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Towards a Science of Health Care Quality: Notes From the Field

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What We All Know

- ...\$.30 to \$.40 of every dollar spent on health care...is spent on costs associated with “overuse, underuse, misuse, duplication, system failures, unnecessary repetition, poor communication, and inefficiency.”

IOM 2008

- 3/4 of patients surveyed described their medical encounter as “a nightmare to navigate”

American Hospital Assoc 2008

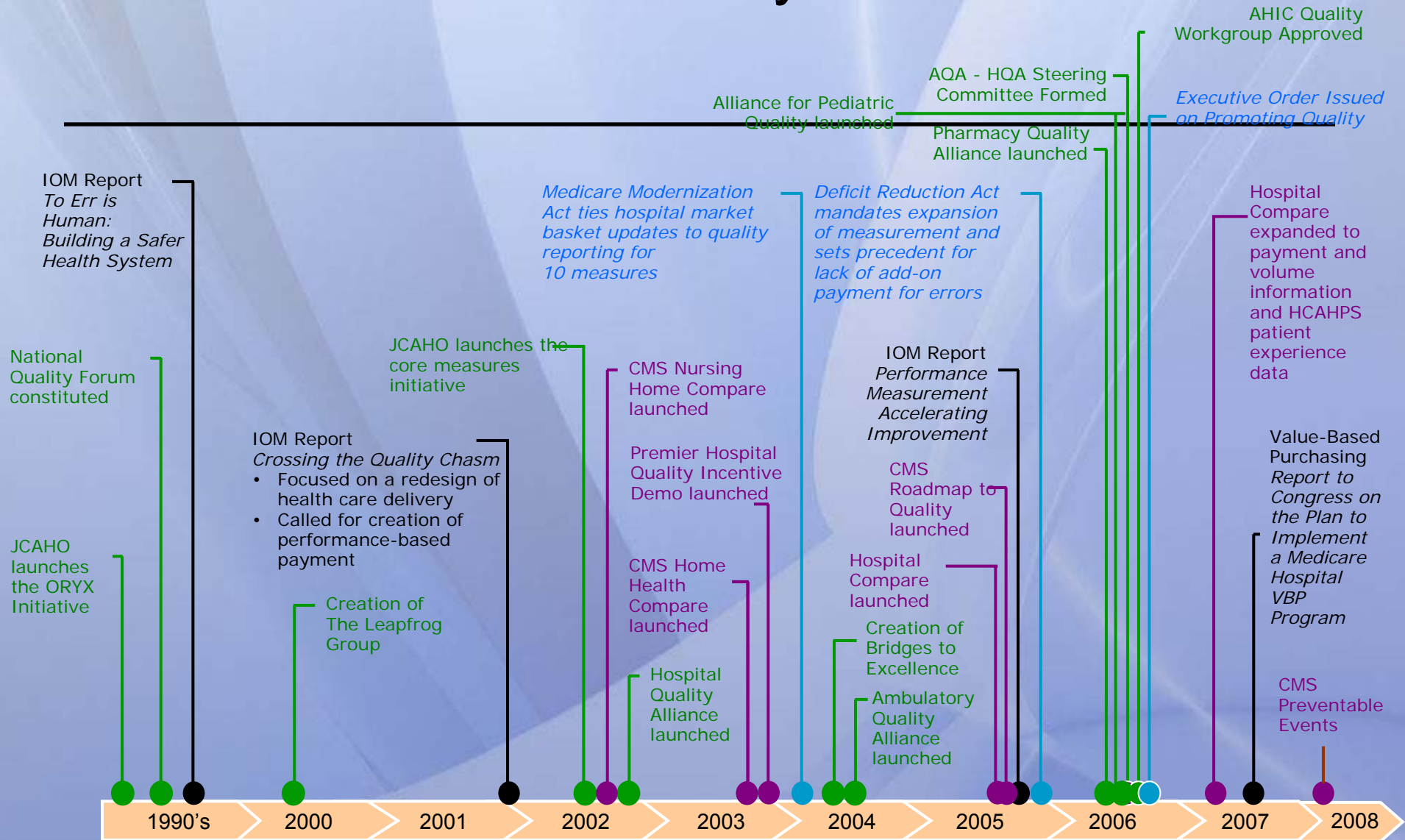
- 80,000 hospital deaths and 150,000 ambulatory deaths occur annually from Error

