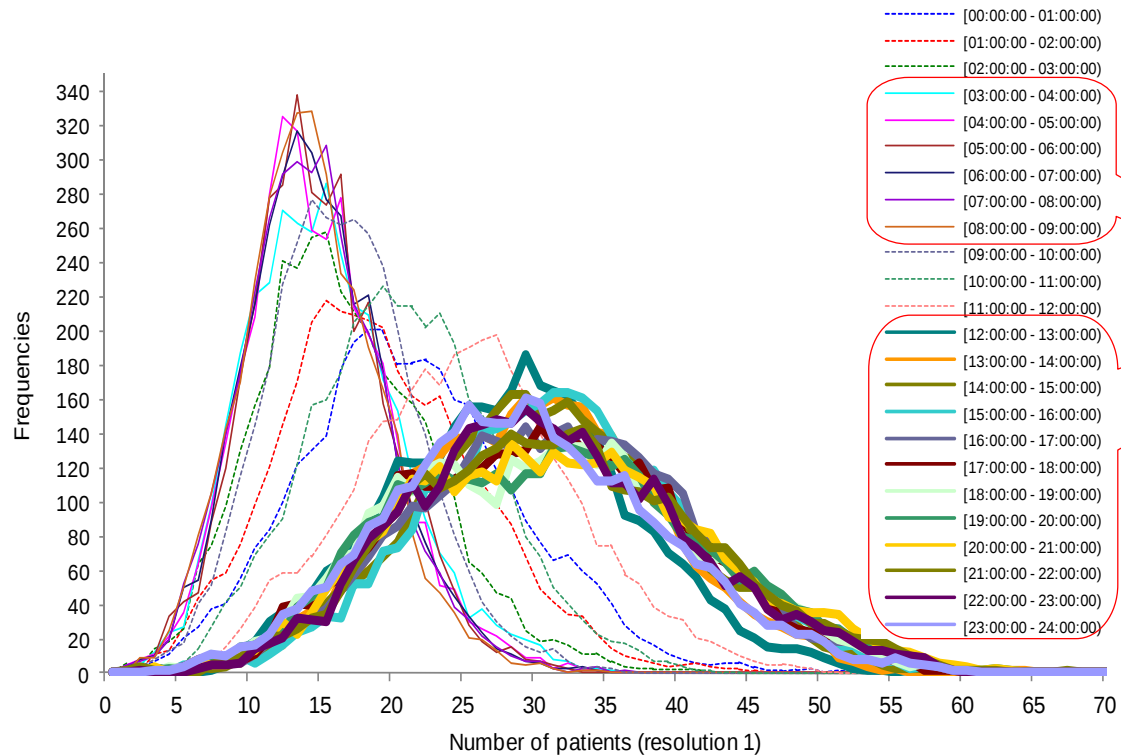
<i>Goodness-of-Fit Tests</i>			
<i>Tests</i>	<i>Statistic</i>	<i>DF</i>	<i>p Value</i>
Residuals Std	0.011		
Kolmogorov-Smirnov	0.028		<.0001
Cramer-von Mises	14953.04		<.0001
Andersen-Darling	96156.09		<.0001
Chi-Square	460741.3	94	<.0001

HomeHospital Time by ED Internal state (sec.)

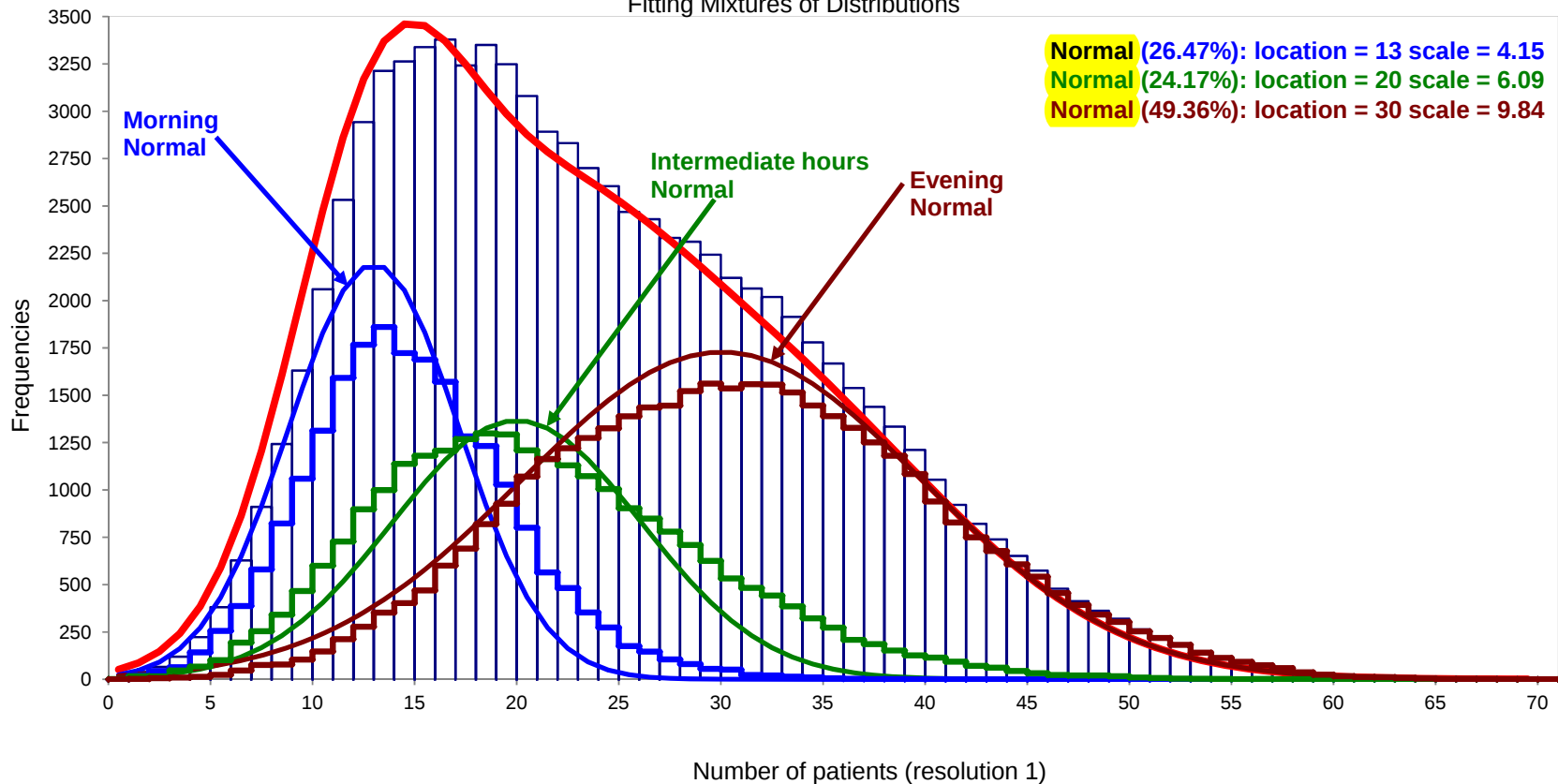
January 2004-October 2007, all days



HomeHospital Time by ED Internal state (sec.) January 2004-October 2007, all days



HomeHospital Time by ED Internal state (sec.)
January 2004-October 2007, all days
Fitting Mixtures of Distributions



Time by ED Internal state (sec.) Statistics

N	120960000
N(average per day)	86400
Mean	23.63
Standard Deviation	10.7
Variance	114.4
Median	22
Minimum	0
Maximum	70
Skewness	0.557
Kurtosis	-0.23178
Standard Error Mean	0.00097
Interquartile Range	16
Mean Absolute Deviation	8.808
Median Absolute Deviation(MAD)	8
Coefficient of Variation (CV) (%)	45.27
L-moment 2 (half of Gini's Mean Difference)	6.031
L-Skewness	0.121
L-Kurtosis	0.0813
Coefficient of L-variation (L-CV)(%) (Gini's Coefficient)	25.53

Parameter Estimates

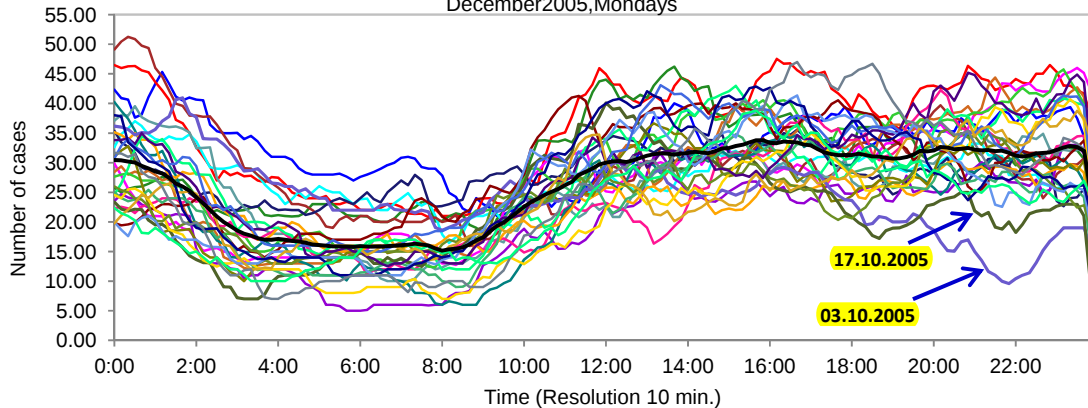
Components	Mixing Proportions (%)	Location	Scale	Shape	Mean	Standard Deviation
1. Normal	26.47	13.01	4.15		13.01	4.15
2. Normal	24.17	20.00	6.09		20.00	6.09
3. Normal	49.36	30.04	9.84		30.04	9.84

