

“Theompirical” OR (IE/OM/MS), or
Theoretical & Empirical Journeys through
Service Systems (Hospitals, Call Centers, Banks,...)

Avishai Mandelbaum
IE&M & SEELab, Technion
[link](#)

Research Partners (30+ years)

Students:

Aldor*, Baron Yonit*, **Carmeli-Yuviler Nitzan**, Carmeli Boaz*, Chen Hong*, **Cohen Izik***, **Feldman Zohar***, **Garnett***, Ghebali*, **Gurvich***, Khudiakov*, Koren*, Maman*, **Marmor***, Reich*, Rosenshmidt*, Shaikhet*, **Senderovich***, Tseytlin*, **Yom-Tov***, Zaied*, Zeltyn*, **Zychlinski**, Zohar Eti*, Zviran*, ...

Theory:

Armony, Atar, Azriel, Chen Hong, Cohen Izik*, Gurvich*, Feigin, Gal, Huang Junfei*, Jelenkovic, Kaspi, Massey, Momcilovic, Reiman, Shimkin, Stolyar, Trichakis, Trofimov, Wasserkrug, Whitt, Yom-Tov*, Zeltyn*, Zhang Jiheng, Zhang Hanqin, ...

Exploratory Data Analysis, Data Sources, Statistics, Projects:

Brown, Gans, Shen Haipeng*, Sakov, Zhao Linda; Zeltyn*; Ritov, Goldberg*; Gurvich*, Huang Junfei*, Liberman*; Liu Nan, Ye Han; Armony, Marmor*, Tseytlin*, Yom-Tov*; Gorfine, Ghebali*; Tezcan; Kim Song-Hee, Won Chul Cha; Feigin, Azriel*; Rafaeli; Momcilovic, Trichakis; Bunnell, Kadish, Leib; ...

Industry:

Mizrahi Bank, Fleet Bank, Rambam Hospital, IBM Research, Hapoalim Bank, Pelephone Cellular, Samsung Hospital, Dana Farber Cancer Institute, LivePerson, CheetahLabs,...

Technion SEE Laboratory (SEELab):

Feigin; Trofimov, Nadjharov, Gavako; Kutsy; Senderovic*, Carmeli*; Liberman*, Koren*, Plonsky*; Research Assistants, Visitors, Postdocs, ...



Note: In many Western countries, there is a short list of popular “first names,” but countless “last names.” In China, it is just the reverse. The list of last names is short, and the number of first names is in the billions (from chinapage.com/biography/lastname.html).

2 Prerequisites: Data, Models

- **Data:** The language of
 - ✓ Multi-disciplinary research (e.g. OR + Psychology + NLP)
 - ✓ Academia-Industry partnerships (e.g. university & bank)
- **Models:** Simple models at the service of complex realities (not too simple)
 - ✓ Insights often rooted in deep mathematics
 - ✓ Stochastic networks: Empirical; QNets, **FNets**, **DNets**; **SimNets**

Service Engineering (& Science)

Background: Math, CS, Stat, OR;
Science, Management, Engineering

Playground: Service Systems (Call Centers, Hospitals, Banks,...)

Theoretical Framework: Asymptotics (of Queueing/Stochastic Networks)

Empirical Support: Technion SEELab

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 (**QNets**), 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: Accuracy vs. Value

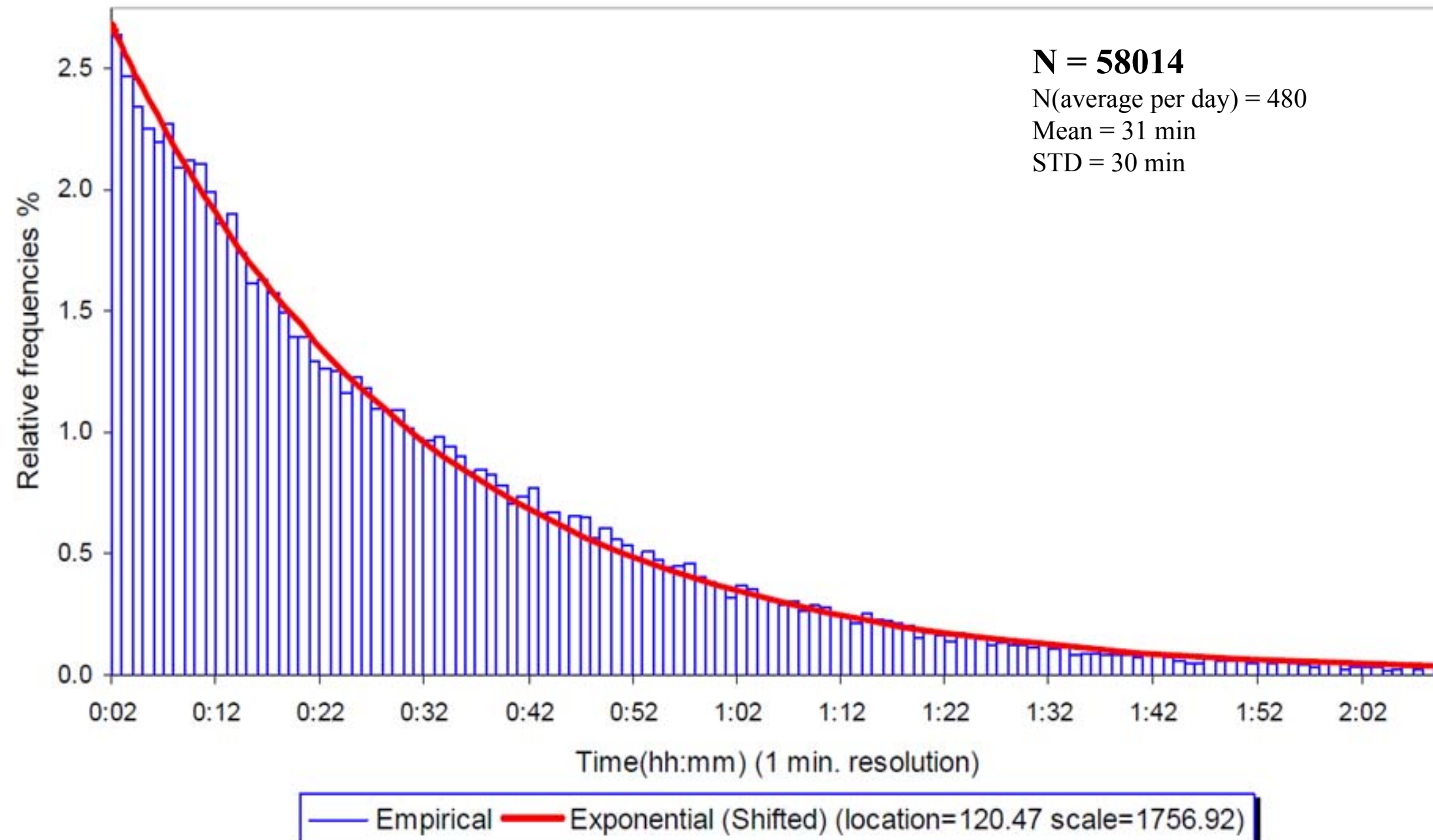
- Has asymptotic research helped one approximate or simulate a service system more efficiently, estimate its parameters more accurately, teach it to our students more effectively, perhaps even manage the system better?
- I am of the opinion that the answer to such questions **could and should have been “yes”** more often than it has been.
- How to change this? And why be optimistic?
My approach has been to **marry theory with data (theompirical research)**, supported by (what I only recently came to realize is) **Process Mining (= of Stochastic Networks**: their building-blocks, structure, protocols; flows and laws).

Pause for a Commercial (Behavioral OR) Mining Service Protocols

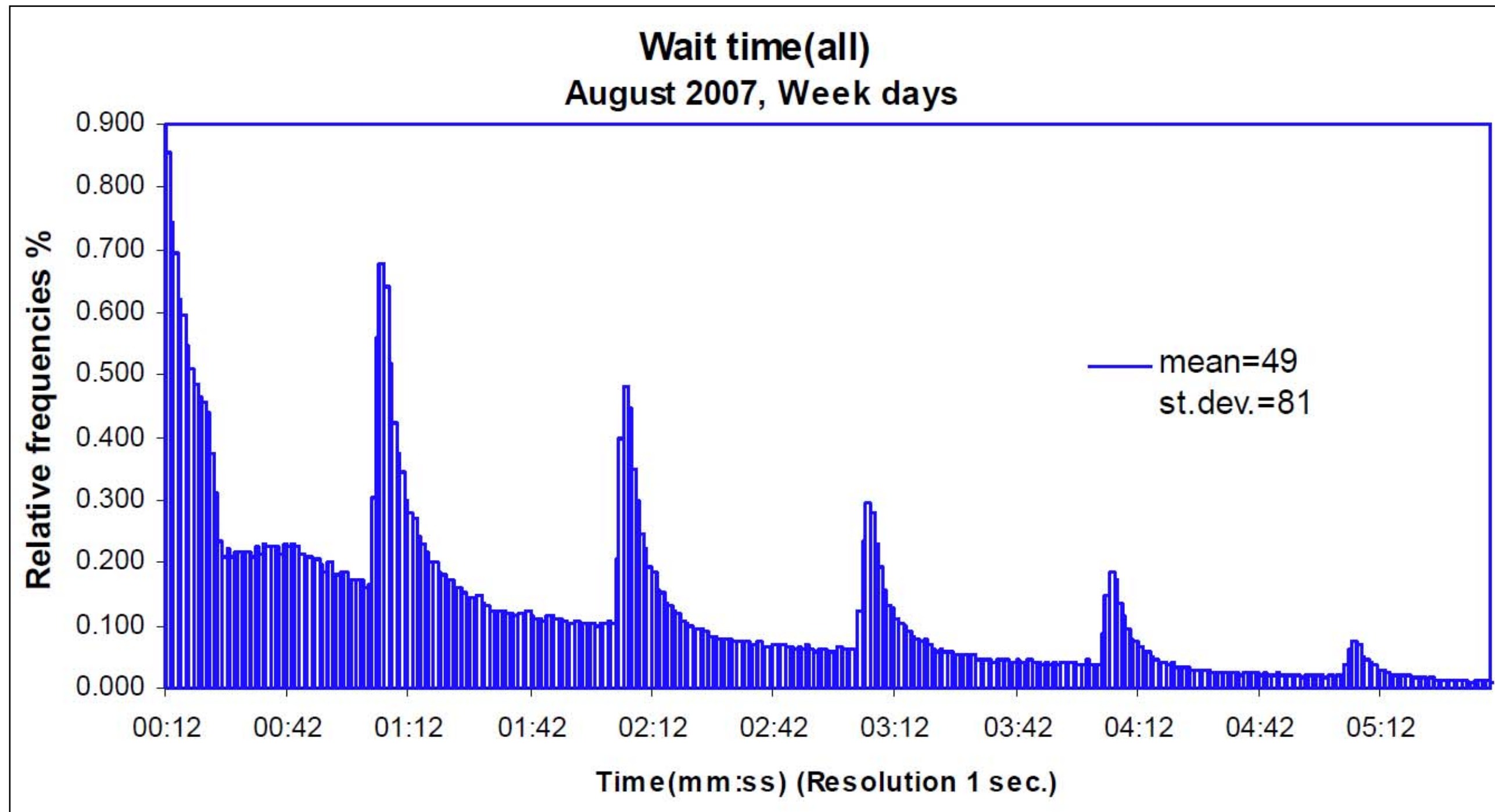
- FCFS
- Dynamic priority
- Random order (mixed with FCFS: exposes data limitations - must be there to figure it out)

Waiting-Time for Exam: Single-Server Queue (in HT)

November 2013–May 2014

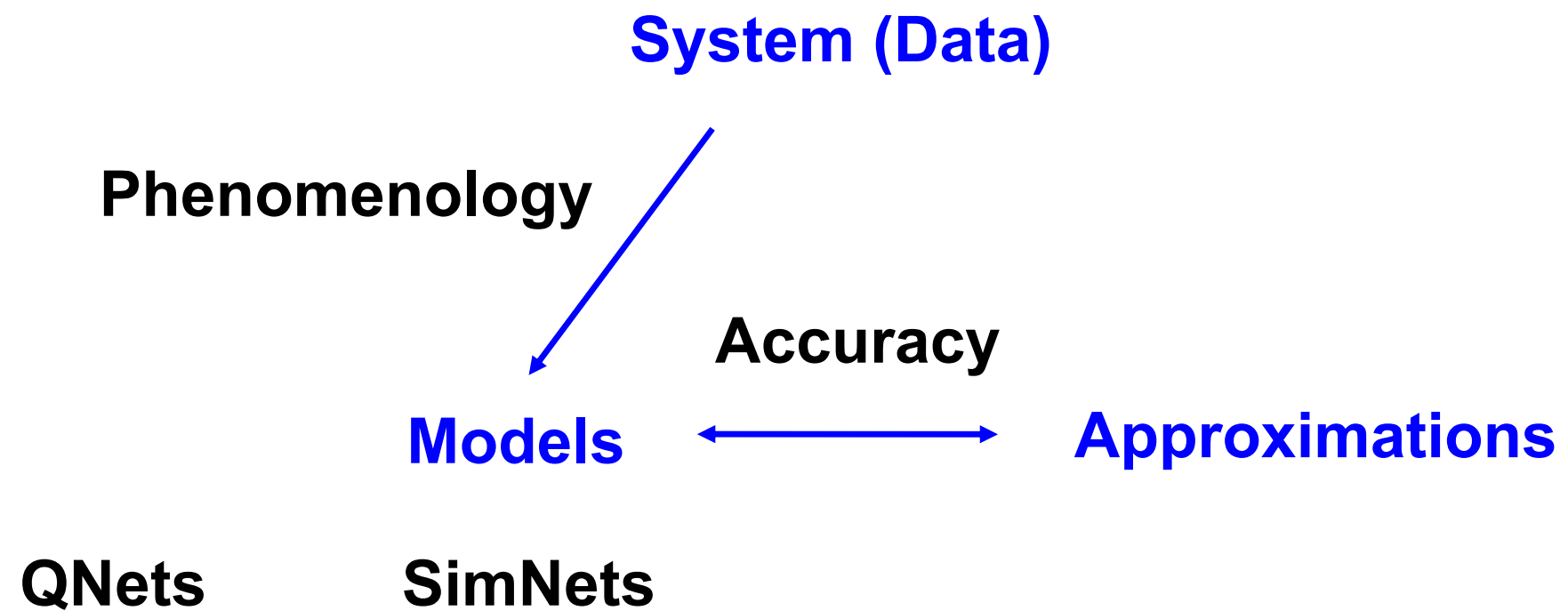


Distribution of Waiting Time on the Phone

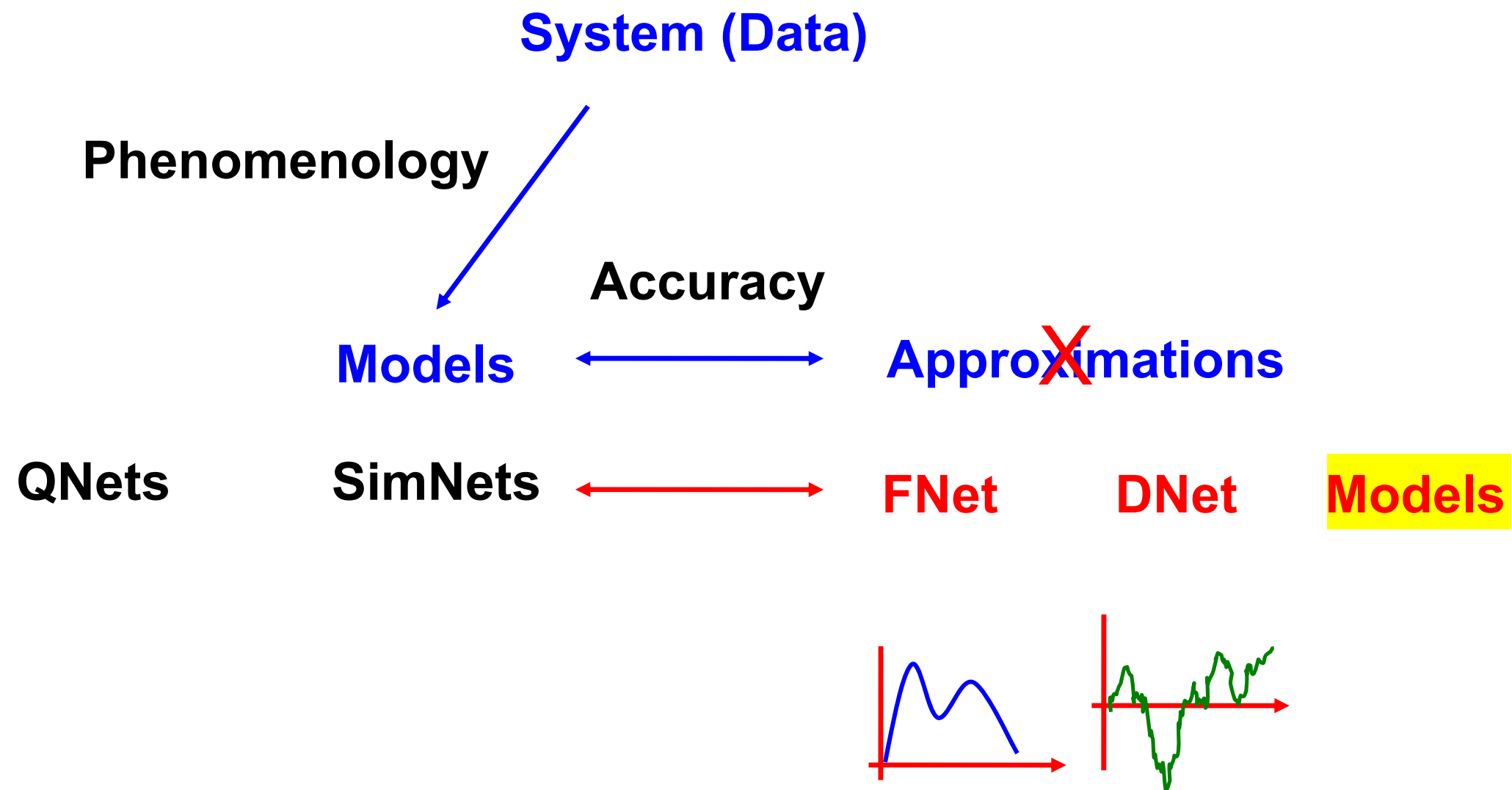


Back to Queueing Asymptotics: Accuracy vs. Value

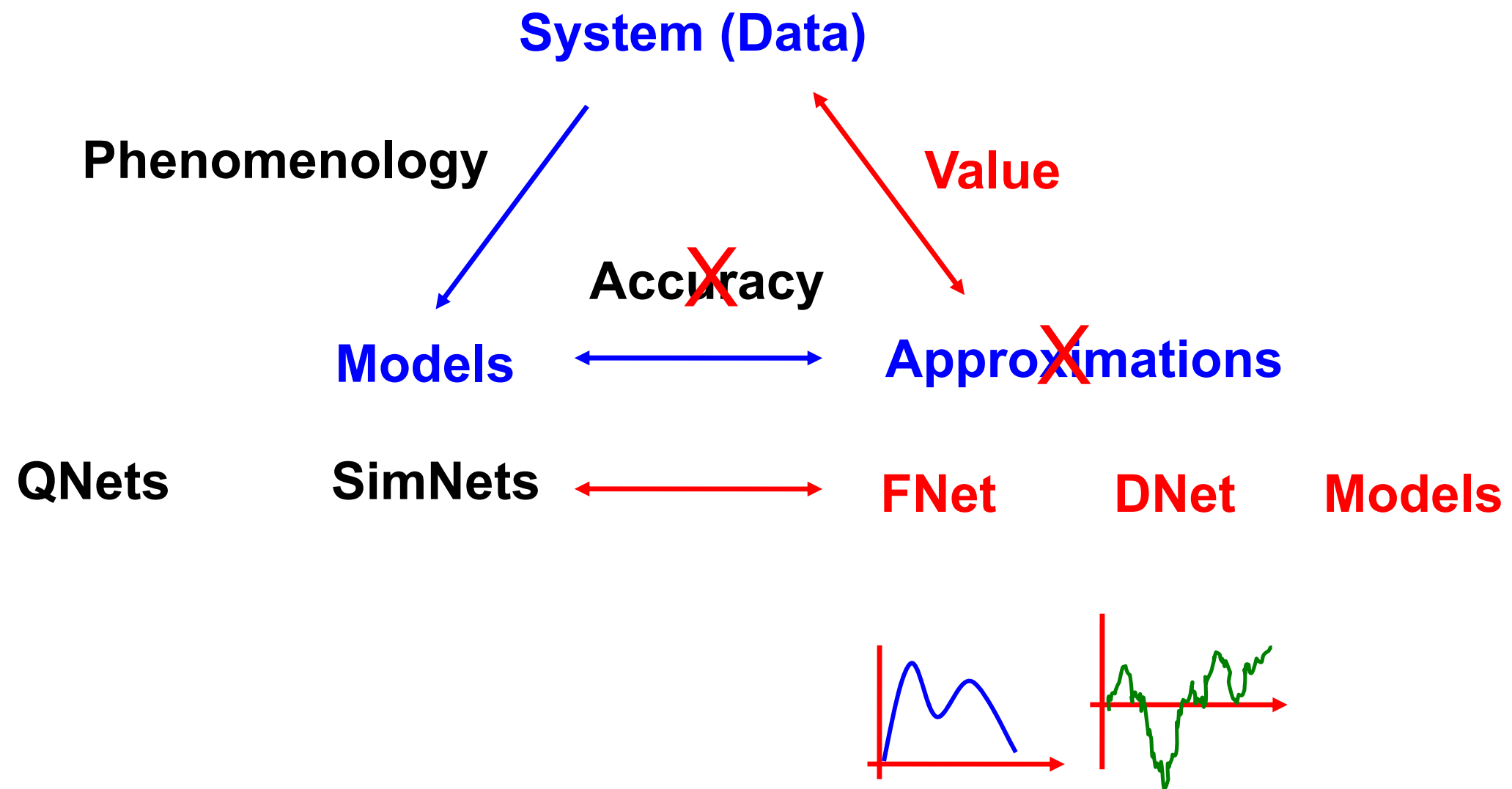
Prevalent Asymptotic Approximations



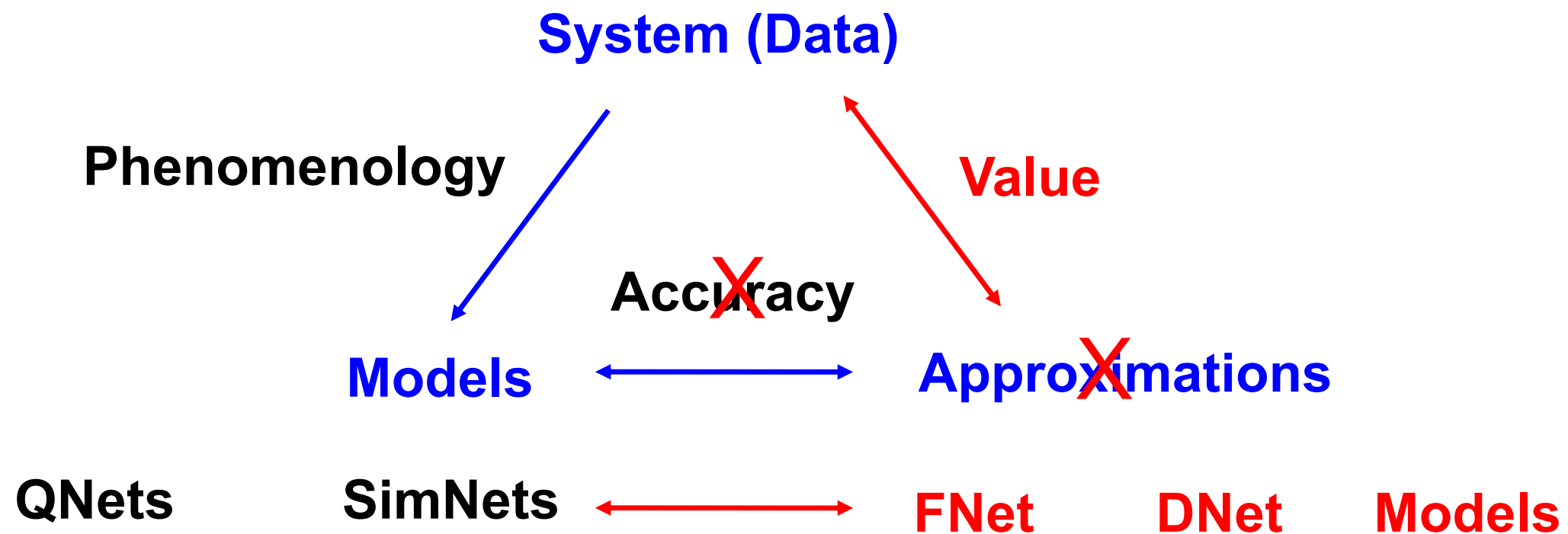
Data-Based ~~Prevalent~~ ~~Asymptotic~~ ~~Approximations~~ **Models**



~~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:

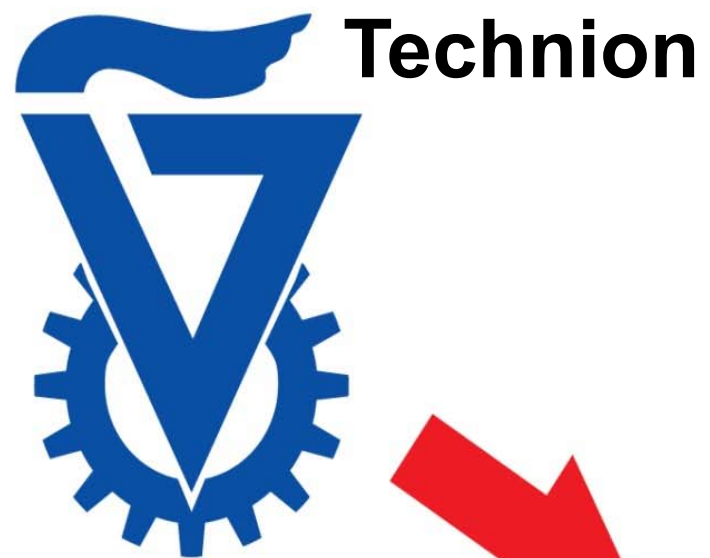
Exceeds Value of originating stylized model

Normal, Brownian Motion

; Bachelier 1900

Poisson

; Poisson 1838



SEELab (2007- ...)
Service Enterprise Engineering



SEELab: Environment for graphical EDA

Operational histories (customers, servers) at the **individual-transaction level**, e.g.

1. ***Bank Anonymous Call-Center**: 1 year, 350K calls by 15 agents - in 2000, **which paved the way to:**
2. ***U.S. Bank Call-Center** : 2.5 years, 220M calls, 40M by 1000 agents
- 3-4. Israeli Cellular Company: 2.5 years, 110M calls, 25M calls by 750 agents; **ILBank (IVR, SBR)**: 2 years
5. Back to Bank Anonymous: Call-Center and more - from January 2010, **daily-deposit at a SEESafe**
6. Service Engineering **internet site**: click-stream data (2 years)
7. ***Home (Rambam) Hospital** : 4 years, 1000 beds, inter-ward patient flow
8. **Emergency Departments (ED)** patient flow:
 - **5 EDs in Israel**: 1-2 years, **late David Sinreich**, ED arrivals & LOS
 - ED in Seoul: 2 months, **K. Song-Hee & W. Cha**, pilot
 - ED in Singapore: 2 years, pilot
9. **U.S. Ambulatory Hospital RTLS (Real-Time Location System)**: Since November 2013
 - **250K events/day (1GB/week)**: **1000** patients, **350** staff (1500 tagged entities), every **3 sec's**
 - Infrastructure: **900 readers (sensors)** over ceilings of 7 (now 8) clinical floors
 - Both actual and planned (**appointment** book of resources: staff, patients, rooms)

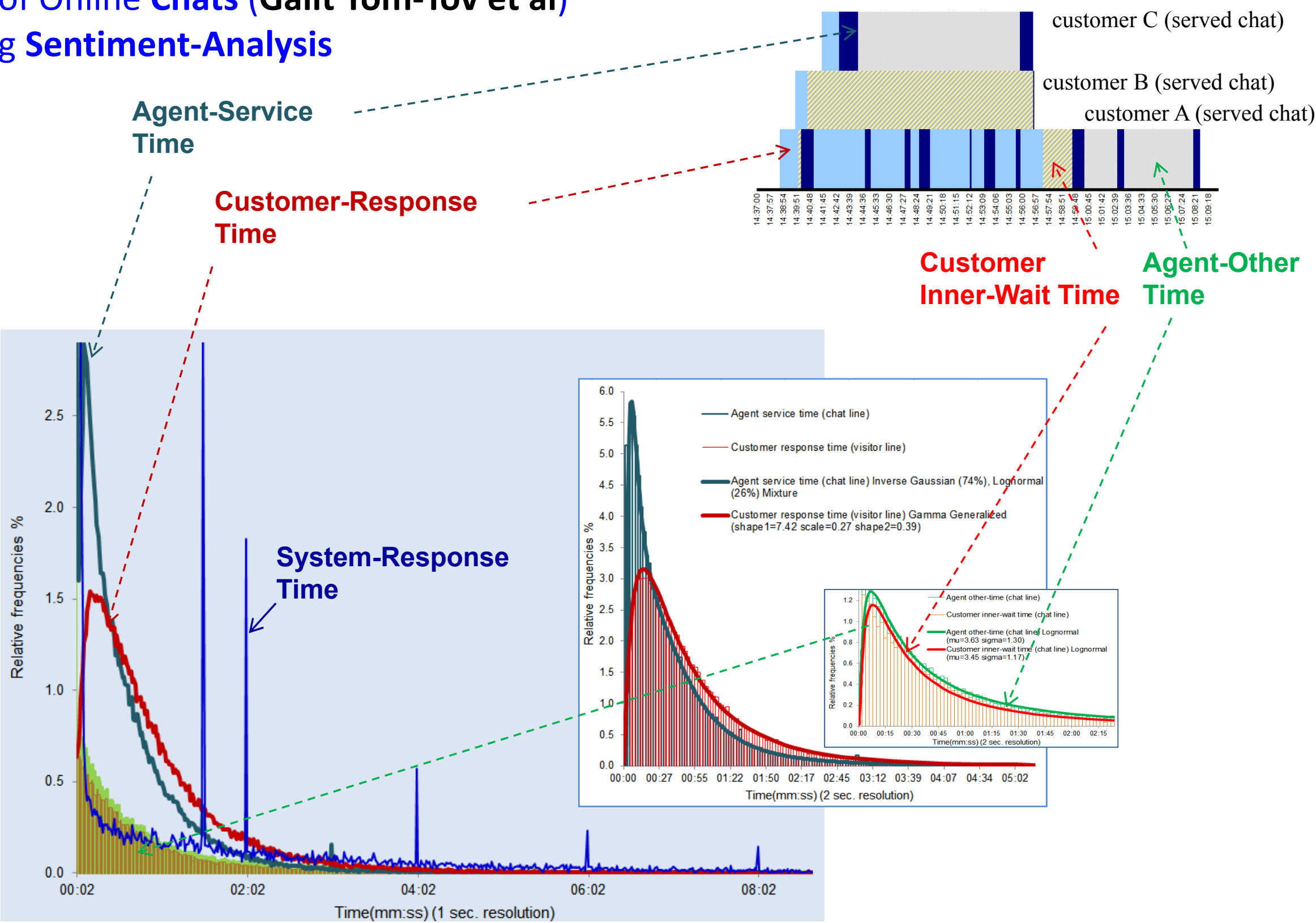
10-13:

Chat Services (Europe); ILBank Warehouse; Smart-City Simulator (Haifa, Boston,...); Justice System