IOM 2006, 2008

O/R Techniques Have Little Impact

- Demand Forecasting has not reduced ED overcrowding, diversions, “sentinel” events or “boarding”
- Scheduling optimization has not reduced OR under-utilization or staff overtime
- EMR’s have nether reduced error (though they’ve changed the pattern), improved data analysis or improved patient –provider communication

Numerous National Quality Initiatives



...Adopting a Number of Quality Improvement Techniques

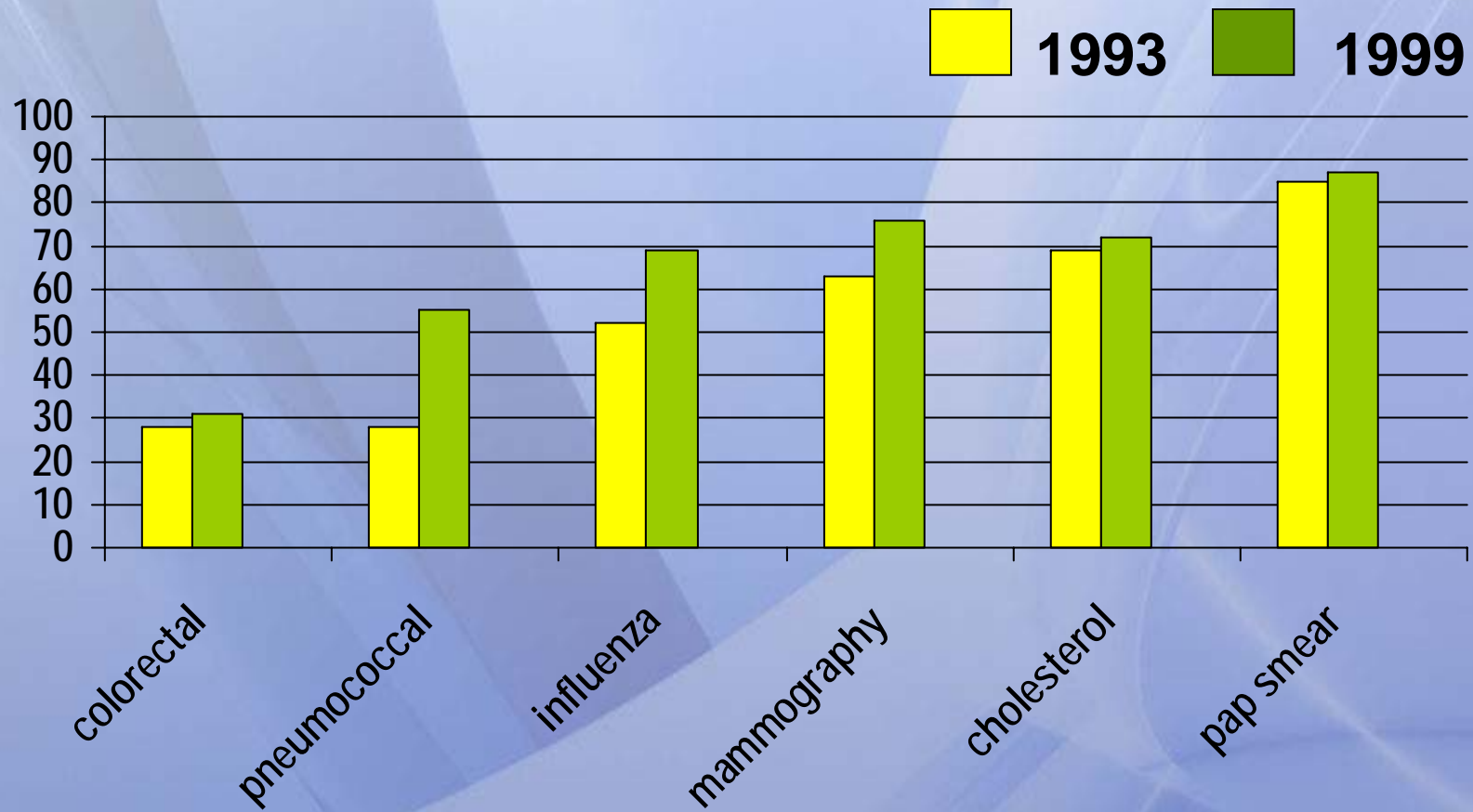
- Total Quality Management
- Continuous Quality Improvement
- Benchmarking
- Evidence-based Medicine
- Clinical Practice Guidelines
- Sigma Six
- Toyota Lean
- Pay-for-Performance