Goodness-of-Fit Tests

Tests	Statistic	DF	p Value
Residuals Std	0.017		
Kolmogorov-Smirnov	0.04		<.0001
Cramer-von Mises	34138.3		<.0001
Andersen-Darling	200657.9		<.0001
Chi-Square	584234.3	65	<.0001

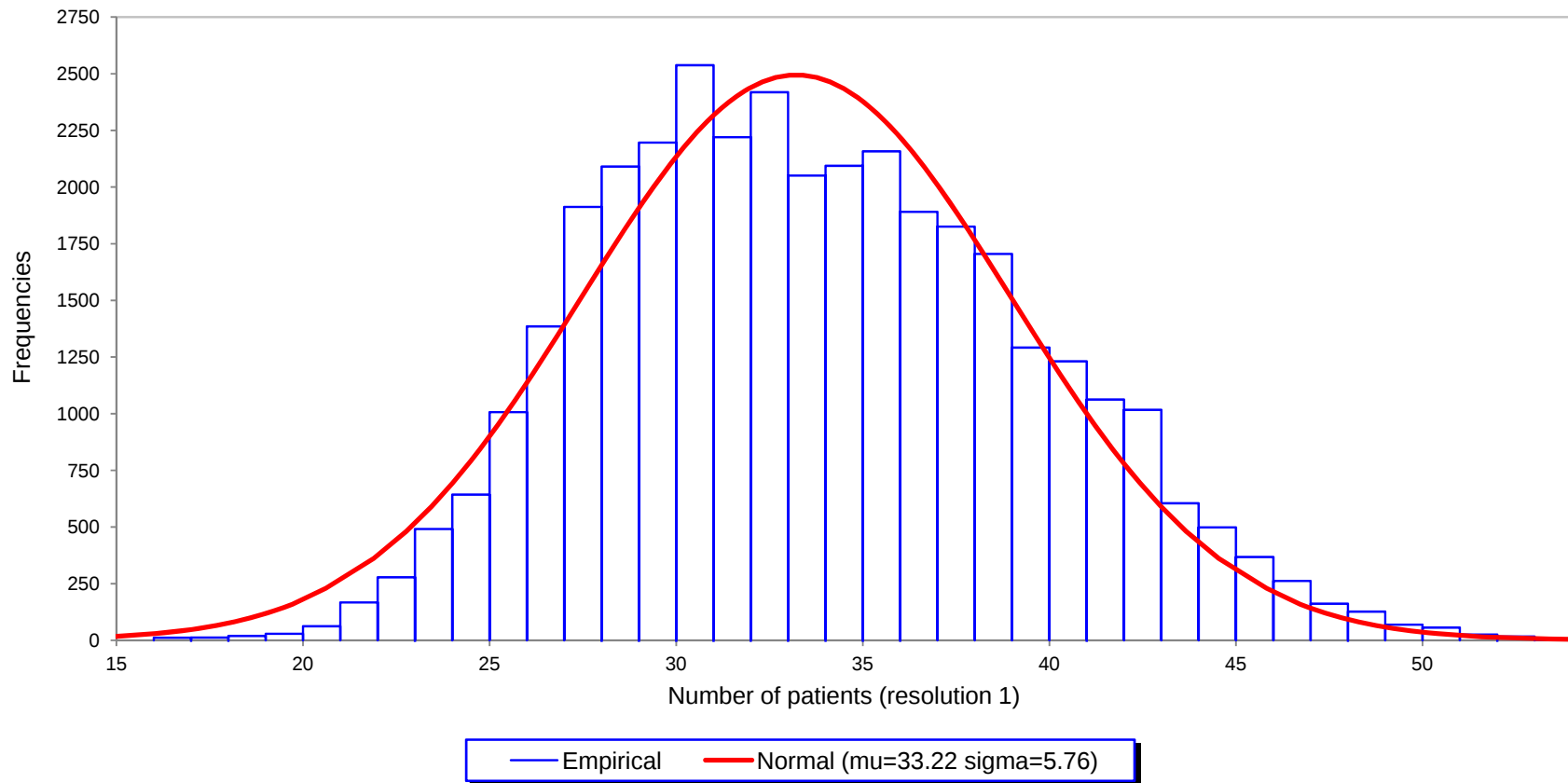
HomeHospital Number of Patients in Emergency Department (average), Emergency Internal Medicine Unit

June2005 July2005 August2005 September2005 October2005 November2005
December2005, Mondays



06.06.2005	20.06.2005	27.06.2005	04.07.2005	11.07.2005	18.07.2005
25.07.2005	01.08.2005	08.08.2005	15.08.2005	22.08.2005	29.08.2005
05.09.2005	12.09.2005	19.09.2005	26.09.2005	03.10.2005	10.10.2005
17.10.2005	24.10.2005	31.10.2005	07.11.2005	14.11.2005	21.11.2005
28.11.2005	05.12.2005	12.12.2005	19.12.2005	26.12.2005	Mondays

HomeHospital Time by ED Internal state (sec.), [13:00-23:00]
Total for January2005 February2005 March2005 April2005 May2005 June2005 July2005 August2005 September2005
November2005 December2005, Mondays



<i>Time by ED Internal state (sec.), [13:00-23:00) Statistics</i>	
N	1656000
N(average per day)	36000
Mean	33.22
Standard Deviation	5.756
Variance	33.13
Median	33
Minimum	16
Maximum	54
Skewness	0.282
Kurtosis	-0.31791
Standard Error Mean	0.00447
Interquartile Range	8
Mean Absolute Deviation	4.712
Median Absolute Deviation(MAD)	4
Coefficient of Variation(CV) (%)	17.33
L-moment 2 (half of Gini's Mean Difference)	3.263
L-Skewness	0.0601
L-Kurtosis	0.0924
Coefficient of L-variation(L-CV)(%) (Gini's Coefficient)	9.82

<i>Parameters for Normal Distribution</i>	
<i>Parameter</i>	<i>Estimate</i>
mu	33.22
sigma	5.76
mean	33.22
std	5.756

<i>Goodness-of-Fit Tests for Normal Distribution</i>			
<i>Test</i>	<i>Statistic</i>	<i>DF</i>	<i>p Value</i>
Residuals Std	0.025		
Kolmogorov-Smirnov	0.069		<.0001
Cramer-von Mises	1068.72		<.0001
Anderson-Darling	6193.27		<.0001
Chi-Square	>1000	34	<.0001

Internal ED, non-holiday Mondays, 13:00-23:00

“Recall”:

- ▶ Arrivals processes to EDs “are” inhomogeneous Poisson ($\approx$ piecewise Poisson)
- ▶ $M/M/\infty$ (ample-server queue) has Poisson steady-state
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- ▶ **U.S. ED census** $\stackrel{d}{=} M/M/B/B$, which can be (has been) used to analyze Ambulance Diversion (ED Blocking)

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- ▶ **U.S. ED census** $\stackrel{d}{=} M/M/B/B$, which can be (has been) used to analyze Ambulance Diversion (ED Blocking)
- ▶ Puzzle (getting ahead): Is the ED an Erlang-A system with $\mu = \theta$? then the “effective number of servers” can be, perhaps, deduced via those LWBS = Left Without Being Seen (or LAMA)

Simple models at the service of complex realities

Emergency-Department Network: Gallery of Models



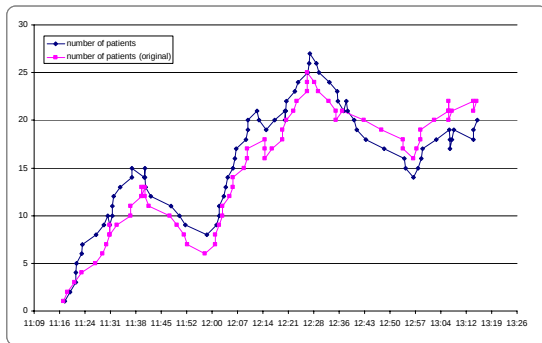
Prerequisite II: Models (FNets)

“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 (Censored LOS)