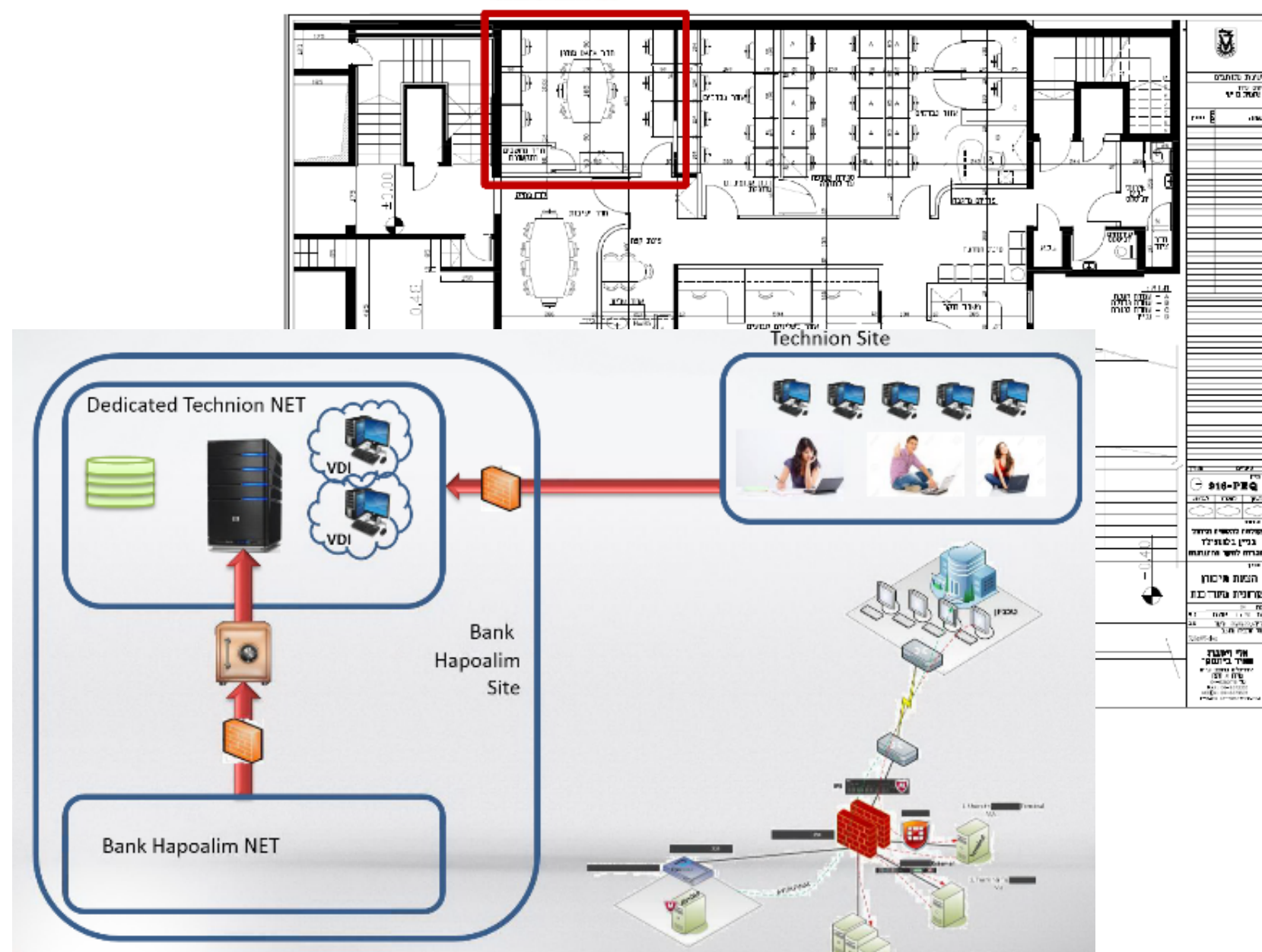
***Open & Free for (reproducible) research and teaching**

Anatomy of Online Chats (Galit Yom-Tov et al)

Supporting Sentiment-Analysis



Data-Room (Secured): Accessing 80% of Bank Warehouse



- T-PADS מעבדה: איך, מי, איפה, ומתי
- מיקוד ראשוני Data Science & Engineering
- אפשרות הרחבה ל Blockchain, IoT, וכו'
- X5 חוזה 5
- בעלות משותפת
- פרסומים משותפים בהסכמה
- דרך פרויקטים, לאימות המחקר POC
- סטטוס: 1ח, ועדת היגוי - 6ח, יום אירוע מרוכז - 1ש

Smart-City “Simulator” (40K sq ft “playground”, 10K offices)

21



Emergency services

Traffic lights

Lightening

Smart parking

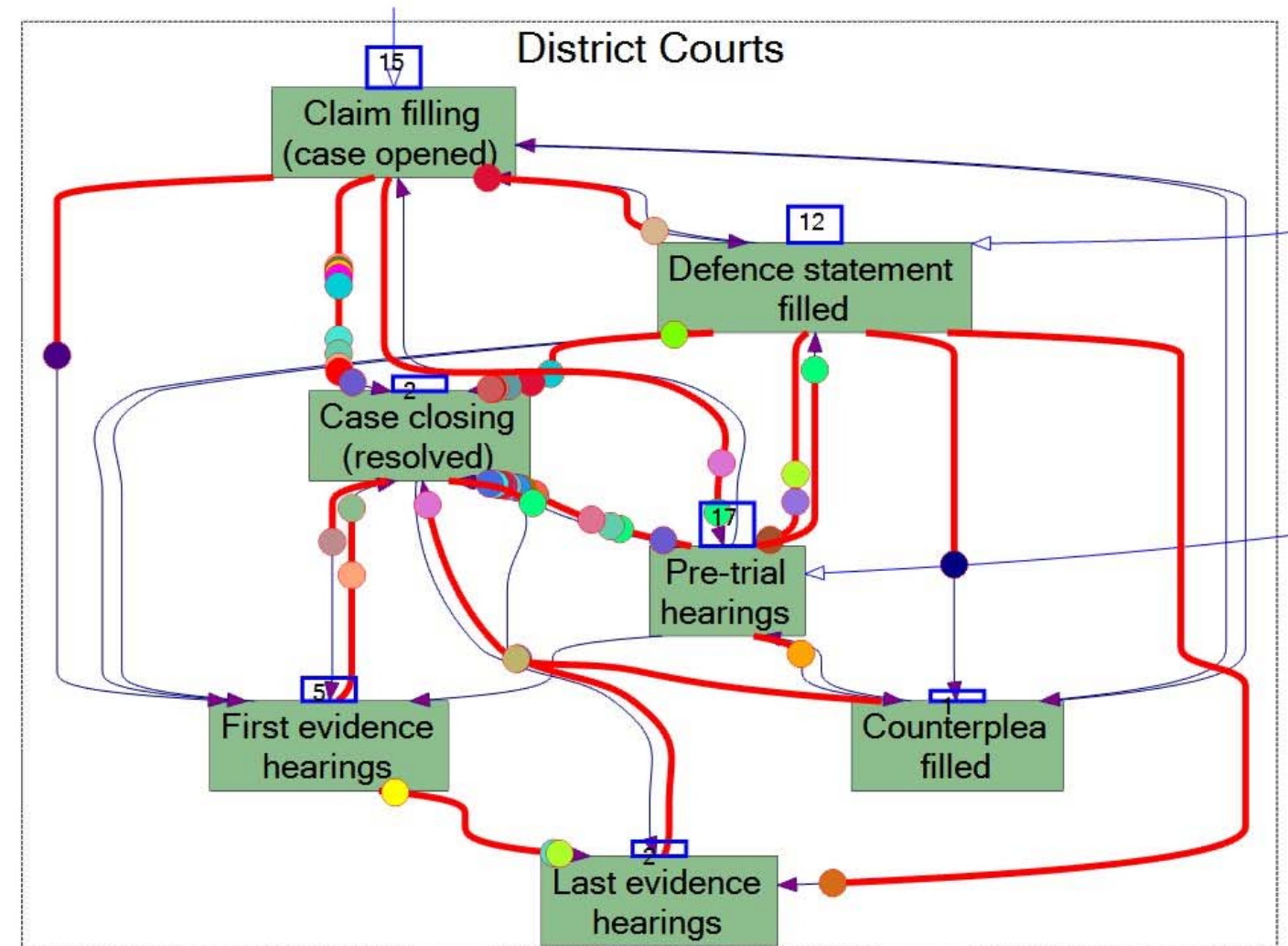
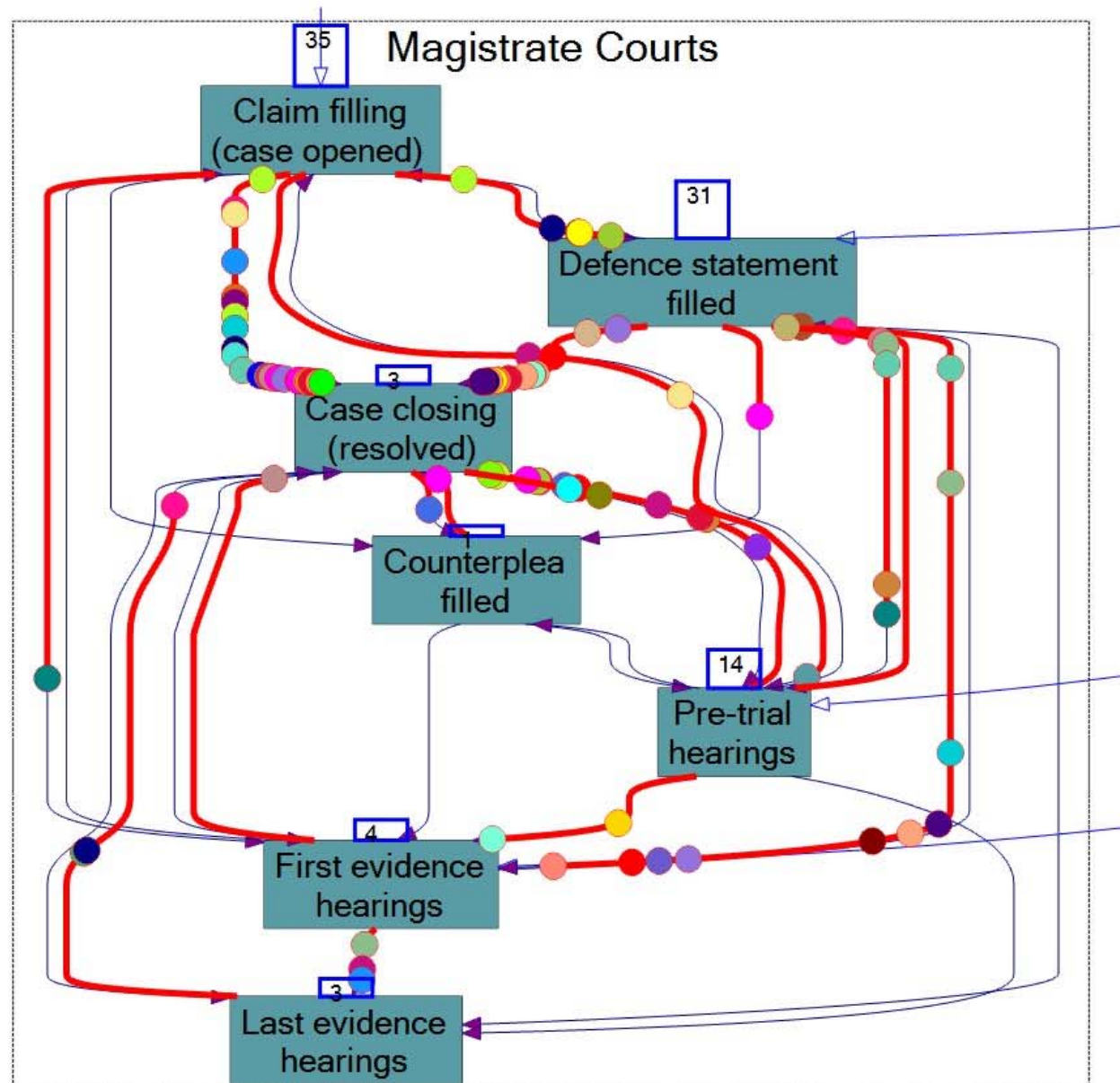
Water, sewage

Energey systems

600 מ"ר משרדים + 2.5 דונם של "מגרש משחקים" + 2 דונם הרחבה עתידית



“Production of Justice”: Case-Flow in the Justice System



Pause for a Commercial: from the Skeptics

The Skeptics

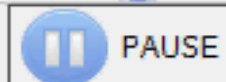
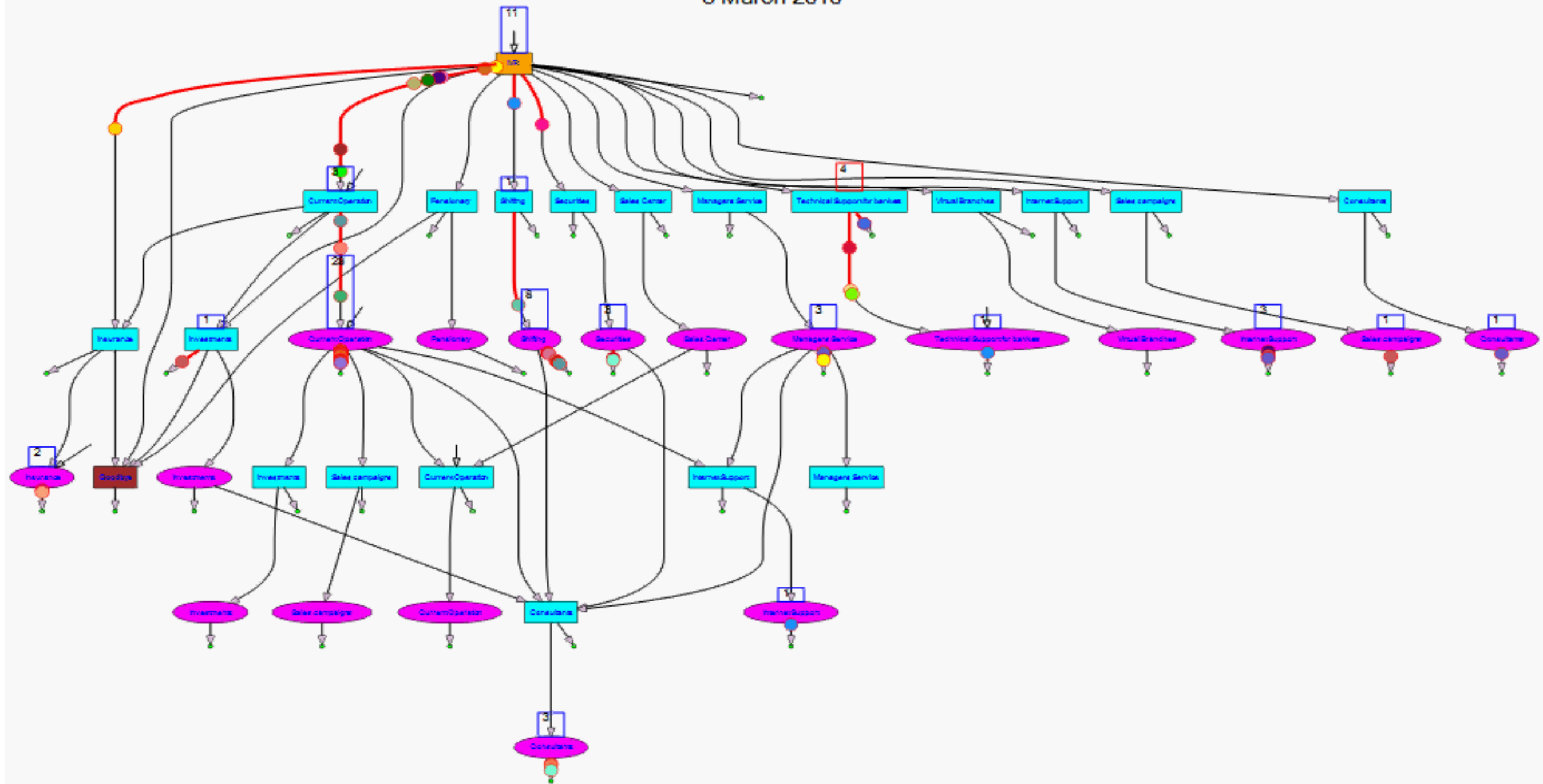
ERC: on using data to motivate theory (as in say Physics, Biology,...)
How can breakthrough mathematics come out of so much data?

NSF: on funding data-collection and maintenance in OR research grants
Finding an interested industry-partner w/ data $\Rightarrow$ problem solved

ISF: on measuring judicial workload (JW)
For the most part, ... the applicant proposes to quantify the unquantifiable and solve the unsolvable, namely JW.
At least since the eighteenth century, there are continuous ...

Still: Nurse-workload in maternity wards: operational+emotional(+cognitive)

Customers flow (ILDUBank) 8 March 2010



10 : 24 : 32 8 March 2010

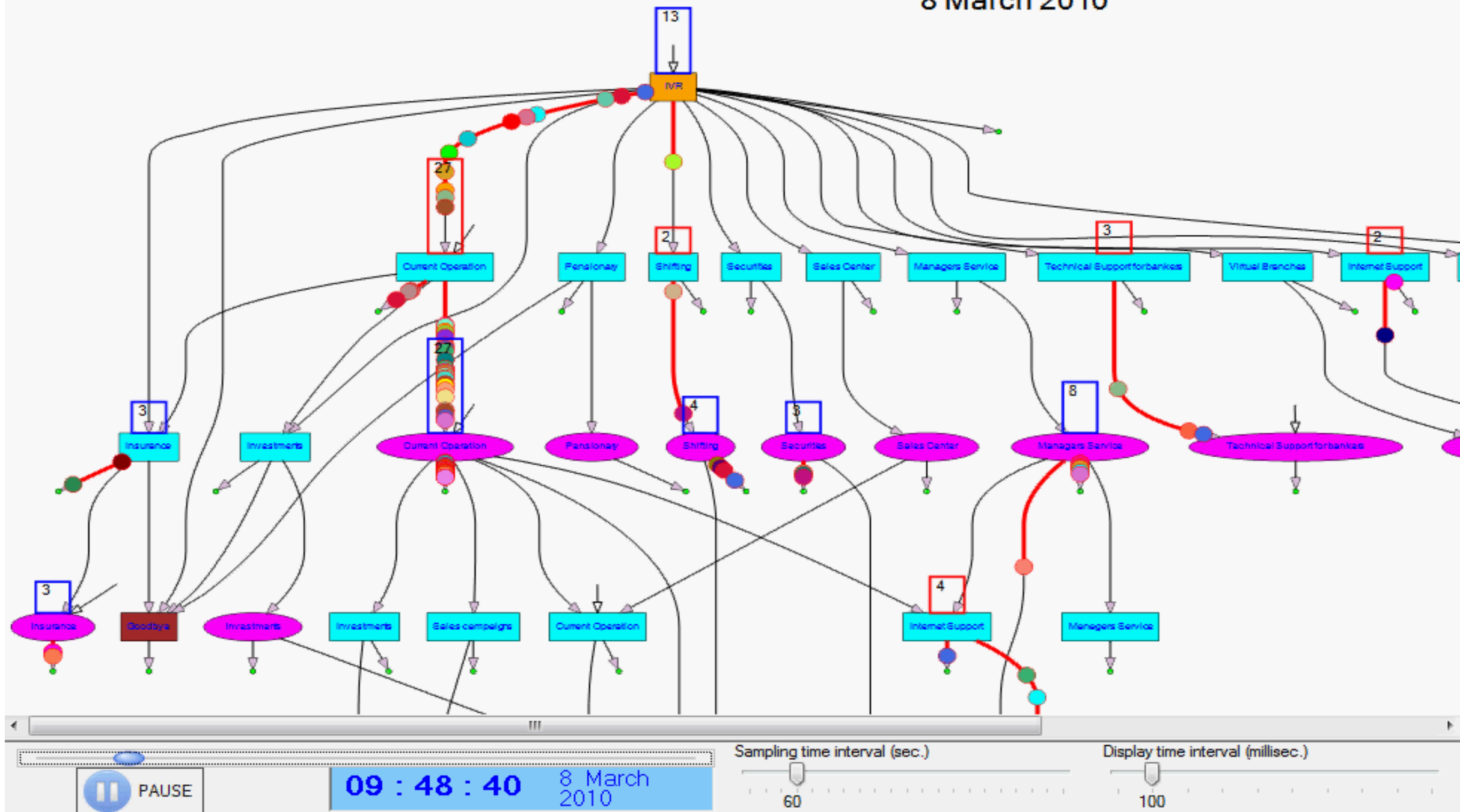
Sampling time interval (sec.)

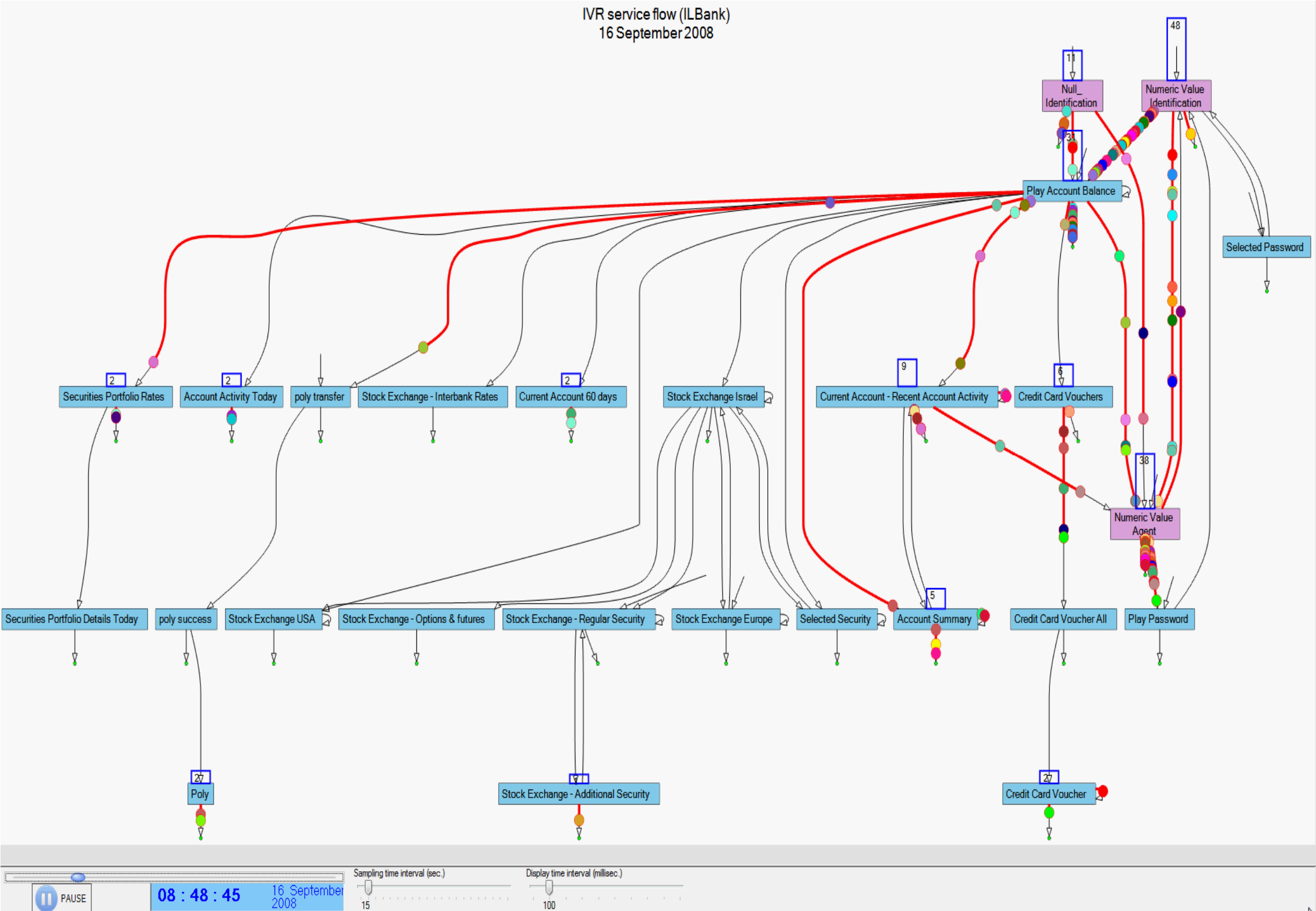
60

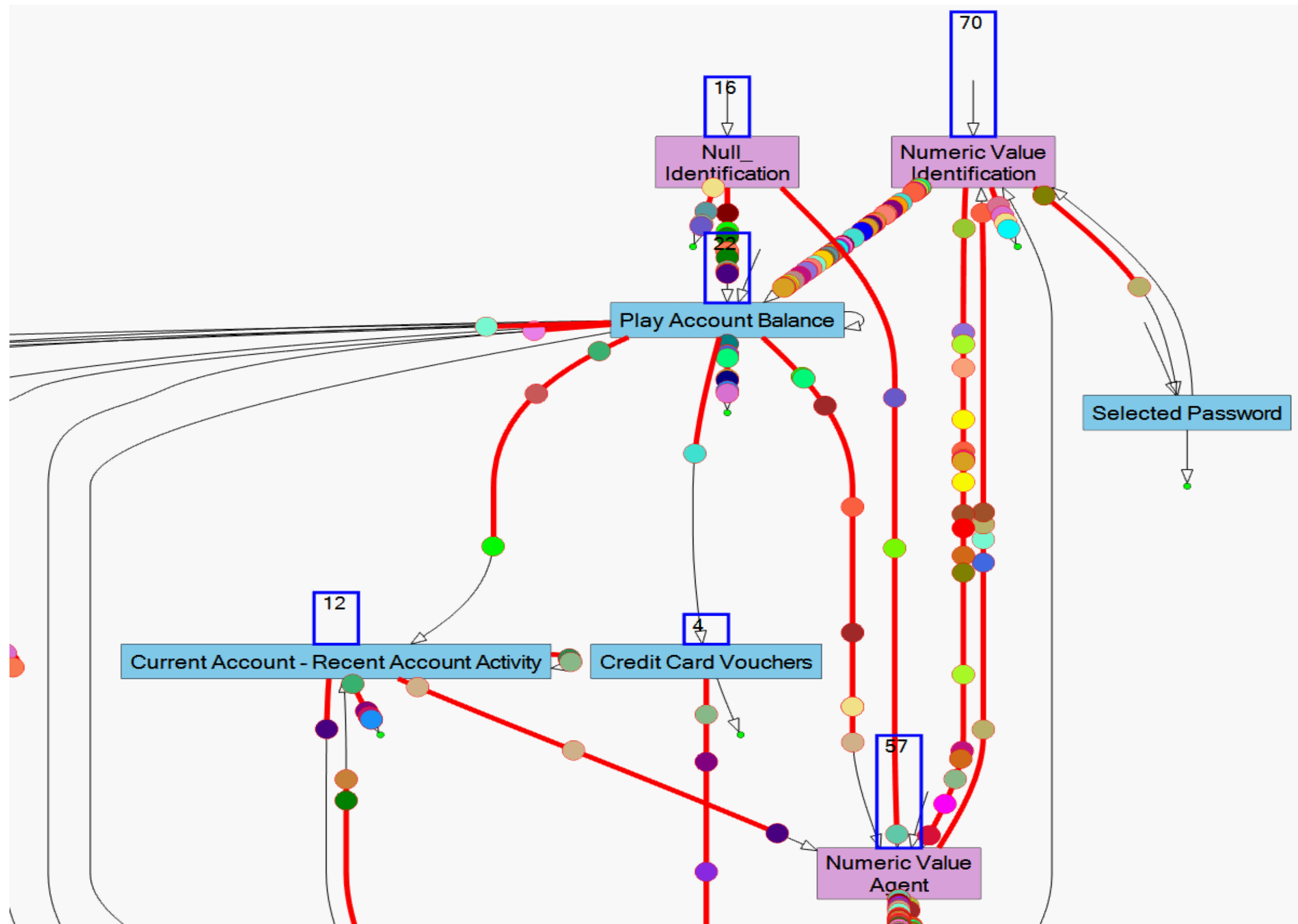
Display time interval (millisec.)

100

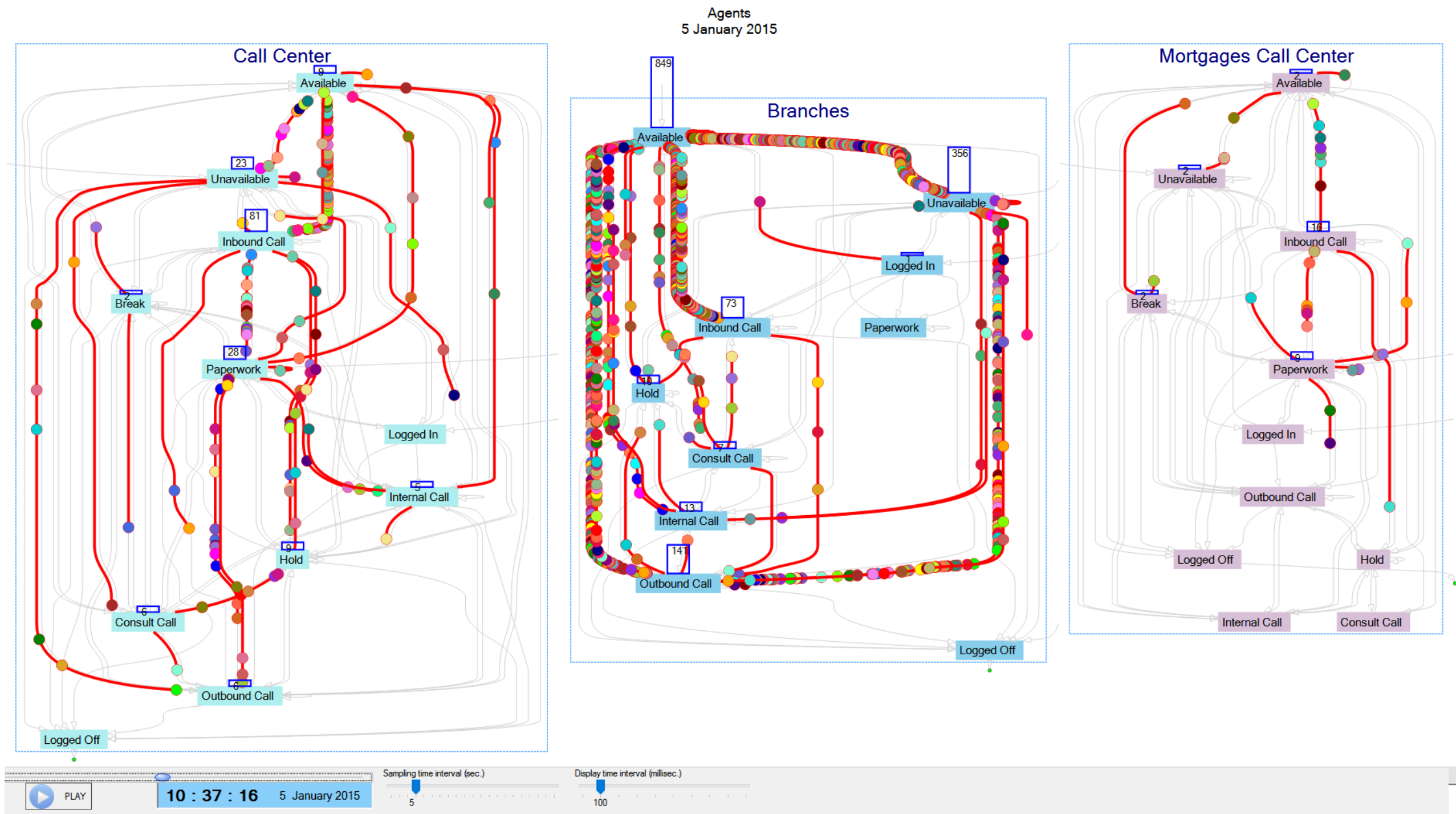
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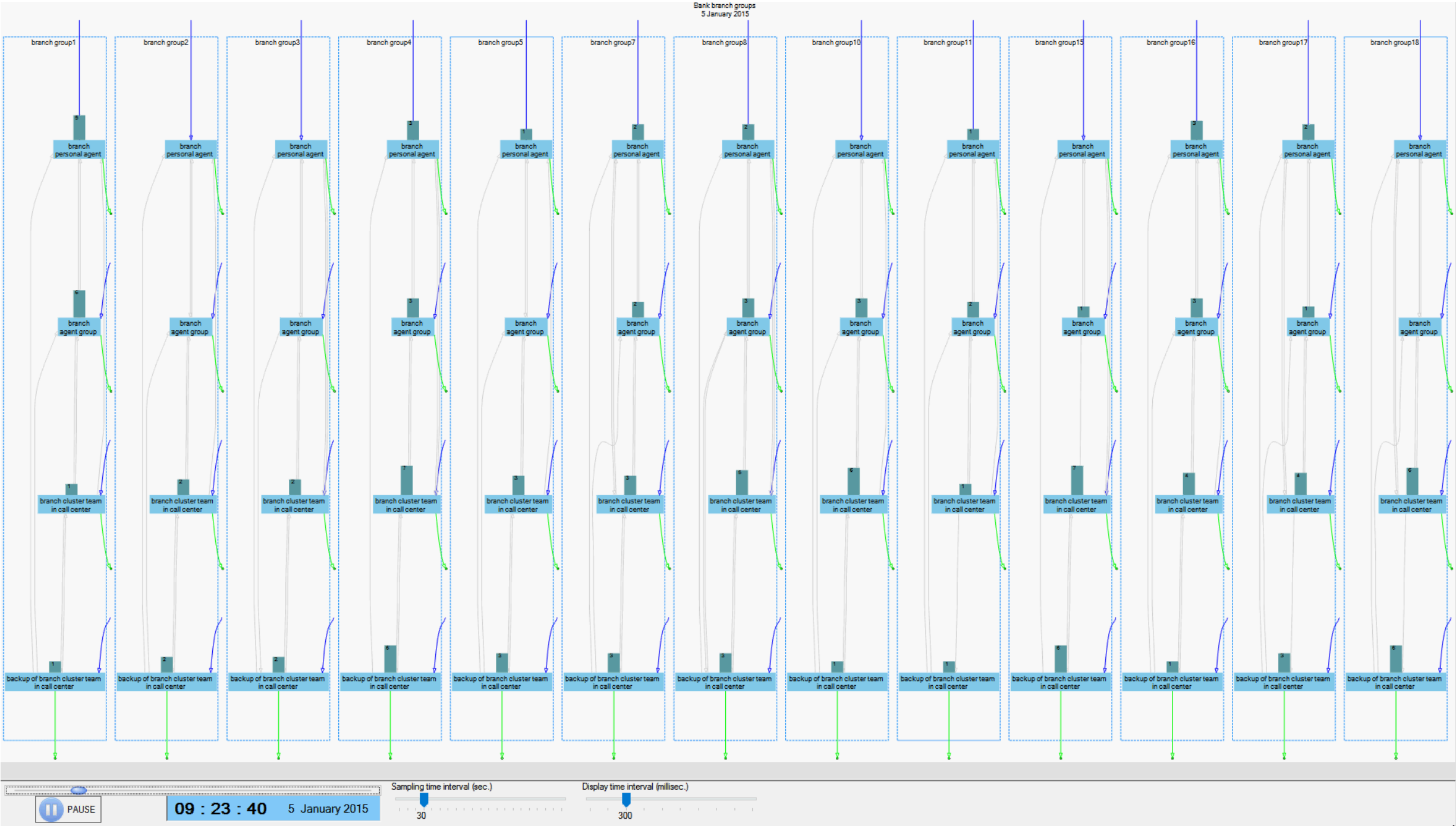