...Using “Rearview Mirror” Quality Measures

- % of 2 years olds fully vaccinated
- % of diabetics with H_{1c} screenings
- % of depressed patients on medications
- Joint Commission LD 3.10 and MM series
- Press-Ganey Patient Satisfaction Scores

**ALL SYSTEM OUTPUTS RATHER
THEN DETERMINANTS OF
QUALITY**

...Yielding



...Begg The Question

- How can a national system that leads the world in technological and biological innovation be so much less than the parts?

The Answer

Medicine lacks a *science* of health care quality

- Incoherent Health Service Research Agenda
- Initiatives treat quality as a black box
- Unable to describe the relationship between satisfaction, outcomes and efficiency

Science of Health Care Quality **Requires...**

- Theory
- Axioms
- Testable Hypothesis
- Field Observations
- Methodologies
- Identification of Key Constraints

The Phenomenon to be Understood

- Patients arrive with
 - Biologic conditions
 - Health beliefs
 - Health experience
 - Health goals/desires
- Providers Arrive with
 - Knowledge of human biological systems
 - Knowledge of treatment options
 - Knowledge of legal and ethical constraints
 - A sense of what is possible

Characteristics Shared with Other Service Industries

- Co-production of value
- perishable, heterogeneous, and intangible (but measurable)
- Inputs are human capital and knowledge
- Production and consumption are inseparable (but the value derived have different temporal gradients)
- Requires significant levels of mass customization and service provider judgment

Characteristics Unique to Health Care

- Consumer has little choice
- Consumer has little ability to objectively judge the quality of the service provided
- Timing of service can be critical
- Consumer satisfaction *NOT* meaningfully defined the same as in other industries
- Severe step-wise cost functions

This Dialectic Process:

(Axioms or First Principles)

1. Is dynamic and transformative
2. information intensive
3. often time-sensitive
4. Predicting and matching service demand to service capacity is critical
5. “outputs “are a customized (but categorical) patient and provider *mutual understanding* of what is possible.

Implications

- A core Service Science tenet posits that clients must take an active role as co-producers of the knowledge based solution to achieve a successful outcome
 - Technically correct care that is either not wanted or not understood erodes the “value” of clinical encounter
- Difficult to specify the terms of the interaction a priori
- Must agree on the expectations and scope of the consultation
 - Which will be dynamic and evolutionary
 - knowledge transfers will be “sticky”

Service Science Challenges

- Enabling technologies for smarter information use
- Reducing and managing uncertainty
- Quality assurance in real time
- When should variability be reduced-and would should it *not*
- Meaningful measurement systems
- Modeling and predicting processes
- Satisfization of resource use

Mutual Understanding Theory of Health Care Quality Frontiers

- The state of what is possible is defined as the health care quality *potential*
- This *potential* must be dynamic to meet changes in medical knowledge, the health system, and patient resources
- The dialectic process is *evolutionary*, as beliefs, knowledge and perceptions are altered by the process
- The interactions amongst the variables in this system are knowable and quantifiable and produce the quality potential *frontier*