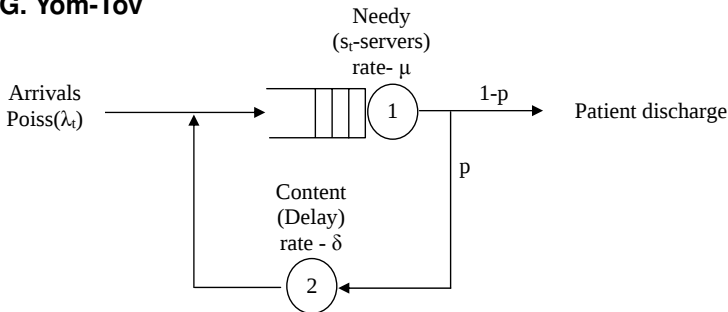


► **Transient Q's:**

- ▶ **Control** of Mass Casualty Events (w/ **I. Cohen, N. Zychlinski**)
- ▶ **Staffing** Chemical MCE (w/ **G. Yom-Tov**)

The Basic Service-Network Model: Erlang-R

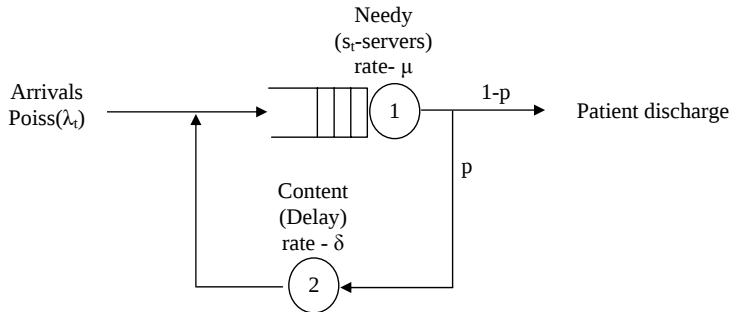
w/ G. Yom-Tov



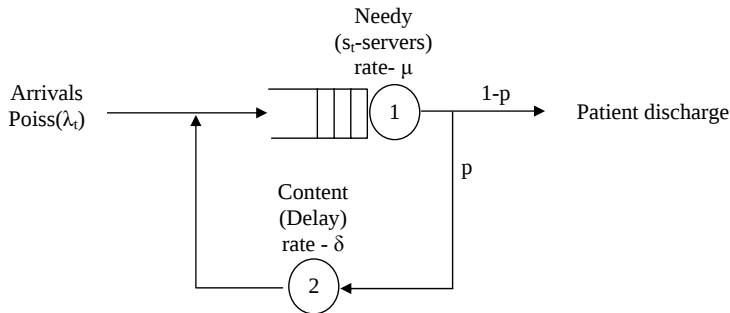
Erlang-R (IE: Repairman Problem 50's; CS: Central-Server 60's) =
2-station "Jackson" Network = (M/M/S, M/M/ ∞) :

- ▶ λ_t – **Time-Varying** Arrival rate
- ▶ S_t – Number of **Servers** (Nurses / Physicians)
- ▶ μ – **Service** rate ($E[\text{Service}] = \frac{1}{\mu}$)
- ▶ p – **ReEntrant** (Feedback) fraction
- ▶ δ – **Content-to-Needy** rate ($E[\text{Content}] = \frac{1}{\delta}$)

Fluid Model $\leftrightarrow$ (Time-Varying) Erlang-R System



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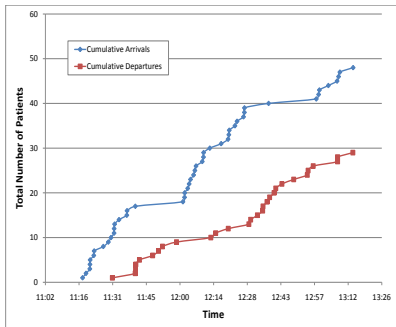
FNet of a 2-station “Jackson” Network:

$$\begin{aligned}\frac{d}{dt}q_t^1 &= \lambda_t - \mu \cdot (q_t^1 \wedge s_t) + \delta \cdot q_t^2, \\ \frac{d}{dt}q_t^2 &= p \cdot \mu \cdot (q_t^1 \wedge s_t) - \delta \cdot q_t^2.\end{aligned}\tag{1}$$

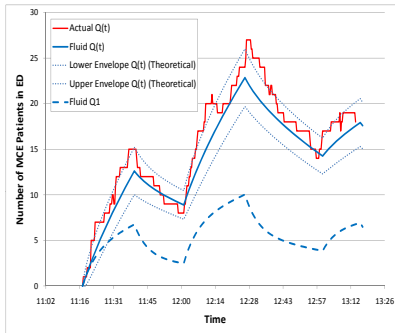
Erlang-R: Fitting a Simple Model to a Complex Reality

Chemical MCE Drill (Israel, May 2010)

Arrivals & Departures (RFID)



Erlang-R (Fluid, Diffusion)

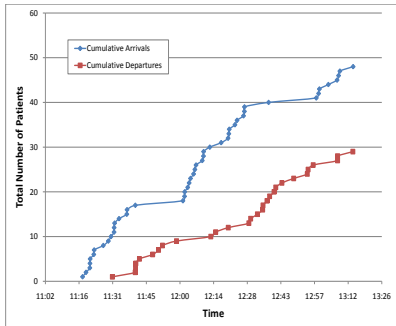


- **Recurrent/Repeated** services in MCE Events: eg. Injection every 15 minutes

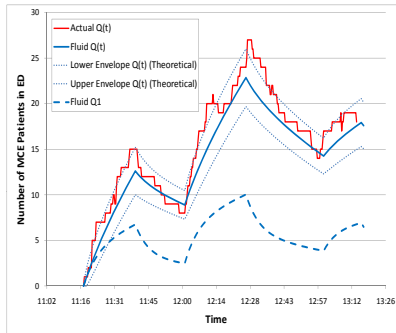
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Erlang-R (Fluid, Diffusion)



- ▶ **Recurrent/Repeated** services in MCE Events: eg. Injection every 15 minutes
- ▶ **Fluid (Sample-path)** Modeling, via Functional Strong Laws of Large Numbers
- ▶ **Stochastic** Modeling, via Functional Central Limit Theorems
 - ▶ ED in **MCE**: Confidence-interval, usefully narrow for **Control**
 - ▶ ED in **normal (time-varying)** conditions: Personnel **Staffing**

An Asymptotic Framework: Erlang-R in the ED

System = Emergency Department (eg. Rambam Hospital)

- ▶ **SimNet** = Customized ED-Simulator (Marmor & Sinreich)
- ▶ **QNet** = Erlang-R (time-varying 2-station Jackson; w/ **Yom-Tov**)
- ▶ **FNets** = 2-dim dynamical system (Massey & Whitt)
- ▶ **DNets** = 2-dim Markovian Service Net (w/ Massey and Reiman)

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Asymptotic Framework

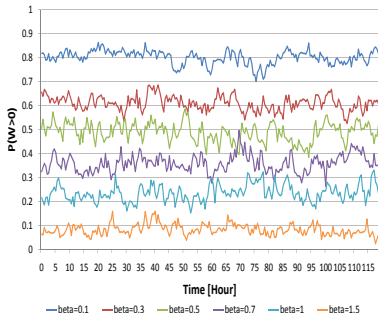
- ▶ Data and Measurements
- ▶ Fit a simple model (time-varying Erlang-R) to a complex reality (ED Physicians)
- ▶ Develop FNets (Offered-Load of Physicians) and (relevant) DNet (if needed)
- ▶ Use FNet / DNet for Design ($\sqrt{\cdot}$ -Staffing), Analysis, ...
- ▶ Simulate reality (ED with $\sqrt{\cdot}$ -staffing of Physicians)
- ▶ Validation: stable performance, confidence intervals, ...

Case Study: Emergency Ward Staffing

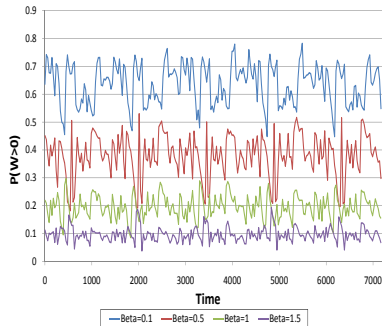
Many-Server ($\uparrow \infty$) Approximations for Small Systems (1-7)

- ▶ Staffing resolution: 1 hour
- ▶ Lower bound: 1 doctor per type
- ▶ Flexible (time-varying square-root) staffing: **Yunan's Lecture**
- ▶ Rounding effects $\Rightarrow$ Not all performance levels achievable

90 Servers

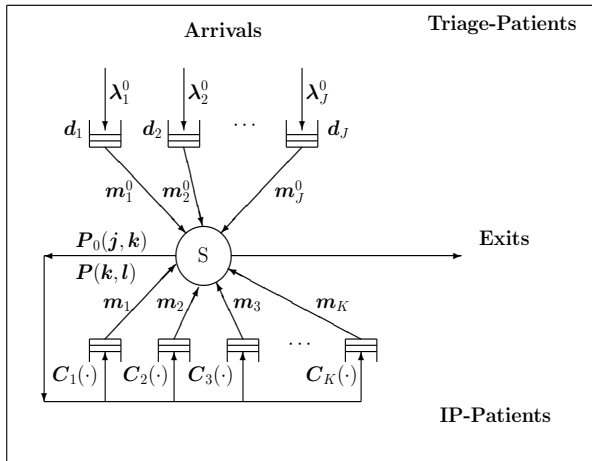


1-7 Doctors



ED Patient Flow: The Physicians View

with J. Huang, B. Carmeli; N. Shimkin



Goal: Adhere to **Triage-Constraints**, then **release In-Process** Patients

Models: Time-Varying (FNet) - in process; Stationary (DNet) - v. soon

Prerequisite II: Models (D Nets, QED 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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- ▶ **Call Centers:** Wait **"seconds"** for **minutes** service;
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- ▶ CCs: **50%** served "immediately" & **90%** utilization $\Rightarrow$ **QED**
- ▶ EDs + IWs: **?**

