

Appointment-Generated Arrival Processes

(as in outpatient scheduling in healthcare)

IEOR 4615, Service Engineering, Professor Whitt

Lecture 18, April 9, 2015

(Song-Hee Kim (PhD14) and Ponni Vel (IEOR14))
[two 2015 papers]

We are also interested in the service and how it is provided, and the associated operational efficiency (queueing analysis).

But...for now

Only look at the *arrival process*

And how it evolves from a *schedule*

Start with a simple deterministic model
of an appointment arrival process

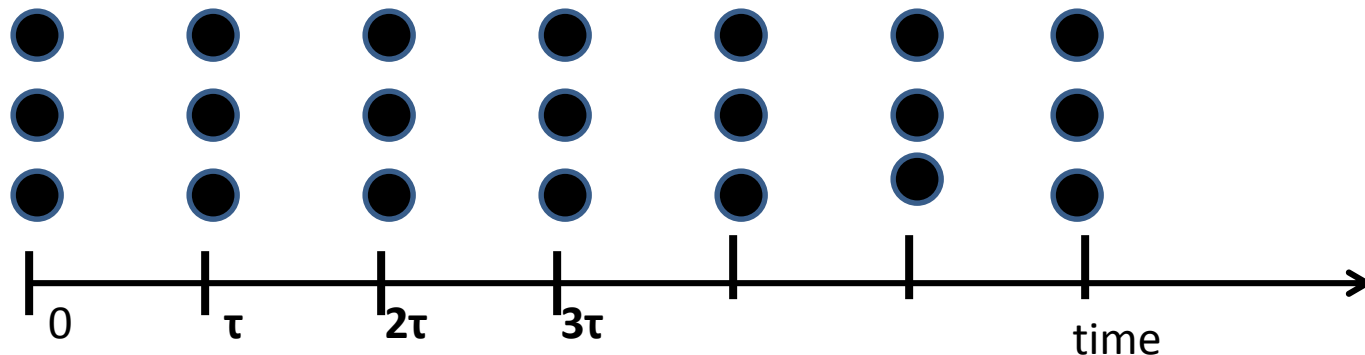
And investigate how realistic it is.

For our example, the evidence indicates that
***the data are neither consistent with the
deterministic model nor an NHPP.***

We have developed new models.

Deterministic Appointment Model

- Often Involving a **computerized appointment system**
- **Batches** of β customers arrive every τ minutes
 - Batch j arrives at time $(j-1)\tau$
 - Total of v batches
 - Daily Total Number of arrivals: $N = v\beta$
 - Total Time: $T = (v-1)\tau$



Random Schedule and Deviations From It

- Random Filling of Schedule over Time
- Random No Shows on the day
- Extra Unscheduled Arrivals
- Random Lateness or Earliness

So...

Is a deterministic model appropriate?

Or... Is an NHPP appropriate?

No, we need something else.

DATA

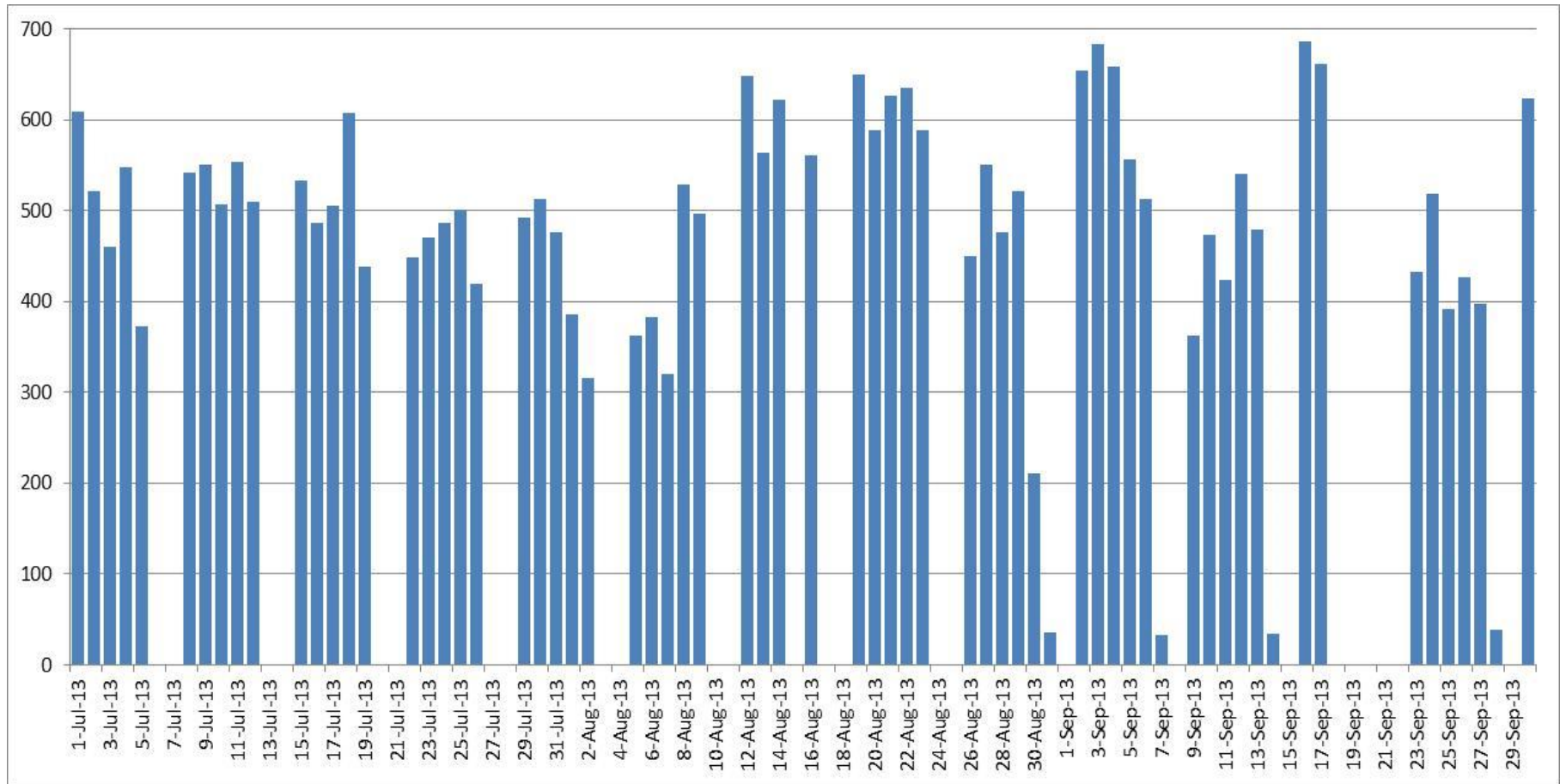
Endocrinology Clinic

Samsung Medical Center, Seoul, Korea

- Endocrinology: deals with the endocrine system (glands which excrete hormones into the blood stream)
- Three Months: 7/1/2013-9/30/2013
- Appointments to see one of 16 doctors
- Outpatient only (within one day)
- 39,253 entries
 - Scheduled day and time (by 10 minutes), when schedule made
 - Schedule evolves over time, including on the target day
 - Define “the schedule” as determined at the end of the previous day
 - **27800 show up**, 8500 cancel in advance, 3000 no shows