Developing Objective Functions for:

Provider Variables

Practice Setting

Arrival Patterns

Informational Needs

Provider Personality

Patient Variables

Patient Health Status

Patient Health Beliefs

Patient Circumstances

Patient Personality

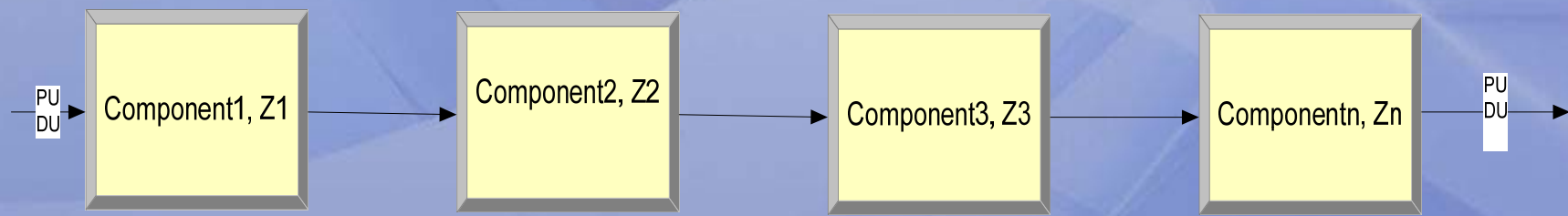
Provider-Patient Interaction



Individualized Desired Outcomes

Beginning the Mathematical Model

- Definition patient understanding array PU , and provider understanding array DU .
- Definition of element variables within understanding array A .
- Definition of component transfer function vector $Z_C^p(PU)$ and $Z_C^d(PU)$
- Definition of component inequality constraints (Potential/Frontier) $P V_{C_i}(A)$
- Definition of component current value $C V_{C_i}(A)$



Key Determinants or Inputs

Ratio of Patient-Provider time to all other time in Service

- Internal Parameters

- (Service Capacity = staffing, exam rooms, scheduling schemas)

- External Parameters

- (Service Demand = stochastic for ED, deterministic for Specialty, mixed for Primary)

- Information Parameters

- (External for Specialty, Internal for ED mixed for Primary)

Developing a Mathematical Model

- Principle Component Analysis
 - Outcomes (*the Gap between what is possible and what has occurred for individual patients*)
 - Patient satisfaction (*rooted in likelihood of compliance with agreed-to treatment recommendations*)
 - Efficiency (*ratio of patient-provider time to all other time in clinic*)



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Examples form the Field

Field Tests

- Direct Observations in
 - Primary care (Pediatric, Emergency Departments, Family and GIM clinics)
 - Rural and Urban
 - Specialty care (e.g. Nephrology, Orthopedics, Cardiology, HIV)
 - Large (V.A. and Kaiser), Medium (15-60 providers) and small (>3 providers) clinics

Field Study 1: Matching Service Demand to Service Capacity in an Emergency Department

- Nearly half of U.S. hospitals report ED crowding on a daily basis
- Seriously ill and injured ED patients can wait 12-48 hours for an inpatient bed
- $> \frac{1}{2}$ million inbound ambulances diverted in 2007
- “Crowding” / imbalance in patient/staff ratio responsible for 35% of “sentinel events