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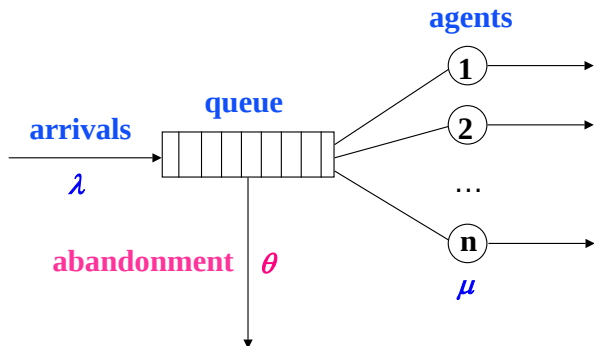
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- ▶ EDs + IWs: **?** Multiple scales! IW-"Beds" (10's) are QED while IW-Doctors (1's) are in conventional heavy-traffic (hours wait for minutes service), hence the bottlenecks

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).

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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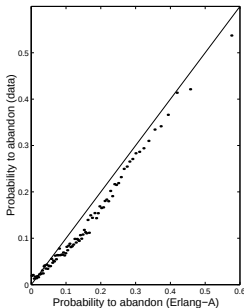
Question: **Is Erlang-A Relevant?**

YES ! Fitting a Simple Model to a Complex Reality, both **Theoretically** and **Practically**

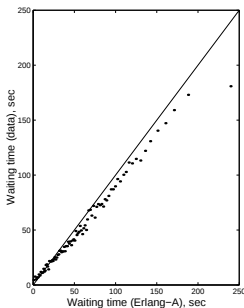
Erlang-A: Fitting a Simple Model to a Complex Reality

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

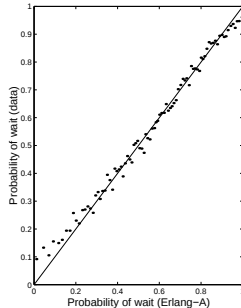
% Abandon



$E[Wait]$



$\% \{Wait > 0\}$



- ▶ Empirically-Based & Theoretically-Supported Estimation of (Im)Patience: $\hat{\theta} = P\{Ab\}/E[W_q]$
- ▶ Small Israeli Bank (more examples in progress)
- ▶ Hourly performance vs. Erlang-A predictions, 1 year: aggregated groups of 40 similar hours

QED Theory (Erlang '13; Halfin & Whitt '81; w/Garnett & Reiman '02)

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

- **QED** $\% \{\text{Cust Wait} > 0\} \approx \alpha, \quad 0 < \alpha < 1;$
or $\% \{\text{Serv Idle} > 0\} \approx 1 - \alpha$
- **Customers** $\{\text{Abandon}\} \approx \frac{\gamma}{\sqrt{N}}, \quad 0 < \gamma;$
- **Agents** $\text{OCC} \approx 1 - \frac{\beta + \gamma}{\sqrt{N}} \quad -\infty < \beta < \infty;$
- **Managers** $N \approx R + \beta\sqrt{R}, \quad R = \lambda \times E(S) \quad \text{not small};$

Here $R = \text{Offered Load}$

eg. $R = 25 \text{ call/min.} \times 4 \text{ min./call} = 100$

Erlang-A: QED Approximations (Examples)

Assume **Offered Load** R not small ($\lambda \rightarrow \infty$).

Let $\hat{\beta} = \beta \sqrt{\frac{\mu}{\theta}}$, $h(\cdot) = \frac{\phi(\cdot)}{1 - \Phi(\cdot)}$ = hazard rate of $\mathcal{N}(0, 1)$.

► **Delay Probability:**

$$P\{W_q > 0\} \approx \left[1 + \sqrt{\frac{\theta}{\mu}} \cdot \frac{h(\hat{\beta})}{h(-\hat{\beta})} \right]^{-1}.$$

► **Probability to Abandon:**

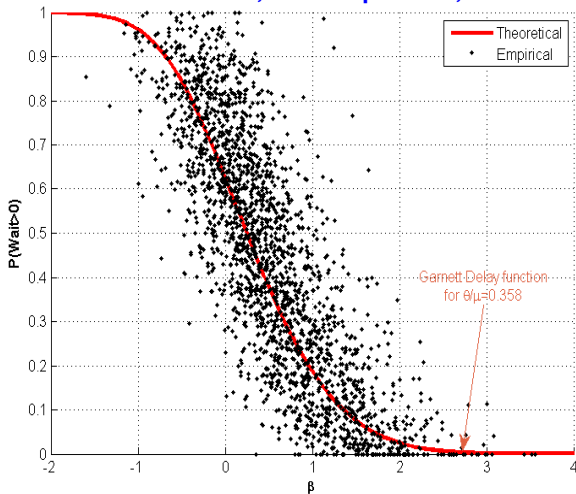
$$P\{\text{Ab} | W_q > 0\} \approx \frac{1}{\sqrt{N}} \cdot \sqrt{\frac{\theta}{\mu}} \cdot [h(\hat{\beta}) - \hat{\beta}].$$

► **$P\{\text{Ab}\} \propto E[W_q]$** , both order $\frac{1}{\sqrt{N}}$:

$$\frac{P\{\text{Ab}\}}{E[W_q]} = \theta \quad (\approx g(0) > 0).$$

QED Theory vs. Data: $P(W_q > 0)$

IL Telecom; June-September, 2004



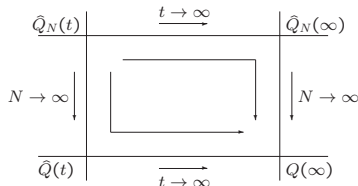
- ▶ 2205 half-hour intervals (13 summer weeks, week-days)
- ▶ Erlang-A approximations for the appropriate $\mu/\theta \approx 3$

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**

$$\hat{V}_N = \sqrt{N} V_N \Rightarrow \hat{V} = \left[\frac{1}{\mu} \hat{Q} \right]^+_{95}$$

QED Intuition: Why $P\{W_q > 0\} \in (0, 1)$?

1. Why **subtle**: Consider a large service system (e.g. call center).
 - Fix λ and let $N \uparrow \infty$: $P\{W_q > 0\} \downarrow 0$, $P(I > 0) \uparrow 1$.

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 - ▶ $\Rightarrow$ (CLT) **Square-root staffing**: $N \approx R + \beta\sqrt{R}$
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 - ▶ Steady-state: $L(M/M/N + M) \stackrel{d}{=} L(M/M/\infty) \stackrel{d}{=} \text{Poisson}(R)$, with $R = \lambda/\mu$ (Offered-Load)

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 - ▶ Steady-state: $L(M/M/N + M) \stackrel{d}{=} L(M/M/\infty) \stackrel{d}{=} \text{Poisson}(R)$, with $R = \lambda/\mu$ (Offered-Load)
 - ▶ $\text{Poisson}(R) \stackrel{d}{\approx} R + Z\sqrt{R}$, with $Z \stackrel{d}{=} N(0, 1)$.

QED Intuition: Why $P\{W_q > 0\} \in (0, 1)$?

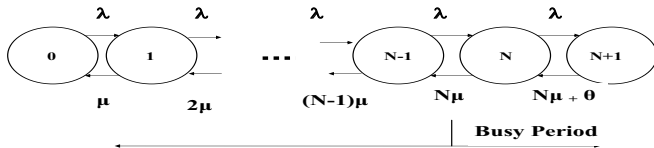
1. Why **subtle**: Consider a large service system (e.g. call center).
 - ▶ Fix λ and let $N \uparrow \infty$: $P\{W_q > 0\} \downarrow 0$, $P\{I > 0\} \uparrow 1$.
 - ▶ Fix N and let $\lambda \uparrow \infty$: $P\{W_q > 0\} \uparrow 1$, $P\{I > 0\} \downarrow 0$.
 - ▶ $\Rightarrow$ **Must** have both λ and N increase simultaneously:
 - ▶ $\Rightarrow$ (CLT) **Square-root staffing**: $N \approx R + \beta\sqrt{R}$
 $(\lambda \approx N\mu - \beta\sqrt{N\mu})$
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 $P\{Z \geq (N - R)/\sqrt{R}\} \stackrel{\sqrt{\cdot} \text{ staffing}}{\approx} P\{Z \geq \beta\} = 1 - \Phi(\beta).$

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3. QED Excursions

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

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

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

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

$$P(\text{Server Wait} > 0) = P(\text{Customer Wait} = 0)$$

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$$P(\text{Server Wait} > 0) = P(\text{Customer Wait} = 0)$$

Special case: $\mu = \theta$ (Impatient):

Then $\mathbf{Q} \stackrel{d}{=} \mathbf{M}/\mathbf{M}/\infty$, since sojourn-time is $\exp(\mu = \theta)$.