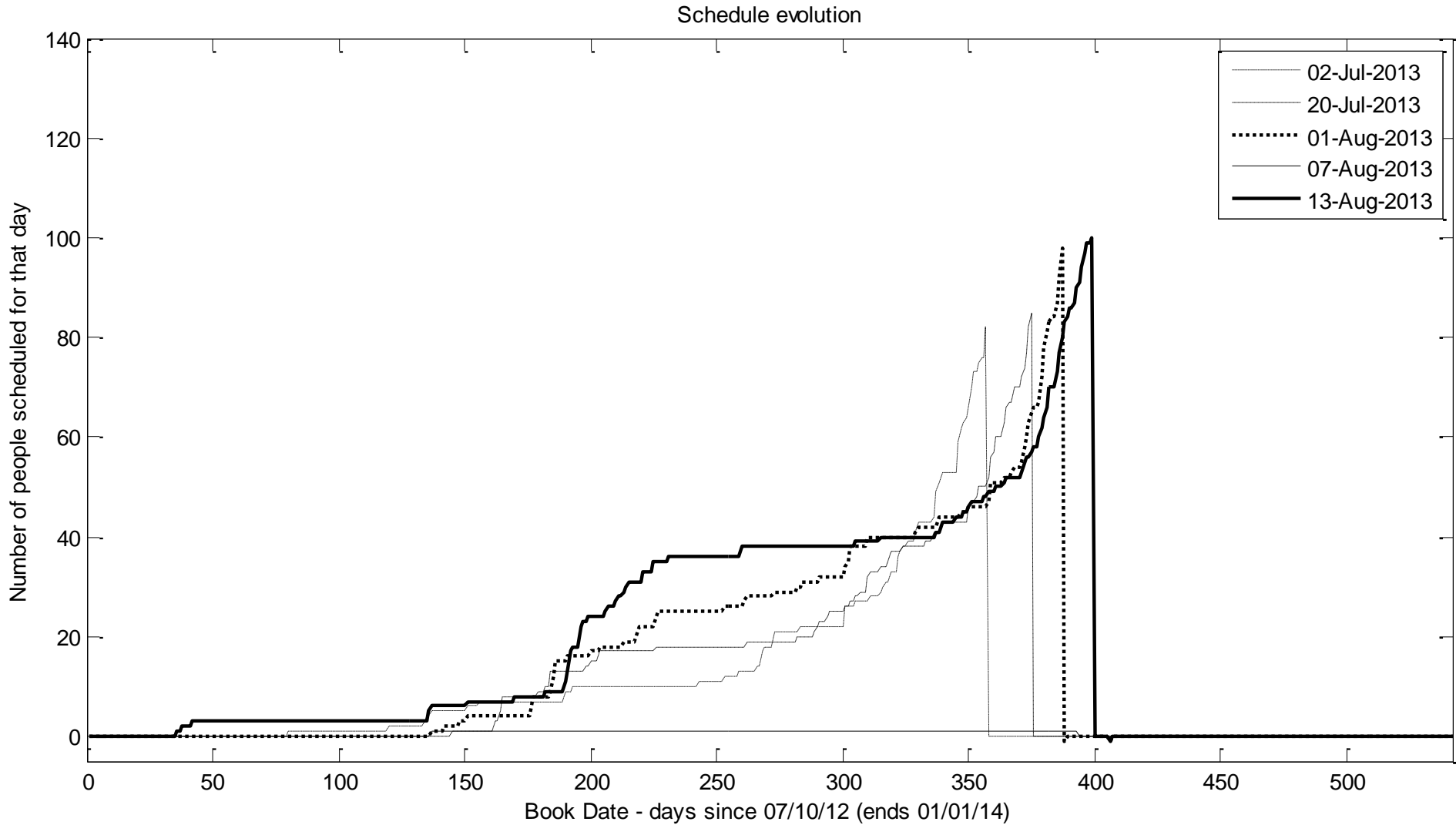
Actual Daily Arrivals: 62 days

14 weeks, 7/1 to 9/30: $448 \times 62 \approx 27,800$

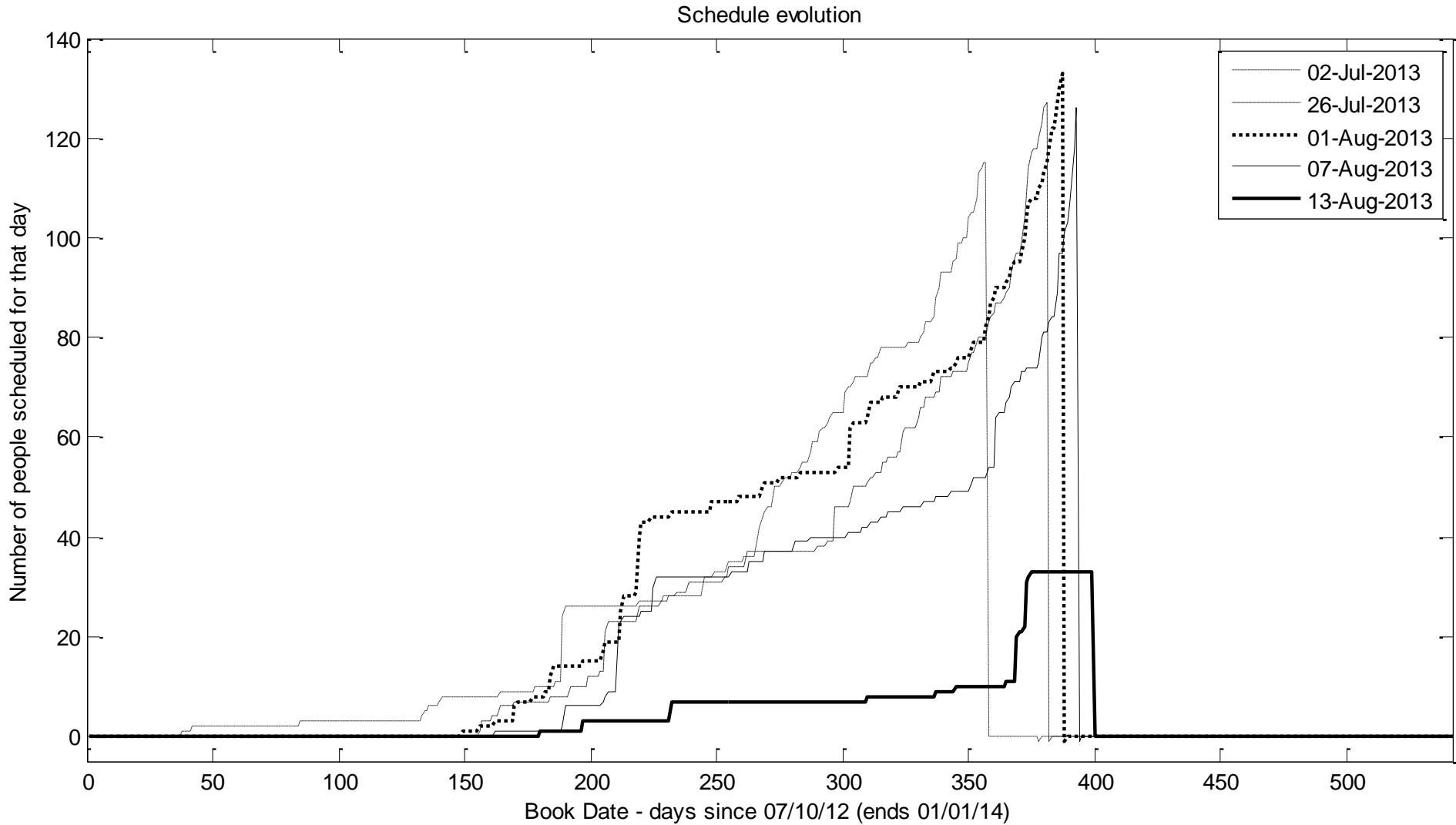


(Standard Deviation if Poisson is $(448)^{1/2} = 21$)

Evolution of the Schedule for Dr. 1



Evolution of the Schedule for Dr. 5



The schedule evolves over time.

About 25% set the last two weeks

About 55% determined 2-13 weeks in advance (many at about 3 months)

About 20% determined more than 13 in advance.