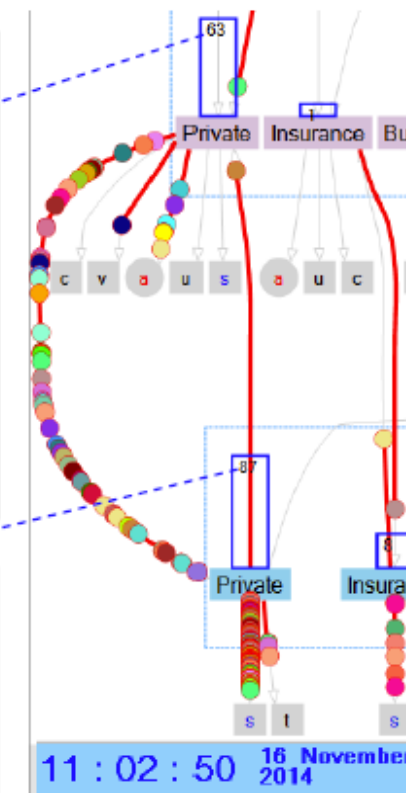
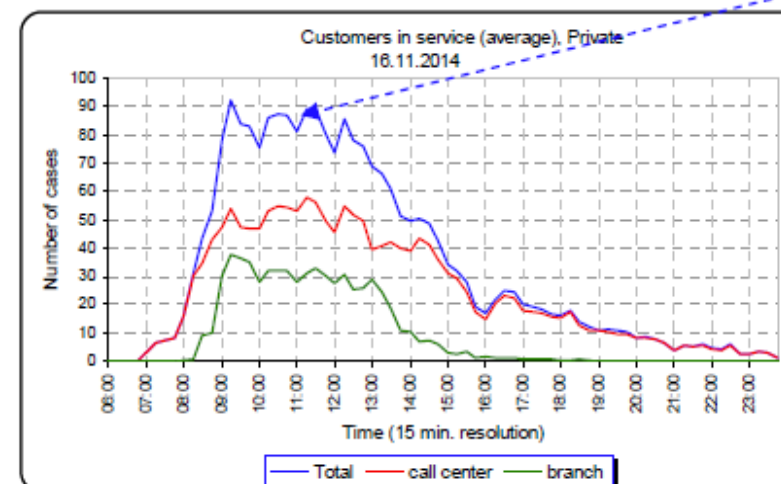
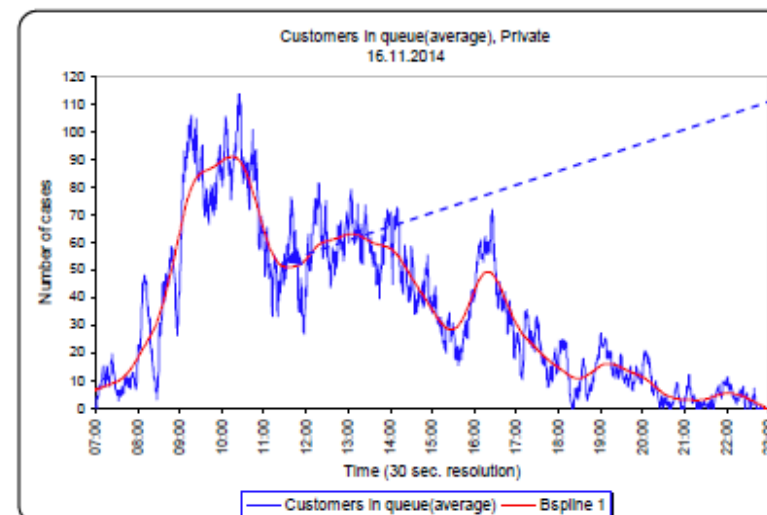
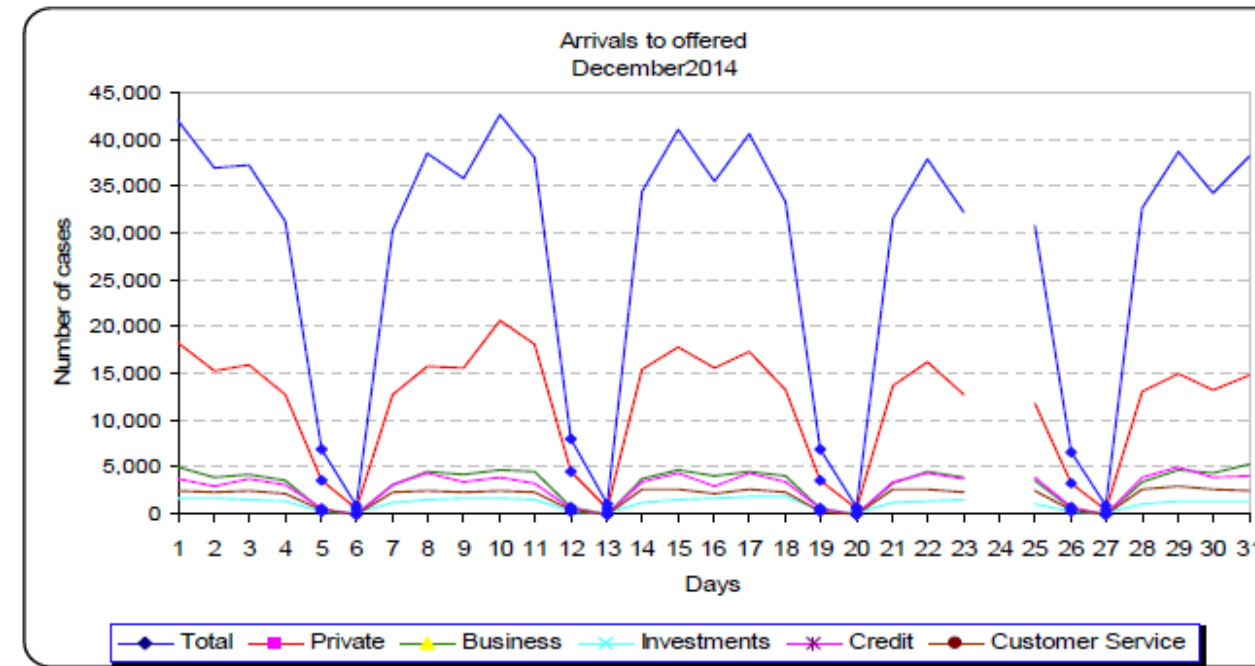
Agents (2000): Branches (1700) and Call Centers - Commercial (270), Mortgages (50)



ILDU Banking: Branch Groups Virtual Queues



ILDU Banking:
branches and call centers
Service Types





Research Goal (within reach)

e.g. Specific Emergency-Department, with ample reliable data

- Real-time: control of patient-flow (bottlenecks); status info and prediction
- Short-term: on Monday, set Tuesday's staffing levels (or next week's); real cost of care for the individual patient (vs. mean/negotiated costs)
- Long-term: capacity allocation, facility/triage design; social network (e.g. correlated w/ outcomes); transformative change (Epic); congestion laws

All above will be enabled by parsimonious (robust) models, created in real-time **mining** (semi- or un-supervised) of ED **processes** and **models** (empirical, simulation = **SimNets**, mathematical = **QNets**, **FNets**, **DNets**,...)

Laws & Models: Data-Based Erlang A/R/S, following B/C

- Little's Law (Steady-State, Transient), State-Space Collapse,...
- Erlang-B (**B**locking) and Erlang-C (?)
 - **Erlang**, Agner Krarup: Queueing Theory was born in 1909, in his paper "The Theory of Probabilities and **Telephone Conversations**"
- 2. Erlang-A
 - **Abandonment**: While waiting for service, does service-value dominate residual-wait-cost
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 - **Return/Feedback**: Customers often return to service (positively, negatively, just needing)
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 - **Servers**: Challenging to manage, and model, no less so than customers

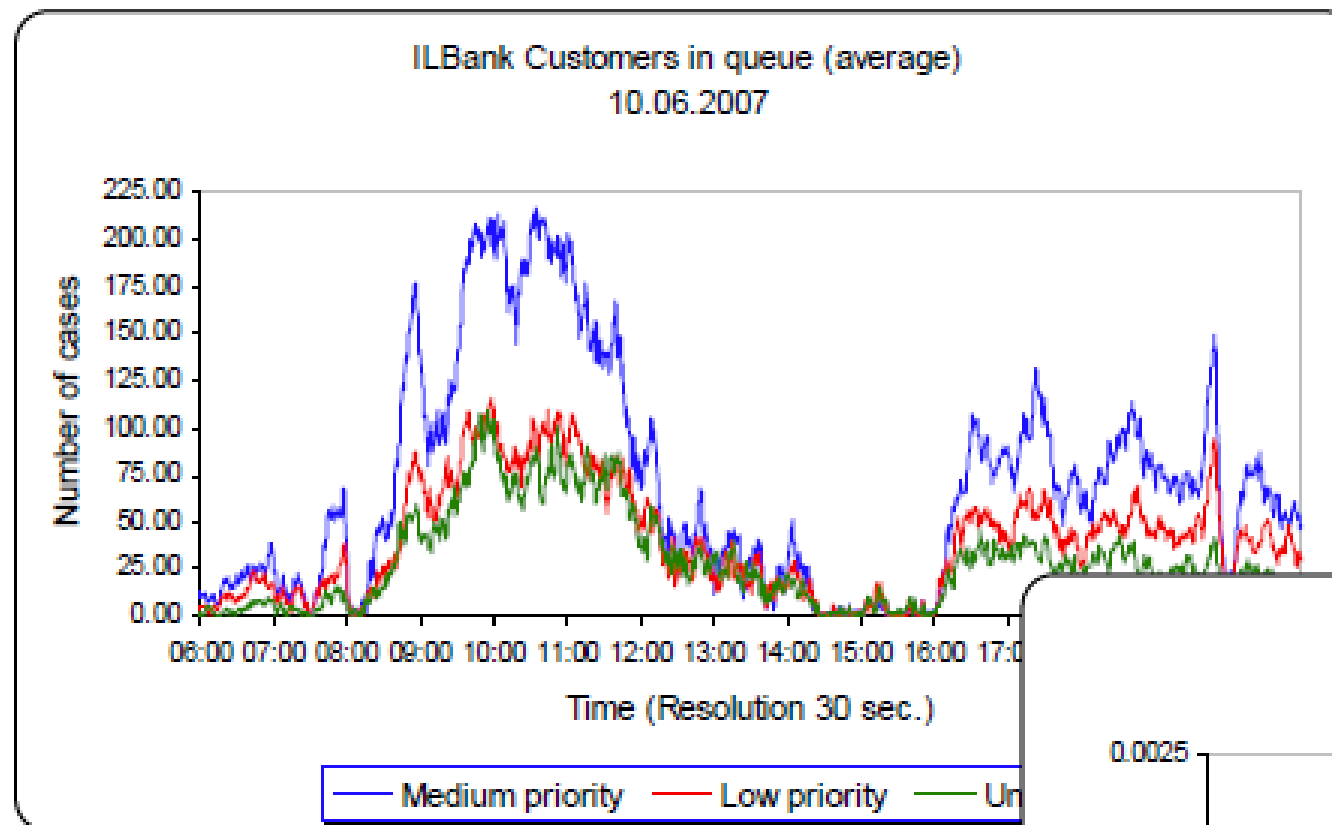
Above: Simple (Parsimonious/Robust) models of complex realities, yet not too simple

For the 3 Models (if time permits):

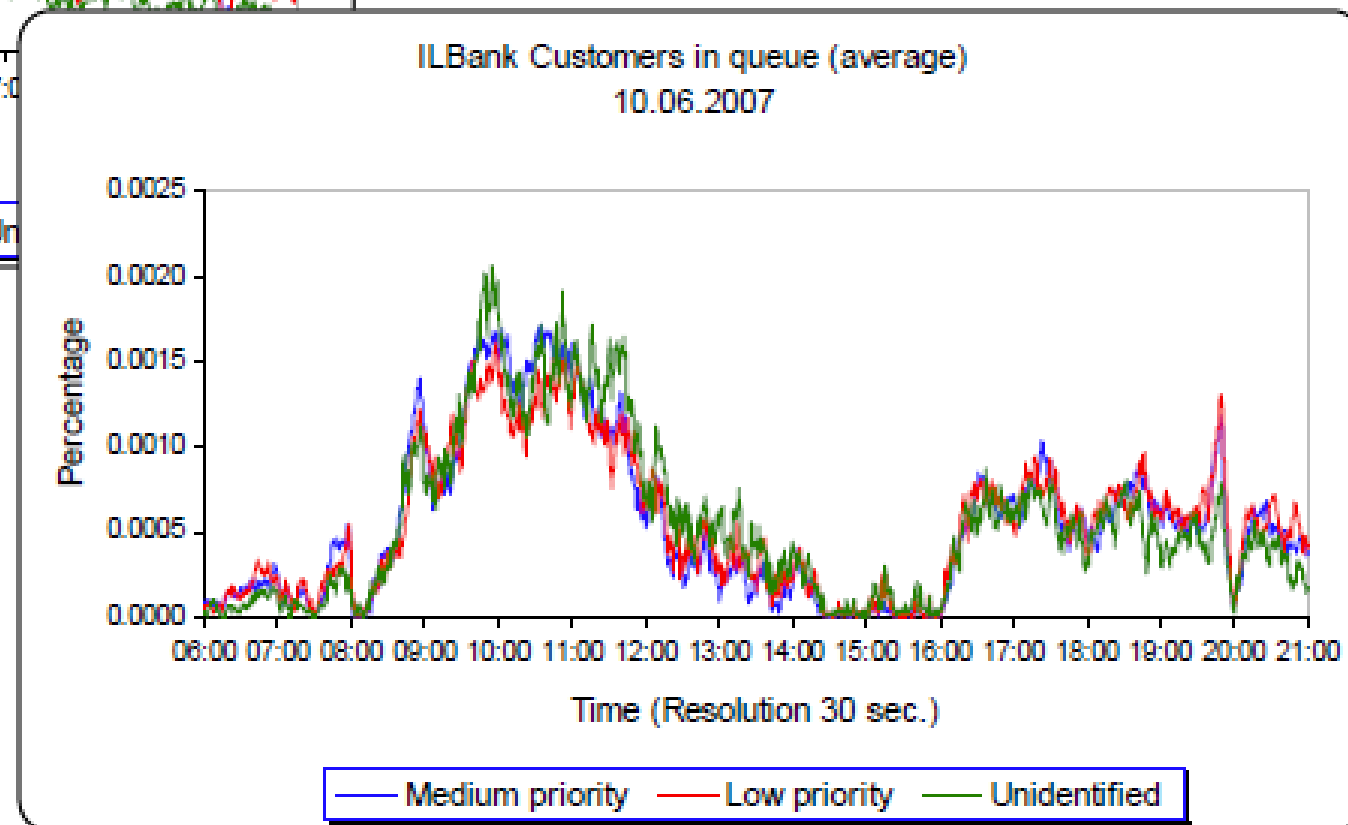
1. Motivating **Phenomenon**, via Data
2. Informal **Description** of Model
3. Example of **Application(s)**
4. Model **Expressiveness** (Strength)
5. Insights, **Extensions**

Dynamics: Parsimonious Models (Congestion Laws)

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



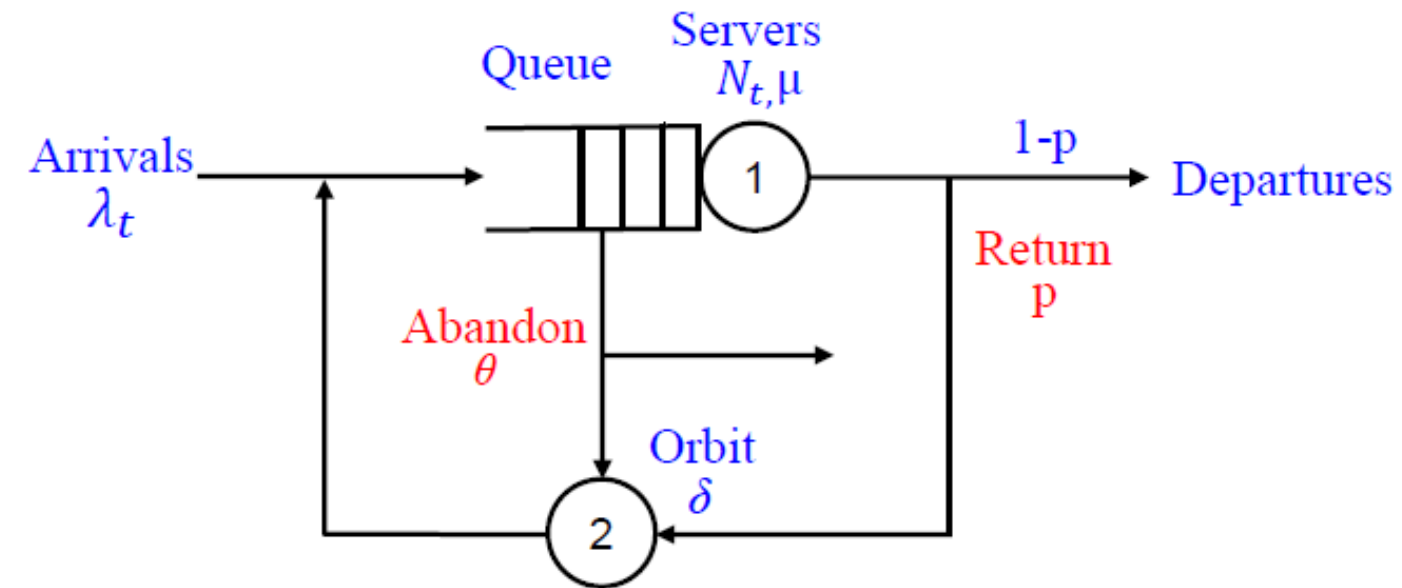
Queue "Shape"



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

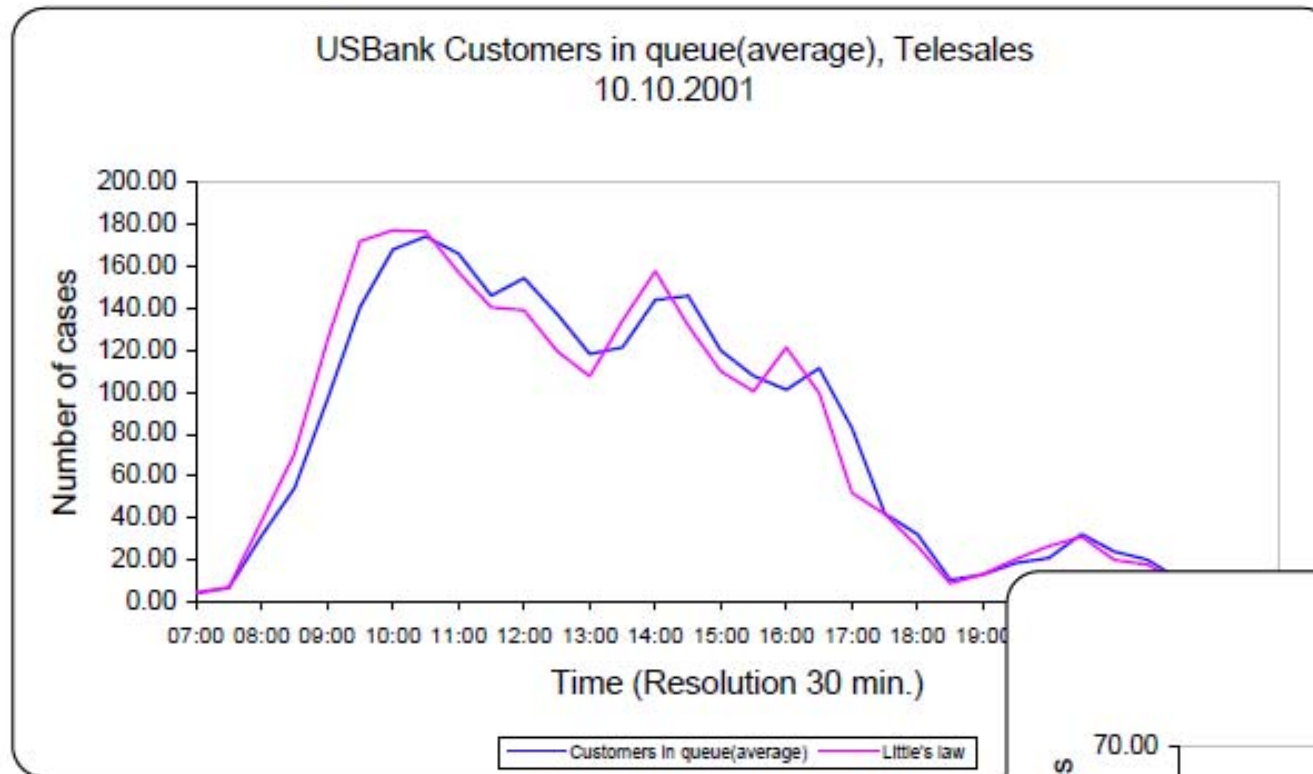
Model Selection: As Simple as Possible but Not Simpler

Service with Retrials and Abandonment; w/ Massey, Reiman, Stolyar



Little's Law $L = \lambda \times W$, in a Time-Varying Environment

Time-Gap: # in System lags behind **Little / 30 min**



Call Center:

⇒ **Piecewise Steady-State**

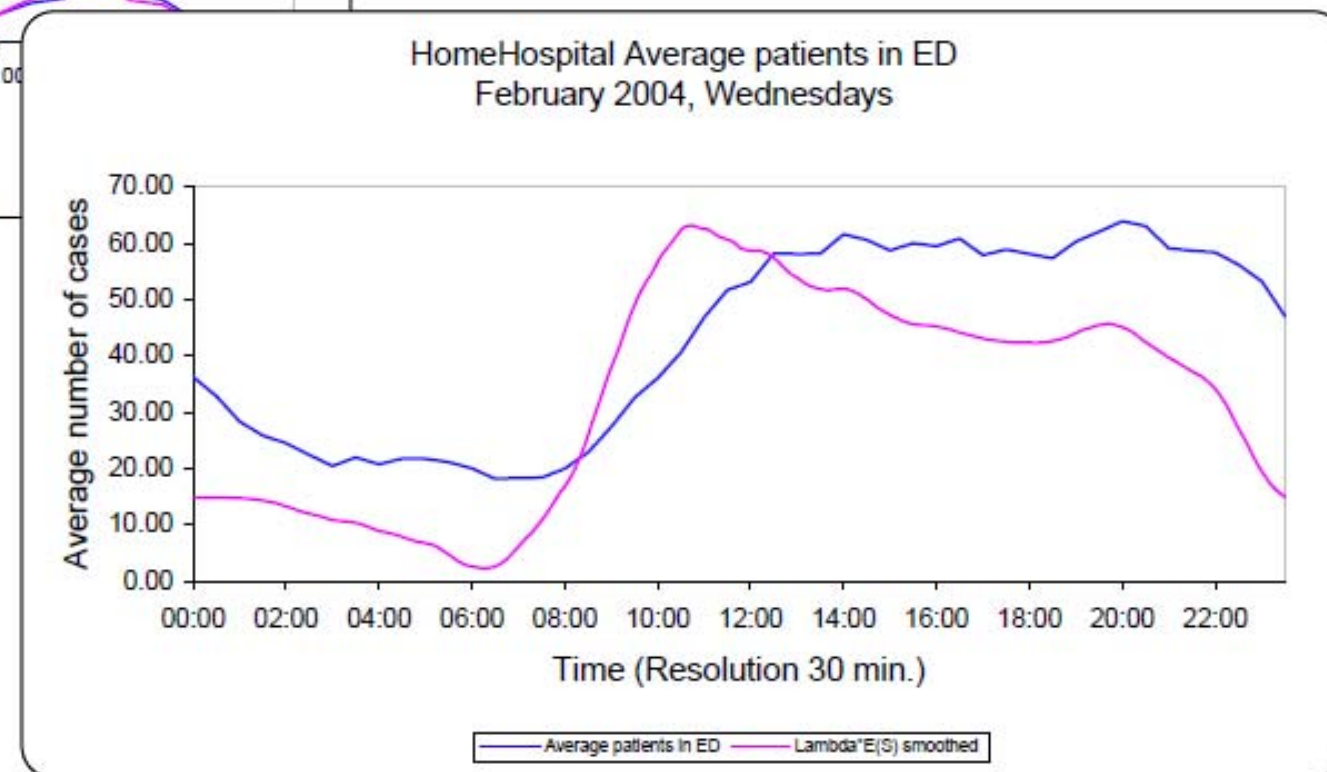
Emergency Dept:

⇒ **Time-Varying Transient**

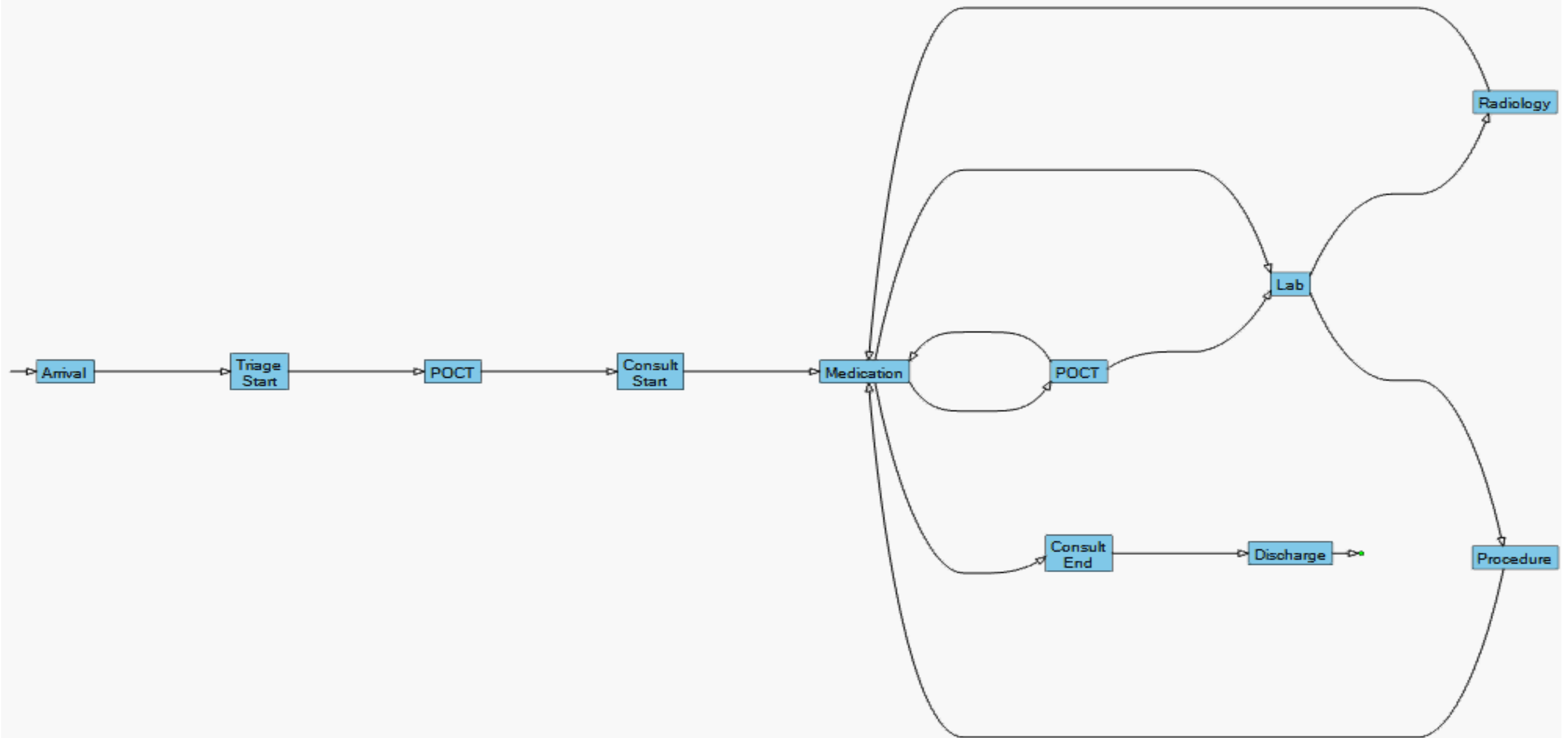
$$EL(t) = \tilde{\lambda}(t) \times EW,$$

$$\tilde{\lambda}(t) = E\lambda(t - W_e).$$

(Bertsimas, Mourtzinou;
Fralix, Riano, Serfozo)



Patient flow (XYHospital)
October 2012



PLAY

17 : 01 : 52

12 October
2012

Sampling time interval (sec.)

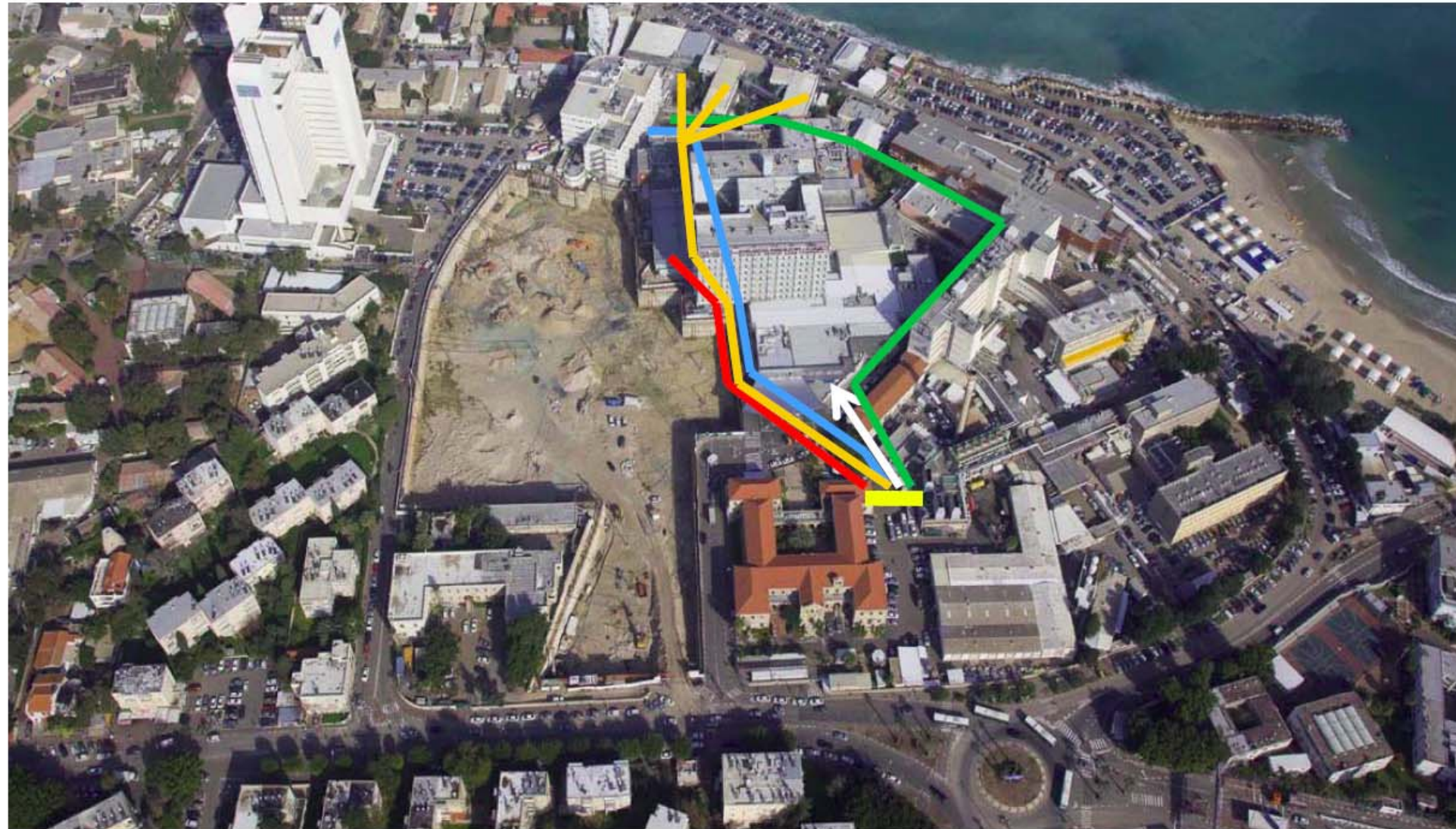
180

Display time interval (millisec.)