Emergency Department Constraints

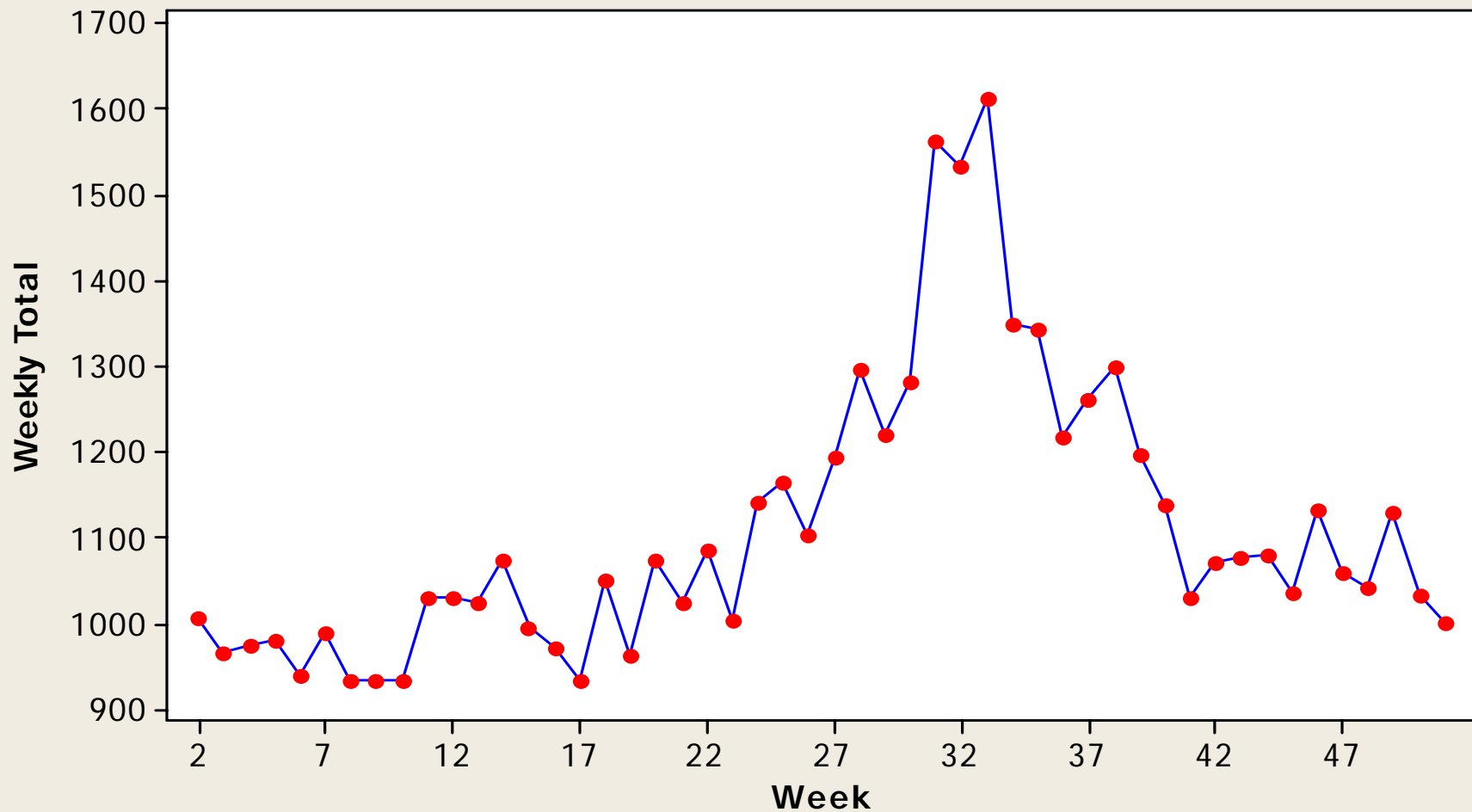
- Must accept all takers
- Legal mandates (with financial penalties) on staffing
 - 1 RN per 3 low acuity
 - 1 RN per 2 medium acuity
 - 1 RN per 1 high acuity
- Joint Commission Requires ED's to examine Patient flow (JCAHO LD.3.15 leadership standard)
- Severe Step-wise cost functions

ED Benchmarks

- Arrival to triage 6 min
- Triage to room 23 min
- Arrival to MD 53 min
- Arrival to admit 253 min
- Arrival to discharge 139 min
- Lab turnaround 49 min
- Radiograph turnaround 47 min
- LWOT 2.3%
- Pts per MD hour 2.25 pts/hr