If also $\beta = 0$ (Prevalent): $P\{\text{Wait} > 0\} \approx 1/2$.

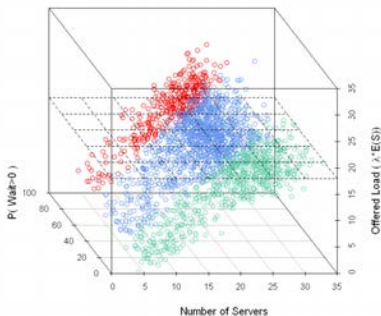
QED Erlang-X (Markovian Q's: Performance Analysis)

- ▶ Pre-History, 1914: **Erlang** (Erlang-B = $M/M/n/n$, Erlang-C = $M/M/n$)
- ▶ Pre-History, 1974: Jagerman (Erlang-B)
- ▶ History Milestone, 1981: **Halfin-Whitt** (Erlang-C, $GI/M/n$)
- ▶ Erlang-A ($M/M/N+M$), 2002: w/ **Garnett** & Reiman
- ▶ Erlang-A with General (Im)Patience ($M/M/N+G$), 2005: w/ Zeltyn
- ▶ Erlang-C (ED+QED), 2009: w/ Zeltyn
- ▶ Erlang-B with Retrial, 2010: Avram, Janssen, van Leeuwen
- ▶ Refined Asymptotics (Erlang A/B/C), 2008-2011: Janssen, van Leeuwen, Zhang, Zwart
- ▶ Production Q's, 2011: Reed & Zhang
- ▶ Universal Erlang-A, ongoing: w/ Gurvich & Huang
- ▶ Queueing Networks:
 - ▶ (Semi-)Closed: Nurse Staffing (Jennings & de Vericourt), CCs with IVR (w/ Khudiakov), Erlang-R (w/ Yom-Tov)
 - ▶ CCs with Abandonment and Retrials: w. Massey, Reiman, Rider, Stolyar
 - ▶ Markovian Service Networks: w/ Massey & Reiman
- ▶ Leaving out:
 - ▶ **Non-Exponential Service Times**: $M/D/n$ (Erlang-D), $G/Ph/n$, $\dots$, $G/GI/n+GI$, Measure-Valued Diffusions
 - ▶ **Dimensioning** (Staffing): $M/M/n$, $\dots$, time-varying Q's, V- and Reversed-V, $\dots$
 - ▶ **Control**: V-network, Reversed-V, $\dots$, SBRNets

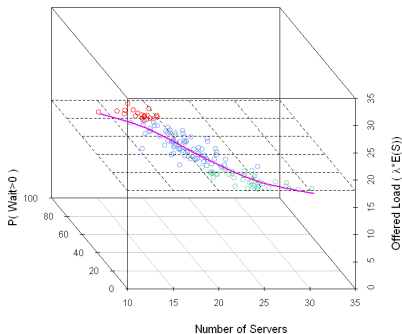
Operational Regimes: Q(uality) vs. E(fficiency)

CC in a Large Israeli Bank

$P\{W_q > 0\}$ vs. (R, N)



R-Slice: $P\{W_q > 0\}$ vs. N



3 Operational Regimes:

- ▶ **QD**: $\leq 25\%$
- ▶ **QED**: $25\% - 75\%$
- ▶ **ED**: $\geq 75\%$

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 **?** **Intuition**, as **R** or **N** increase unilaterally.

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QD Regime: $N \approx R + \delta R$, $0.1 < \delta < 0.25$ (eg. $N = 115$)

- ▶ 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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- ▶ Wait same order as service-time; $\gamma\%$ Abandon (10-25%).

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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.

Asymptotic Landscape: 9 Operational Regimes, and then some

Erlang-A, w/ I. Gurvich & J. Huang

Erlang-A μ & θ fixed	Conventional scaling			Many-Server scaling			NDS scaling		
	Sub	Critical	Over	QD	QED	ED	Sub	Critical	Over
Offered load per server	$\frac{1}{1+\delta}$	$1 - \frac{\beta}{\sqrt{n}}$	$\frac{1}{1-\gamma}$	$\frac{1}{1+\delta}$	$1 - \frac{\beta}{\sqrt{n}}$	$\frac{1}{1-\gamma}$	$\frac{1}{1+\delta}$	$1 - \frac{\beta}{n}$	$\frac{1}{1-\gamma}$
Arrival rate λ	$\frac{\mu}{1+\delta}$	$\mu - \frac{\beta}{\sqrt{n}}\mu$	$\frac{\mu}{1-\gamma}$	$\frac{n\mu}{1+\delta}$	$n\mu - \beta\mu\sqrt{n}$	$\frac{n\mu}{1-\gamma}$	$\frac{n\mu}{1+\delta}$	$n\mu - \beta\mu$	$\frac{n\mu}{1-\gamma}$
# servers	1			n			n		
Time-scale	n			1			n		
Impatience rate	θ/n			θ			θ/n		
Staffing level	$\frac{\lambda}{\mu}(1+\delta)$	$\frac{\lambda}{\mu}(1 + \frac{\beta}{\sqrt{n}})$	$\frac{\lambda}{\mu}(1-\gamma)$	$\frac{\lambda}{\mu}(1+\delta)$	$\frac{\lambda}{\mu} + \beta\sqrt{\frac{\lambda}{\mu}}$	$\frac{\lambda}{\mu}(1-\gamma)$	$\frac{\lambda}{\mu}(1+\delta)$	$\frac{\lambda}{\mu} + \beta$	$\frac{\lambda}{\mu}(1-\gamma)$
Utilization	$\frac{1}{1+\delta}$	$1 - \sqrt{\frac{\theta}{\mu}} \frac{h(\hat{\beta})}{\sqrt{n}}$	1	$\frac{1}{1+\delta}$	$1 - \sqrt{\frac{\theta}{\mu}} \frac{h(\hat{\beta})}{\sqrt{n}}$	1	$\frac{1}{1+\delta}$	$1 - \sqrt{\frac{\theta}{\mu}} \frac{h(\hat{\beta})}{n}$	1
$E(Q)$	$\frac{1}{\delta(1+\delta)}$	$\sqrt{n}g(\hat{\beta})$	$\frac{n\mu\gamma}{\theta(1-\gamma)}$	$\frac{1}{\delta} \varrho_n$	$\sqrt{n}g(\hat{\beta})\alpha$	$\frac{n\mu\gamma}{\theta(1-\gamma)}$	$o(1)$	$ng(\hat{\beta})$	$\frac{n^2\mu\gamma}{\theta(1-\gamma)}$
$P(Ab)$	$\frac{1}{n} \frac{1}{\delta} \frac{\theta}{\mu}$	$\frac{\theta}{\sqrt{n}\mu} g(\hat{\beta})$	γ	$\frac{1}{n} \frac{(1+\delta)}{\delta} \frac{\theta}{\mu} \varrho_n$	$\frac{\theta}{\sqrt{n}\mu} g(\hat{\beta})\alpha$	γ	$o(\frac{1}{n^2})$	$\frac{\theta}{n\mu} g(\hat{\beta})$	γ
$P(W_q > 0)$	$\frac{1}{1+\delta}$	≈ 1		ϱ_n	$\alpha \in (0, 1)$	≈ 1	≈ 0	≈ 1	
$P(W_q > T)$	$\frac{1}{1+\delta} e^{-\frac{\delta}{1+\delta}\mu T}$	$1 + O(\frac{1}{\sqrt{n}})$	$1 + O(\frac{1}{n})$	≈ 0		$f(T)$	≈ 0	$\frac{\Phi(\hat{\beta} + \sqrt{\theta}\mu T)}{\Phi(\hat{\beta})}$	$1 + O(\frac{1}{n})$
Congestion $\frac{EW_q}{ES}$	$\frac{1}{\delta}$	$\sqrt{n}g(\hat{\beta})$	$n\mu\gamma/\theta$	$\frac{1}{n} \frac{(1+\delta)}{\delta} \varrho_n$	$\frac{\alpha}{\sqrt{n}} g(\hat{\beta})$	$\frac{\mu\gamma}{\theta}$	$o(\frac{1}{n})$	$g(\hat{\beta})$	$n\mu\gamma/\theta$

- **Conventional:** Ward & Glynn (03, $G/G/1 + G$)
- **Many-Server:**
 - QED: Halfin-Whitt (81), Garnett-M-Reiman (02)
 - ED: Whitt (04)
 - NDS: Atar (12)