100

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

Drill: Chemical MCE, Rambam Hospital, May 2010



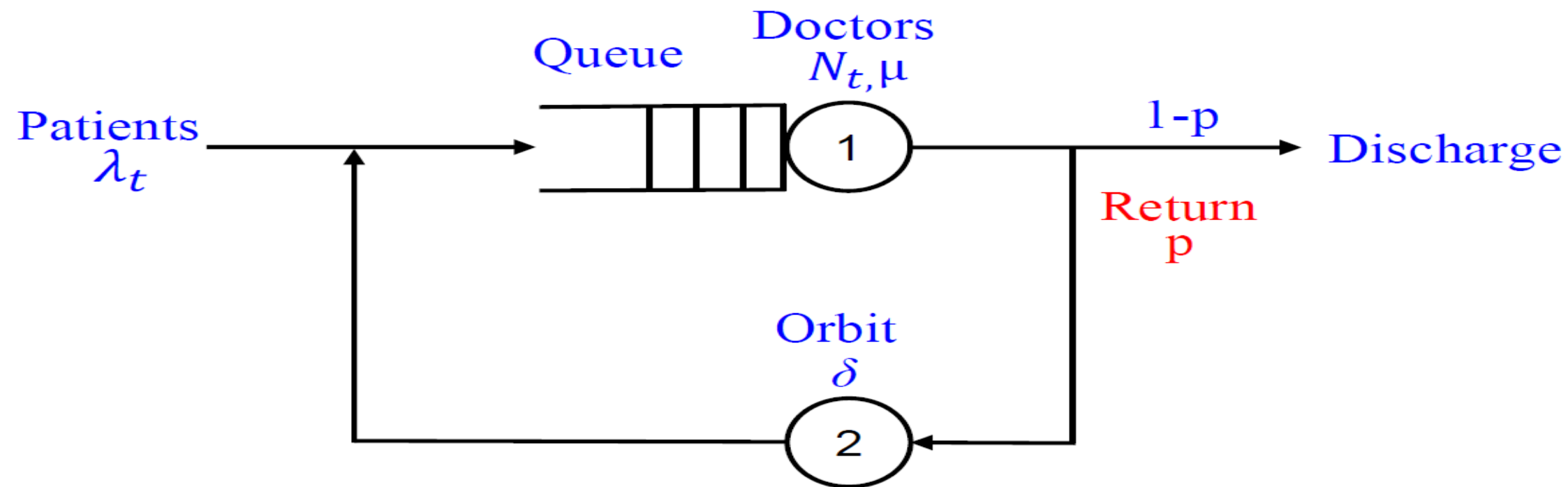
Focus on **severely wounded** casualties (≈ 40 in drill)

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

Single patient history; the study period November 2013-May 2014 (DayHospital)



Erlang-R $\leftrightarrow$ Fluid Model, w/ Galit Yom-Tov



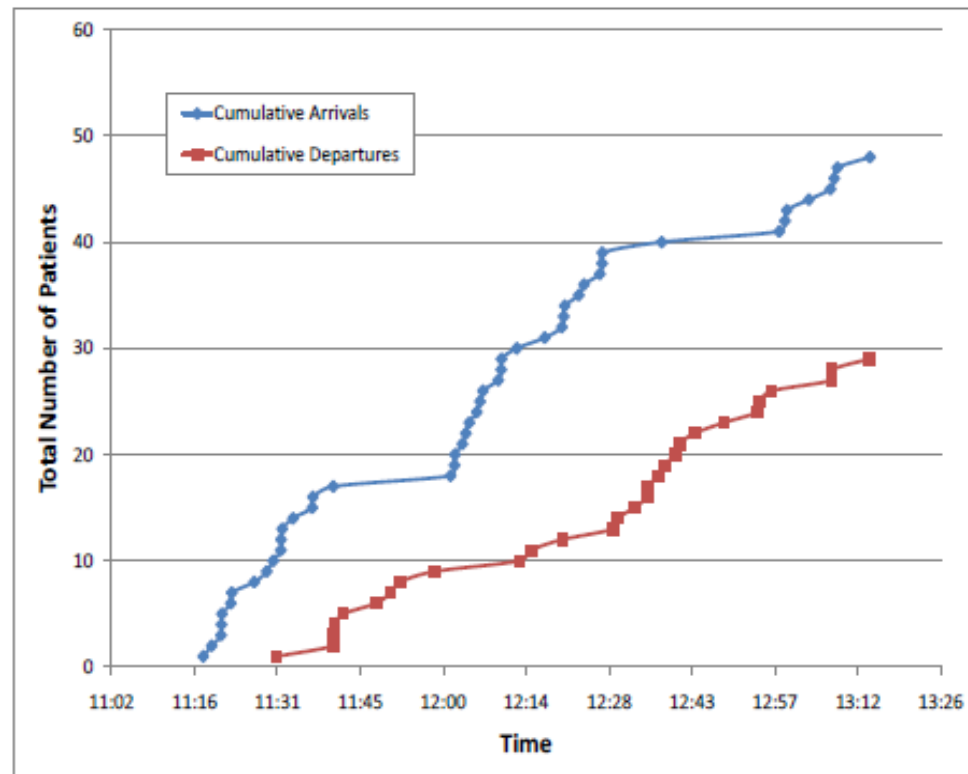
Functional Strong Law of Large Numbers, for a 2-station QNet. BUT **FNet** = ODE: derived **directly** (no QNet), spreadsheet "solution"

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

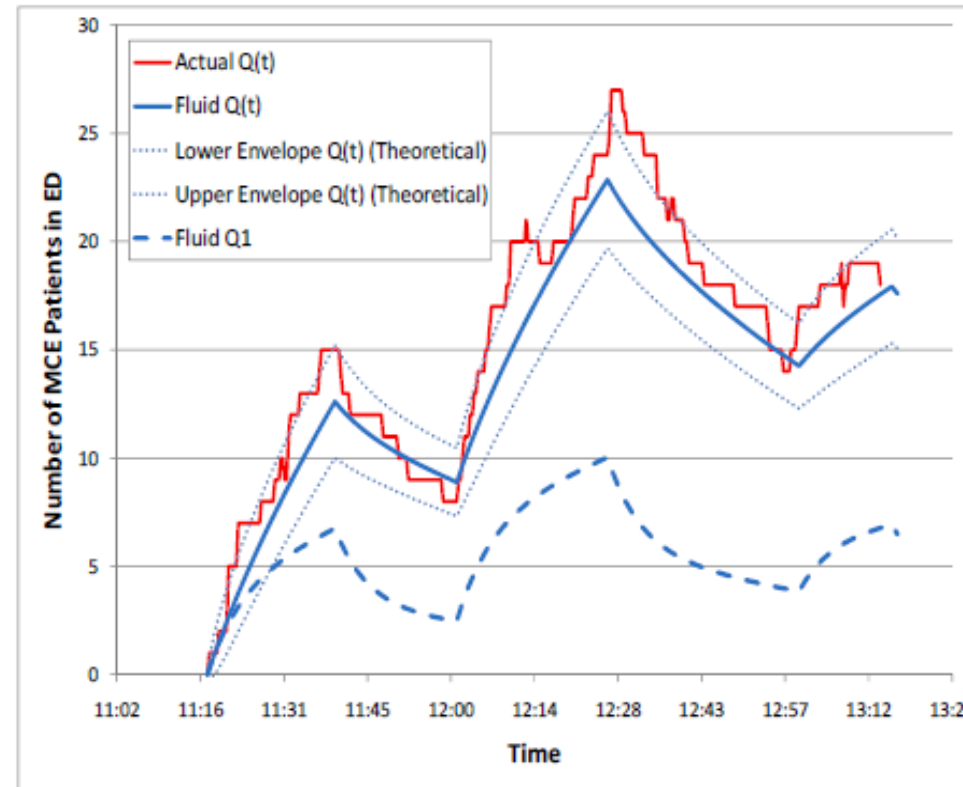
Erlang-R Value: FNet vs. Data

Chemical MCE Drill (Israel, May 2010, 11:00-13:00)

Arrivals & Departures (RFID)



Erlang-R (Fluid, Diffusion)



- ▶ **Recurrent/Repeated** services in Chemical MCE: injection every 15/30/60 min
- ▶ **Fluid** = ODE
- ▶ **Diffusion** (confidence band), via F. Central Limit Theorem: Usefully narrow

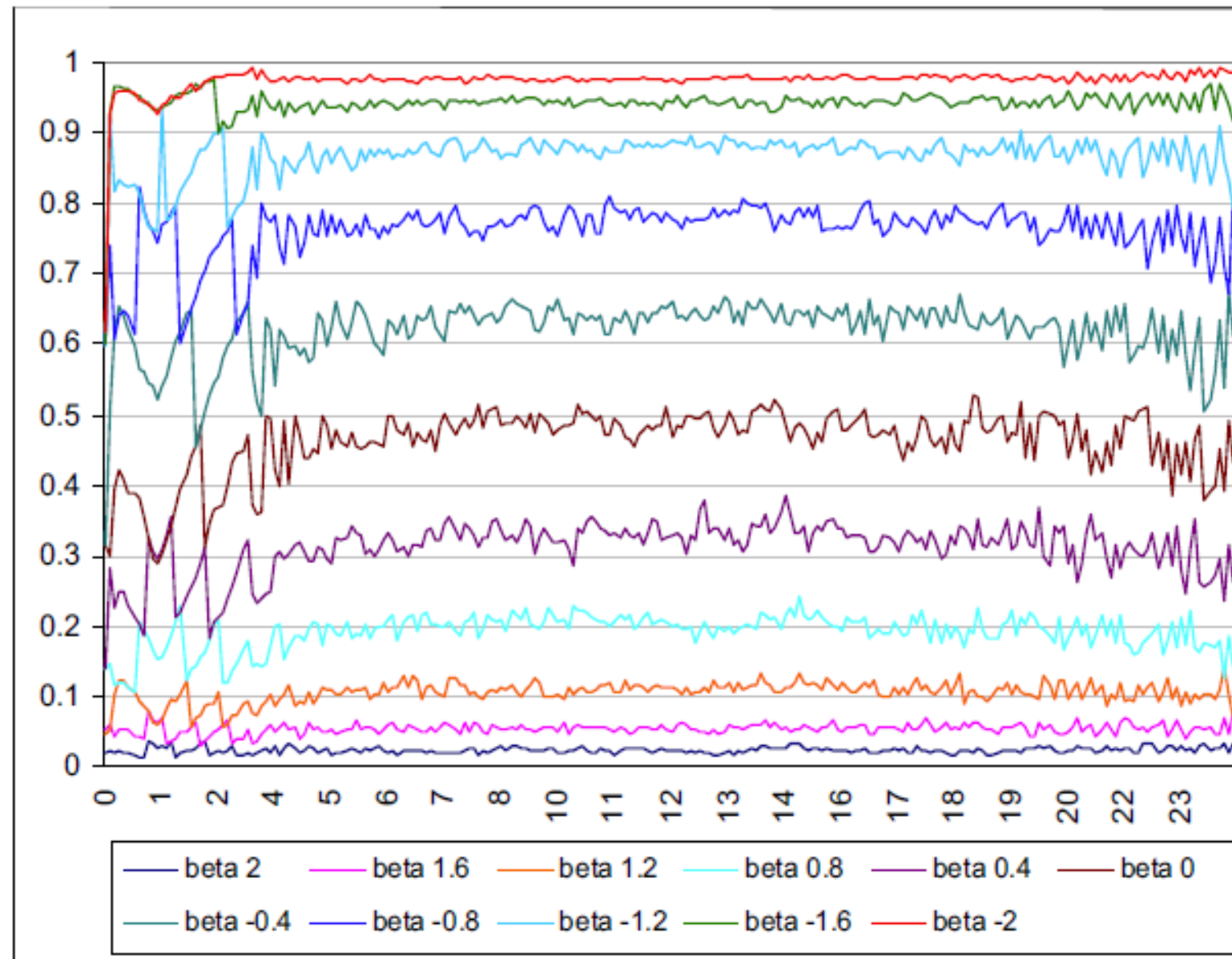
A Data-Based Framework, or “Erlang-R in the ED”

System = e.g. Emergency Department

- ▶ **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)
- ▶ **SimNet** = Customized ED-Simulator (Marmor & Sinreich)

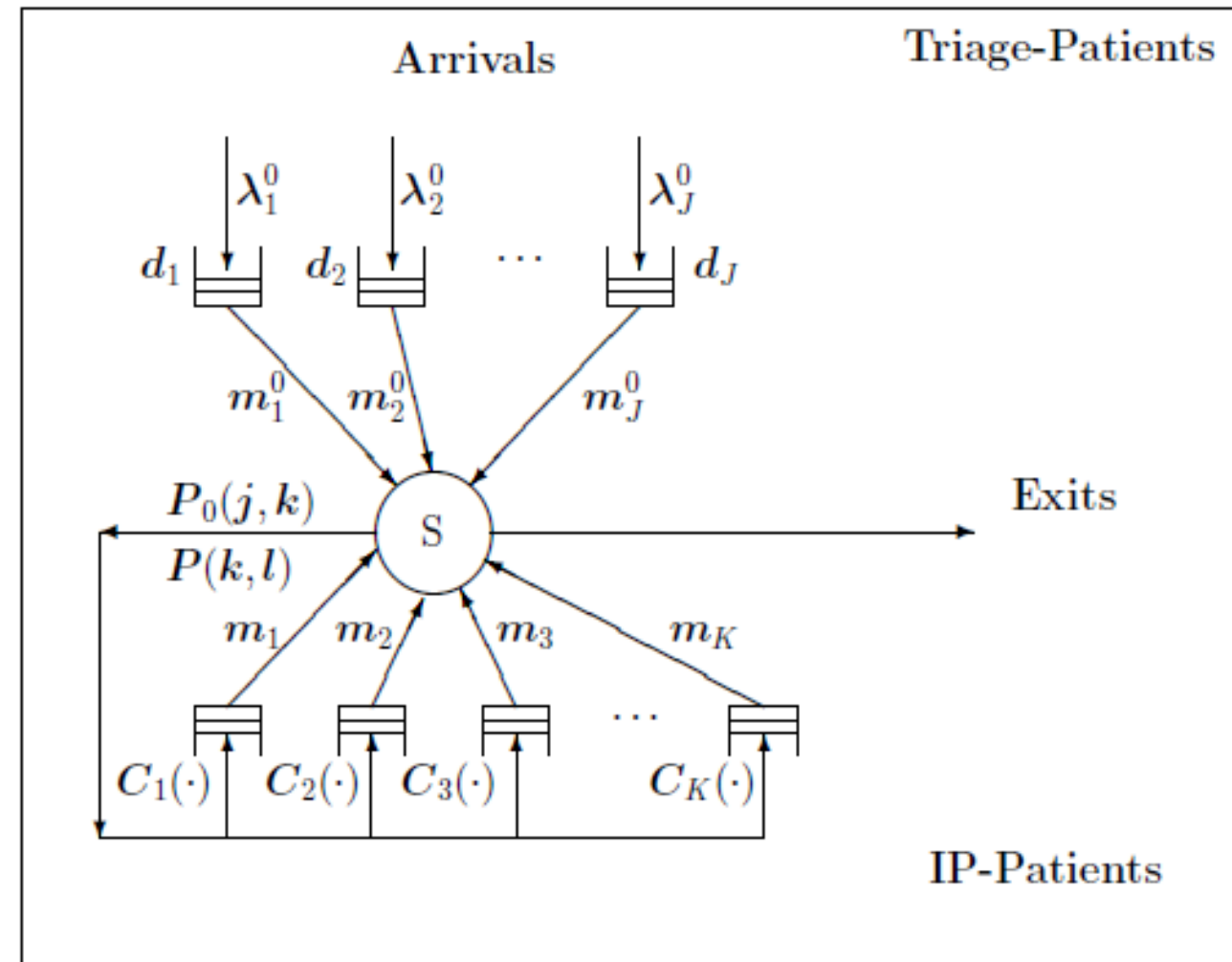
Time-Stable Performance of Time-Varying Systems

Delay Probability = As in the Stationary Erlang-A / R



ED Patient Flow: The Physicians View

with J. Huang, B. Carmeli, S. Israelit

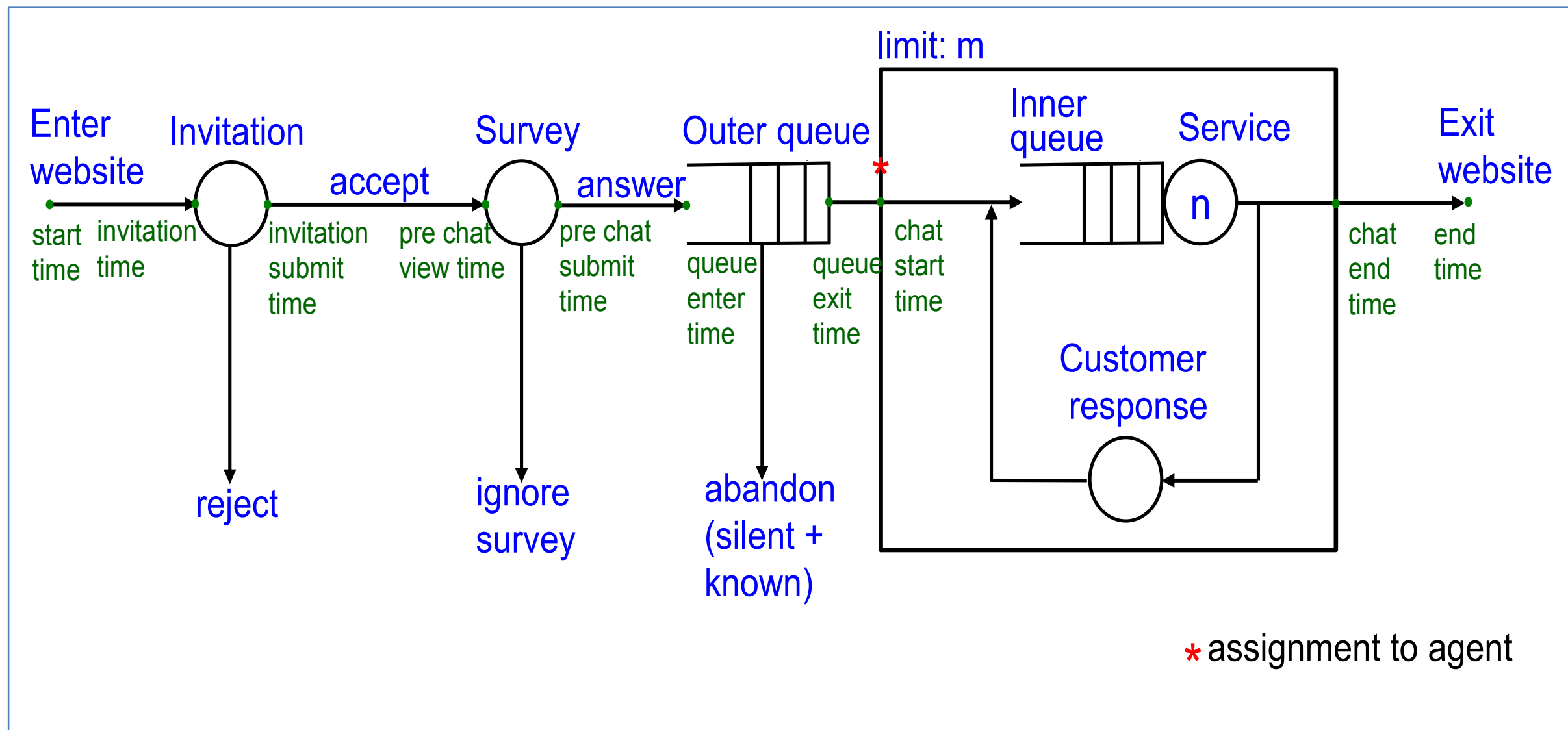


Goal: Adhere to **Triage-Constraints**, then **release In-Process** Patients

Online Chats – Returns: from Internet-Prompts to Chat-Sessions

Empirical: Galit Yom-Tov & Anat Rafaeli & graduate (OB) students

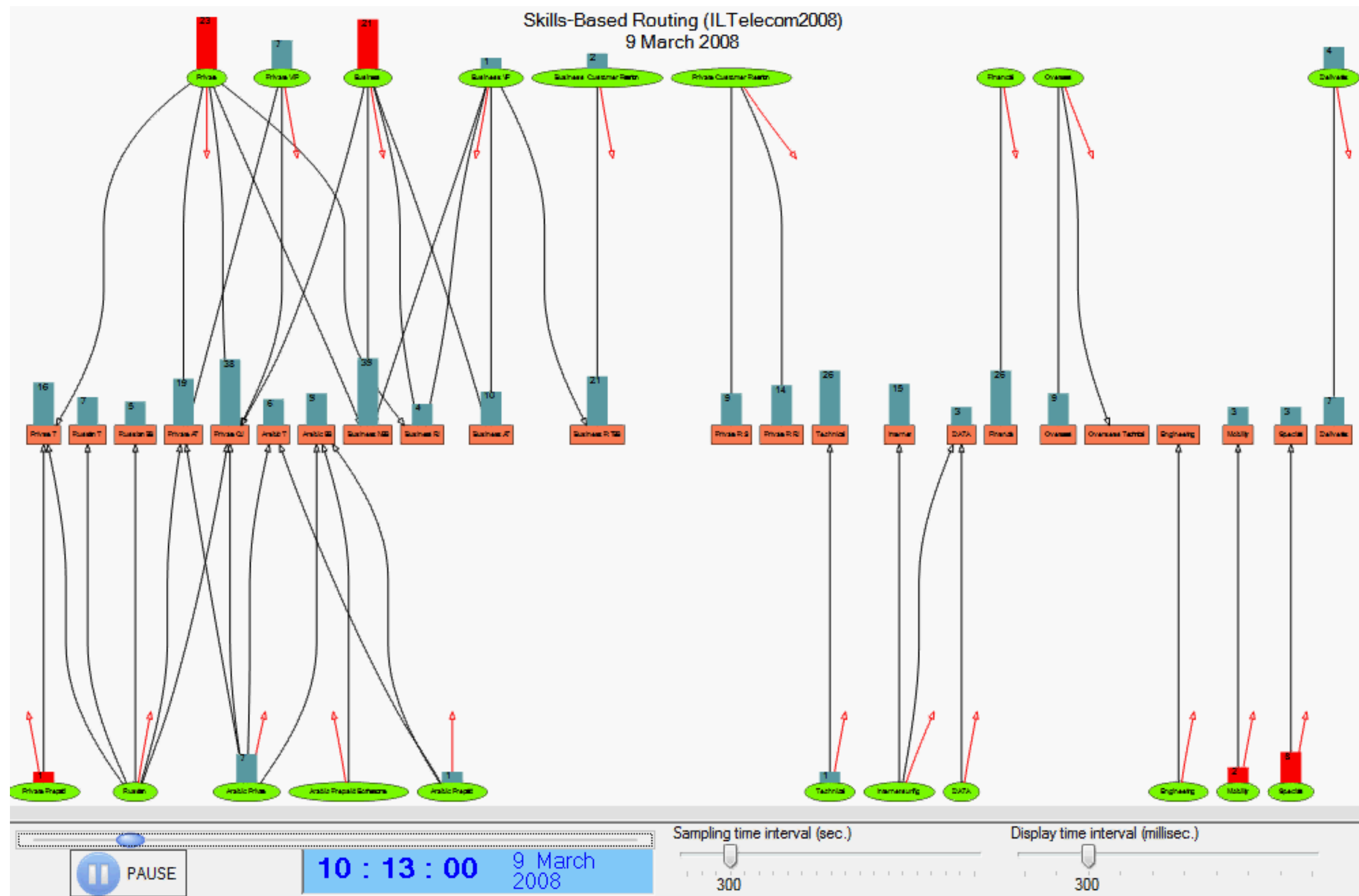
Theory: Tolga Tezcan & Jiheng Zhang, Routing & Staffing, OR 2014



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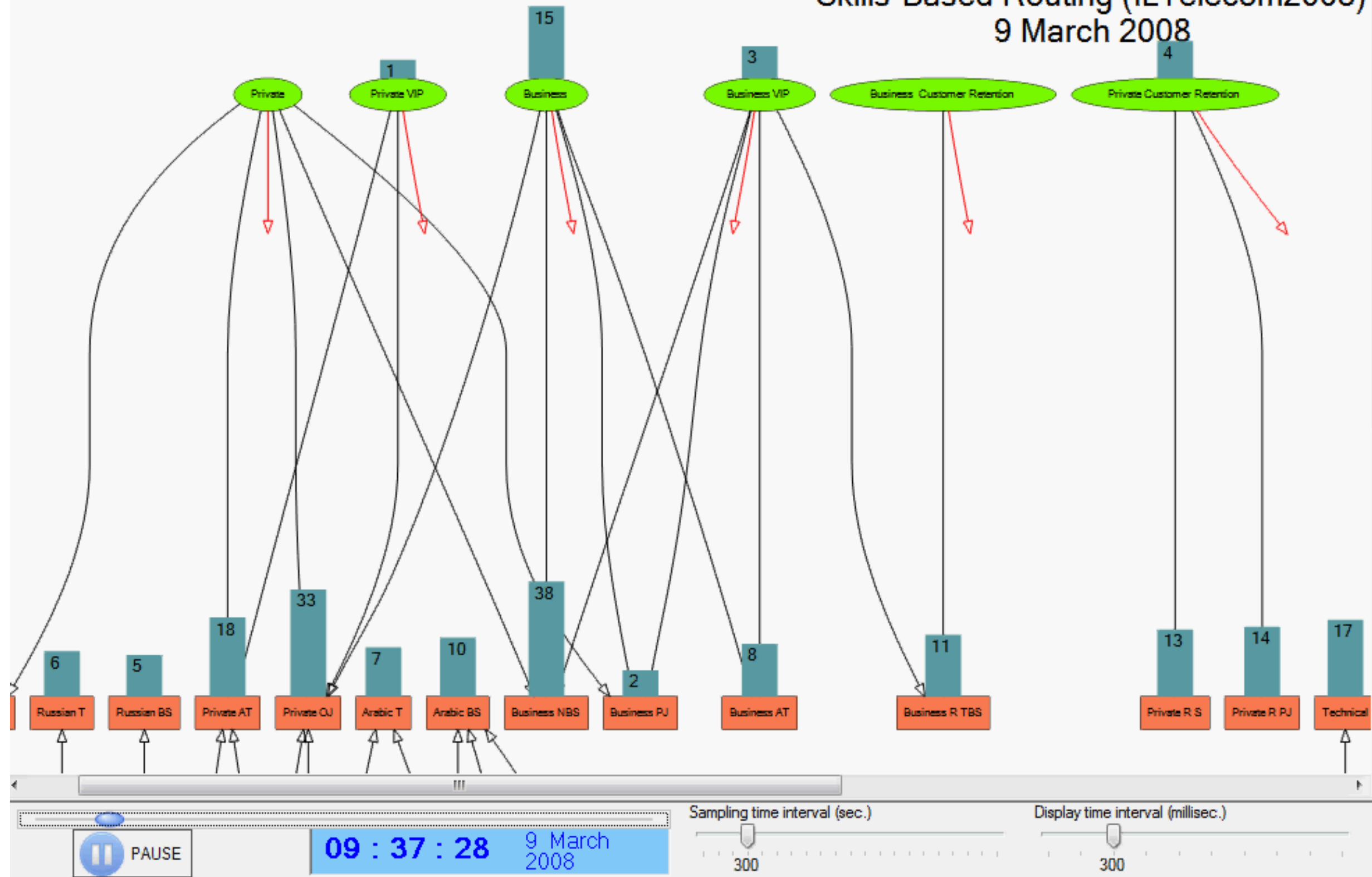
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Above: Simple (Parsimonious) models of complex realities, yet not too simple (Robust)



Skills-Based Routing (ILTelecom2008)

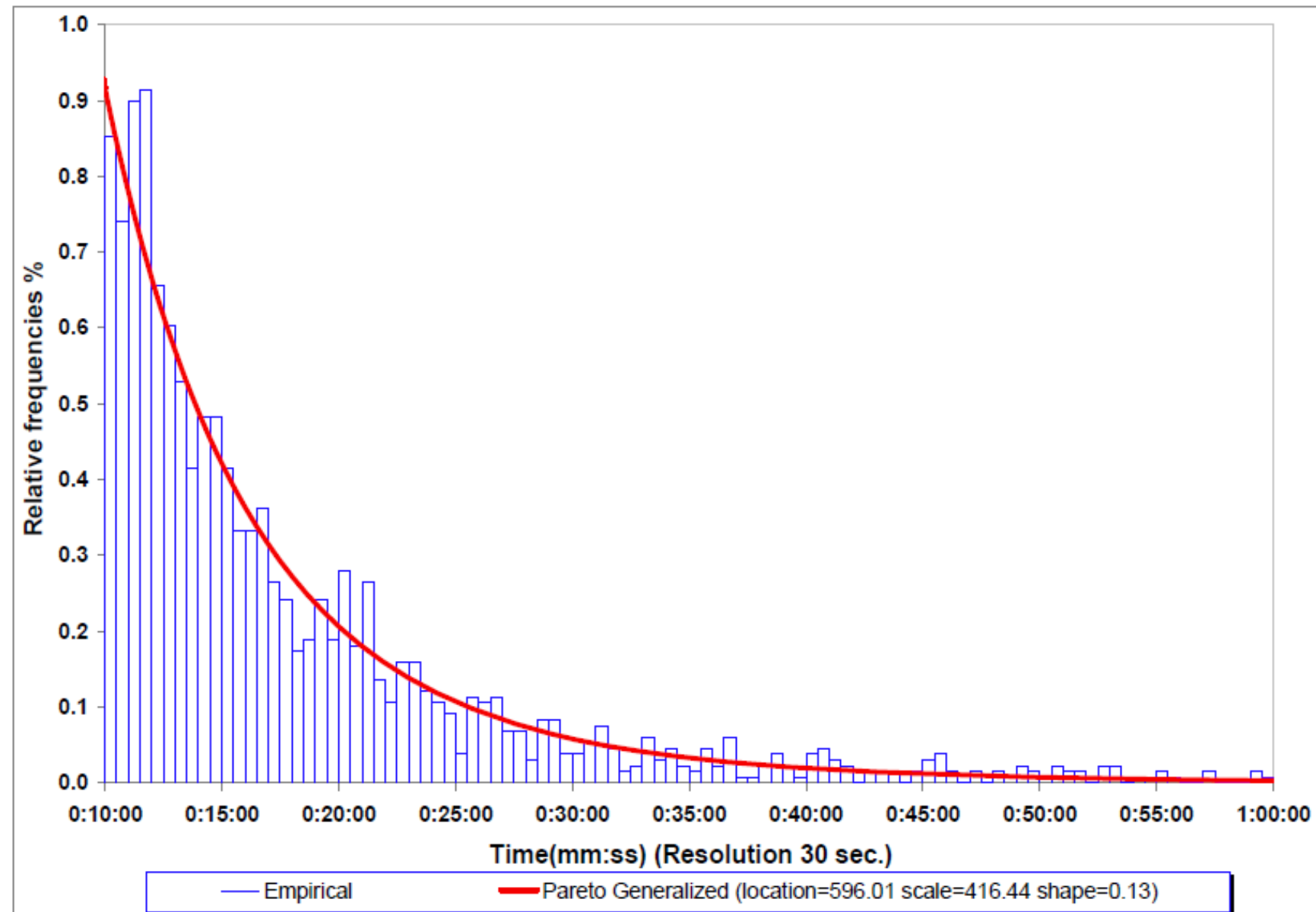
9 March 2008



Primitives: (Im)Patience

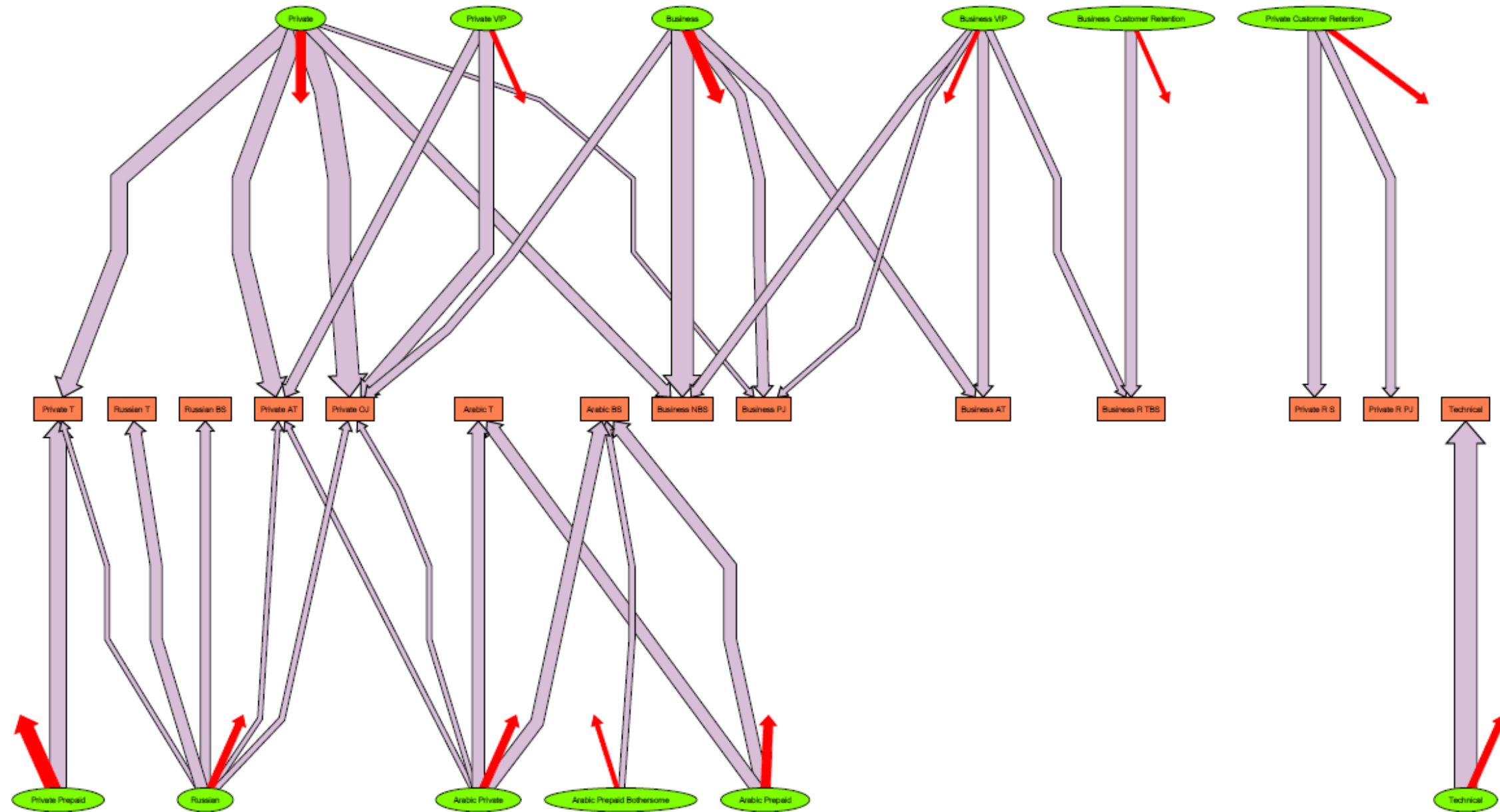
Israeli Bank: Uncensored 13,000 Customers, 24/11/2008

Patience $\geq 10min$: Why Pareto Tail?