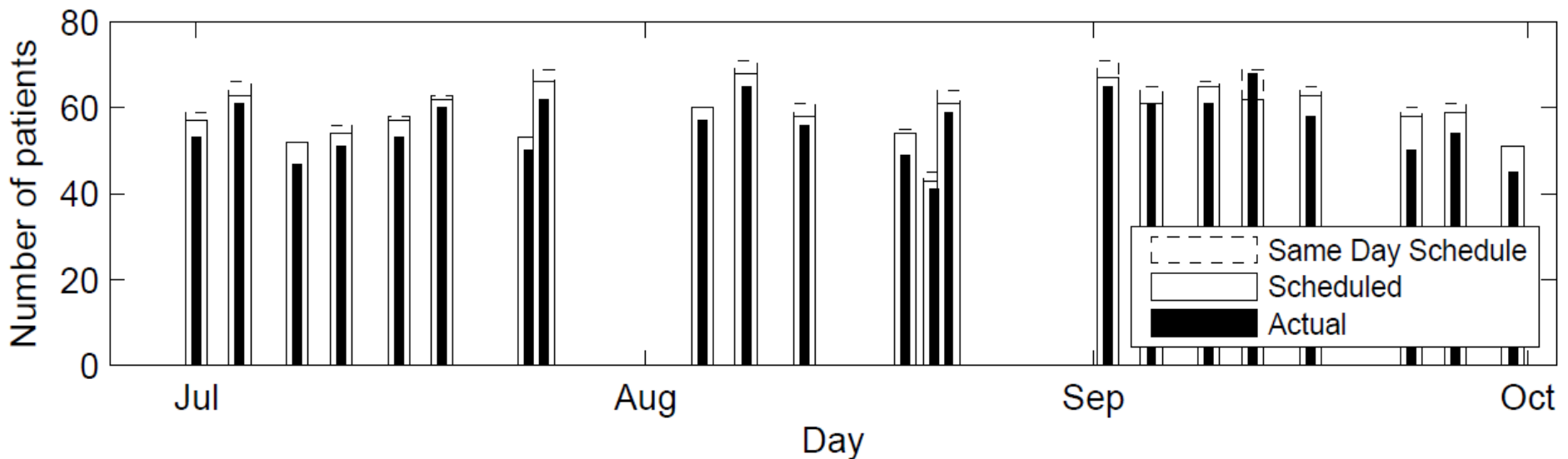
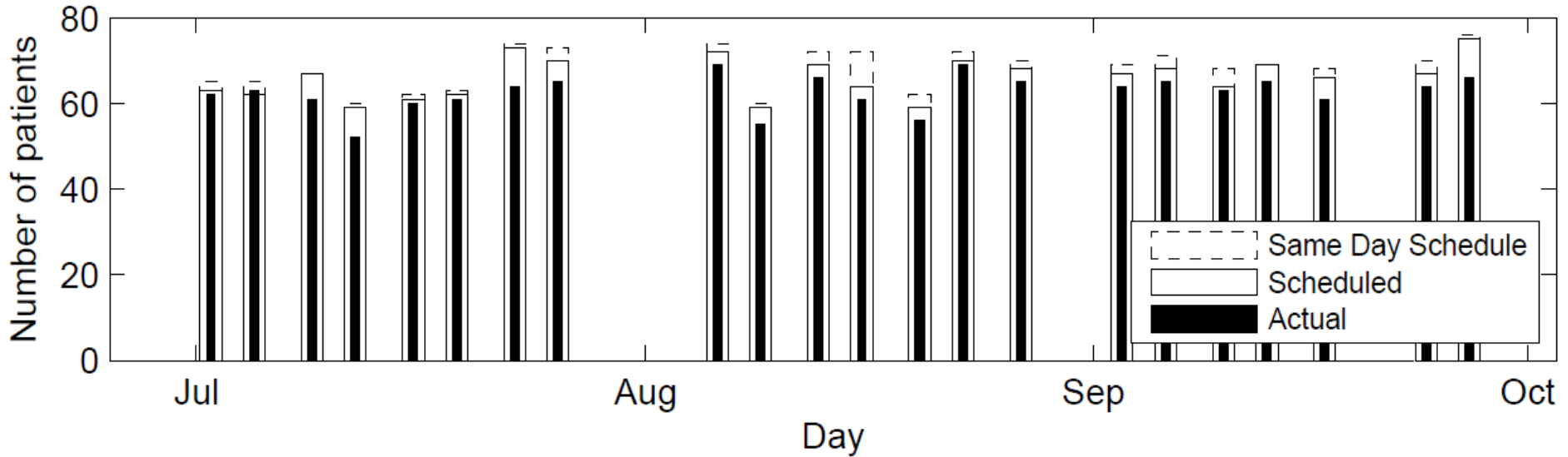
Focus on ONE DOCTOR:

Doctor 9

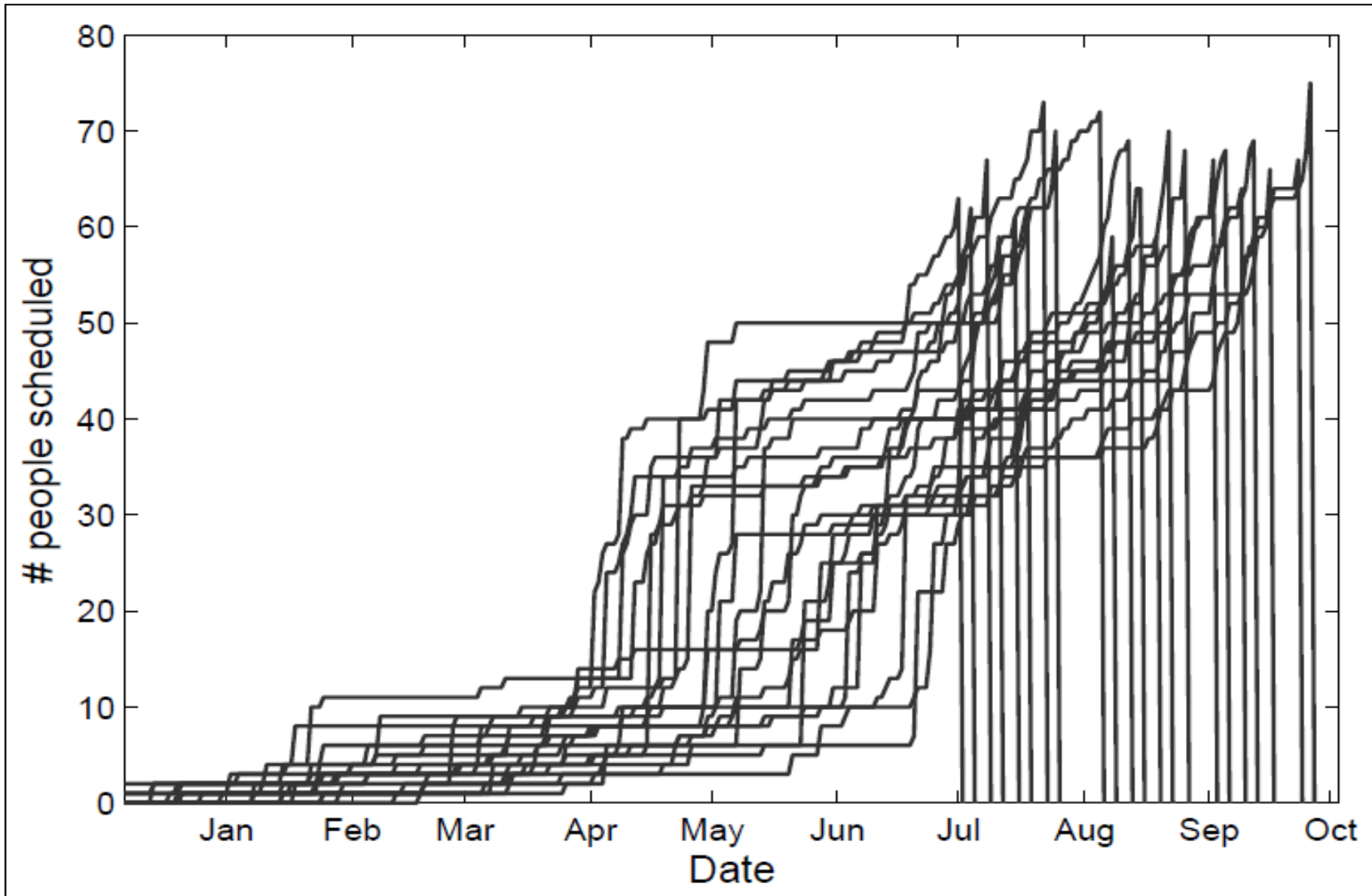
22 morning shifts

8:00am-1:00pm

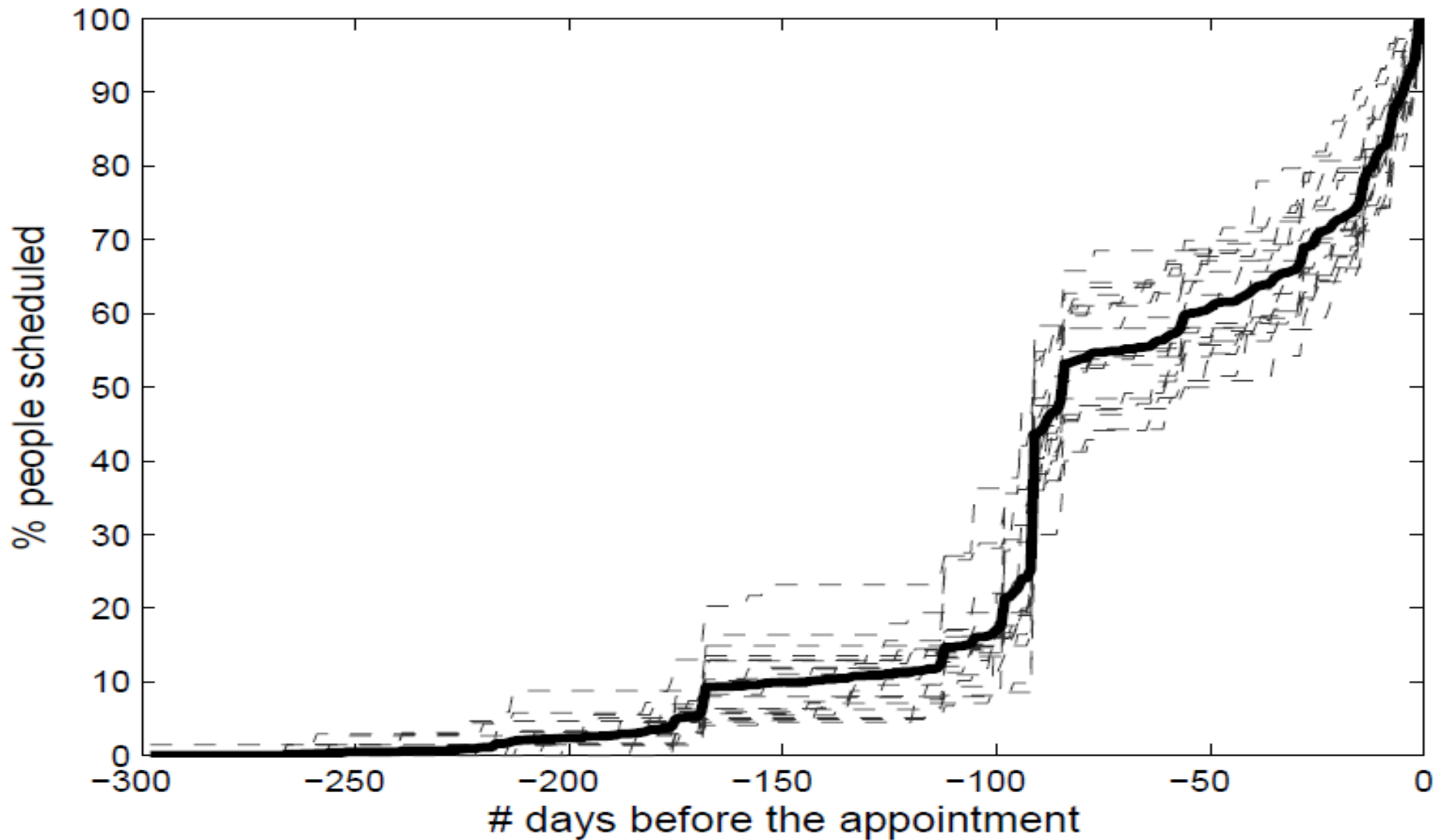
Daily totals for Dr. 9: 22 am shifts (top), 22 pm shifts (bottom)



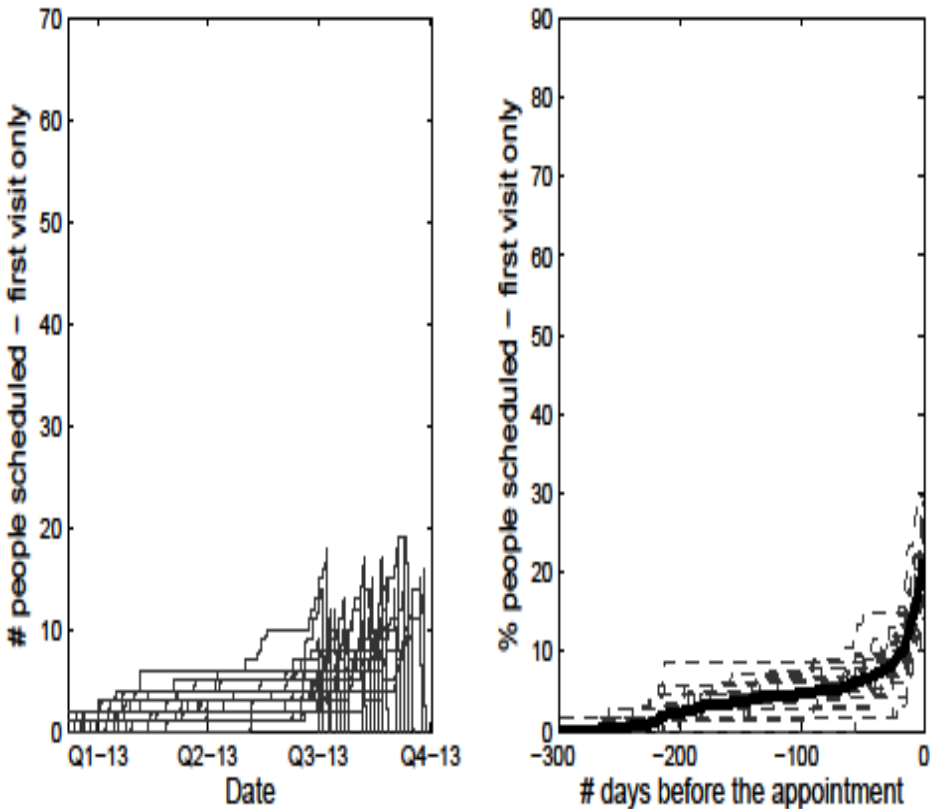
Evolution of the Schedule for Dr. 9 Over a Year: 22 am shifts



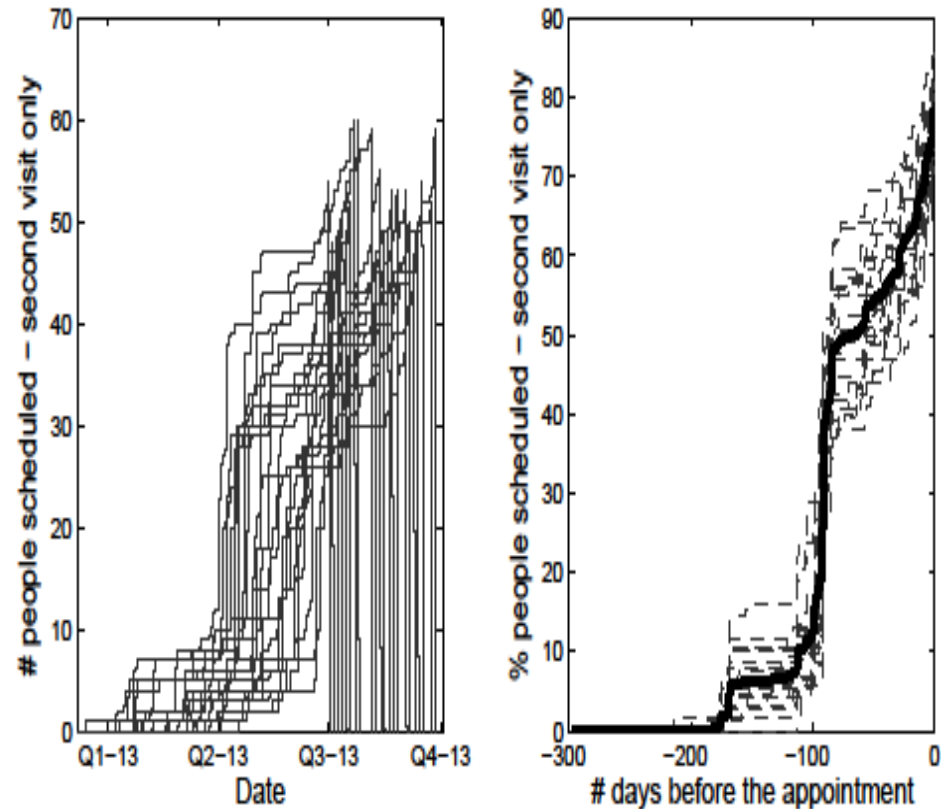
When the Schedule Is Made: Number of Days Before



Evolution of Schedule for First Visits 22%



Evolution of Schedule for Repeat Visits 78%



Let ***the schedule*** be ***defined*** by its state at the end of the previous day.