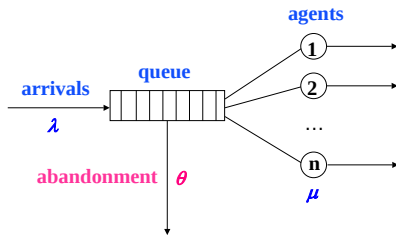
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Arrival rate λ	$\frac{\mu}{1+\delta}$	$\mu - \frac{\beta}{\sqrt{n}}\mu$	$\frac{\mu}{1-\gamma}$	$\frac{n\mu}{1+\delta}$	$n\mu - \beta\mu\sqrt{n}$	$\frac{n\mu}{1-\gamma}$	$\frac{n\mu}{1+\delta}$	$n\mu - \beta\mu$	$\frac{n\mu}{1-\gamma}$
# servers	1			n			n		
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Staffing level	$\frac{\lambda}{\mu}(1+\delta)$	$\frac{\lambda}{\mu}(1 + \frac{\beta}{\sqrt{n}})$	$\frac{\lambda}{\mu}(1-\gamma)$	$\frac{\lambda}{\mu}(1+\delta)$	$\frac{\lambda}{\mu} + \beta\sqrt{\frac{\lambda}{\mu}}$	$\frac{\lambda}{\mu}(1-\gamma)$	$\frac{\lambda}{\mu}(1+\delta)$	$\frac{\lambda}{\mu} + \beta$	$\frac{\lambda}{\mu}(1-\gamma)$
Utilization	$\frac{1}{1+\delta}$	$1 - \sqrt{\frac{\theta}{\mu}} \frac{h(\hat{\beta})}{\sqrt{n}}$	1	$\frac{1}{1+\delta}$	$1 - \sqrt{\frac{\theta}{\mu}} \frac{h(\hat{\beta})}{\sqrt{n}}$	1	$\frac{1}{1+\delta}$	$1 - \sqrt{\frac{\theta}{\mu}} \frac{h(\hat{\beta})}{n}$	1
$\mathbb{E}(Q)$	$\frac{1}{\delta(1+\delta)}$	$\sqrt{ng}(\hat{\beta})$	$\frac{n\mu\gamma}{\theta(1-\gamma)}$	$\frac{1}{\delta}\varrho_n$	$\sqrt{ng}(\hat{\beta})\alpha$	$\frac{n\mu\gamma}{\theta(1-\gamma)}$	$o(1)$	$ng(\hat{\beta})$	$\frac{n^2\mu\gamma}{\theta(1-\gamma)}$
$\mathbb{P}(Ab)$	$\frac{1}{n} \frac{1}{\delta} \frac{\theta}{\mu}$	$\frac{\theta}{\sqrt{n\mu}} g(\hat{\beta})$	γ	$\frac{1}{n} \frac{(1+\delta)}{\delta} \frac{\theta}{\mu} \varrho_n$	$\frac{\theta}{\sqrt{n\mu}} g(\hat{\beta})\alpha$	γ	$o(\frac{1}{n^2})$	$\frac{\theta}{n\mu} g(\hat{\beta})$	γ
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Congestion $\frac{EW_q}{ES}$	$\frac{1}{\delta}$	$\sqrt{ng}(\hat{\beta})$	$n\mu\gamma/\theta$	$\frac{1}{n} \frac{(1+\delta)}{\delta} \varrho_n$	$\frac{\alpha}{\sqrt{n}} g(\hat{\beta})$	$\frac{\mu\gamma}{\theta}$	$o(\frac{1}{n})$	$g(\hat{\beta})$	$n\mu\gamma/\theta$

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 - ▶ ED: Whitt (04)
 - ▶ NDS: Atar (12)
- ▶ **“Missing”:** ED+QED; Hazard-rate scaling ($M/M/N+G$); Time-Varying, Non-Parametric; Moderate- and Large-Deviation; Networks; Control

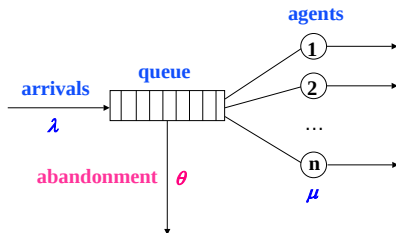
Universal Approximations: Erlang-A (M/M/N + M)

w/ I. Gurvich & J. Huang



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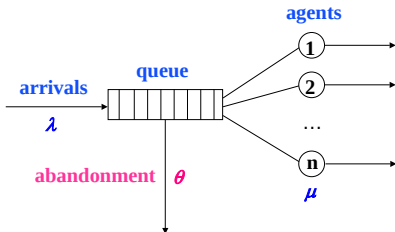


- **QNet:** Birth & Death Queue, with B - D rates

$$F(q) = \lambda - \mu \cdot (q \wedge n) - \theta \cdot (q - n)^+, \quad q = 0, 1, \dots$$

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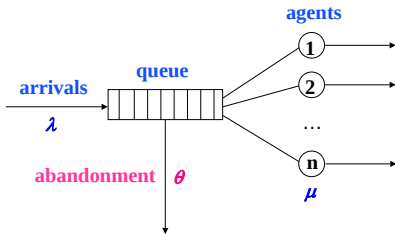
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- **FNet:** Dynamical (Deterministic) System – ODE

$$dx_t = F(x_t)dt, \quad t \geq 0$$

Universal Approximations: Erlang-A (M/M/N + M)

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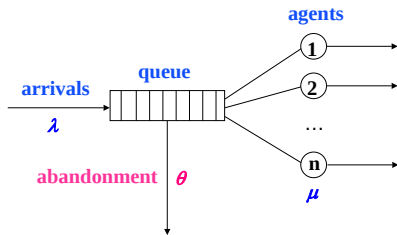
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$$dY_t = F(Y_t)dt + \sqrt{2\lambda} dB_t, \quad t \geq 0$$

Universal Approximations: Erlang-A (M/M/N + M)

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eg. $\mu = \theta$: $\dot{x} = \lambda - \mu \cdot x$, $Y = \text{OU process}$

Accuracy increases as $\lambda \uparrow \infty$ (no additional assumptions)

Value of Universal Approximation

- ▶ **Tractable** - closed-form stable expressions
- ▶ **Accurate** - more than heavy traffic limits
- ▶ **Robust** - all many-server regimes, and beyond, with hardly any assumptions
- ▶ **Value**
 - ▶ Performance Analysis
 - ▶ Optimization (Staffing)
 - ▶ Inference (**w/ G. Pang**)
 - ▶ Simulation (**w/ J. Blanchet**)
- ▶ Limitation: Steady-State (but working on it)

Why does it **work so well**?

Coupling **“Busy” + “Idle” Excursions** of B&D and the corresponding Diffusion (durations order $\frac{1}{\sqrt{\lambda}}$)

Universal Diffusion: Tractability

- Density function of $Y(\infty) - n$:

$$\pi(x) = \begin{cases} \frac{\sqrt{\mu}}{\sqrt{\lambda}} \frac{\phi(\sqrt{\mu}(x/\sqrt{\lambda} + \beta/\mu))}{\Phi(\beta/\sqrt{\mu})} p(\beta, \mu, \theta), & \text{if } x \leq 0, \\ \frac{\sqrt{\theta}}{\sqrt{\lambda}} \frac{\phi(\sqrt{\theta}(x/\sqrt{\lambda} + \beta/\theta))}{1 - \Phi(\beta/\sqrt{\theta})} (1 - p(\beta, \mu, \theta)), & \text{if } x > 0, \end{cases}$$

Here $\beta := (n\mu - \lambda)/\sqrt{\lambda}$
and

$$p(\beta, \mu, \theta) = \left[1 + \sqrt{\frac{\mu}{\theta}} \frac{\phi(\beta/\sqrt{\mu})}{\Phi(\beta/\sqrt{\mu})} \frac{1 - \Phi(\beta/\sqrt{\theta})}{\phi(\beta/\sqrt{\theta})} \right]^{-1}.$$

Universal Approximation: Accuracy

- ▶ Δ^λ is the “balancing” state, obtained by solving

$$\lambda = \mu(n \wedge \Delta^\lambda) + \theta(\Delta^\lambda - n)^+.$$

Solution: $\Delta^\lambda = \frac{\lambda}{\mu} - \left(\frac{\lambda}{\mu} - n\right)^+ \left(1 - \frac{\mu}{\theta}\right).$

Specifically: QD = $\frac{\lambda}{\mu}$; ED = $n + \frac{1}{\theta}(\lambda - n\mu)$; QED = $n + \mathcal{O}(\sqrt{\lambda})$

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Specifically: QD = $\frac{\lambda}{\mu}$; ED = $n + \frac{1}{\theta}(\lambda - n\mu)$; QED = $n + \mathcal{O}(\sqrt{\lambda})$

- ▶ Centered processes (excursions):

$$\tilde{Q}^\lambda(\cdot) = Q(\cdot) - \Delta^\lambda, \quad \tilde{Y}^\lambda(\cdot) = Y(\cdot) - \Delta^\lambda.$$

Theorem

For f bounded by an m -degree polynomial ($m \geq 0$):

$$\mathbb{E}f(\tilde{Q}^\lambda(\infty)) - \mathbb{E}f(\tilde{Y}^\lambda(\infty)) = \mathcal{O}(\sqrt{\lambda}^{m-1}).$$

- ▶ **Accuracy:** higher than heavy-traffic *limits*

Universal Approximation: Why 2λ ?