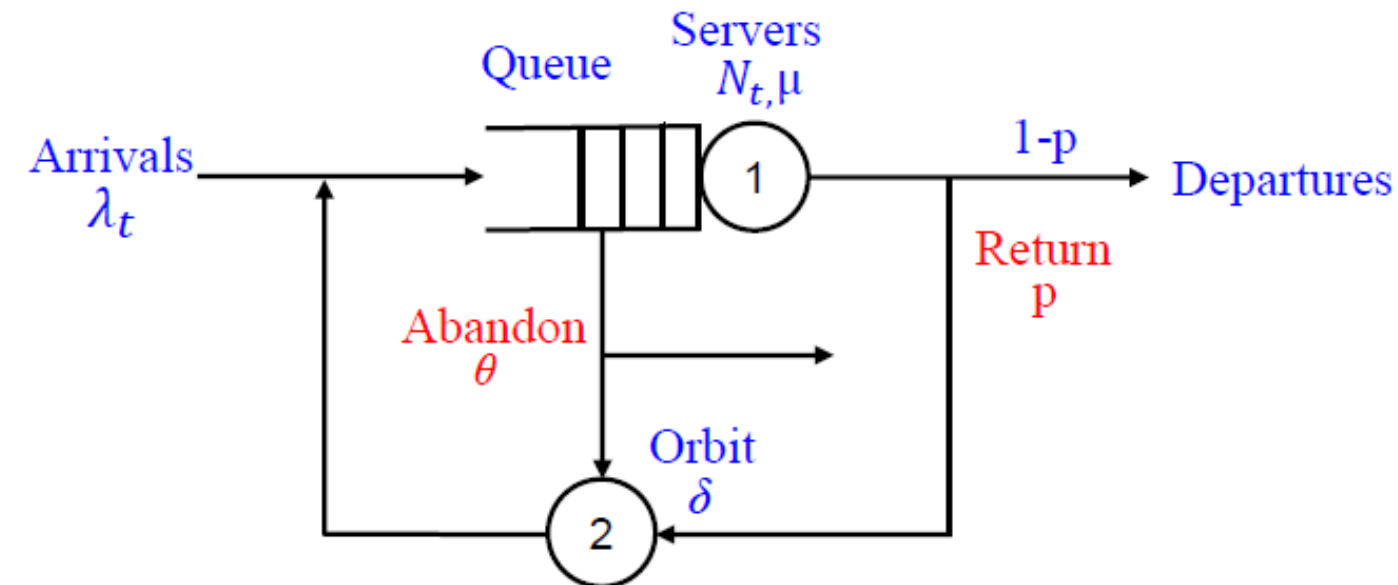
Impatient Customers - Isolate or Aggregate

ILTelecom 9/3/2008



Model Selection: As Simple as Possible but Not Simpler

Service with Retrials and Abandonment; w/ Massey, Reiman, Stolyar

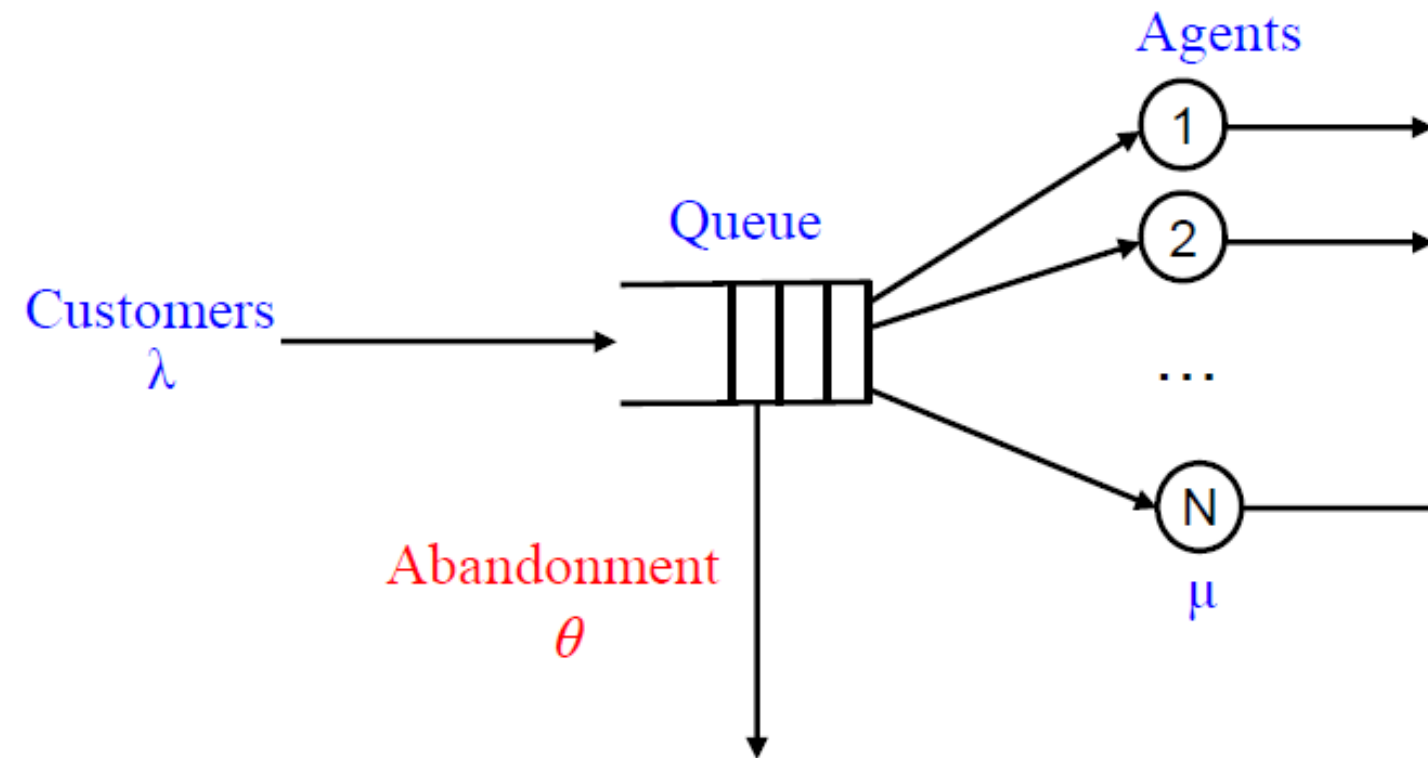


- ▶ Call centers: Visit durations naturally measured in minutes
 - ▶ Arrival rates are "constant" during visit
 - ▶ Returns occur hours after visit

⇒ "Select" Base Model (of 1/2 hour):

Stationary, Abandonment

A Basic Staffing Model: Erlang-A



w/ O. Garnett

“**Birth & Death**” Queue = M/M/N + M (Palm 1940's):

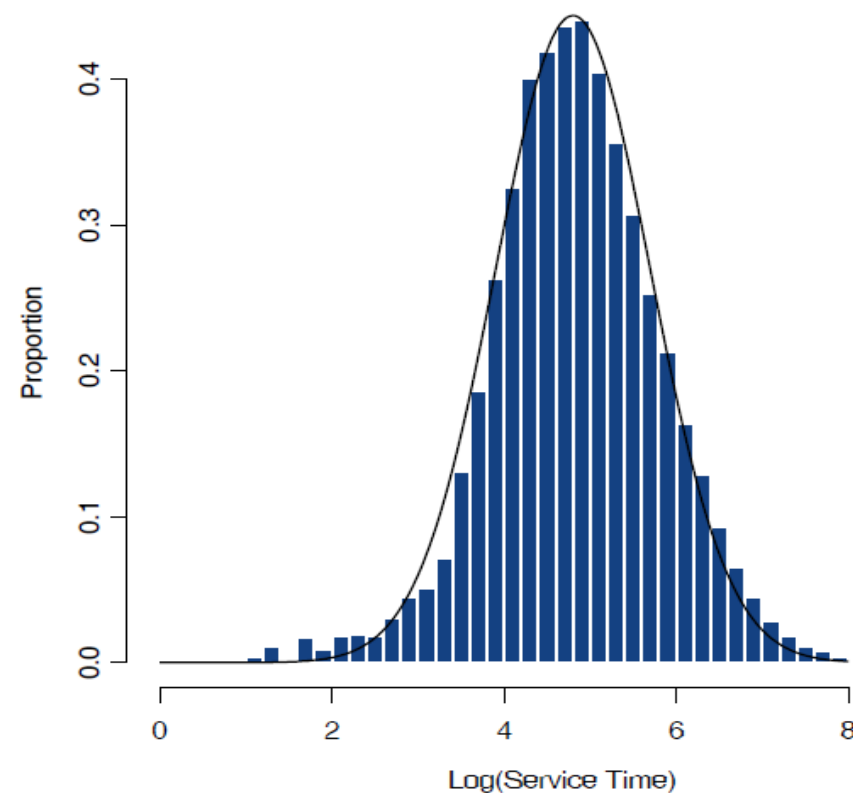
- ▶ λ – **Arrival** rate (Poisson)
- ▶ μ – **Service** rate (Exponential; $E[S] = \frac{1}{\mu}$)
- ▶ θ – **Patience** rate (Exponential, $E[\text{Patience}] = \frac{1}{\theta}$)
- ▶ N – Number of **Servers** (Agents).

Pause for a Commercial: Service Duration in a Call Center

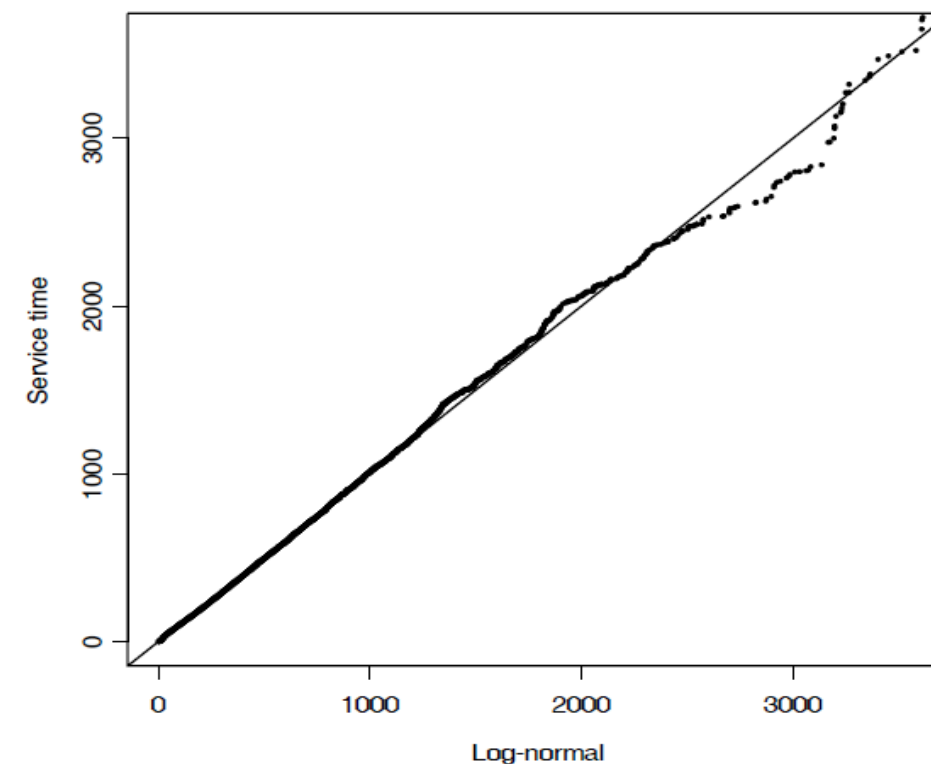
Durations: Phone Calls (Surprises)

Israeli Call Center, Nov–Dec, 1999

Log(Service Times)



LogNormal QQPlot



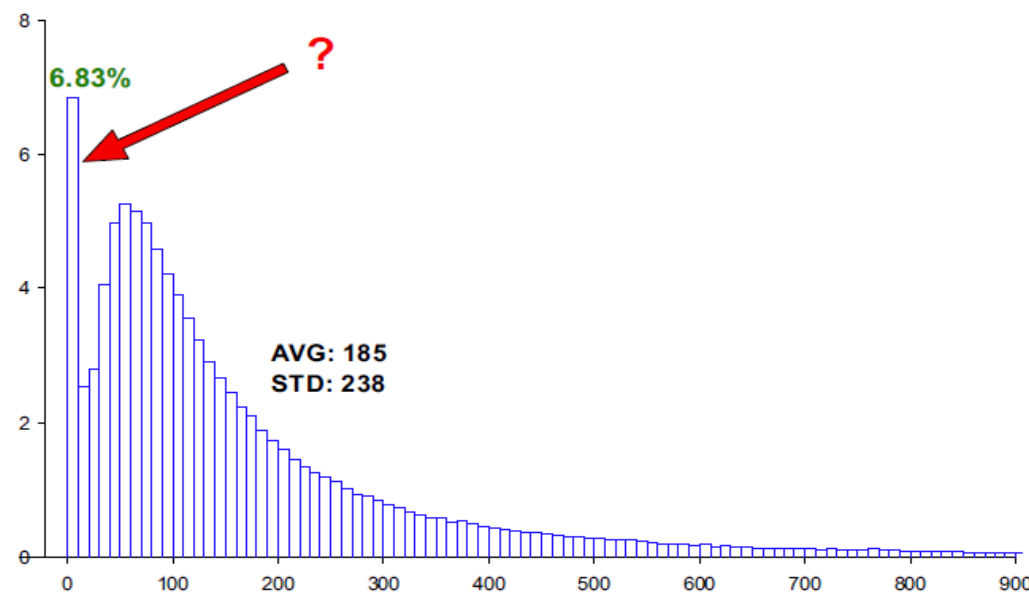
- ▶ **Practically Important:** (mean, std)(log) characterization
- ▶ **Theoretically Intriguing:** Why LogNormal ? Naturally multiplicative but, in fact, also **Infinitely-Divisible** (Generalized Gamma-Convolutions)

Building Blocks of a Model: Service Durations

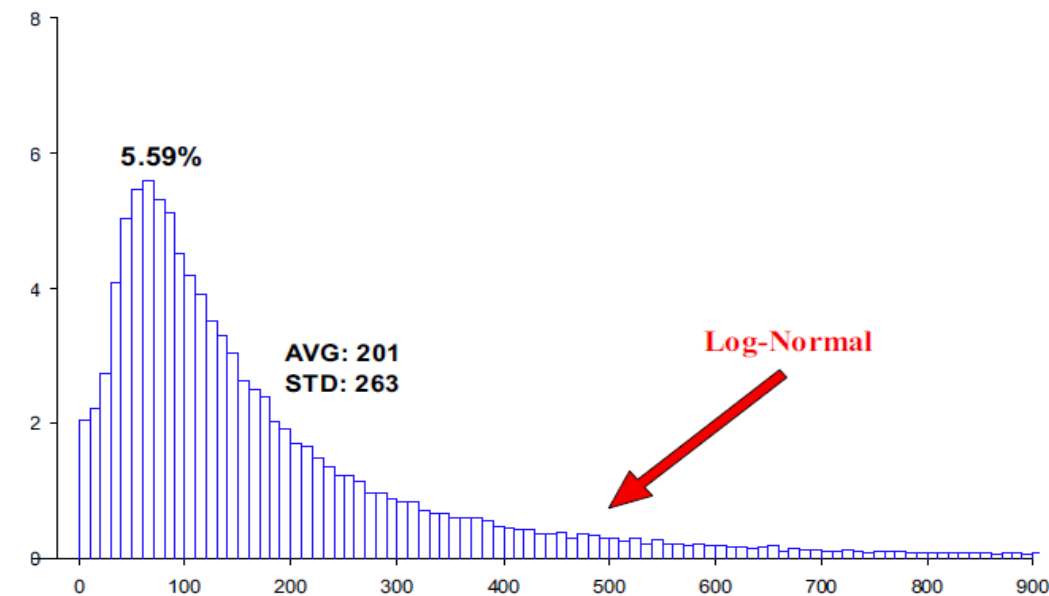
Histogram of Service-Duration in an Israeli Call Center, 1999

Why short services? Why LogNormal?

January-October



November-December



- ▶ January-October: **6.8% Short-Services** (≤ 10 seconds) ?
- ▶ November-December: **LogNormal** durations (common) ?

Erlang-A: Is it Relevant?

Experience:

- ▶ Arrival process **not pure Poisson** (time-varying, σ^2 too large)
- ▶ Service times **not Exponential** (typically close to LogNormal)
- ▶ Patience times **not Exponential** (behavior-dependent).
- ▶ Building Blocks need **not be independent** (eg. long wait associated with long service; w/ **M. Reich & Y. Ritov**)
- ▶ Customers and Servers **not homogeneous** (classes, skills):
w/ **R. Atar, G. Shaikhet; R. Atar, I. Gurvich, ...**
- ▶ Customers return for service (after busy, abandonment; dependently:
P. Khudiakov, R. Ghebali, M. Gorfine, P. Feigin)
- ▶ ..., and more.

Question: **Is Erlang-A Relevant? Robust enough? YES !**

- ▶ **Practice**: Staffing engine of Work-Force Management software
- ▶ **Theory**: Theoretical engine of Operational Regimes
QD, ED, QED

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	$\frac{\bar{\Phi}(\hat{\beta} + \sqrt{\theta\mu}T)}{\bar{\Phi}(\hat{\beta})}$	$1 + O(\frac{1}{n})$
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), w/ Garnett & Reiman (02)
 - 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 (multi-regimes)

Pause for a Commercial: Finally a Theorem (QED)

Prerequisite II: Models (D**N**ets, 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.

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

- ▶ **Call Centers:** Wait "**seconds**" for **minutes** service;
- ▶ **Transportation:** Search "**minutes**" for **hours** parking;
- ▶ **Hospitals:** Wait "**hours**" in ED for **days** hospitalization in IW's.

Moreover, a significant fraction not delayed in queue: e.g. in well-run

- ▶ CCs: **50%** served "immediately" & **90%** utilization $\Rightarrow$ **QED**
- ▶ 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

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, ...
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 = 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:**

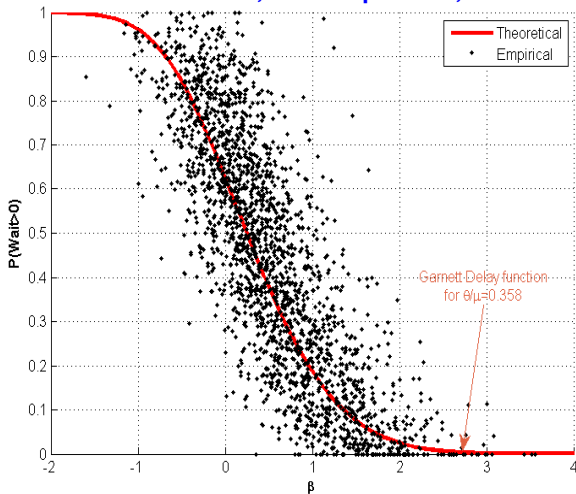
$$P\{\text{Ab} | W_q > 0\} \approx \frac{1}{\sqrt{N}} \cdot \sqrt{\frac{\theta}{\mu}} \cdot [h(\hat{\beta}) - \hat{\beta}].$$

► $P\{\text{Ab}\} \propto E[W_q]$, 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



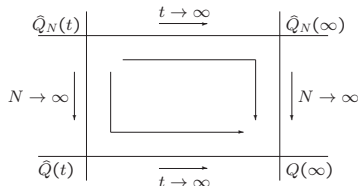
- ▶ 2205 half-hour intervals (13 summer weeks, week-days)
- ▶ Erlang-A approximations for the appropriate $\mu/\theta \approx 3$

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**

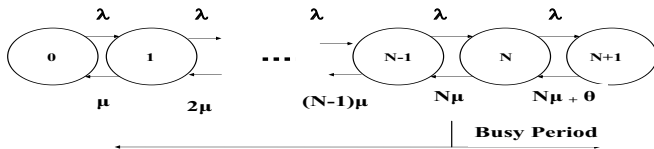
$$\hat{V}_N = \sqrt{N} V_N \Rightarrow \hat{V} = \left[\frac{1}{\mu} \hat{Q} \right]^+_{95}$$

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 $N \uparrow \infty$: $P\{W_q > 0\} \downarrow 0$, $P\{I > 0\} \uparrow 1$.
 - ▶ 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**: $N \approx R + \beta\sqrt{R}$
 $(\lambda \approx N\mu - \beta\sqrt{N\mu})$
2. **Erlang-A** (M/M/N+M), with parameters $\lambda, \mu, \theta; N$, in which $\mu = \theta$:
(Im)Patience and Service-times are equally distributed.
 - ▶ 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 $Z \stackrel{d}{=} N(0, 1)$.
 - ▶ $P\{W_q(M/M/N + M) > 0\} \stackrel{\text{PASTA}}{=} P\{L(M/M/N + M) \geq N\} \stackrel{\mu=\theta}{=} P\{L(M/M/\infty) \geq N\} \approx P\{R + Z\sqrt{R} \geq N\} =$
 $P\{Z \geq (N - R)/\sqrt{R}\} \stackrel{\sqrt{\cdot} \text{ staffing}}{\approx} P\{Z \geq \beta\} = 1 - \Phi(\beta).$

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

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

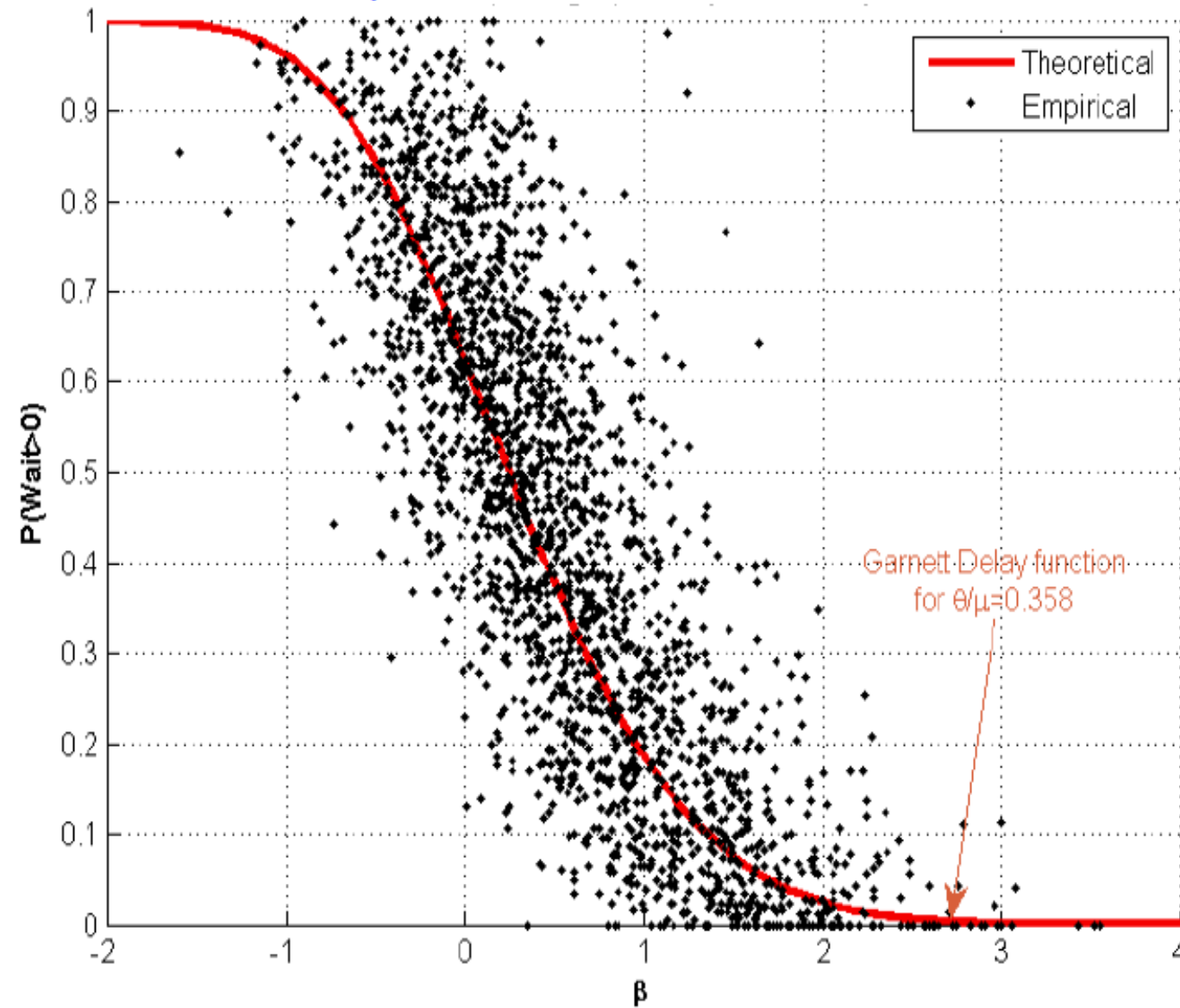
Special case: $\mu = \theta$ (Impatient):

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

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

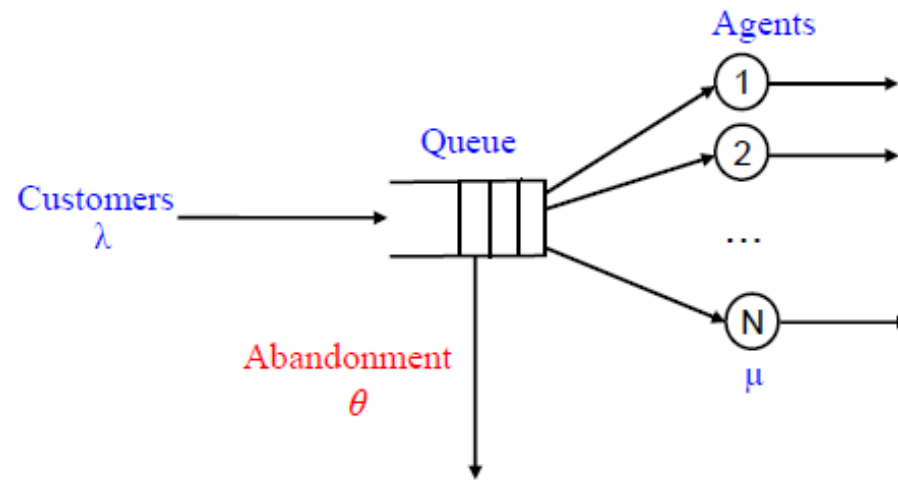
Erlang-A Value: DNet $P(W_q > 0)$ vs. Data

IL Telecom; June-September, 2004 (2205 30min intervals, weekdays)



- Approximations, w/ Patience $\approx 3 \times$ Service-Duration ($\mu/\theta \approx 3$)

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



w/ I. Gurvich & J. Huang

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

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

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

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

- **DNet**: Universal (Stochastic) Approximation – SDE

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

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

Parsimonious (Tractable, Robust), Accurate, Valuable

Abandonment: Further Applications

- **Personalized Queueing**: e.g. when Impatience has been estimated (personalized), exactly or approximately, choosing who to serve next will benefit from taking this information into account (e.g. Shortest-Patience-First); **w/ P. Momcilovic**
- **CCs**: Control of congestion via announcement / information that monitors the queue, **w/ Junfei Huang, Hanqin Zhang, Jiheng Zhang**
- **IVRs** (Self-Services): identify via Mixture Fitting, **w/ N. Carmeli, H. Kaspi**
- **EDs**: Left-Without-Being-Seen. Theory exists if All-Unknown – Current-Status Data. Otherwise, namely when having both Right-Censoring and Current-Status data, theory developed **w/ Y. Yefenof, Y. Goldberg, Y. Ritov**
- **Chats**: Before or within session (registered vs. silent abandonment)

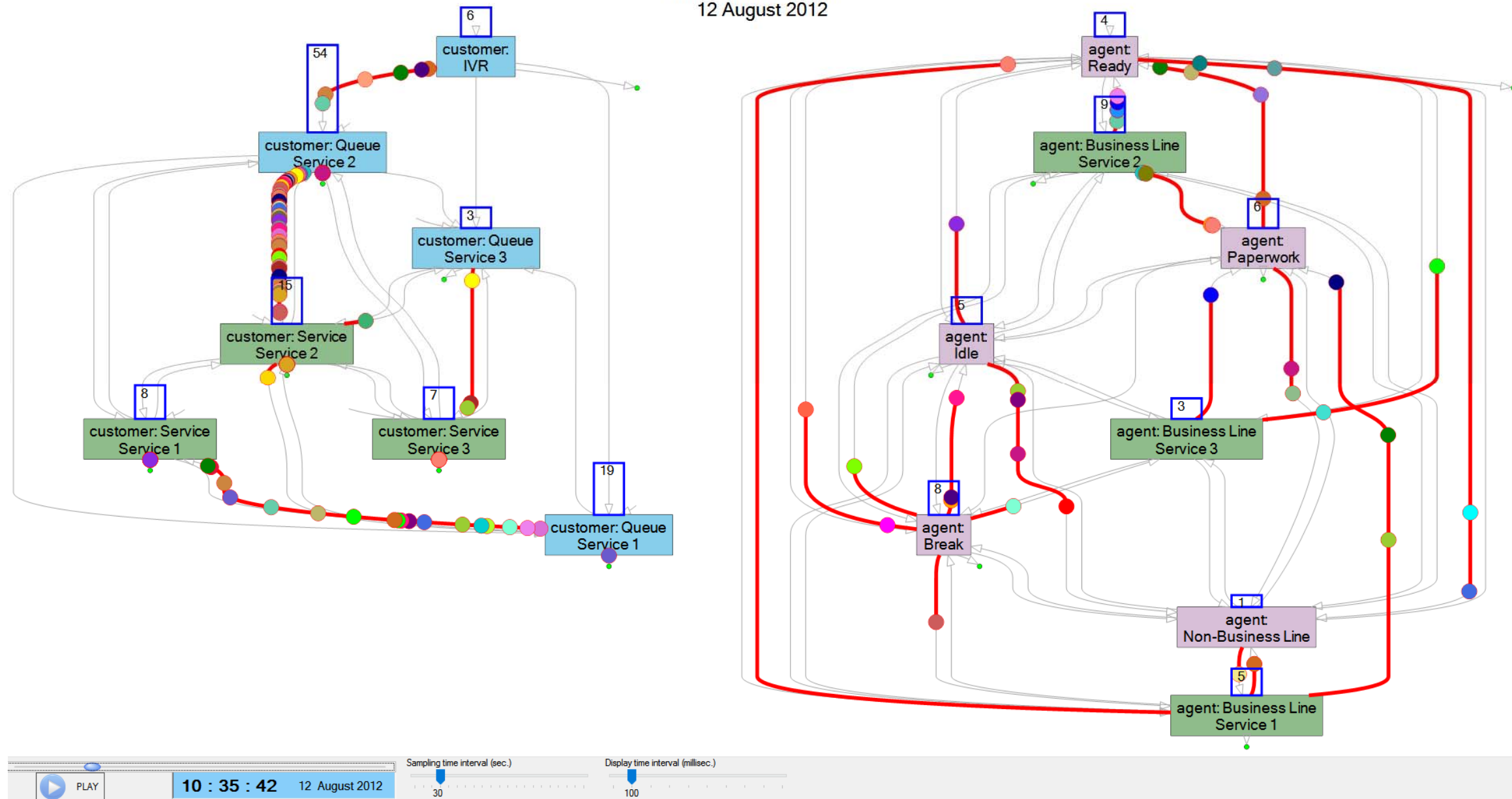
Laws & Models: Data-Based Erlang A/R/S, following B/C