The schedule: Batches B_{sj} scheduled to arrive in slot j at time $(j-1)\tau$

No shows B_{nj}

Additional unscheduled arrivals B_{uj}
(scheduled on the same day)

Then $B_{aj} = B_{sj} - B_{nj} + B_{uj}$
(not counting deviations: earliness or lateness)

Schedule for Doctor 9 for 22 Morning Shifts

time slot	22 days in July-October 2013																					Avg	Var	Var/Avg	
7:50	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0.00		
8:00	0	0	0	0	0	0	1	1	1	0	1	0	0	0	1	0	0	0	0	0	1	1	0.32	0.23	0.71
8:10	0	0	0	0	0	0	0	1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0.05	0.05	1.00
8:20	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0.00		
8:30	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0.00		
8:40	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0.00		
8:50	3	4	5	4	4	4	4	4	4	1	3	2	1	4	2	4	4	2	4	5	4	3	3.41	1.30	0.38
9:00	3	4	2	3	3	2	3	3	3	3	3	2	2	2	2	3	4	3	2	3	4	2	2.77	0.47	0.17
9:10	3	3	3	2	2	2	4	2	2	3	2	3	2	3	3	3	2	2	3	2	3	3	2.59	0.35	0.13
9:20	2	2	4	2	3	2	3	2	2	3	3	3	2	3	2	3	3	3	3	2	3	2	2.59	0.35	0.13
9:30	3	2	3	4	3	3	4	3	3	3	3	3	1	3	2	2	2	2	3	3	3	3	2.77	0.47	0.17
9:40	3	3	3	2	2	2	2	3	3	2	2	3	2	3	2	2	2	2	3	2	2	2	2.36	0.24	0.10
9:50	3	3	3	3	2	3	3	3	3	3	3	2	2	3	3	3	3	3	2	2	3	3	2.77	0.18	0.07
10:00	3	2	3	3	2	3	2	3	2	3	3	3	3	3	3	3	4	4	3	3	3	3	2.91	0.28	0.10
10:10	3	3	3	3	3	3	3	3	2	2	3	3	3	3	3	3	3	3	3	3	3	3	2.91	0.09	0.03
10:20	2	3	3	3	3	3	3	2	3	3	2	3	3	3	3	2	3	2	3	4	3	3	2.82	0.25	0.09
10:30	3	2	3	3	3	2	4	2	3	2	3	3	3	3	3	2	3	3	2	4	3	3	2.82	0.35	0.12
10:40	3	1	3	3	3	1	3	2	3	2	3	3	2	3	2	1	3	2	3	3	3	2	2.45	0.55	0.22
10:50	2	3	3	3	1	2	3	2	3	3	2	3	3	3	3	3	3	3	2	3	3	3	2.68	0.32	0.12
11:00	3	2	3	2	3	2	3	2	2	4	4	4	2	3	3	3	3	3	4	3	4	4	2.95	0.52	0.18
11:10	3	3	3	1	3	3	3	3	2	3	3	2	3	2	1	3	2	3	3	3	3	3	2.64	0.43	0.16
11:20	2	3	3	3	3	3	3	3	3	3	3	3	3	2	2	3	3	3	3	3	3	4	2.91	0.18	0.06
11:30	3	3	2	3	3	3	3	3	3	3	3	3	3	3	3	3	3	2	2	2	3	2	2.77	0.18	0.07
11:40	3	2	3	3	2	3	3	3	3	1	2	3	3	2	3	3	3	3	3	3	2	3	2.68	0.32	0.12
11:50	3	3	3	3	3	2	2	3	3	2	3	2	4	3	3	3	2	2	3	3	1	3	2.68	0.42	0.16
12:00	2	3	3	2	3	3	4	3	3	2	3	3	3	3	3	3	3	2	2	3	4	4	2.86	0.31	0.11
12:10	3	3	3	2	3	3	2	3	2	3	2	3	3	3	4	3	1	2	3	2	3	3	2.68	0.42	0.16
12:20	2	4	3	2	3	3	3	3	4	3	3	3	2	2	3	1	3	1	4	3	3	3	2.77	0.66	0.24
12:30	2	1	0	0	0	3	3	3	3	2	2	2	2	3	3	3	2	4	3	1	2	3	2.14	1.27	0.59
12:40	0	0	0	0	0	2	2	4	3	0	3	2	1	2	3	3	4	2	3	0	0	3	1.68	2.13	1.27
12:50	0	0	0	0	0	0	0	1	4	0	0	0	0	3	4	0	2	0	4	0	0	4	1.00	2.67	2.67
13:00	1	0	0	0	1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0.09	0.09	0.95
Daily Total	63	62	67	59	61	62	73	70	72	59	69	64	59	70	68	67	68	64	69	66	67	75	66.09	21.32	0.32
[8:50, 12:20] Total	60	61	67	59	60	57	67	60	61	57	63	60	56	62	57	61	60	58	59	65	64	64	60.82	9.77	0.16
All slot avg	2.0	2.0	2.2	1.9	2.0	2.0	2.4	2.3	2.3	1.9	2.2	2.1	1.9	2.3	2.2	2.2	2.2	2.1	2.2	2.1	2.2	2.4	2.07	1.73	0.84
All slot var	1.5	1.9	2.2	1.9	1.8	1.5	1.8	1.3	1.5	1.7	1.5	1.6	1.5	1.3	1.7	1.8	1.6	1.6	2.2	1.8	1.6	(across all days)			
All slot var/avg	0.7	1.0	1.0	1.0	0.9	0.8	0.8	0.6	0.6	0.9	0.7	0.7	0.8	0.6	0.6	0.8	0.8	0.8	0.7	1.1	0.8	0.7			
[8:50, 12:20] avg	2.7	2.8	3.0	2.7	2.7	2.6	3.0	2.7	2.8	2.6	2.9	2.7	2.5	2.8	2.6	2.8	2.7	2.6	2.7	3.0	2.9	2.9	2.76	0.42	0.15
[8:50, 12:20] var	0.2	0.6	0.3	0.5	0.4	0.4	0.4	0.3	0.4	0.5	0.2	0.3	0.5	0.3	0.4	0.4	0.7	0.3	0.4	0.7	0.4	0.4	(across all days)		
[8:50, 12:20] var/avg	0.1	0.2	0.1	0.2	0.1	0.2	0.1	0.1	0.1	0.2	0.1	0.1	0.2	0.1	0.2	0.1	0.3	0.1	0.2	0.2	0.1	0.1			

A High-Demand Service System with Overloaded and At-Capacity Days

Distribution of Batch Size B_s in the Main interval [8:50,12:20]

number k	$\hat{P}(B_s = k)$				
	1	2	3	4	5
10 at-capacity days	0.04	0.25	0.63	0.07	0.01
12 overloaded days	0.02	0.27	0.63	0.08	
All days	0.03	0.26	0.63	0.08	0.004

Distribution of Number After the Main interval, N_o

$\hat{P}(N_o = k)$										
0	1	2	3	4	5	6	7	8	9	10
0.30	0.20	0.30	0.10	0.10						
					0.25	0.17		0.25		0.33
0.14	0.09	0.14	0.05	0.05	0.14	0.09		0.14		0.18