- ▶ Semi-martingale representation of the B&D process:
Fluid + Martingale
- ▶ Predictable quadratic variation:

$$\int_0^t [\lambda + \mu(Q_s \wedge n) + \theta(Q_s - n)^+] ds$$

- ▶ In **steady-state**, arrival rate $\equiv$ departure rate:

$$\lambda = \mathbb{E}[\mu(Q_s \wedge n) + \theta(Q_s - n)^+]$$

- ▶ Expectation of the predictable quadratic variation:

$$\mathbb{E} \int_0^t [\lambda + \mu(Q_s \wedge n) + \theta(Q_s - n)^+] ds = 2\lambda t$$

- ▶ $d\text{Martingale}_t \approx \sqrt{2\lambda} \cdot d\text{Brownian}_t$

Reconciling Steady-State and Time-Varying Models

- ▶ **Challenge:** Accommodate time-varying demand (routine)
- ▶ **Prerequisite:** **Flexible Capacity**
 - ▶ As in Call Centers and to a degree in Healthcare,
 - ▶ In contrast to **rigid** (fixed) staffing level during a shift: doomed to alternate between overloading and underloading

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- ▶ **History:**
 - ▶ **Jennings, M., Reiman, Whitt** (1996): Emergence of the phenomenon, via infinite-server heuristics
 - ▶ **Feldman, M., Massey, Whitt** (2008): Stabilize delay probability via QED staffing (justified theoretically only for Erlang-A with $\mu = \theta$)
 - ▶ **Liu and Whitt** (ongoing): Stabilize abandonment probability by ED staffing, via a corresponding network, theoretically and empirically
 - ▶ **Huang, Gurvich, M.** (ongoing): QED theory

The Offered-Load $R(t), t \geq 0$ ($R(t) \leftrightarrow R$)

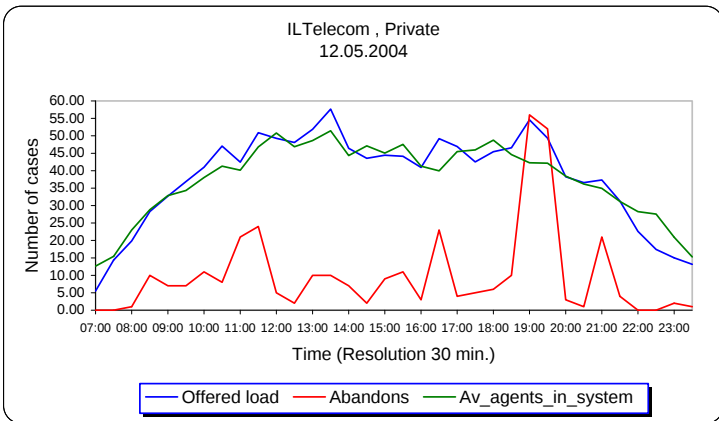
Empirically (in SEESat):

- ▶ Process: $L(\cdot) =$ **Least** number of **servers** that guarantees **no delay**.
- ▶ **Offered-Load** Function $R(\cdot) = E[L(\cdot)]$

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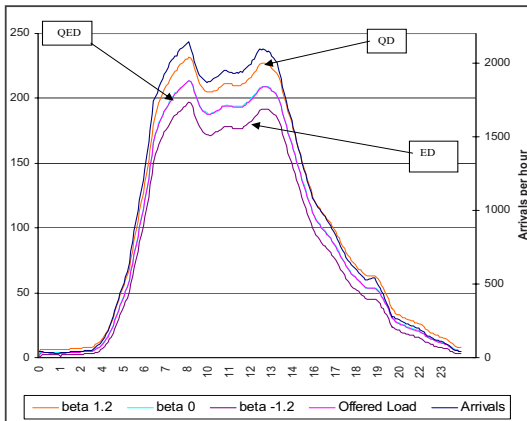
Time-Varying Arrival Rates

Square-Root Staffing:

$$N(t) = R(t) + \beta \sqrt{R(t)}, \quad -\infty < \beta < \infty.$$

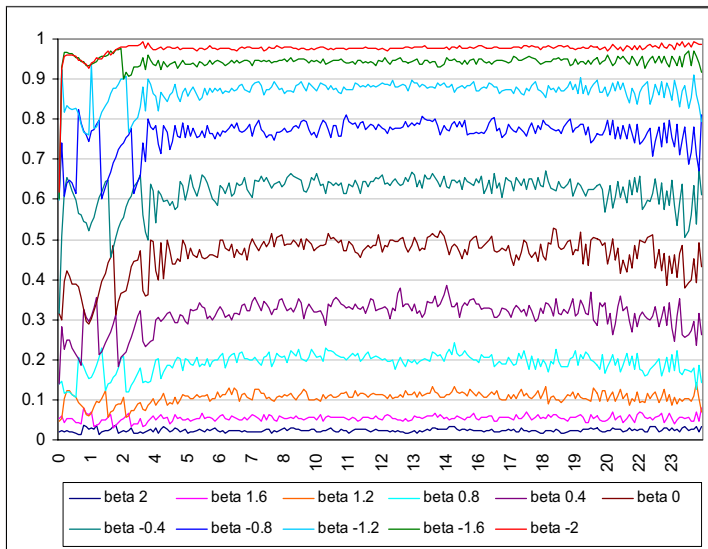
$R(t)$ is the **Offered-Load** at time t ($R(t) \neq \lambda(t) \times E[S]$)

Arrivals, Offered-Load and Staffing



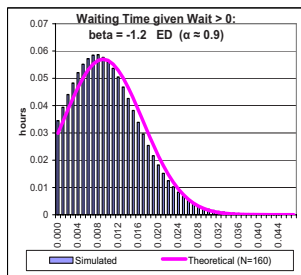
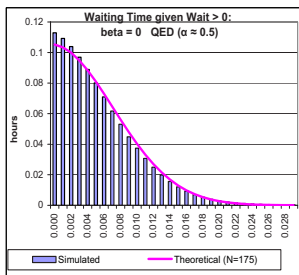
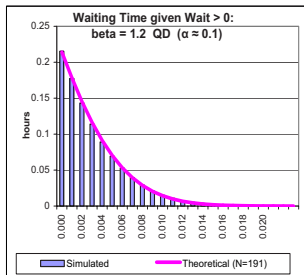
Time-Stable Performance of Time-Varying Systems

Delay Probability = as in the **Stationary Erlang-A / R**



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 (some **Averaging** Principle?)
- **Generalizes** up to a single-station within a complex network (eg.
Doctors in an Emergency Department, modeled as Erlang-R).

Calculating the Offered-Load $R(t)$, Theoretically

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Think $M_t/G/N_t^? + G$ vs. $M_t/G/\infty$: **Ample-Servers**.

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

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

- ▶ $\{A(t), t \geq 0\}$ Arrival-Process, rate $\lambda(\cdot)$;
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Stationary models: $\lambda(t) \equiv \lambda$ then $R(t) \equiv \lambda \times E[S]$.

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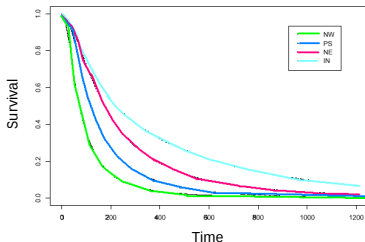
QED-c: $N_t = R_t + \beta R_t^c$, $1/2 \leq c < 1$; ($c = 1$ separate analysis).

Estimating / Predicting the Offered-Load

Must account for “**service times of abandoning customers**”.

- ▶ Prevalent Assumption: Services and (Im)Patience independent.
- ▶ But recall Patient VIPs: Willing to wait longer for more services.

Survival Functions by Type (Small Israeli Bank)



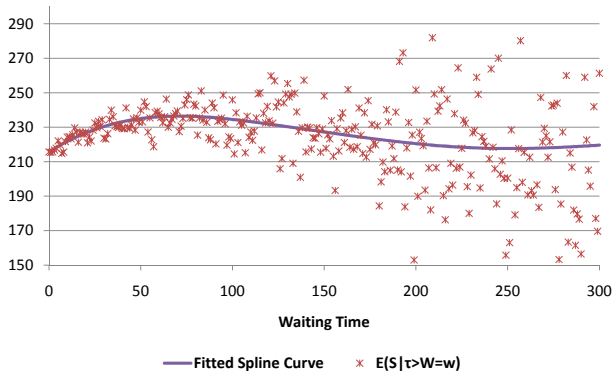
Service times stochastic order: $S_{New}^{st} < S_{Reg}^{st} < S_{VIP}^{st}$

Patience times stochastic order: $\tau_{New}^{st} < \tau_{Reg}^{st} < \tau_{VIP}^{st}$

Dependent Primitives: Service- vs. Waiting-Time

Average Service-Time as a function of Waiting-Time

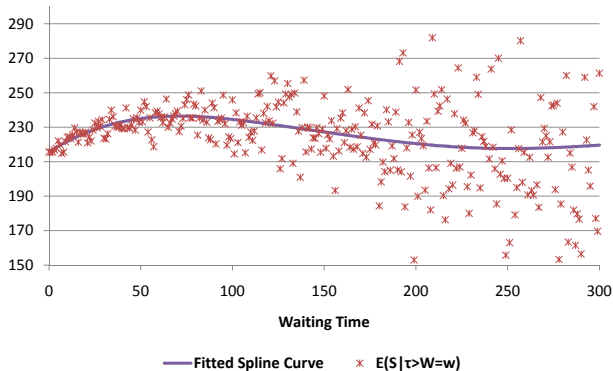
U.S. Bank, Retail, Weedays, January-June, 2006



Dependent Primitives: Service- vs. Waiting-Time

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⇒ Focus on (**Patience, Service-Time**) jointly , w/ Reich and Ritov.
 $E[S | \text{Patience} = w], w \geq 0$: **Service-Time of the Unserved.**

(Imputing) Service-Times of Abandoning Customers

w/ M. Reich, Y. Ritov:

1. **Estimate** $g(w) = E[S \mid \text{Patience} > \text{Wait} = w], w \geq 0$:

Mean service time of those **served after waiting exactly** w units of time (via non-linear regression: $S_i = g(W_i) + \varepsilon_i$);

2. **Calculate**

$$E[S \mid \text{Patience} = w] = g(w) - \frac{g'(w)}{h_\tau(w)};$$

$h_\tau(w)$ = hazard-rate of (im)patience (via un-censoring);

3. **Offered-load** calculations: Impute $E[S \mid \text{Patience} = w]$ (or the conditional distribution).