- Little's Law (Steady-State, Transient), State-Space Collapse,...
- Erlang-B (**B**locking) and Erlang-C (**?**)
 - **Erlang**, Agner Krarup: Queueing Theory was born in 1909, in his paper "The Theory of Probabilities and **Telephone Conversations**"
- 2. Erlang-A
 - **Abandonment**: While waiting for service, does service-value dominate residual-wait-cost
- 1. Erlang-R
 - **Return/Feedback**: Customers often return to service (positively, negatively, just needing)
- 3. Erlang-S
 - **Servers**: Challenging to manage, and model, no less so than customers

Above: Simple (Parsimonious) models of complex realities, yet not too simple (Robust)

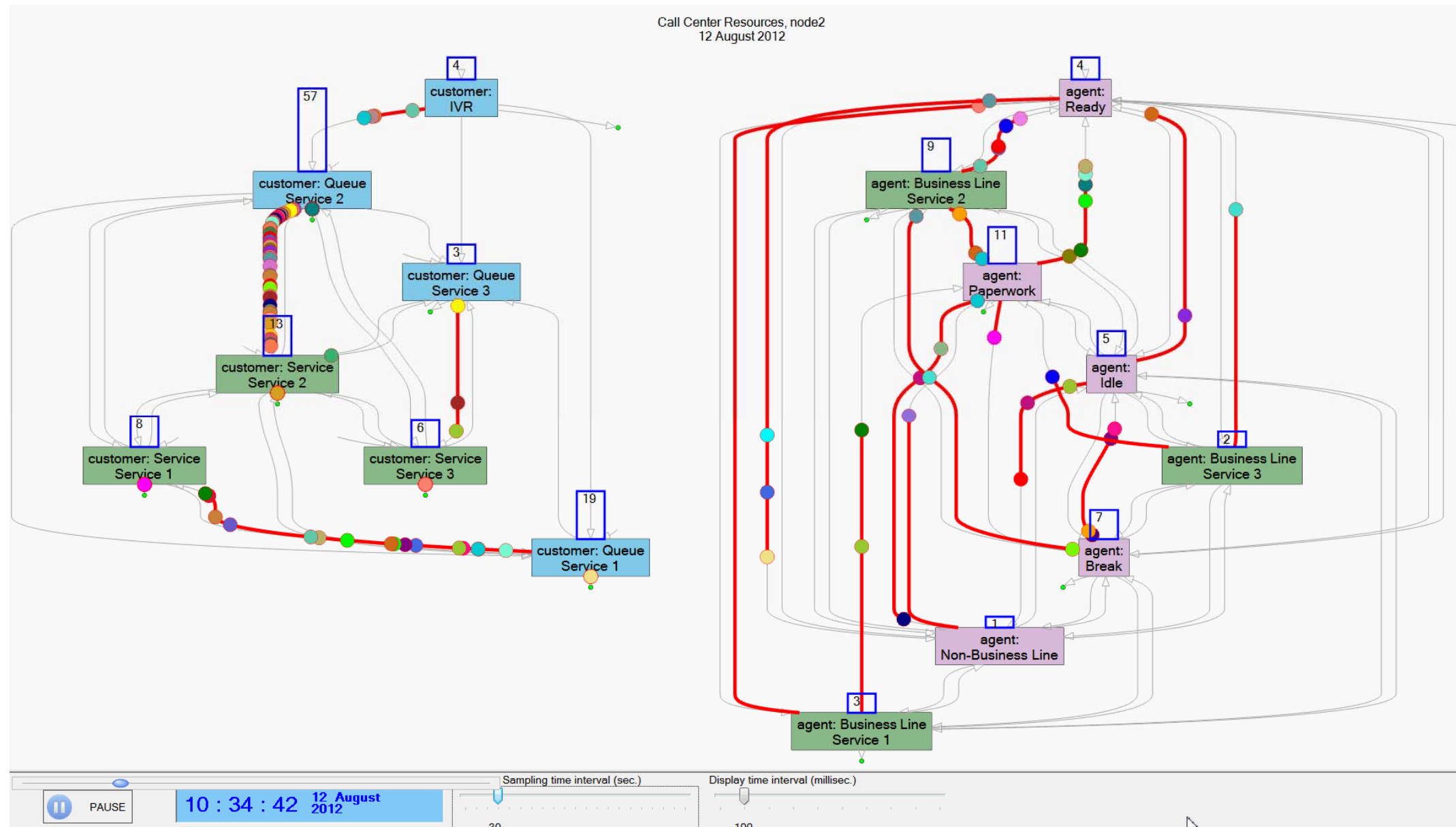
Telephone Services: Customers and Servers - Symmetric Viewpoint

Call Center Resources
12 August 2012



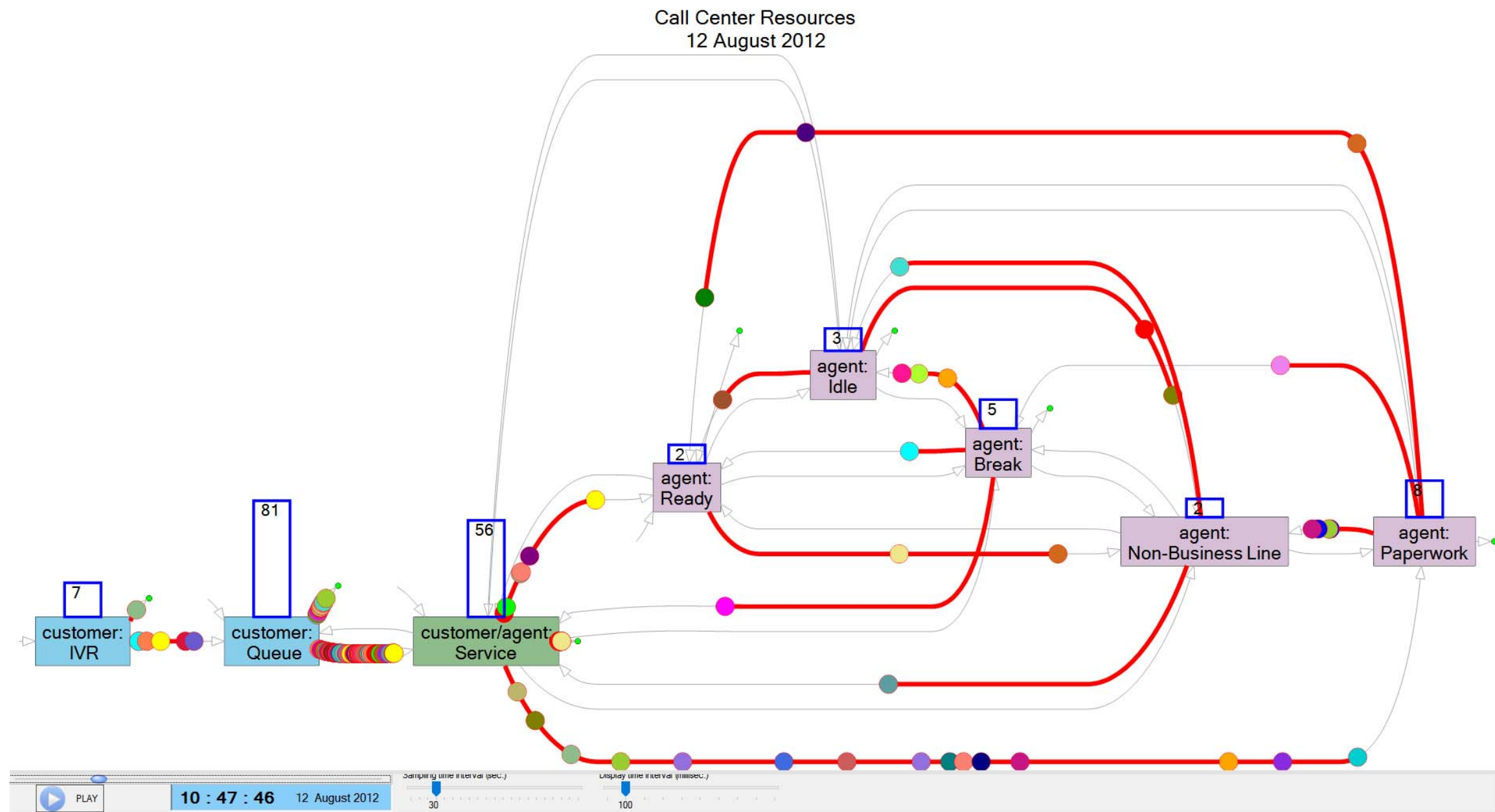
Telephone Services: Customers and Servers

Symmetric Viewpoint



Symmetric View: Customers & Servers

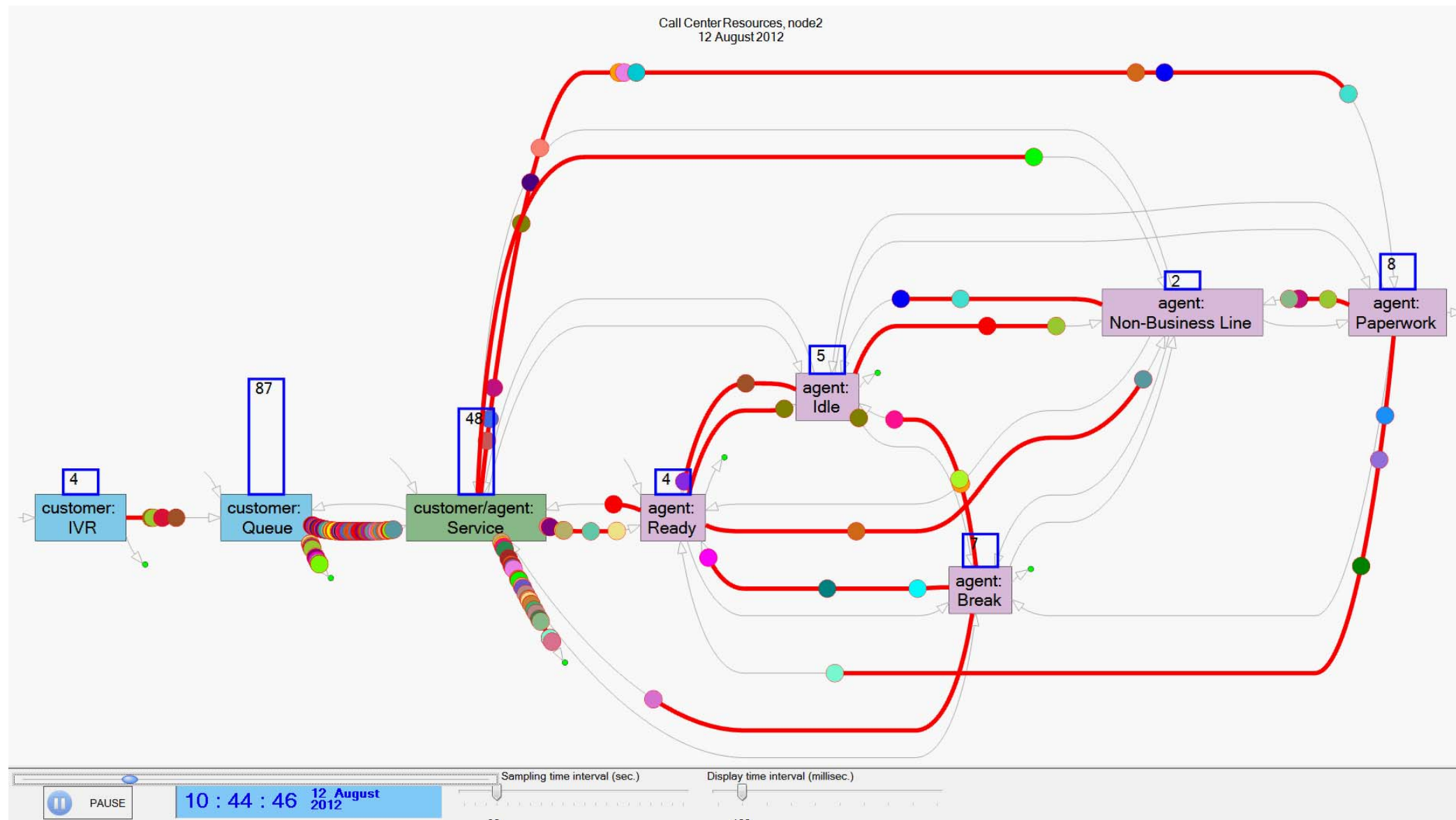
Originates in Resource-Driven Activity-Networks
w/ Armony, Carmeli, Momcilovic



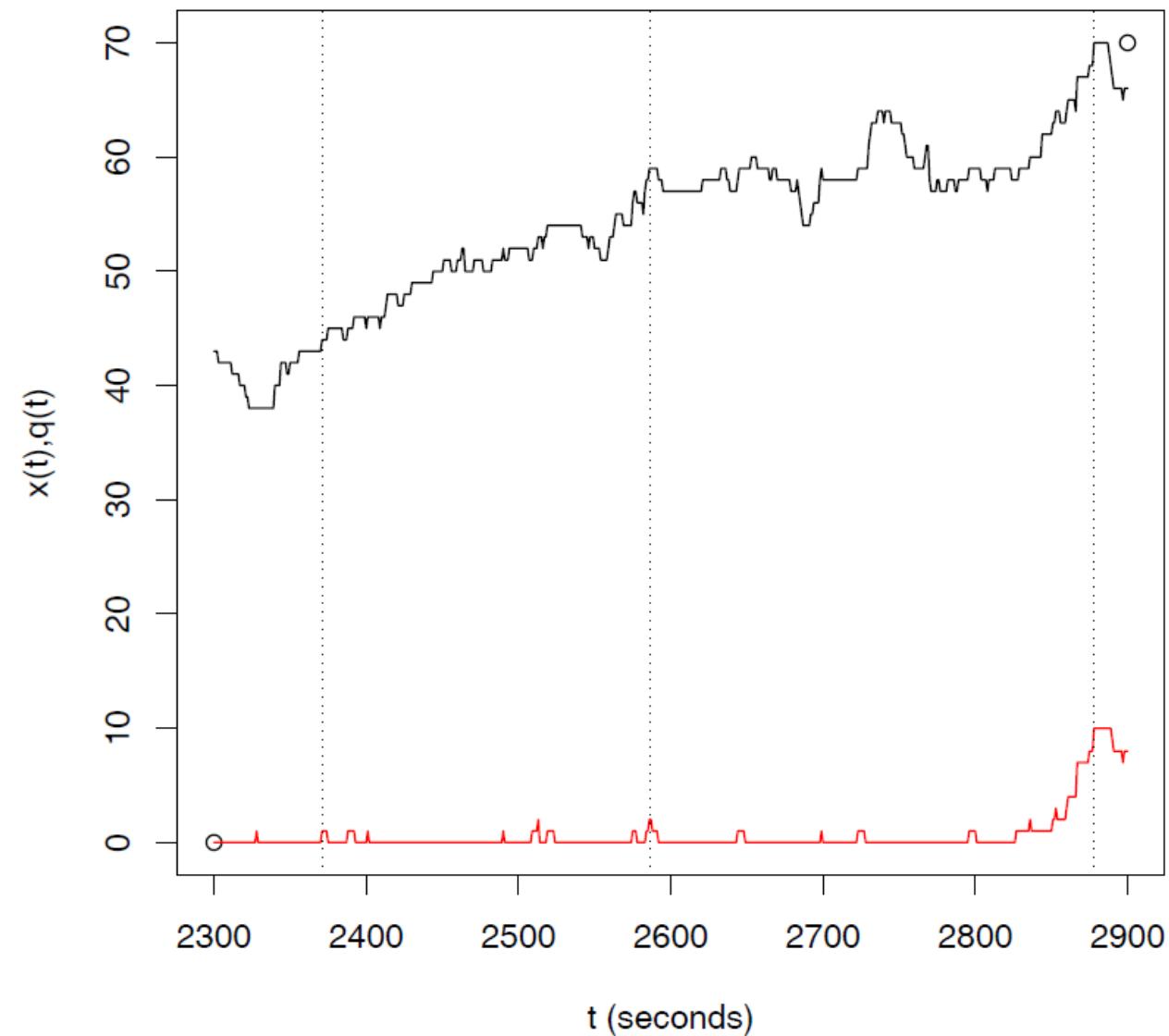
Symmetric View: Customers and Servers

Originates in Resource-Driven Activity-Networks:

w/ Armony, Carmeli, Momcilovic



Number of Servers: Present (=Constant) vs. Available (Random)

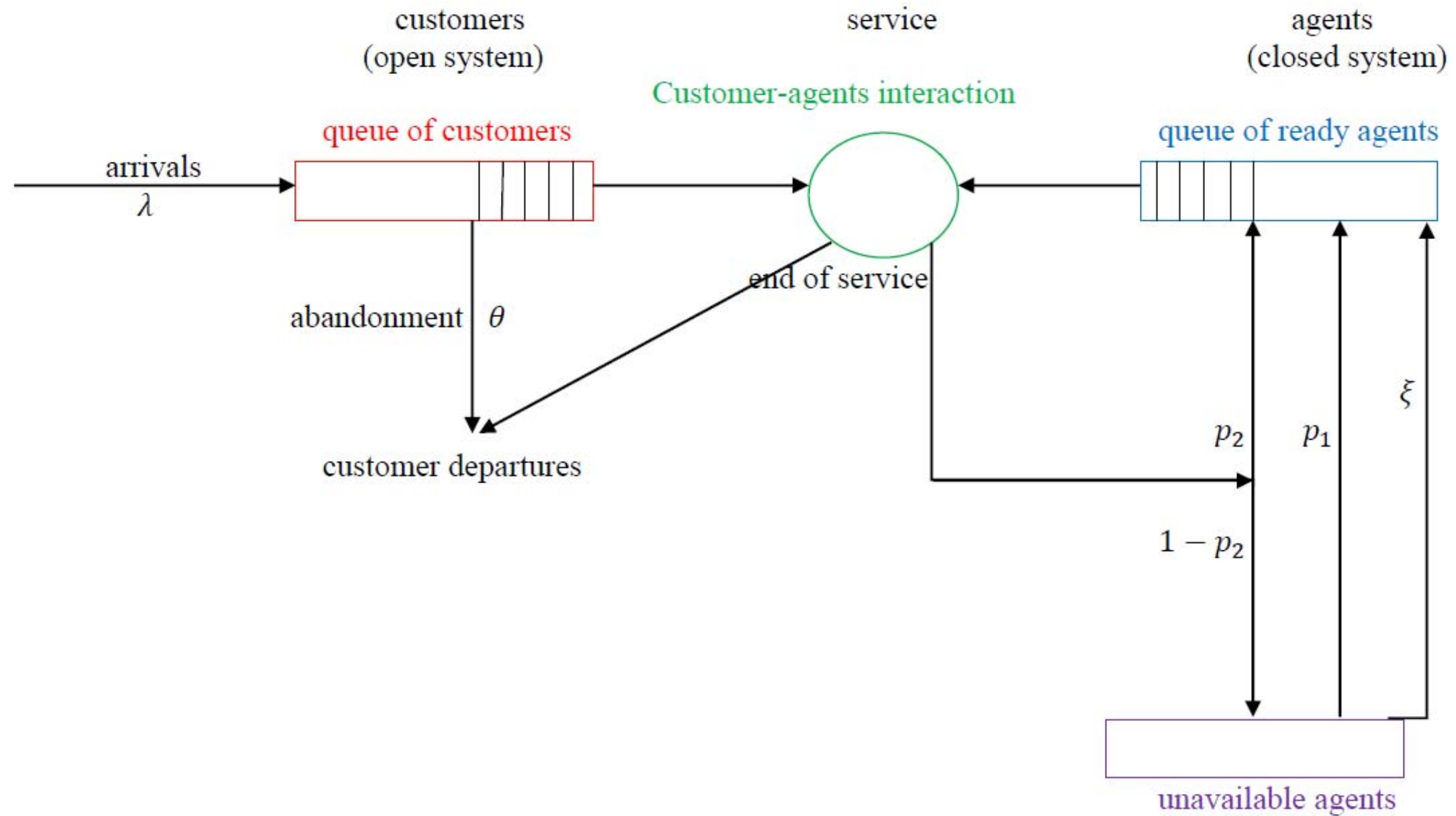


$x(t)$, $q(t)$ in real data (U.S. Bank; telesales; 12/1/2002)

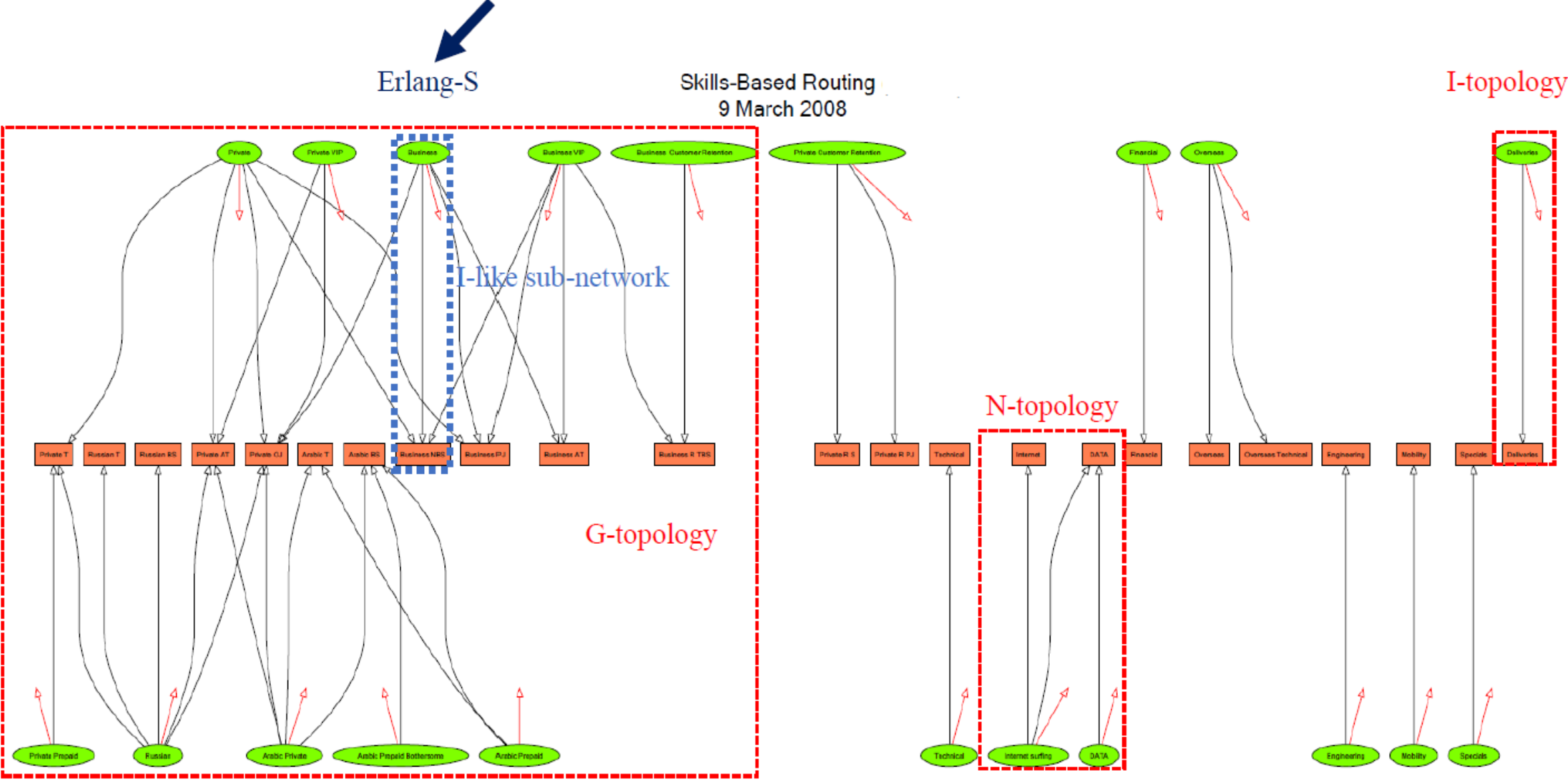
$n(t) = (43, 57, 60)$ @ $t = (10:36:54, 10:43:07, 10:47:23)$:
17 agents became available within 11 minutes

Erlang-S

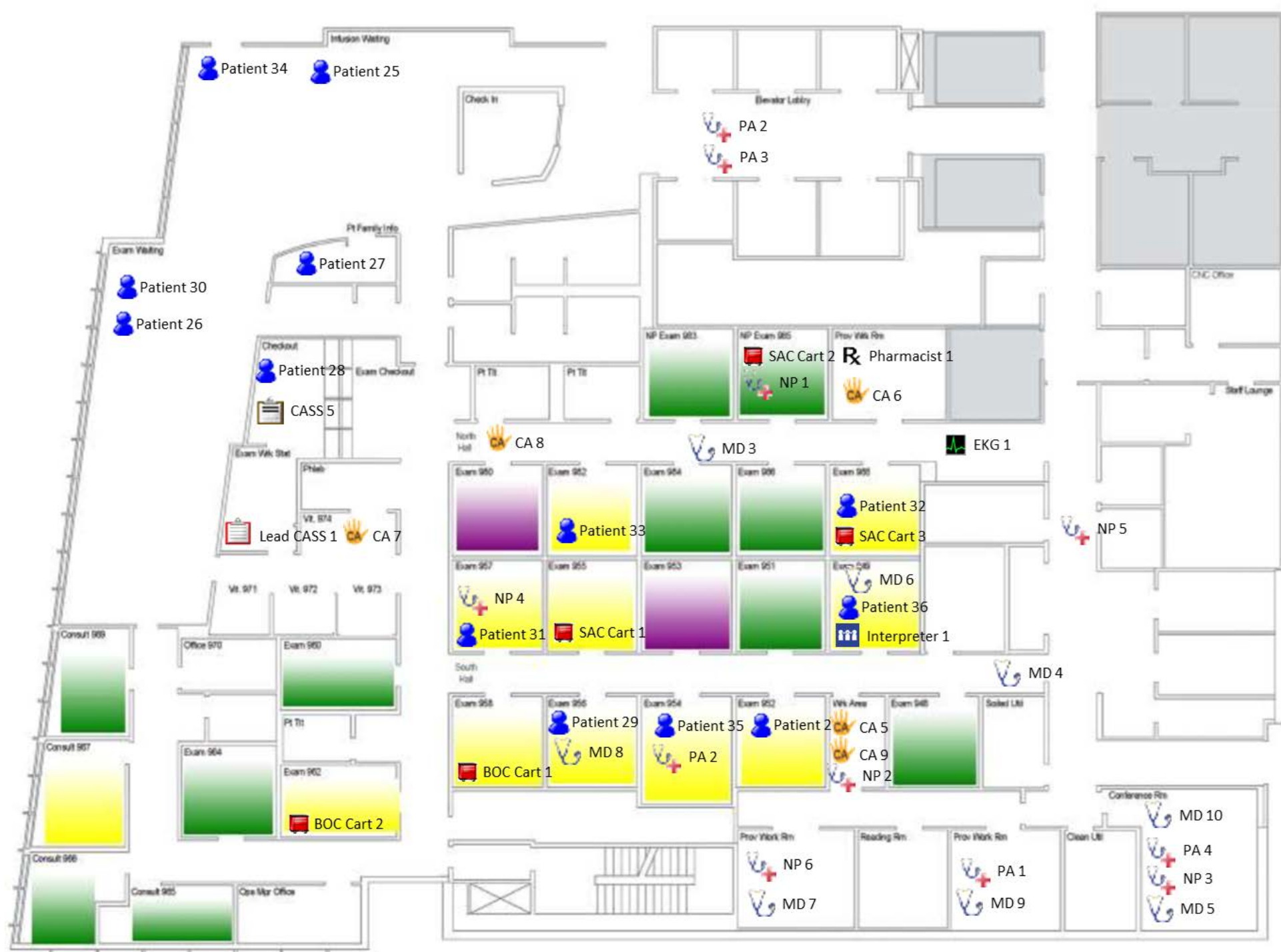
w/ David Azriel & Paul Feigin



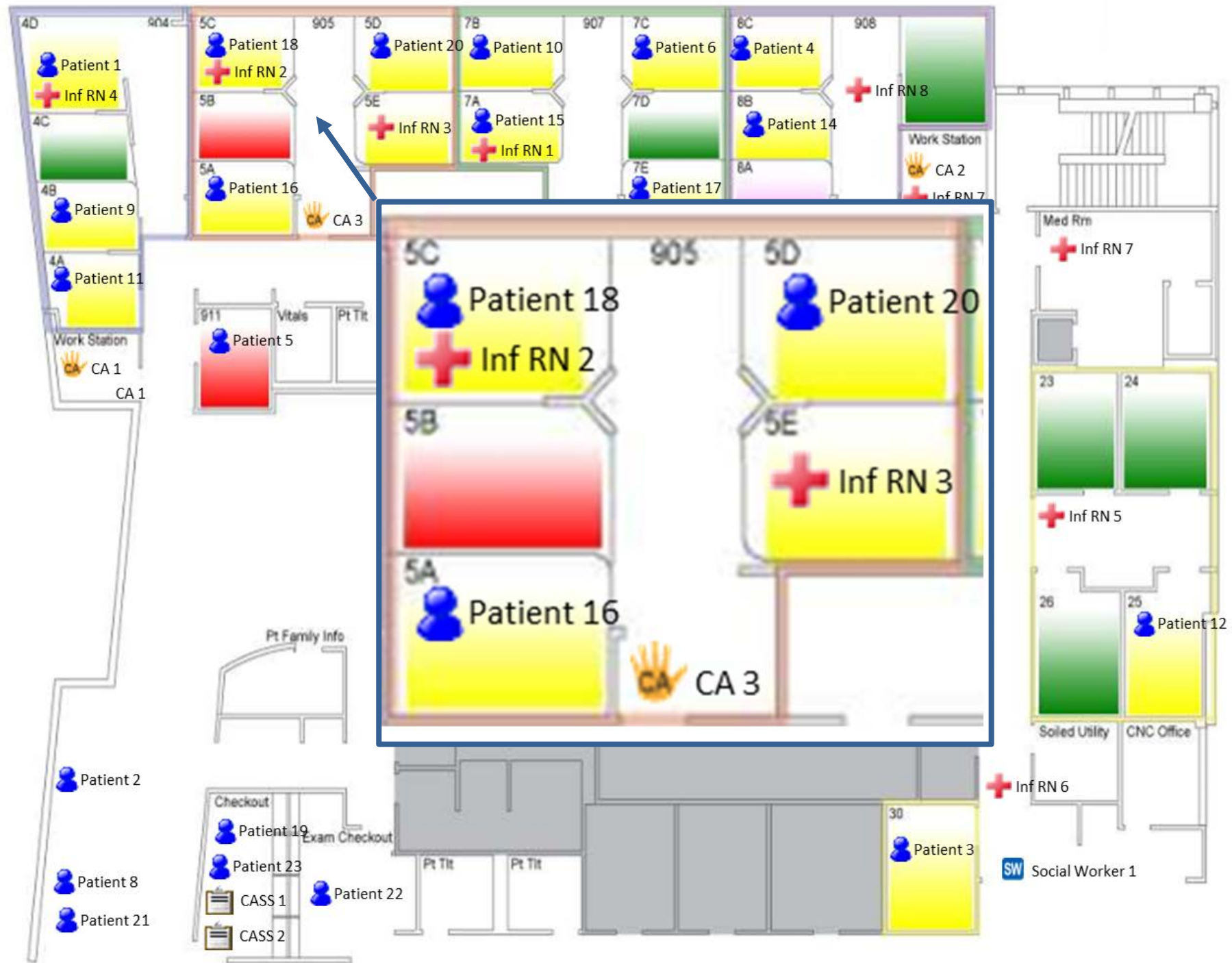
Expressiveness of Erlang-S (Zoom): Servers in Qnets



Topology of a call center:
Server-queues are in the **rectangles** and customer-queues are in the **ovals**