No-Shows for Doctor 9 for 22 Morning Shifts

time slot	22 days in July-October 2013																				Avg	Var	Var/Avg		
7:50	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0.00			
8:00	0	0	0	0	0	0	1	1	1	0	1	0	0	0	1	0	0	0	0	0	1	1	0.32	0.23	0.71
8:10	0	0	0	0	0	0	0	1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0.05	0.05	1.00
8:20	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0.00		
8:30	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0.00		
8:40	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0.00		
8:50	1	0	0	1	0	0	2	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	0.23	0.28	1.23
9:00	0	1	0	1	0	0	0	0	0	1	0	0	0	0	0	0	0	0	0	1	0	0	0.18	0.16	0.86
9:10	0	0	0	1	0	0	2	0	0	0	1	0	1	0	0	0	1	0	0	0	0	0	0.27	0.30	1.11
9:20	0	0	0	0	0	0	0	0	0	0	0	1	0	0	2	0	0	1	0	0	0	0	0.18	0.25	1.38
9:30	0	0	0	2	0	0	1	1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0.18	0.25	1.38
9:40	0	0	2	0	0	0	0	0	0	0	0	1	0	0	0	0	0	0	0	0	1	0.18	0.25	1.38	
9:50	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0.00		
10:00	0	0	0	0	0	0	0	0	0	1	0	2	1	0	0	0	1	0	1	0	0	0	0.27	0.30	1.11
10:10	0	0	0	1	0	0	0	1	0	0	1	1	0	0	0	0	1	0	0	0	2	0	0.32	0.32	1.01
10:20	0	0	1	1	0	0	0	0	1	0	0	1	0	0	0	0	0	0	1	0	0	0	0.23	0.18	0.81
10:30	0	0	0	0	0	0	0	1	1	0	0	0	0	0	1	0	0	0	0	0	1	0	0.18	0.16	0.86
10:40	0	1	0	0	0	0	0	0	0	0	1	0	0	0	0	0	0	1	0	1	0	1	0.23	0.18	0.81
10:50	0	0	1	0	0	1	1	0	0	0	0	0	0	1	0	0	0	0	0	0	0	2	0.27	0.30	1.11
11:00	0	0	0	1	1	0	0	0	0	1	1	0	1	0	0	1	0	0	0	1	0	0	0.32	0.23	0.71
11:10	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	0	0	1	1	0	1	0.18	0.16	0.86
11:20	0	0	0	0	0	1	0	0	0	0	0	1	0	0	1	0	1	0	0	1	0	0	0.23	0.18	0.81
11:30	0	0	0	0	0	0	0	0	0	0	0	1	1	0	0	1	1	1	0	0	0	0	0.23	0.18	0.81
11:40	0	0	1	0	0	0	1	1	1	0	0	0	0	0	0	0	0	0	0	1	0	0	0.23	0.18	0.81
11:50	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	0	1	1	0	0	0	0.14	0.12	0.90
12:00	0	0	1	0	0	0	1	0	1	0	1	2	0	0	1	2	0	0	0	1	1	1	0.55	0.45	0.83
12:10	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	0	0	0.05	0.05	1.00
12:20	1	0	0	0	0	0	0	0	0	0	0	1	0	0	0	0	0	0	0	0	0	0	0.09	0.09	0.95
12:30	0	0	0	0	0	0	1	1	0	0	1	0	0	0	0	0	0	0	0	0	0	0	0.14	0.12	0.90
12:40	0	0	0	0	0	0	0	0	0	0	0	0	0	1	0	0	0	0	0	0	0	1	0.09	0.09	0.95
12:50	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	0.05	0.05	1.00
13:00	1	0	0	0	1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0.09	0.09	0.95
Daily Total	3	2	6	8	2	2	10	7	5	4	6	10	6	2	5	5	6	5	4	7	5	10	5.45	6.35	1.17
[8:50, 12:20] Total	2	2	6	8	1	2	8	4	4	4	4	10	6	1	4	5	6	5	4	7	4	7	4.73	5.64	1.19
All slot avg	2.0	2.0	2.2	1.9	2.0	2.0	2.4	2.3	2.3	1.9	2.2	2.1	1.9	2.3	2.2	2.2	2.2	2.1	2.2	2.1	2.2	2.4	0.17	0.17	1.00
All slot var	1.5	1.9	2.2	1.9	1.8	1.5	1.8	1.3	1.5	1.7	1.5	1.5	1.6	1.5	1.3	1.7	1.8	1.6	1.6	2.2	1.8	1.6	(across all days)		
All slot var/avg	0.7	1.0	1.0	1.0	0.9	0.8	0.8	0.6	0.6	0.9	0.7	0.7	0.8	0.6	0.6	0.8	0.8	0.8	0.7	1.1	0.8	0.7	(across all days)		
[8:50, 12:20] avg	2.7	2.8	3.0	2.7	2.7	2.6	3.0	2.7	2.8	2.6	2.9	2.7	2.5	2.8	2.6	2.8	2.7	2.6	2.7	3.0	2.9	2.9	0.21	0.21	0.98
[8:50, 12:20] var	0.2	0.6	0.3	0.5	0.4	0.4	0.4	0.3	0.4	0.5	0.2	0.3	0.5	0.3	0.4	0.4	0.7	0.3	0.4	0.7	0.4	0.4	(across all days)		
[8:50, 12:20] var/avg	0.1	0.2	0.1	0.2	0.1	0.2	0.1	0.1	0.1	0.2	0.1	0.1	0.2	0.1	0.2	0.1	0.3	0.1	0.2	0.2	0.1	0.1	(across all days)		

Unscheduled Arrivals for Doctor 9 on the 22 Morning Shifts

time slot	22 days in July-October 2013																				Avg	Var	Var/Avg		
7:50	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0.00			
8:00	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	0	0.05	0.05	1.00
8:10	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0.00		
8:20	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0.00		
8:30	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0.00		
8:40	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0.00		
8:50	0	0	0	0	0	0	0	0	0	0	1	0	0	1	0	0	1	0	0	0	0	0	0.14	0.12	0.90
9:00	0	0	0	0	0	0	0	0	0	0	1	0	0	0	0	0	0	0	0	0	1	0.09	0.09	0.95	
9:10	0	0	0	0	0	0	0	0	1	0	1	0	0	0	0	1	0	0	0	0	0	0.14	0.12	0.90	
9:20	1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0.05	0.05	1.00	
9:30	0	1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0.05	0.05	1.00	
9:40	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	1	0	0	0	0	0	0.09	0.09	0.95	
9:50	0	0	0	0	0	0	0	0	0	0	1	1	0	0	0	0	0	0	0	0	0	0.09	0.09	0.95	
10:00	0	0	0	0	0	0	0	0	1	0	0	0	0	0	0	0	0	0	0	0	0	0.05	0.05	1.00	
10:10	0	0	0	0	0	0	0	0	0	1	0	0	0	0	0	0	0	0	0	0	0	0.05	0.05	1.00	
10:20	0	0	0	0	0	0	0	0	0	0	1	0	0	0	0	0	1	0	0	1	0	0.14	0.12	0.90	
10:30	0	0	0	0	0	0	0	0	0	0	1	0	0	0	0	0	1	0	0	0	0	0.09	0.09	0.95	
10:40	0	1	0	0	0	1	0	0	0	0	0	0	0	0	1	0	0	0	0	0	0	0.14	0.12	0.90	
10:50	0	0	0	0	1	0	0	1	0	0	0	0	0	0	0	0	0	0	0	0	0	0.09	0.09	0.95	
11:00	0	1	0	0	0	0	0	0	0	0	0	0	1	0	0	0	0	0	0	0	0	0.09	0.09	0.95	
11:10	0	0	0	1	0	0	0	0	0	0	0	0	0	1	0	0	0	0	0	0	0	0.09	0.09	0.95	
11:20	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	0	0	0	0	0	0	0.05	0.05	1.00	
11:30	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0.00			
11:40	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0.00			
11:50	0	0	0	0	0	0	1	0	0	0	0	0	0	0	0	0	0	0	0	1	0	0.09	0.09	0.95	
12:00	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0.00			
12:10	0	0	0	0	0	0	0	0	0	0	1	0	0	0	0	1	0	0	1	0	0	0.14	0.12	0.90	
12:20	1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0.05	0.05	1.00	
12:30	0	0	0	0	0	0	1	0	0	0	1	0	0	0	0	0	0	0	1	0	0	0.14	0.12	0.90	
12:40	0	0	0	0	0	0	0	0	0	1	2	0	0	0	0	0	0	0	0	0	0	0.14	0.22	1.60	
12:50	0	0	0	0	0	0	0	0	0	0	0	1	0	0	0	1	0	0	0	0	0	0.09	0.09	0.95	
13:00	0	0	0	0	0	0	1	0	0	0	0	1	0	0	0	0	0	0	0	0	0	0.09	0.09	0.95	
Daily Total	2	3	0	1	1	1	1	3	2	1	3	8	3	2	2	2	3	4	0	2	3	1	2.18	2.82	1.29
[8:50, 12:20] Total	2	3	0	1	1	1	1	1	2	1	2	5	2	1	2	2	3	3	0	1	2	1	1.68	1.27	0.76