Challenges: Stable and accurate inference of g, g', h_τ .

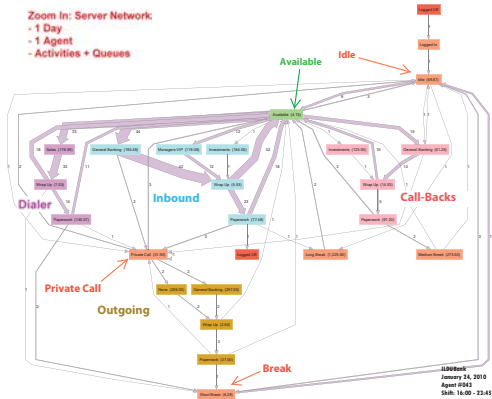
Extending the Notion of the “Offered-Load”

- ▶ **Business** (Banking Call-Center): Offered **Revenues**
- ▶ **Healthcare** (Maternity Wards): Fetus in stress
 - ▶ 2 patients (Mother + Child) = high **operational** and **cognitive** load
 - ▶ Fetus dies $\Rightarrow$ **emotional** load dominates
- ▶ $\Rightarrow$
 - ▶ Offered **Operational** Load
 - ▶ Offered **Cognitive** Load
 - ▶ Offered **Emotional** Load
 - ▶ $\Rightarrow$ **Fair** Division of Load (Routing) between 2 Maternity Wards:
One attending to complications before birth, the other to complications after birth, and both share normal birth

Server Networks

w/ A. Senderovic: **Planning**

then joined A. Gal, M. Weid, Y. Goldberg: **Process Mining**



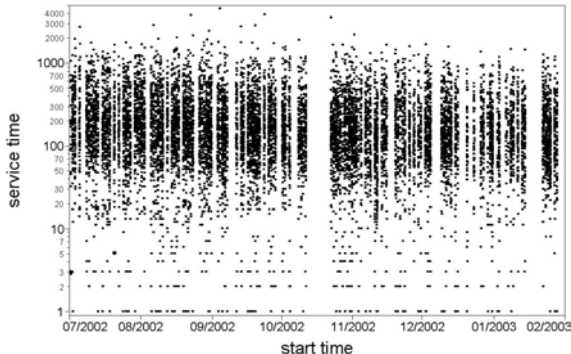
- ▶ As **Challenging** - in theory and practice - as Customer Networks
- ▶ **Uncharted territory**: e.g. Gnedenko/Palm's Machine Repair model, **Armony and Ward** (Asymptotic Little & ASTA)

Individual Agents: Service-Duration, Variability

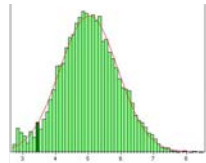
w/ Gans, Liu, Shen & Ye

Agent 14115

Service-Time Evolution: 6 month



Log(Service-Time)

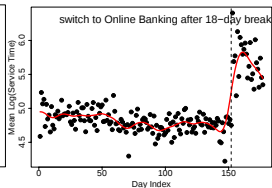
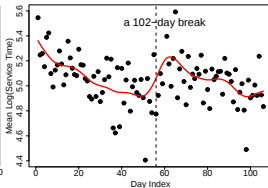
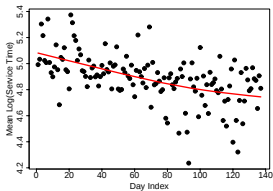


- ▶ **Learning**: Noticeable decreasing-trend in service-duration
- ▶ **LogNormal** Service-Duration, individually and collectively

Individual Agents: Learning, Forgetting, Switching

Daily-Average Log(Service-Time), over 6 months

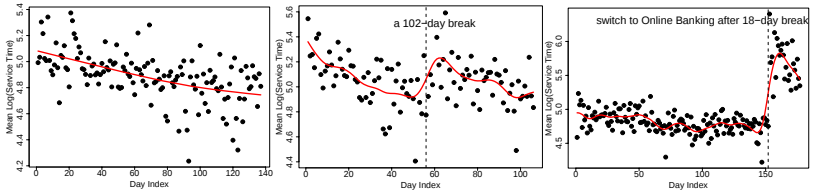
Agents 14115, 14128, 14136



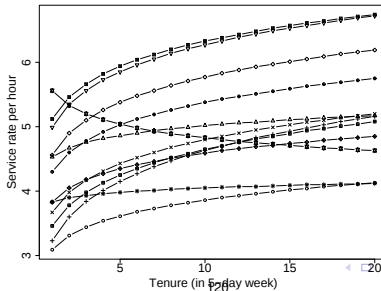
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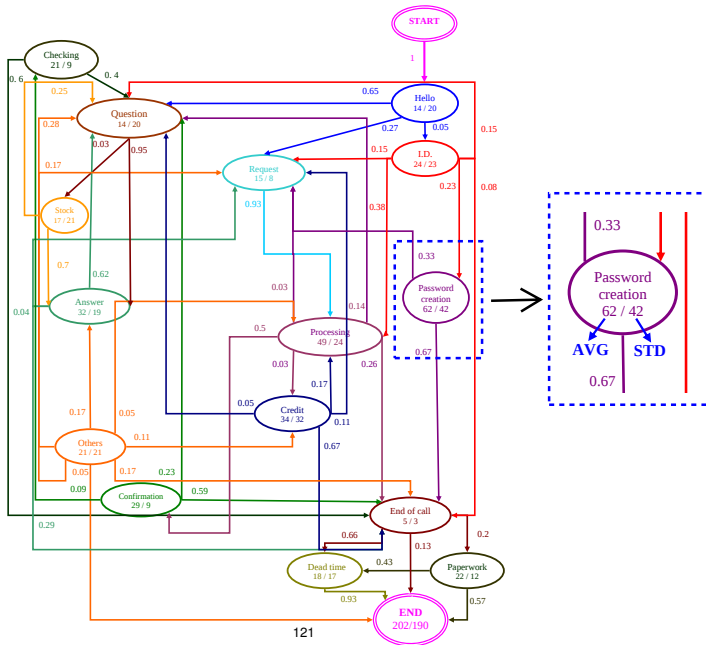
Weakly Learning-Curves for 12 Homogeneous(?) Agents



Tele-Service Process (Beyond present SEE Data)

Retail
Service
(Israeli
Bank)

Statistics
OR
IE
Psychol.
MIS



Why Bother?

- ▶ Server Networks: Uncharted Research Territory
- ▶ In large call centers:
+One Second to Service-Time implies **+Millions** in costs, annually
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Analysis yields (congestion-dependent) cross-selling protocols
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- ▶ **IVR-Process Model**: Customer-Machine Interaction
75% bank-services, poor design, yet scarce research;
Same approach, automatic (easier) data (w/ **N. Yuwiler**)

ServNets: Data-Based Online Automatic Creation

w/ [V. Trofimov](#), E. Nadjharov, I. Gavako = Technion SEELab

- ▶ **ServNets = QNets, SimNets, FNets, DNets**
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- ▶ **Moreover, address the Reproducibility Crisis in Scientific Research (Computation, Massive-data, ...)**

Data-Based Creation ServNets: some Technicalities

- ▶ ServNets = **QNets, SimNets, F Nets, DNets**
- ▶ **Graph Layout:** Adapted from but significantly extends Graphviz (AT&T, 90's); eg. *edge-width*, which must be restricted to *poly-lines*, since there are “no parallel Bezier (Cubic) curves ($B_n(p) = E_p F[B(n, p)], 0 \leq p \leq 1$)
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- ▶ Draws data directly from **SEELab** data-bases:
 - ▶ Relational DBs (Large! eg. USBank Full Binary = 37GB, Summary Tables = 7GB)
 - ▶ Structure: Sequence of events/states, which (due to size) partitioned (yet integrated) into days (eg. call centers) or months (eg. hospitals)
 - ▶ Differs from industry DBs (in call centers, hospitals, websites)