System Badge Type Icons

STAFF



Physician



Nurse Practitioner / Physician Assistant



Fellow



Clinic Assistant



Inf Nurse



Research RN



Program RN



Clinical Nurse Coordinator



Phlebotomist



Pharmacist



Social Worker



CASS



CPC



Ambassador



Manager



IV Nurse



Lead CA



Charge RN



Charge Pharm



Pharm Tech



Scheduler



Interpreter



Nutritionist



Respiratory Therapist



Genetic Counselor

Genetic Counselor



Lead CASS

Lead CASS



CRC

CRC



Exam Practice RN

Exam Practice RN



Massage Therapist

Massage Therapist

PATIENT, FAMILY & OTHER



Patient

Patient



Family

Family



Versus

Versus Staff



SysAdmin

System Administrator



Other Staff

Other Staff

EQUIPMENT



Thermometer

Thermometer



IV Pump

IV Pump



CAS Monitor

CAS Monitor



Stretcher Chair

Stretcher Chair



O2 Sat Monitor

O2 Sat Monitor



AccuVein

AccuVein



Chart

Chart



Phone

Phone



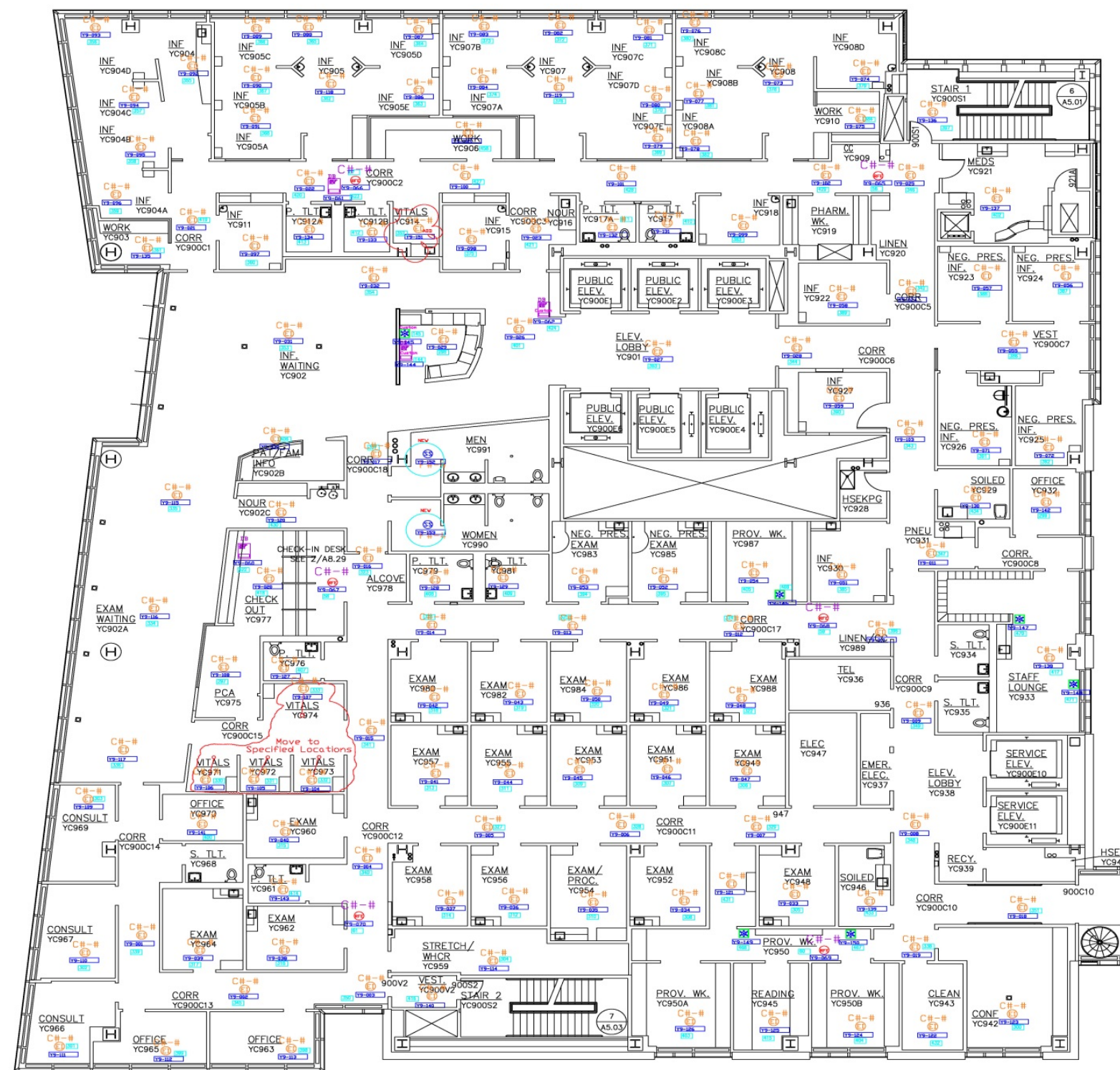
Cart

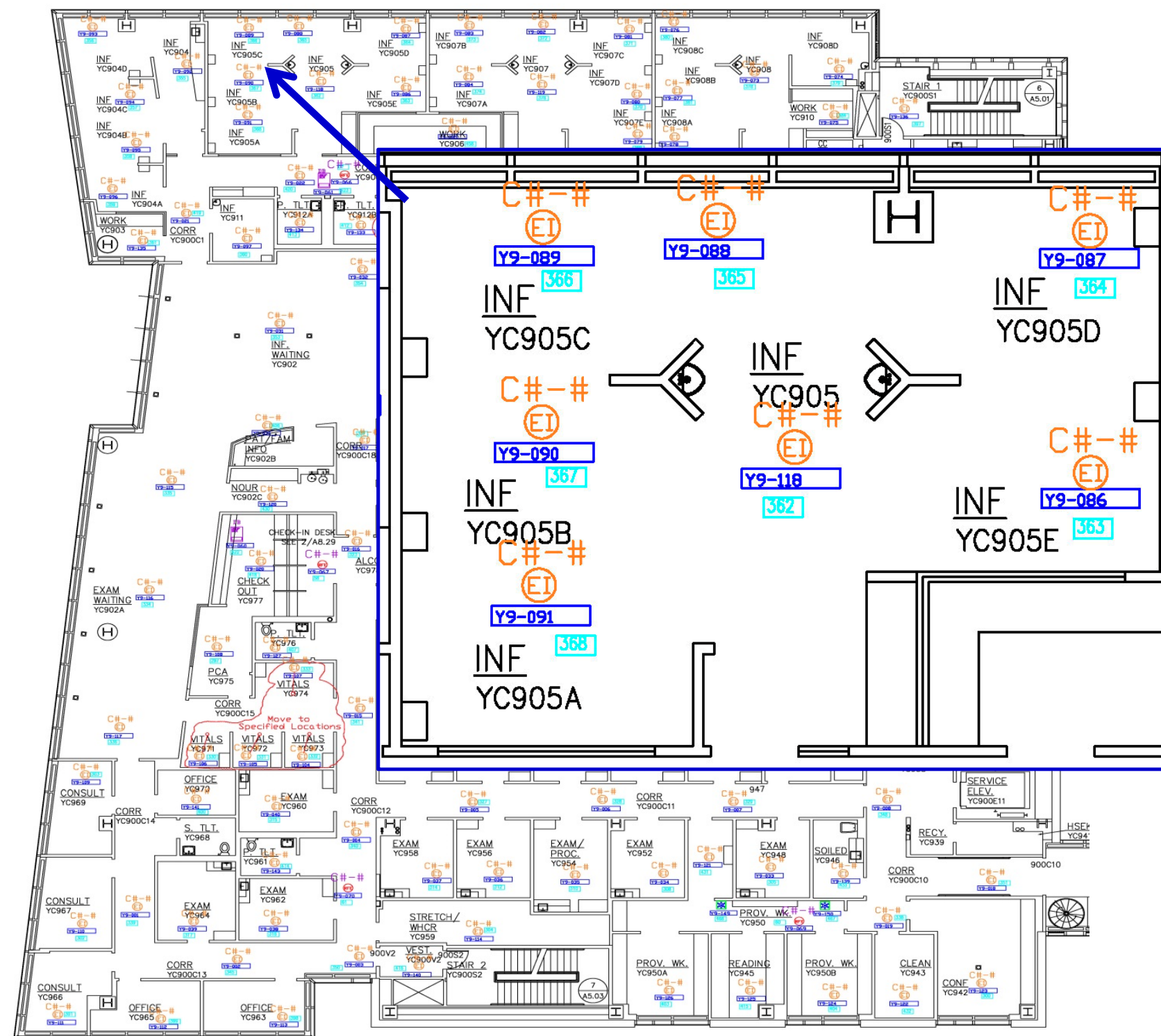
Cart



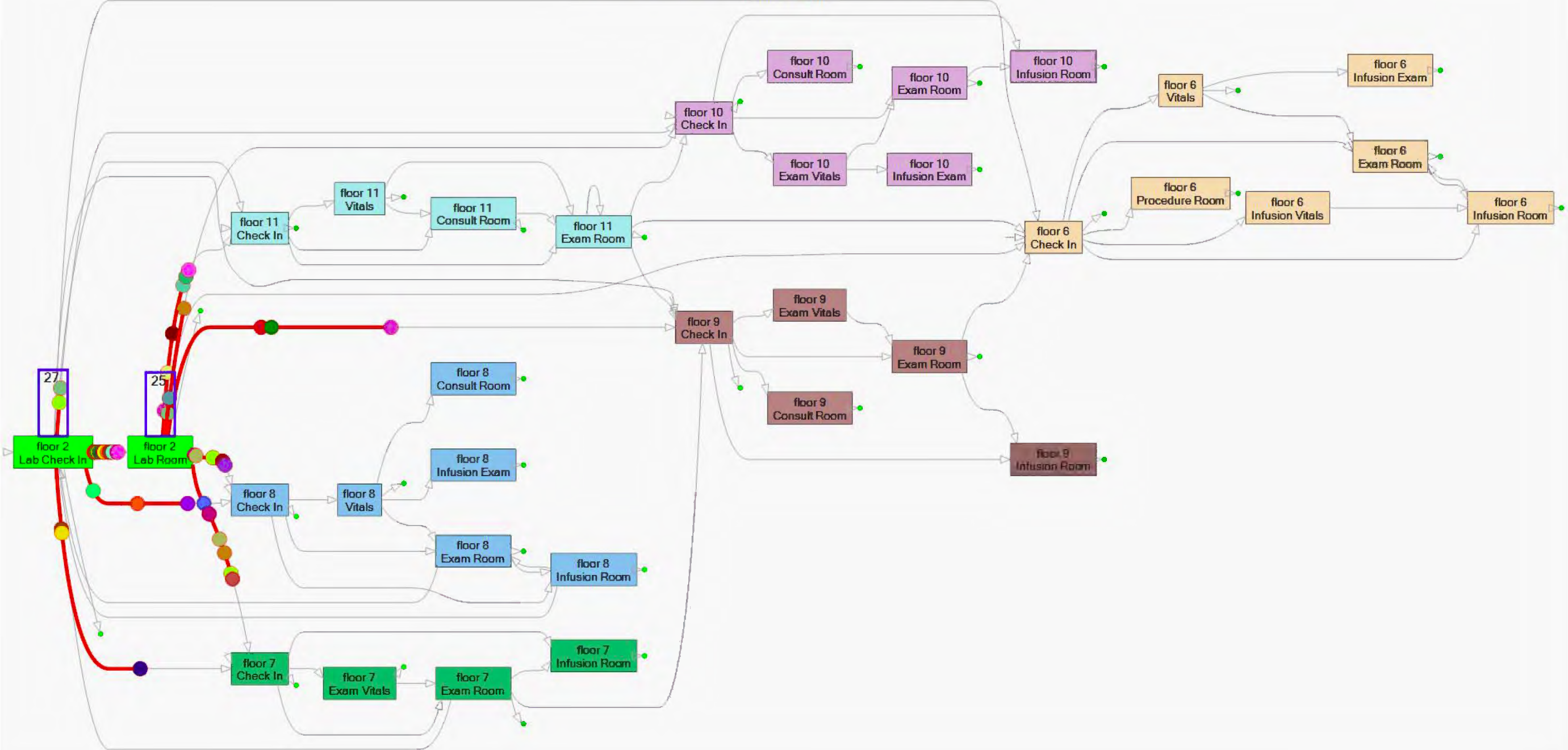
EKG

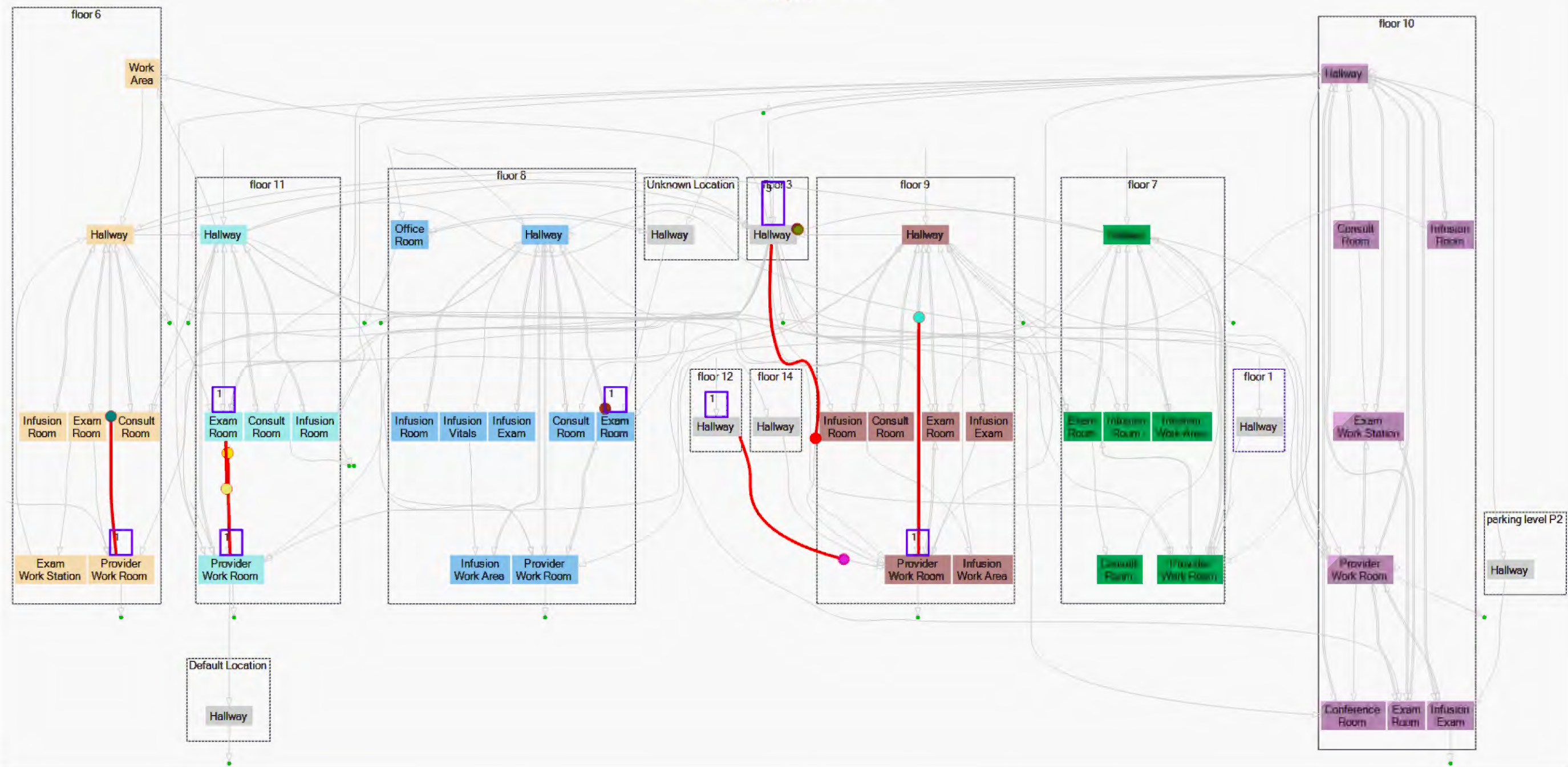
EKG





Patients flow (by floors), Process Mining + Operations Research view (DayHospital)
3 December 2014





Applications in DFCL

Control: rooms status, physicians location, long wait times

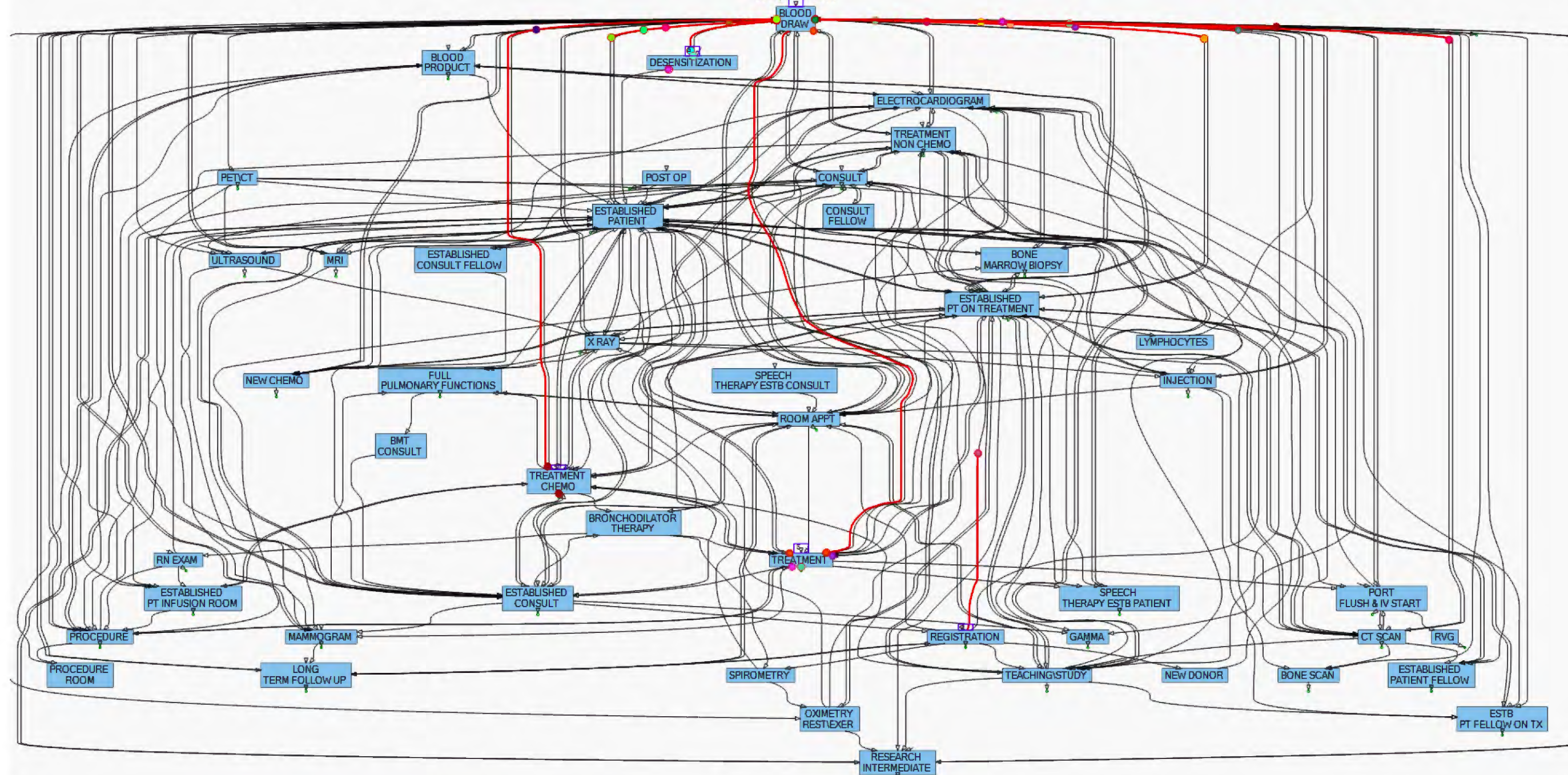
Planning: number infusion chairs, load-balancing among floors

Management: evidence-based

Design/Plan vs. Performance: exam durations (no 15-minutes)

Motivating improvement: room for physician vs. room for patient

Patients appointments (Day/Hospital)
30 September 2014



PAUSE

07 : 04 : 30 30 September 2014

time interval (sec.)

30

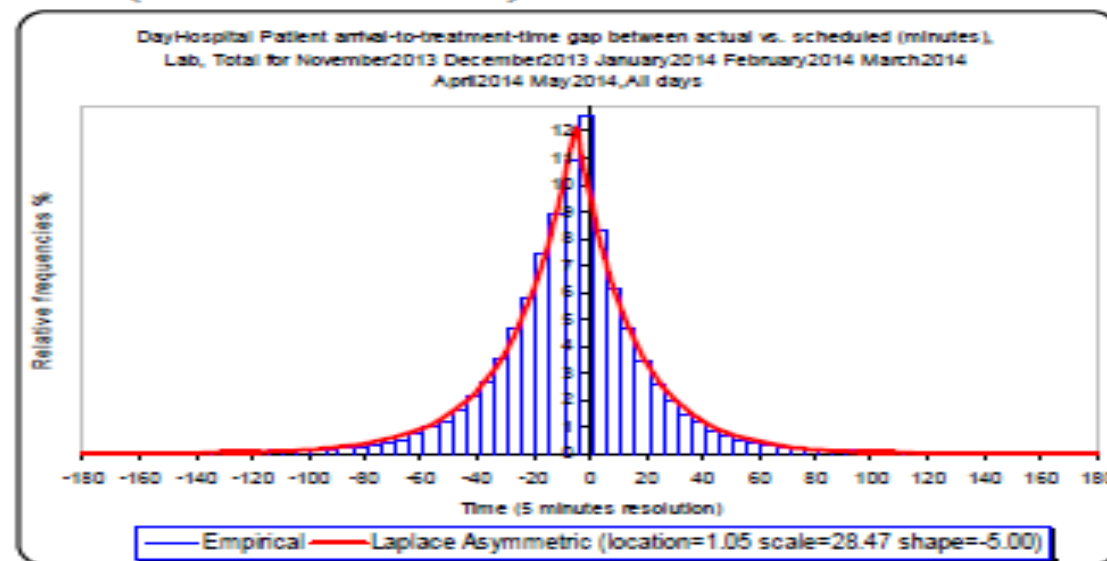
Display time interval (millisec.)

100

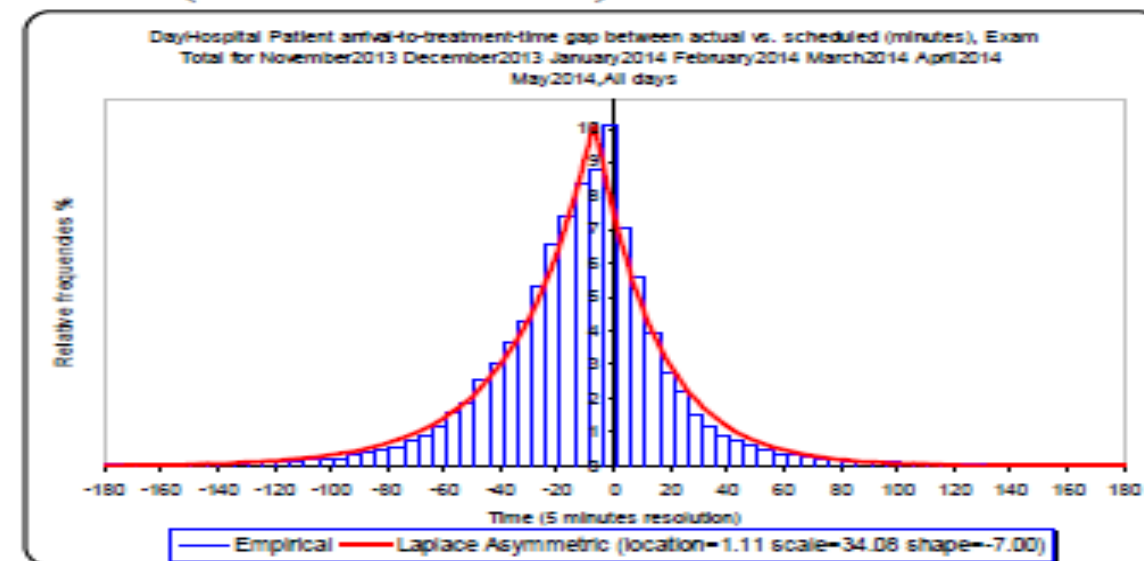
Planned vs. Actual: Improve **Plan**, Reduce **Gap**, then **Control**

w/ A. Senderovich & A. Gal

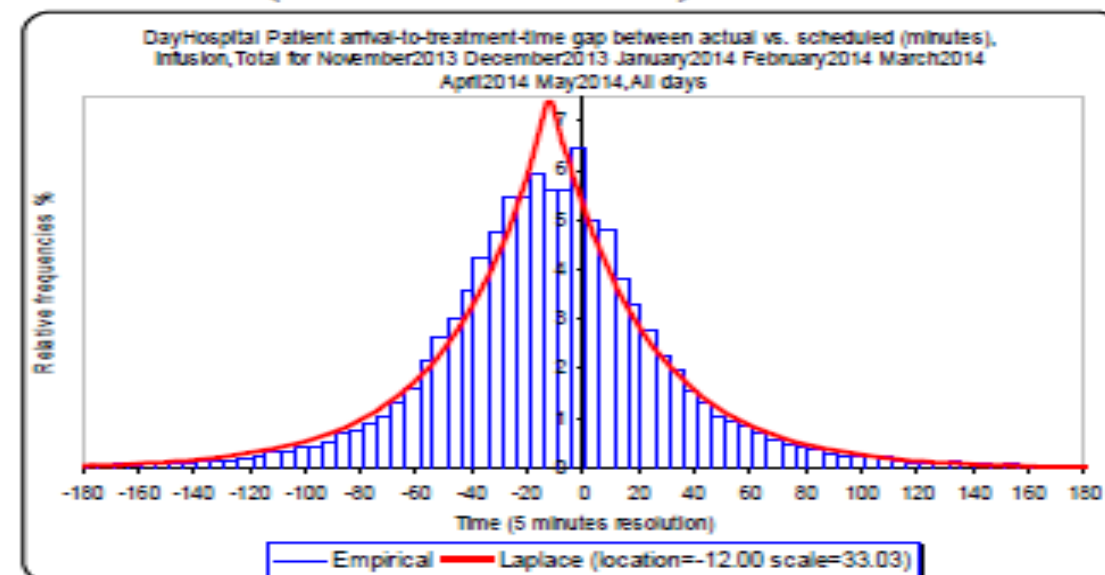
Lab (mean = -7 min)



Exam (mean = -12 min)



Infusion (mean = -12 min)



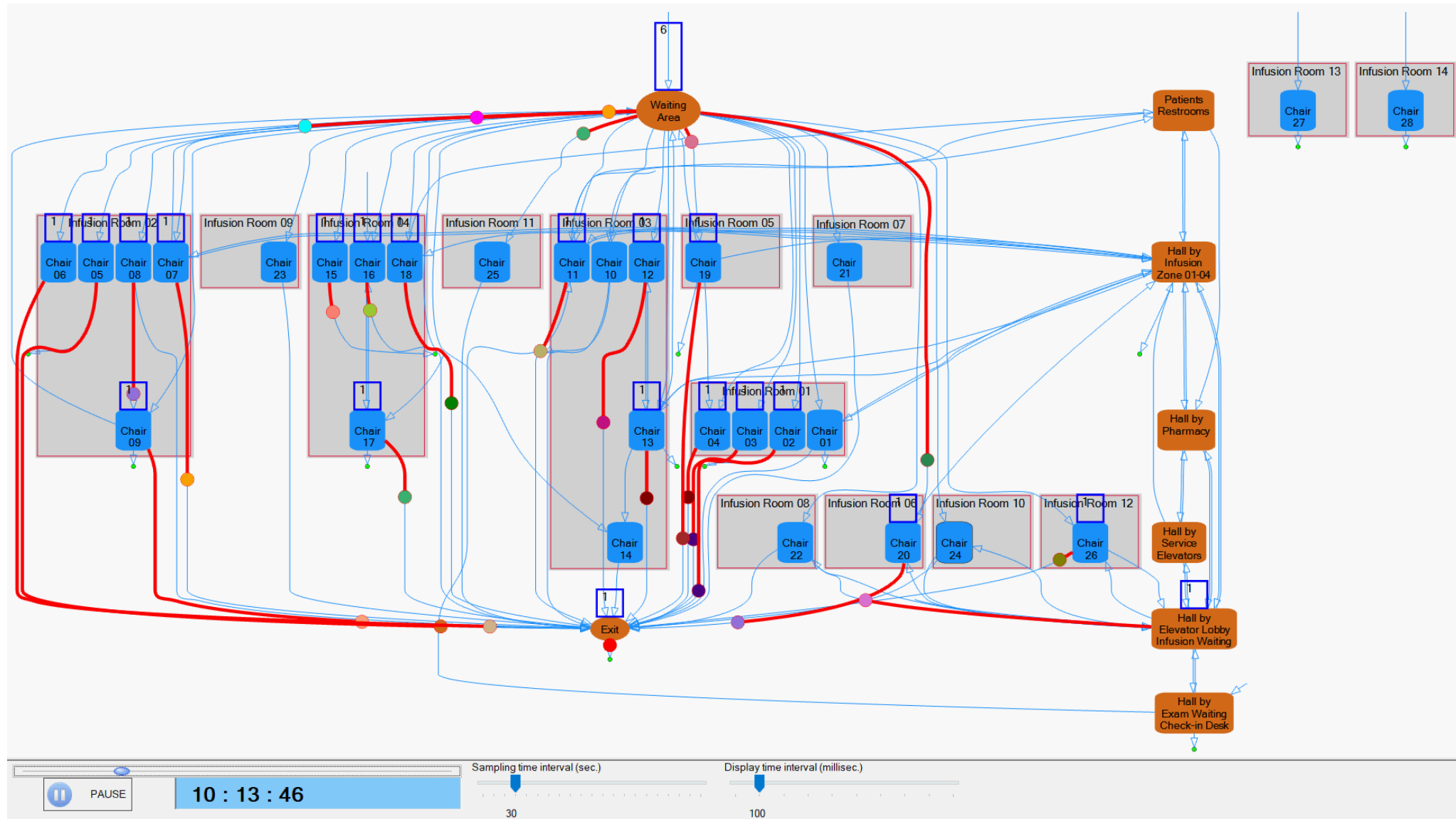
**Appointments dominate healthcare systems.
Such systems are intrinsically stochastic, yet
appointment systems (too) often view them deterministically:**

DFCI Sample Infusion Schedule on Sep 3, 2014

time	patient_id	duration (min)	link_flag	floor_id	disease_center
15:00	01	60	unlinked	9	breast onc.
12:30	02	120	unlinked	9	breast onc.
10:30	03	180	linked	9	genitourinary onc.
12:30	04	60	linked	9	breast onc.
12:00	05	120	linked	9	genitourinary onc.
07:00	06	60	unlinked	9	breast onc.
⋮	⋮	⋮	⋮	⋮	⋮

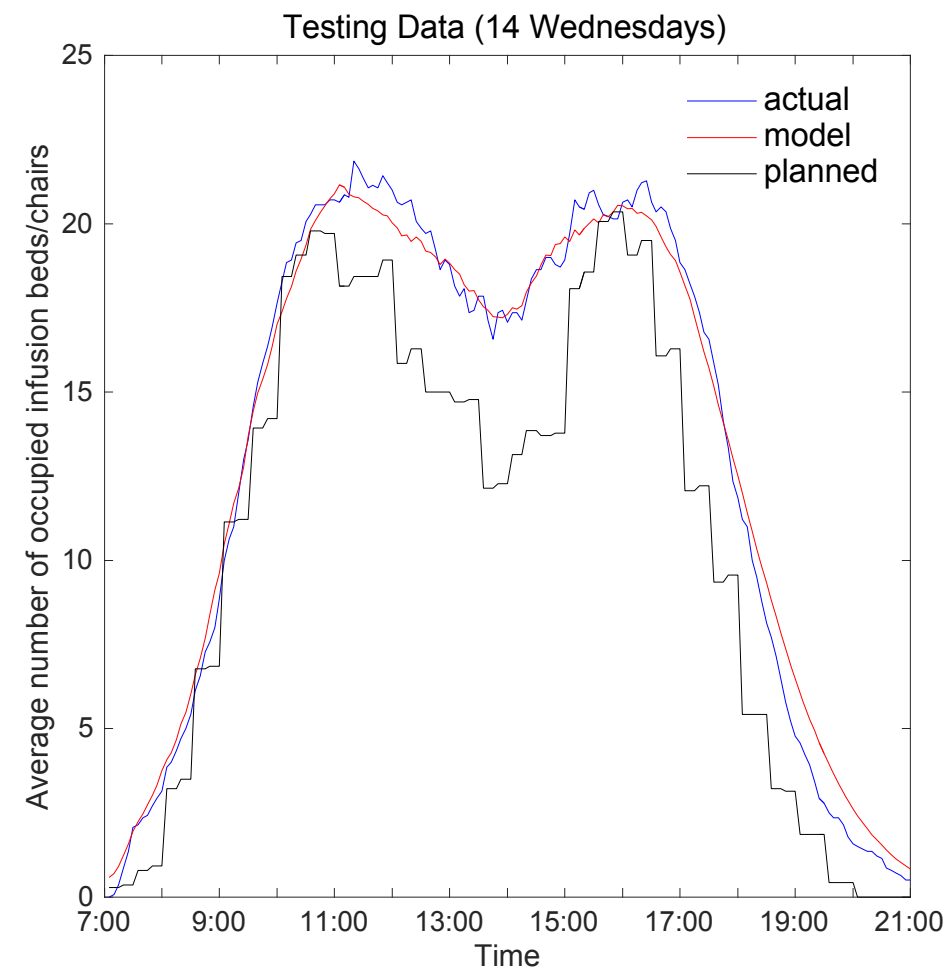
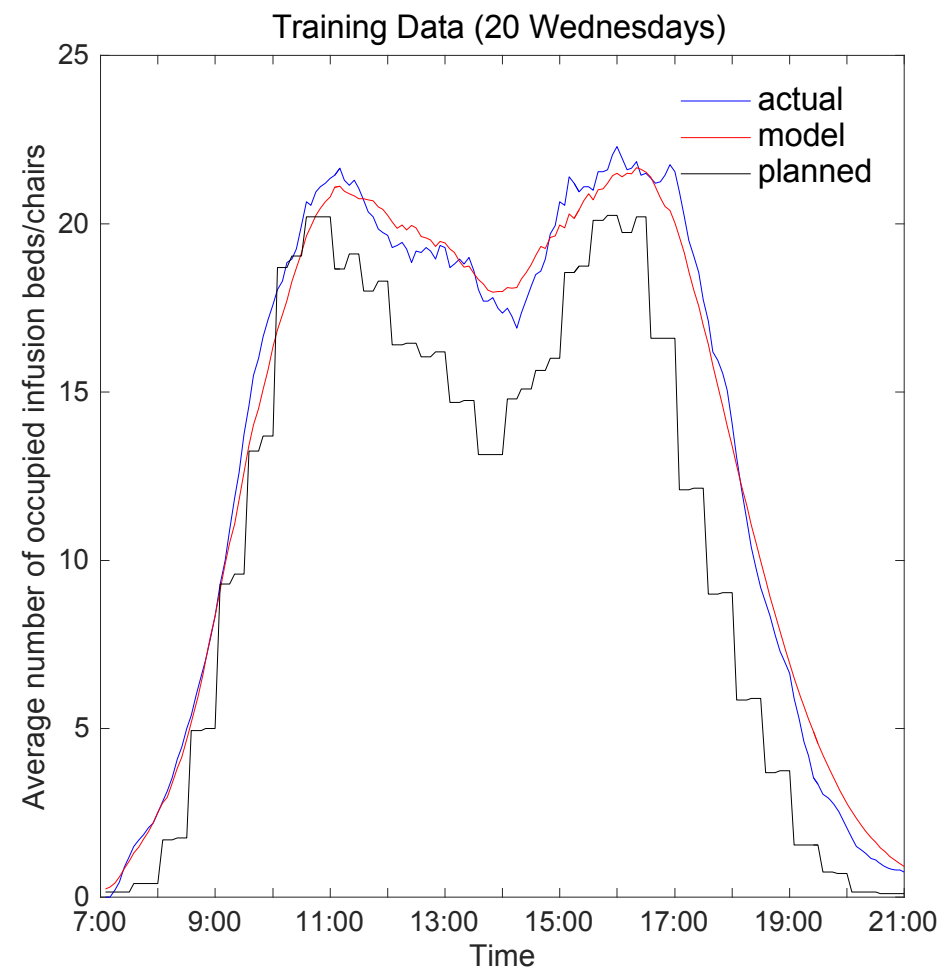
w/ P. Momcilovic & N. Trichakis: Develop methodologies for appointment scheduling, in **multi-server** environments, that take into account **stochastic** punctuality and service-durations