Daily Totals

N_s = number scheduled, N_a = number arrived

N_n = number no-shows, N_u = number unscheduled

Full am shift: $N_a = N_s - N_n + N_u$

$E[N_s] = 66.1$, $E[N_n] = 5.5$, $E[N_u] = 2.0$, $E[N_a] = 62.6$

$\text{Var}(N_s) = 21.9 > \text{Var}(N_a) = 17.4$

Dispersion: $D = \text{Var}/\text{Mean}$

$D_s = 21.9/66.1 = 0.33$, $D_a = 17.4/62.6 = 0.278$

Main time interval [8:50, 12:20]:

$E[N_s] = 60.8$ while $\text{Var}(N_s) = 9.8$

View of Daily Totals

With appointments: Not constant, but predictable variability,
so **Under-Dispersion**
(less variable than Poisson: $D=V/M = 0.3 < 1$)

With Call Centers: often strong day-to-day variation,
so **Over-Dispersion**
(more variable than Poisson: $D=V/M > 1$)

Random Earliness or Lateness

Let X be the amount of time late.

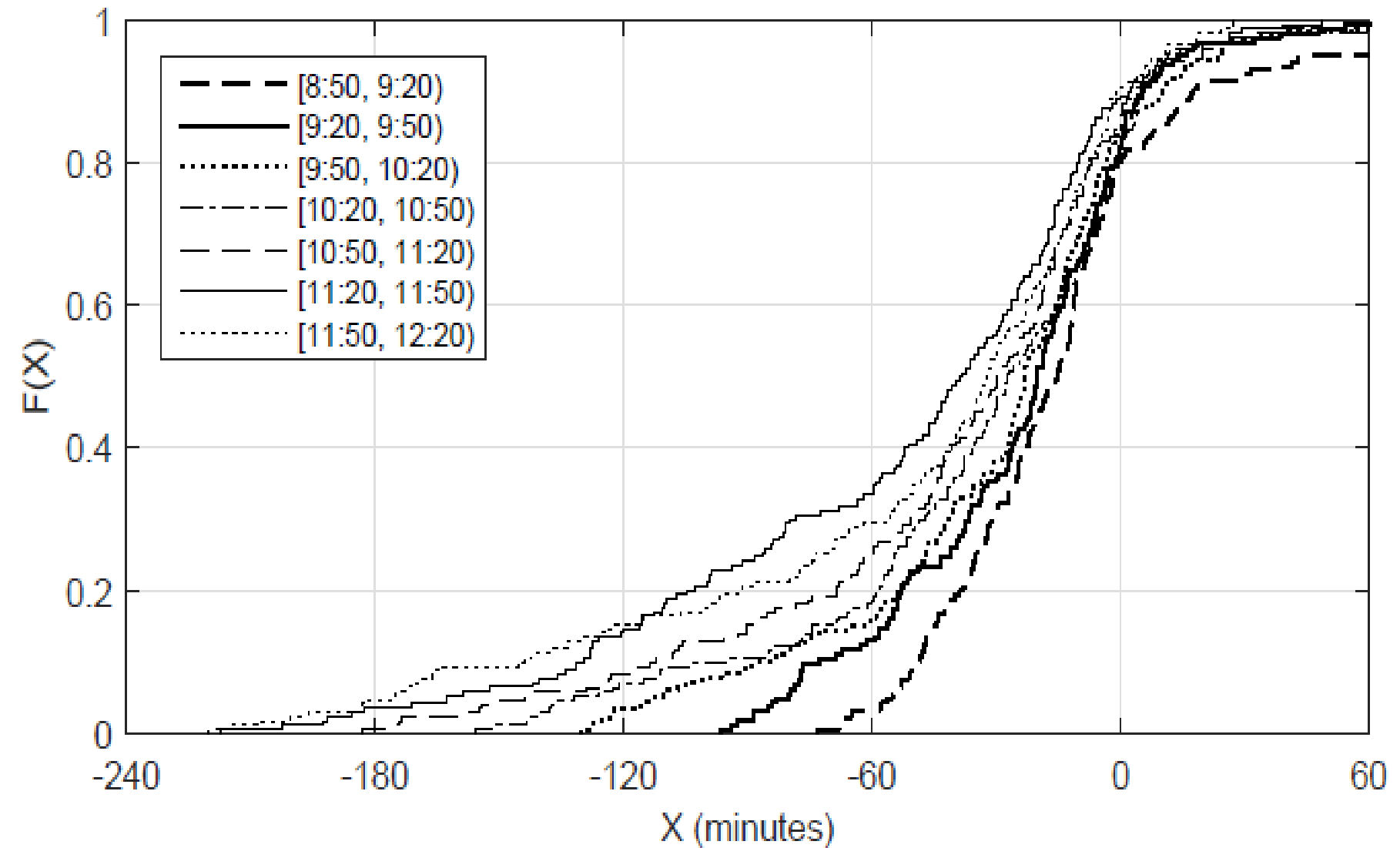
Let $F(t) = P(X \leq t)$ be the cdf.

($X < 0$ means arrive early)

**Average numbers for Dr. 9 in each 30-minute subinterval
within the main time interval [8:50,12:20]**

Interval	Avg # Scheduled	% No-show	% Late	% (Late > 15 min)
[8:50, 9:20)	8.8±0.7	7.9±4.8	21.2±6.9	12.3±5.5
[9:20, 9:50)	7.7±0.5	6.9±4.6	16.7±6.1	4.8±3.4
[9:50, 10:20)	8.6±0.4	6.8±4.4	15.0±6.7	6.4±3.4
[10:20, 10:50)	8.1±0.6	7.9±3.2	17.6±5.0	3.3±2.9
[10:50, 11:20)	8.3±0.5	9.0±3.9	13.6±4.4	5.4±3.9
[11:20, 11:50)	8.4±0.3	7.9±3.7	10.4±4.7	3.9±3.0
[11:50, 12:20)	8.2±0.5	9.3±4.1	9.5±5.4	3.8±3.5
[8:50, 12:20)	58.0±1.3	8.0±1.7	15.0±1.5	5.8±1.6

Dr. 9: Lateness empirical cdf's in each 30-minute interval



Within the Day

Conditional on the daily total

If there is significant variability due to no shows and lateness,
Then the arrival process may be NHPP within the day,
conditional on the daily total.

[Key math: generalization of conditional uniform property]

Appointments might be like call centers in that way.

Even though daily totals are very different.

Appointment arrival data passed the KS
Tests of NHPP when viewed properly.

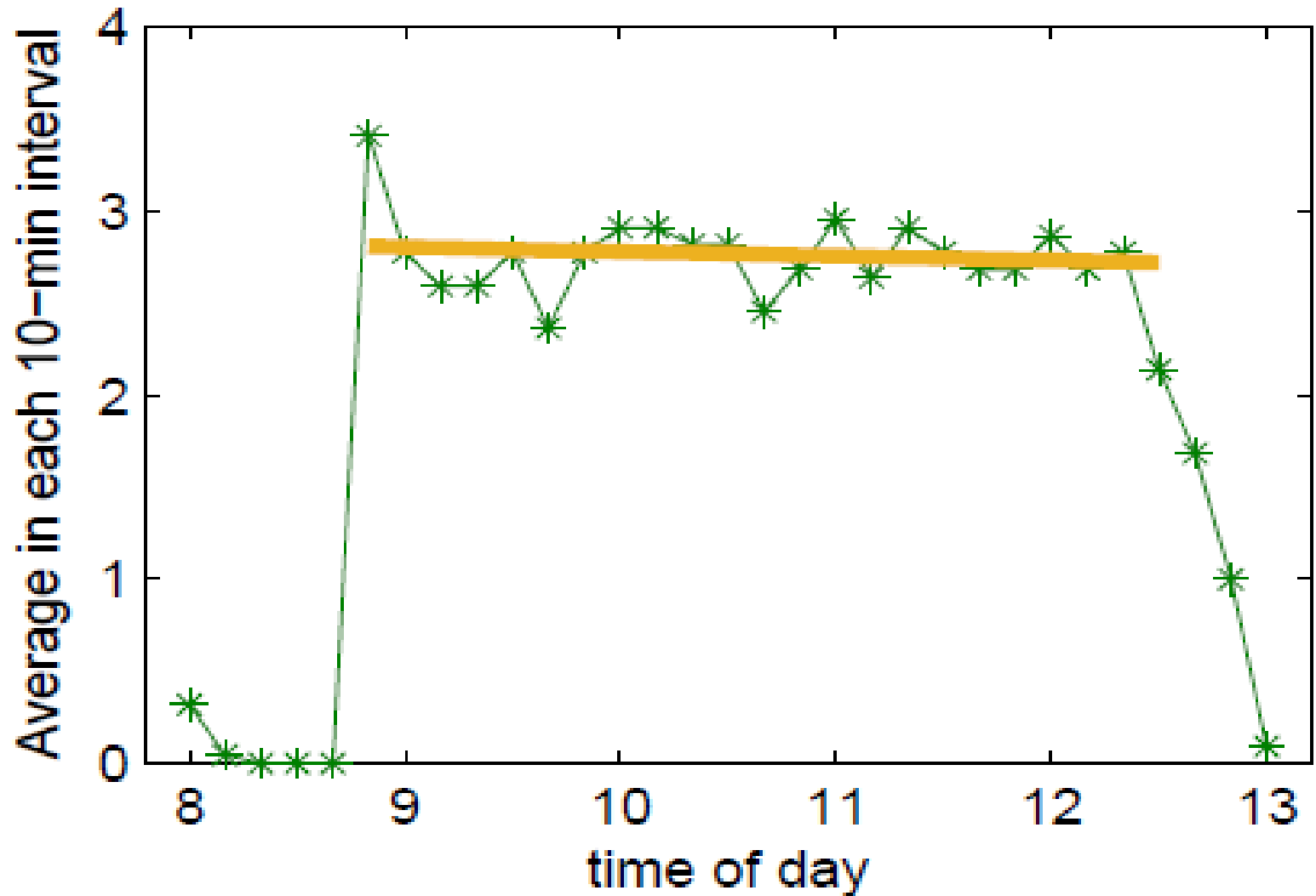
Next Move to the arrival rate function

Overall Arrival Process Model

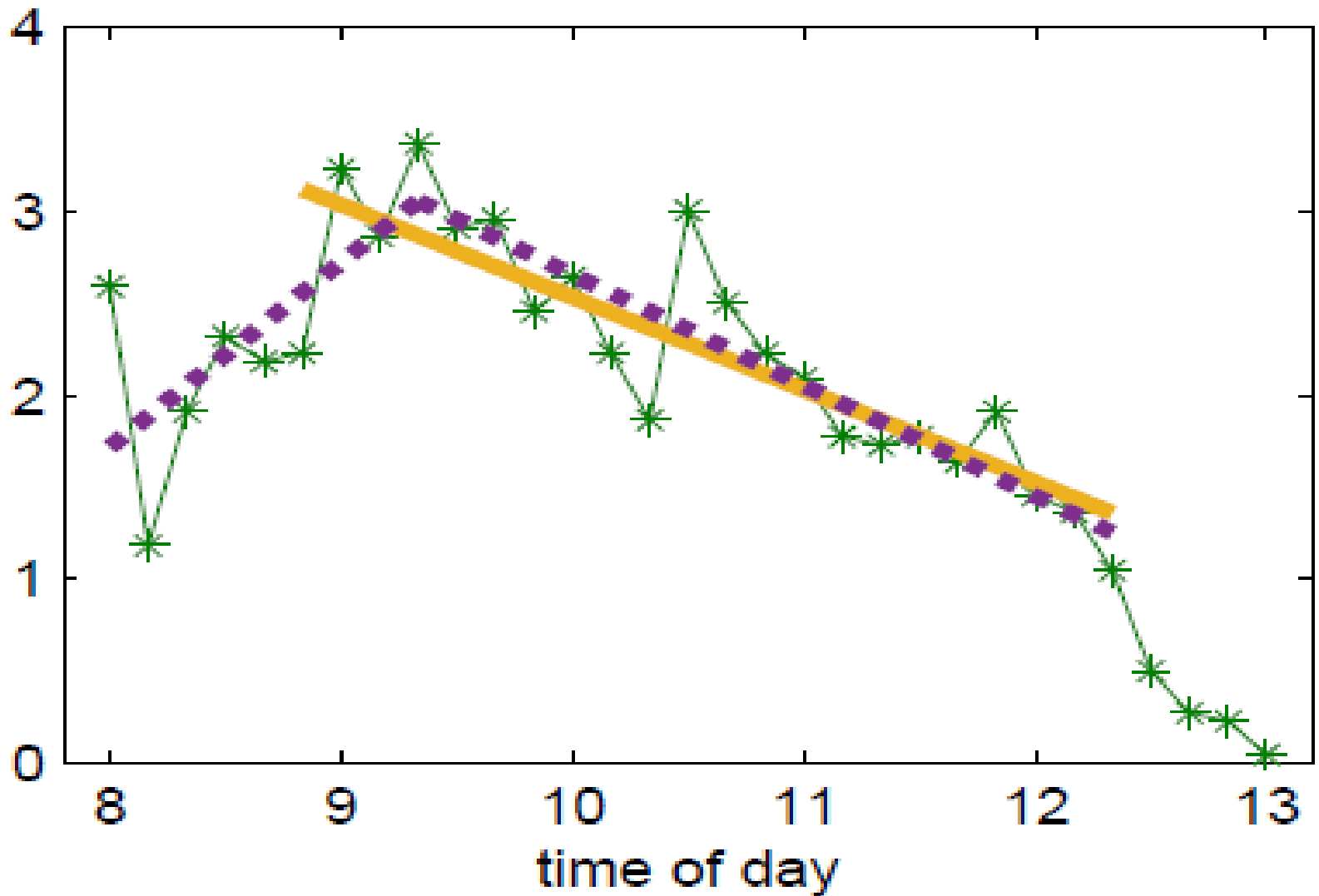
Daily Totals : Binomial or Gaussian with given mean and variance

Within the day, conditional on the daily total:
NHPP with deterministic rate

Average number of patients *scheduled* for Dr. 9
in each ten-minute interval of the 22 am shifts



Average number of patient *arrivals* for Dr. 9
in each ten-minute interval of the 22 am shifts



References

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2. Song-Hee Kim, Ward Whitt and Won Chul Cha. 2015. **A Data-driven model of an appointment-generated arrival process at an outpatient clinic.** Working paper, Submitted to *Operations Research*.

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1. T. Cayirli, E. Veral. 2003. **Outpatient scheduling in healthcare: a review of the literature.** *POMS* 12(4) 519-549.
2. L. Green, S. Savin. 2008. **Reducing delays for medical appointments: a queueing approach.** *Oper. Res.* 56(6) 1526-1538.
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