Patients Flow: Nth Floor, Infusion Unit w/ P. Momcilovic & N. Trichakis



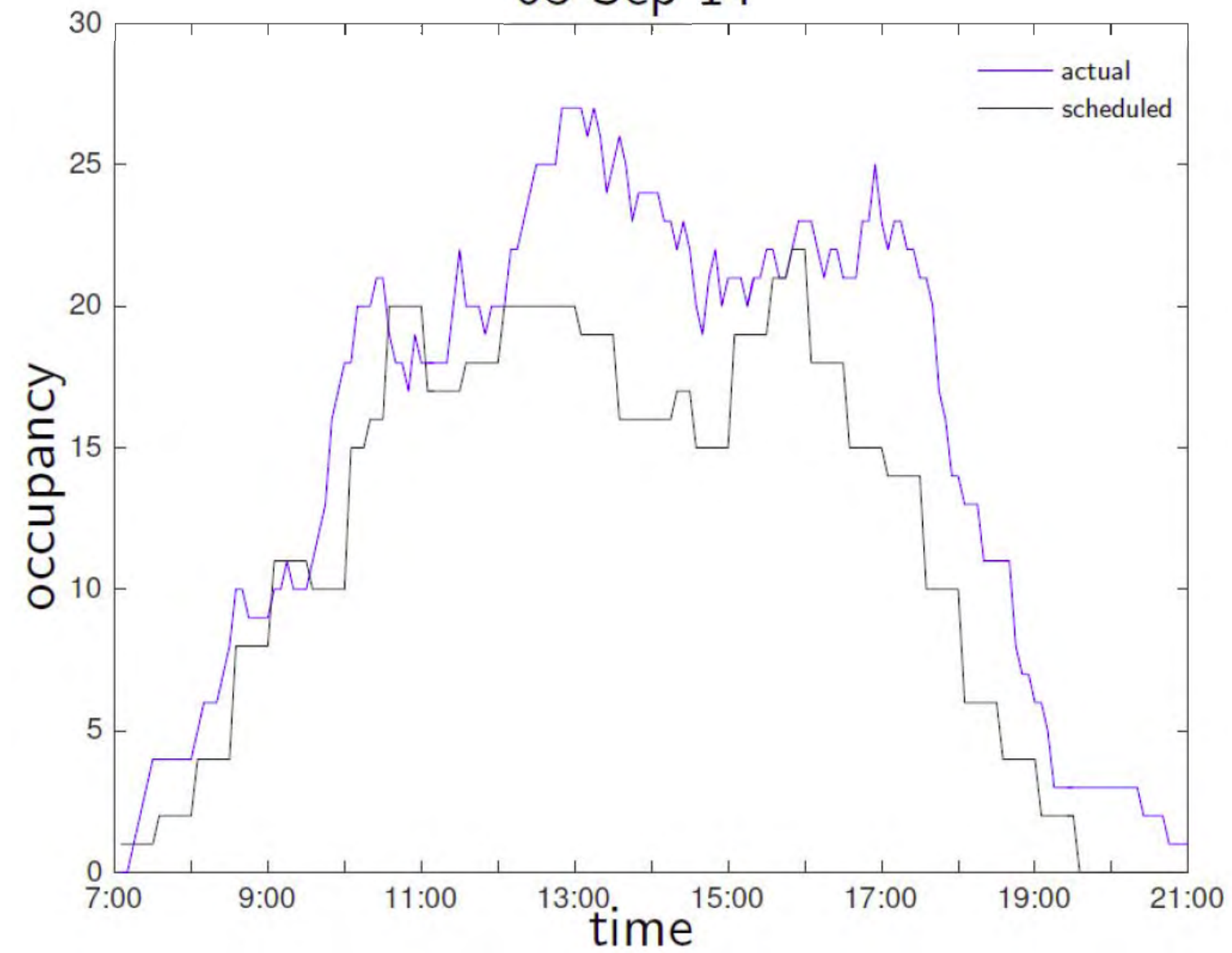
Appointment Scheduling: Approach & Validation

- Infinite-server model
- CLT approximation
- Benchmark based on Robust-Optimization framework



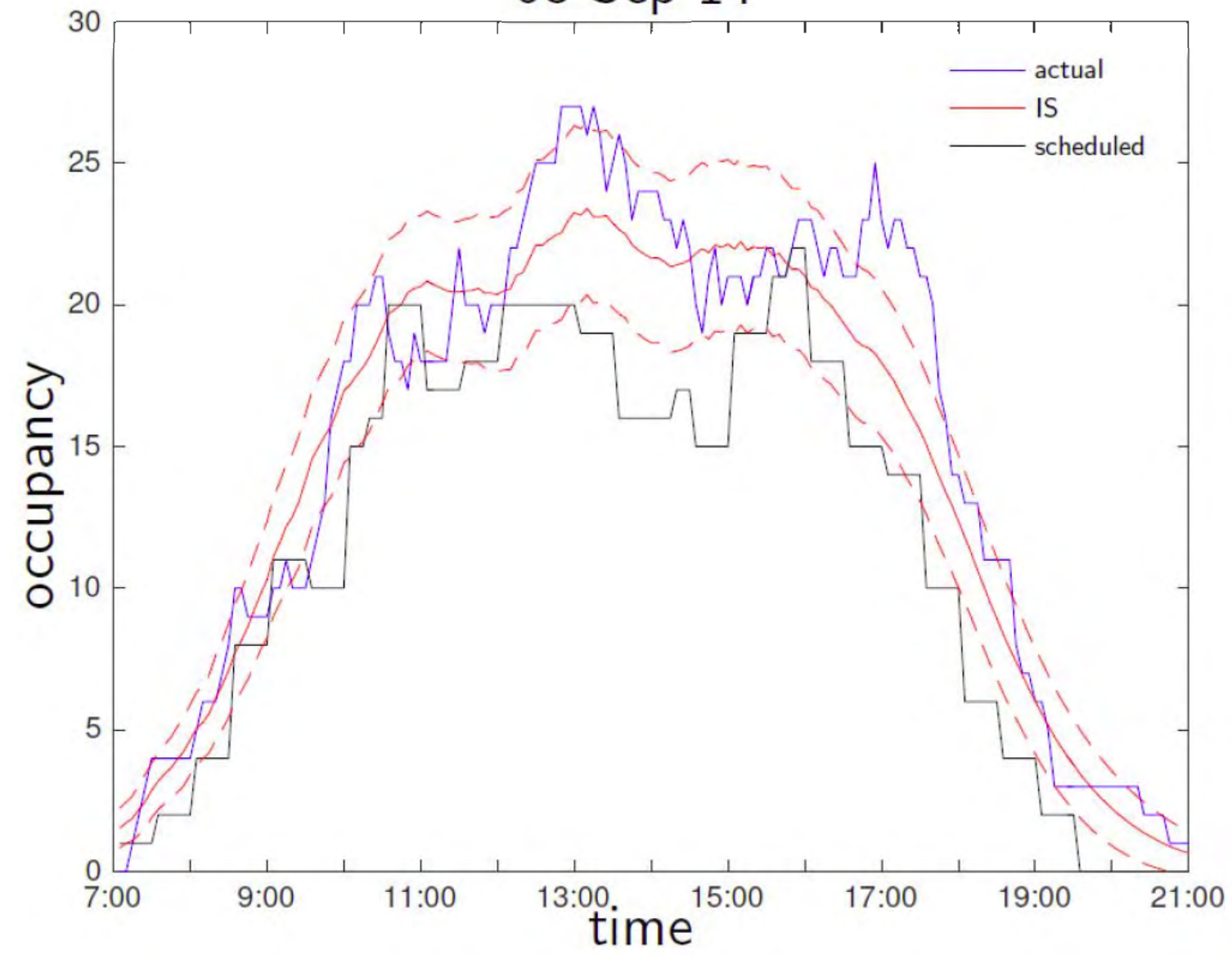
Scheduled **vs** Actual

03-Sep-14



Validation with DFCI Data

03-Sep-14



Comparison with DFCI Practice

- infusion service on floors 8, 9
- $c_t \approx 25$, $n \approx 90$
- compare IS with “means-based” scheduling
- consider 85% and 95% utilization levels

Reduce costs (waiting, overtime) by approximately 30%

Some Observations on “Academia-Industry Partnerships” (not MSR or the Original-Bell-Labs)

- Intriguingly: RTLS successful at a **leading research** hospital?
- Intriguingly: initiative of **Physicians** at DFCI (as opposed to managers)?
- **Conjectures?**

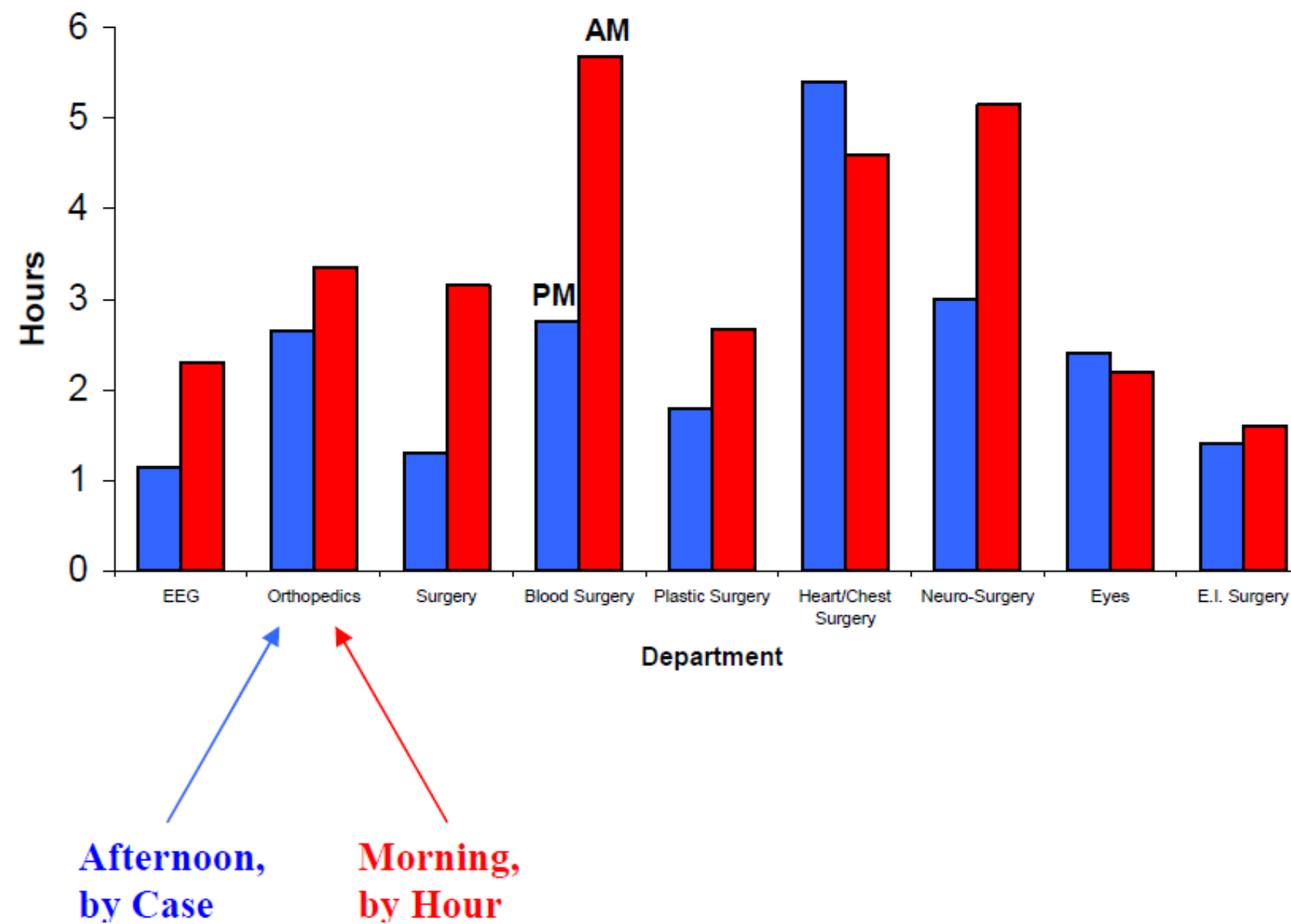
Partnership with “Strong” partners, who appreciate Research/Evidence-Based-Management, and it is naturally based on **Knowledge** partnering with **Data**

Data enables **SYMMETRIC Partnerships**, caters to goals of both partners (even IP)

Last Commercial: Key Missing Piece – Incentives
or: Even doctors (professors,...) can manage

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

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



Lessons from my Journeys

- **Data** is the language of
 - ✓ Multi-disciplinary research (e.g. OR + Psychology + NLP)
 - ✓ Academia-industry partnerships (e.g. university & bank)
- Parsimonious (robust) **models** at the service of complex realities
- **OR** needs Machine-Learning/Process-Mining/... to overcome its “curse of parsimony” (build high-stakes confidence - think “clinical-trials” via simulation);
 - ✓ The latter need OR to overcome their “curse of dimensionality” (overfitting); go beyond data-limits = predicting an existing system (think “ED counterfactuals”)
- **Academia & Industry** must acknowledge (organize for, fund) data-professionalism
- **Theompirical** (data-based OR) research is feasible (1st data acquisition most challenging), important, “dangerous” before tenure, but mostly it is **fun & worth it**

The Technion SEE Center / Laboratory

Data-Based Service Science / Engineering



Mixture Fitting

The mixture distribution function:

$$p(x) = \sum_{j=1}^k w_j \varphi(x; \theta_j); \quad \sum_{j=1}^k w_j = 1; \quad w_j > 0$$

θ_j – distribution parameters (shape, scale or location parameters)

φ – given distribution function Normal, Lognormal, Gamma, Exponential, Weibull or Invers Gaussian)

k – given number of components of the mixture

w_j – weighting coefficient or mixing proportions

Task: To find the optimal w_j and θ_j for sample X and given k and φ .

Solution:

- Specify initial values for w_j by using grid.
- For each point in a grid calculate mean, standard deviation (specify initial values for θ_j) and objective function (sum of squares of difference between empirical and fitted cumulative distribution functions).
- Select points with smallest values of objective function (best points).
- For each best point compute fitted distribution by using optimization algorithm.
- The best fit with smallest value of objective function is selected.

Distribution Fitting

Algorithms of distribution fitting (SEESat software):

1. *Maximum likelihood estimates*

(11 distributions)

2. *L-moments estimates* (linear combination of order statistics)

(31 distributions)

3. *Maximization of likelihood function*

(7 distributions)

4. *Optimization of the object function* (sum of squares of differences between empirical and fitted cumulative distribution functions)

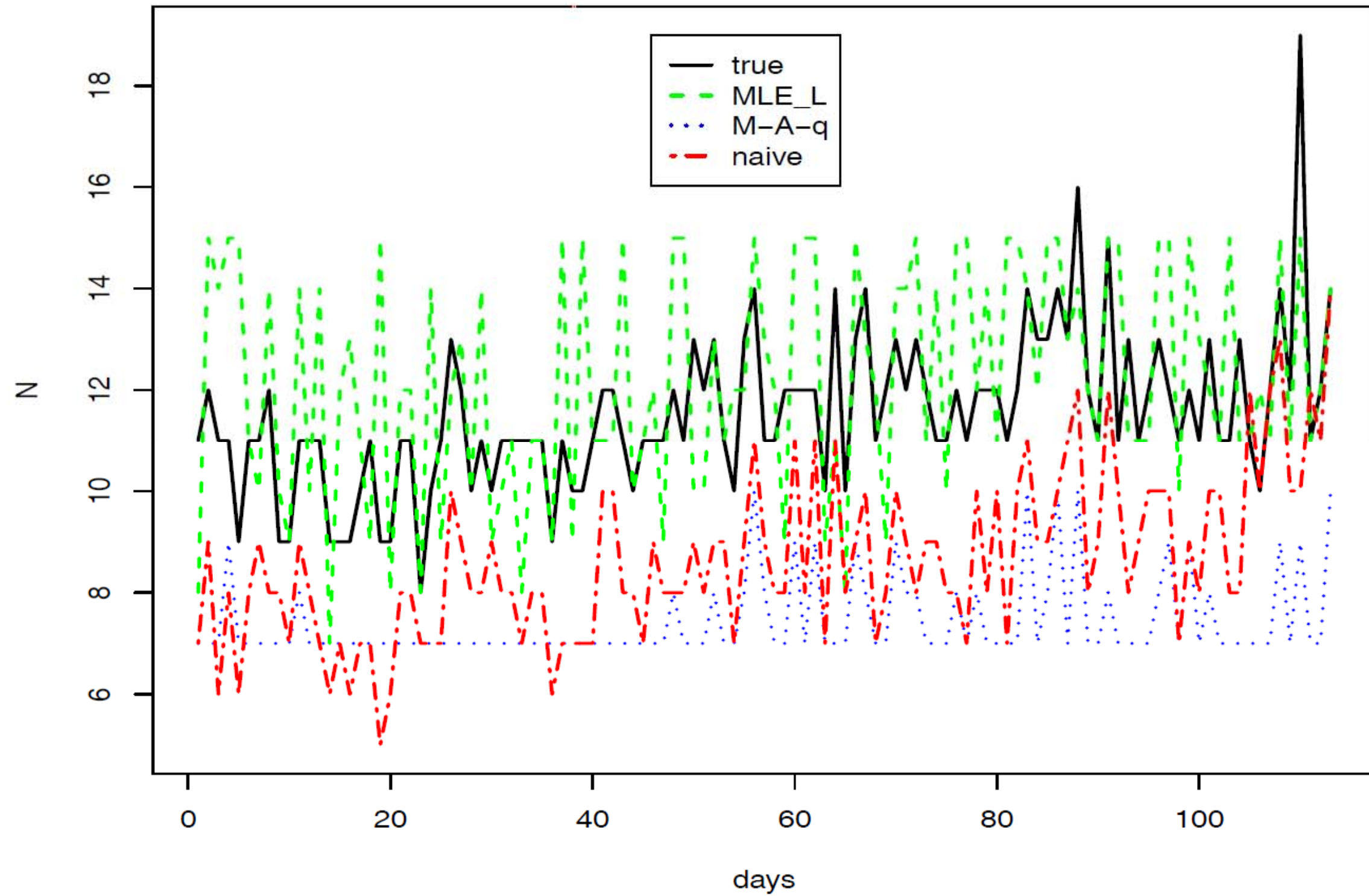
(1 distribution)

Data-Based Creation of ServNets: some Technicalities

- ▶ **ServNets** = QNets, SimNets, FNetS, 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 ($B_n(p) = E_p F[B(n, p)], 0 \leq p \leq 1$)
- ▶ **Algorithm**: Dot Layout (but with cycles), based on Sugiyama, Tagawa, Toda ('81): “Visual Understanding of **Hierarchical** System Structures”
- ▶ 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)

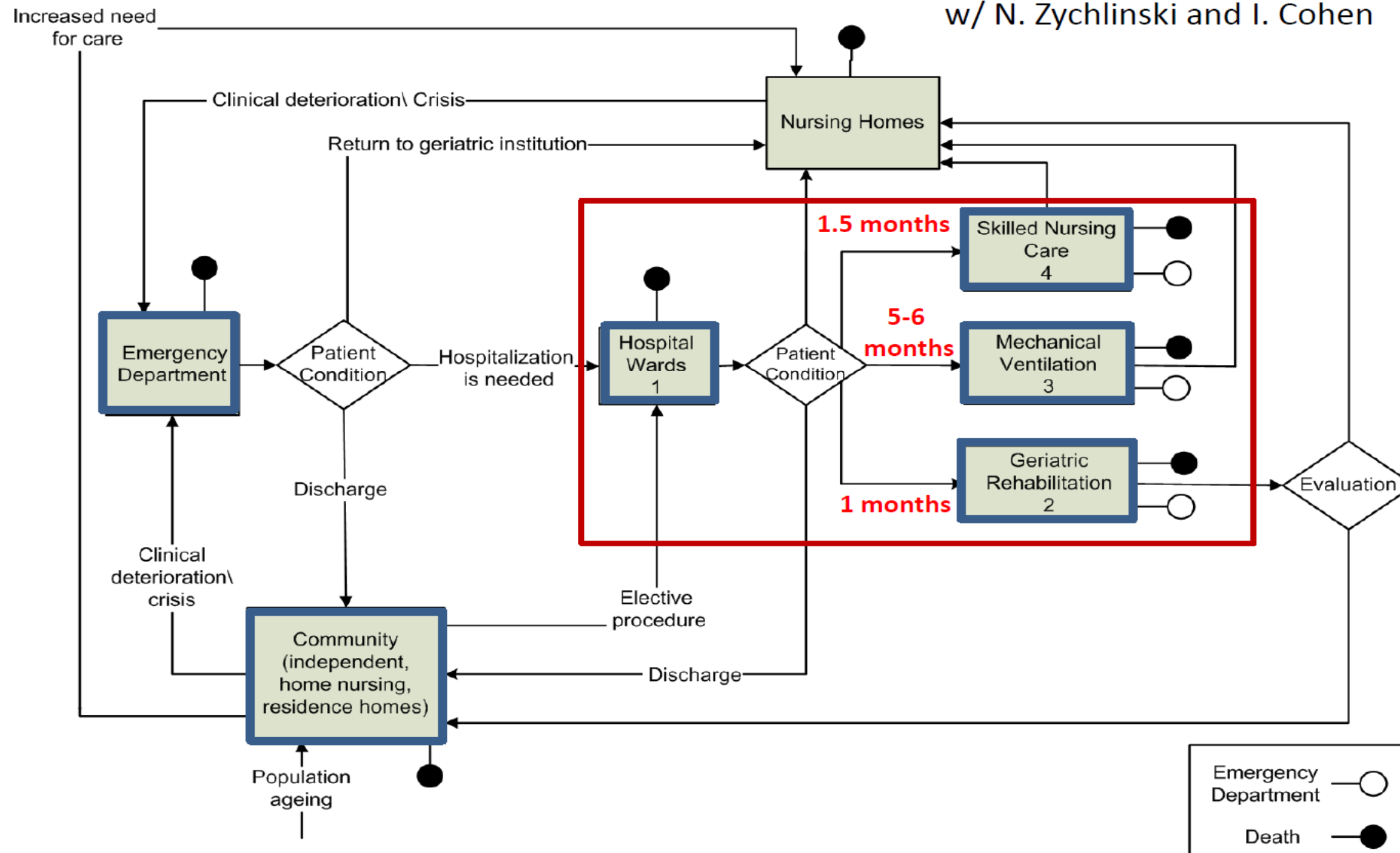
Estimating the Number of Present Servers

2/1/2005 – 27/6/2005



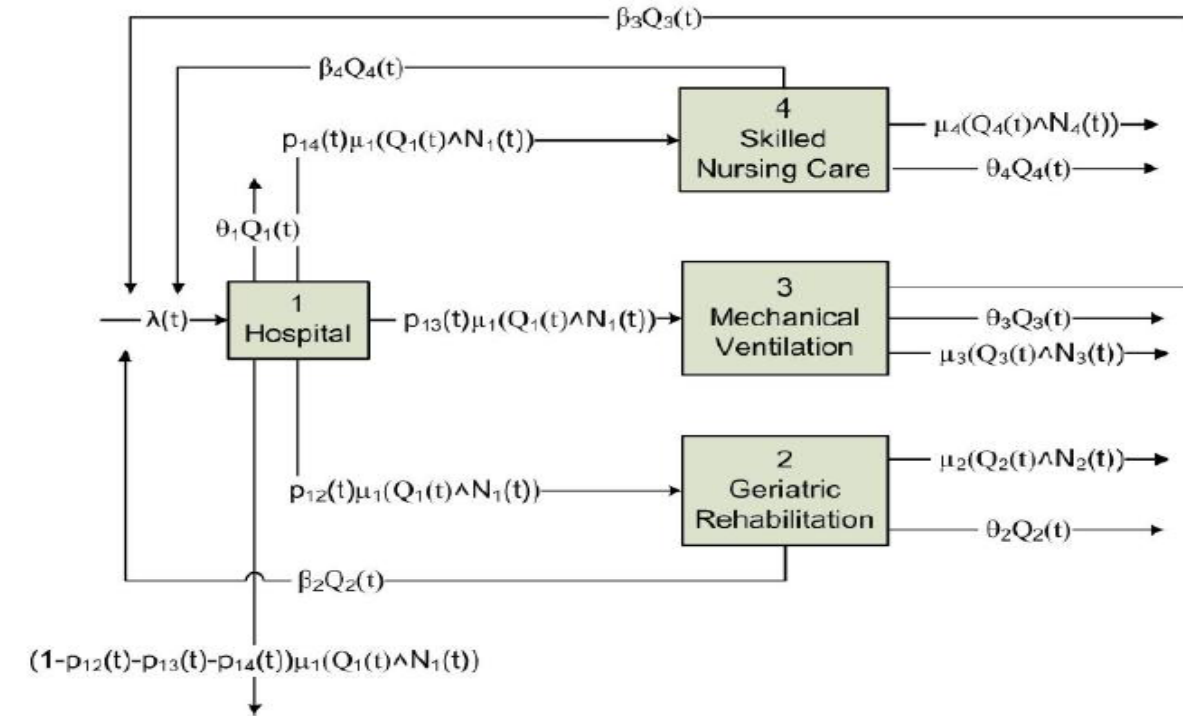
Patient Flow: ED to Wards to Nursing/Geriatric Institutions

w/ N. Zychlinski and I. Cohen



A Fluid Model

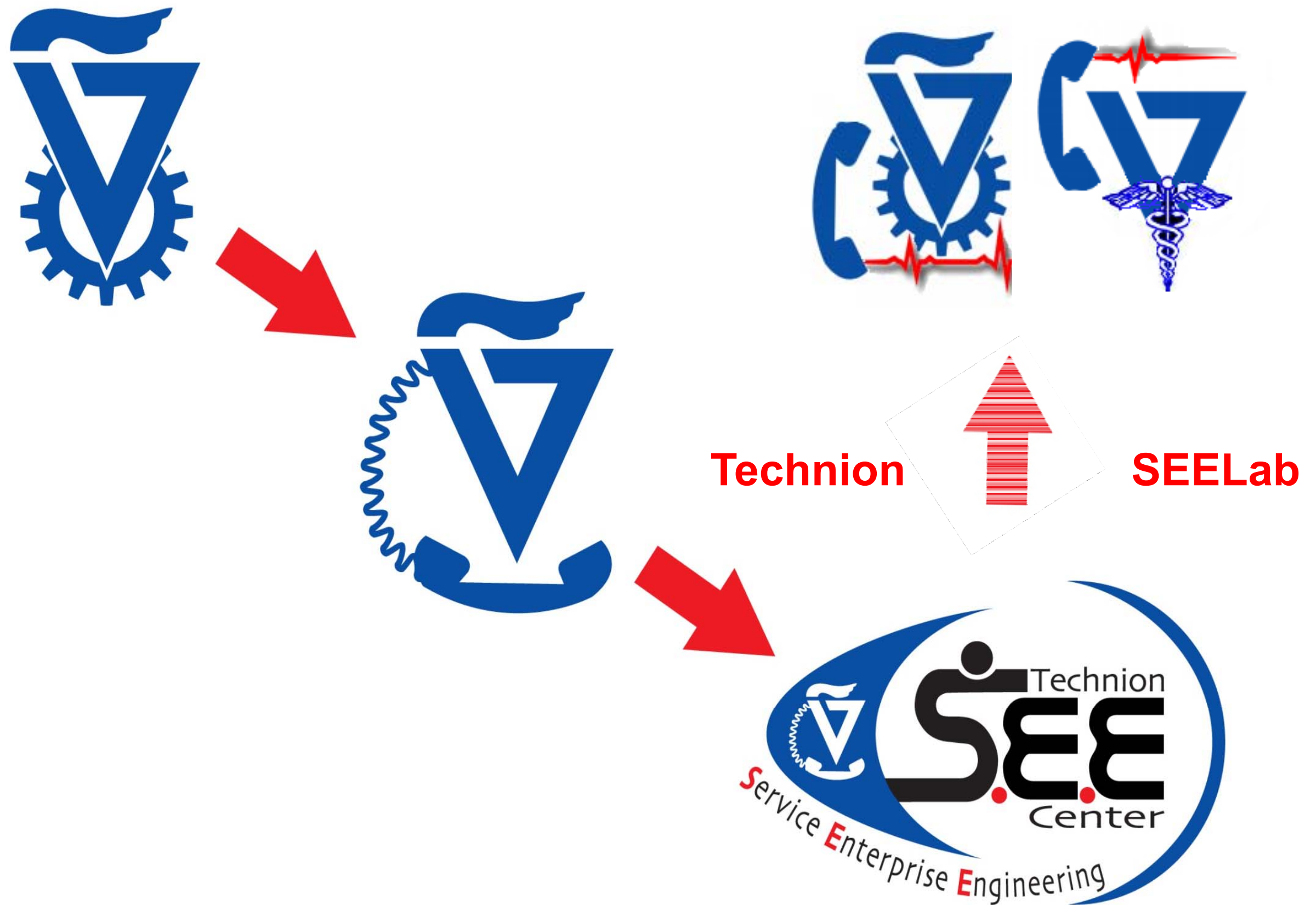
Stations 2,3 and 4



$$\dot{Q}_2(t) = (p_{12}(t)\mu_1 \cdot (Q_1(t) \wedge N_1)) \wedge (N_2 - Q_2(t))^+ - \theta_2 Q_2(t) - \beta_2 Q_2(t) - \mu_2 \cdot (Q_2(t) \wedge N_2)$$

$$\dot{Q}_3(t) = (p_{13}(t)\mu_1 \cdot (Q_1(t) \wedge N_1)) \wedge (N_3 - Q_3(t))^+ - \theta_3 Q_3(t) - \beta_3 Q_3(t) - \mu_3 \cdot (Q_3(t) \wedge N_3)$$

$$\dot{Q}_4(t) = (p_{14}(t)\mu_1 \cdot (Q_1(t) \wedge N_1)) \wedge (N_4 - Q_4(t))^+ - \theta_4 Q_4(t) - \beta_4 Q_4(t) - \mu_4 \cdot (Q_4(t) \wedge N_4).$$



Data $\Leftrightarrow$ Questions, Directions, Territories

► Open Questions

- Why are service durations **Log-Normal**?
- Staffing to **stabilize** time-varying performance: w/ Huang & Gurvich
- **Fork-join** networks (Petri Nets): w/ Atar & Zviran; Zaeid & Kaspi

► New Directions

- Multi-channel **design**:
self-service (eg. IVR, w/ Carmeli & Kaspi), tele-medicine
- **Protocol** Inference: w/ Liberman & Meilijson; w/ Senderovic, Weidlich, Gal
- **Personalized** Queueing Theory: w/ Momcilovic (LPF)
- Offered-Load/Capacity calculations:
operational, cognitive, **emotional**; **financial**
- All models “born” **equal**:
eg. inference via diffusions (w/ Pang), simulating FNetS

► Uncharted Territories

- **Server Nets**: w/ Armony & Momcilovic; Azriel & Feigin; Senderovic & Gal
- **RTLS**-based research (OR/Stat/IS, e.g. real-time control)
- (Semi)Automatic **creation** of SNetS (OR + Process Mining)
- **Integrated** healthcare:
from onset of symptoms through the hospital to recovery