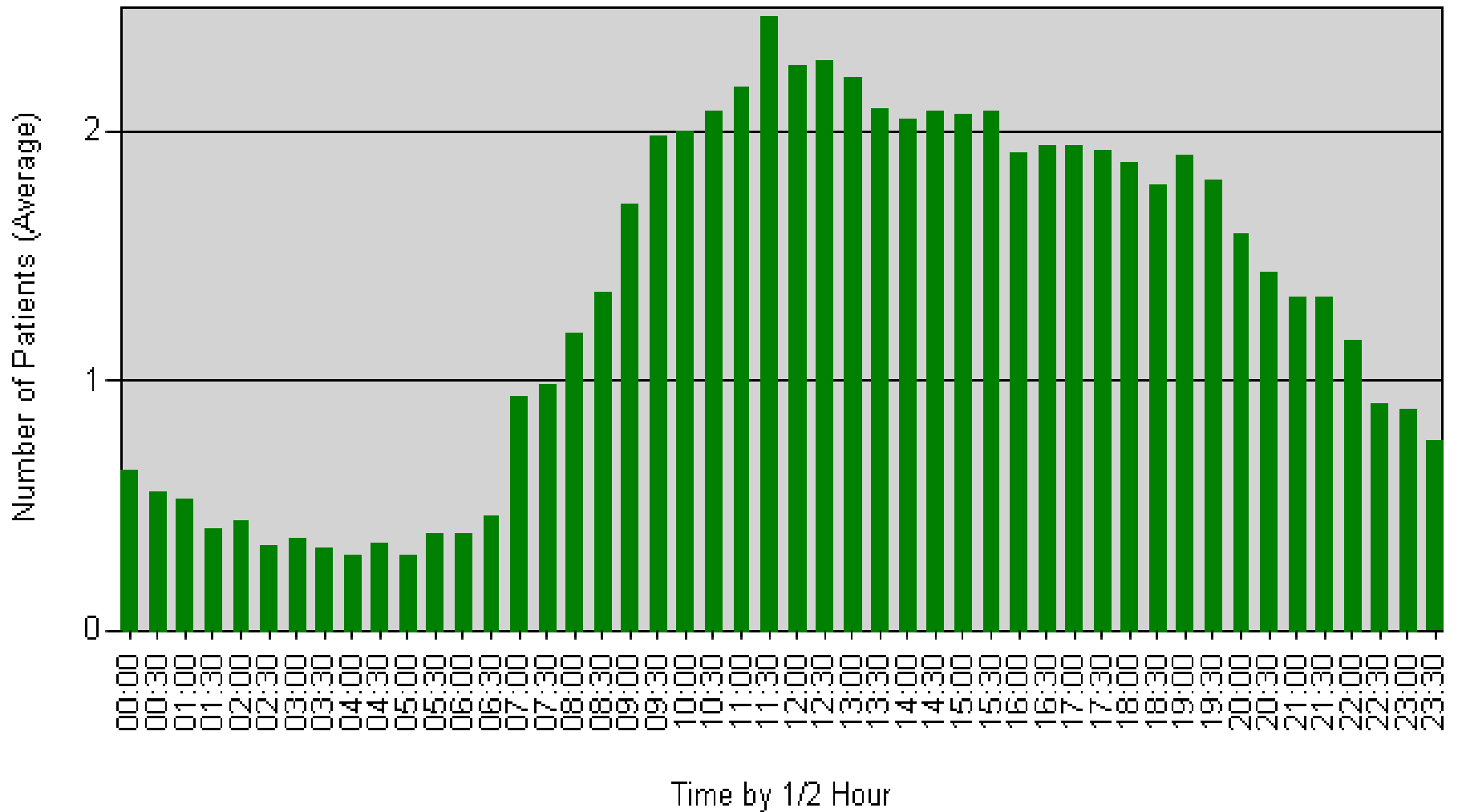
Annual Service Demand Mapped to Week of Year

Time Series Plot of Weekly Total



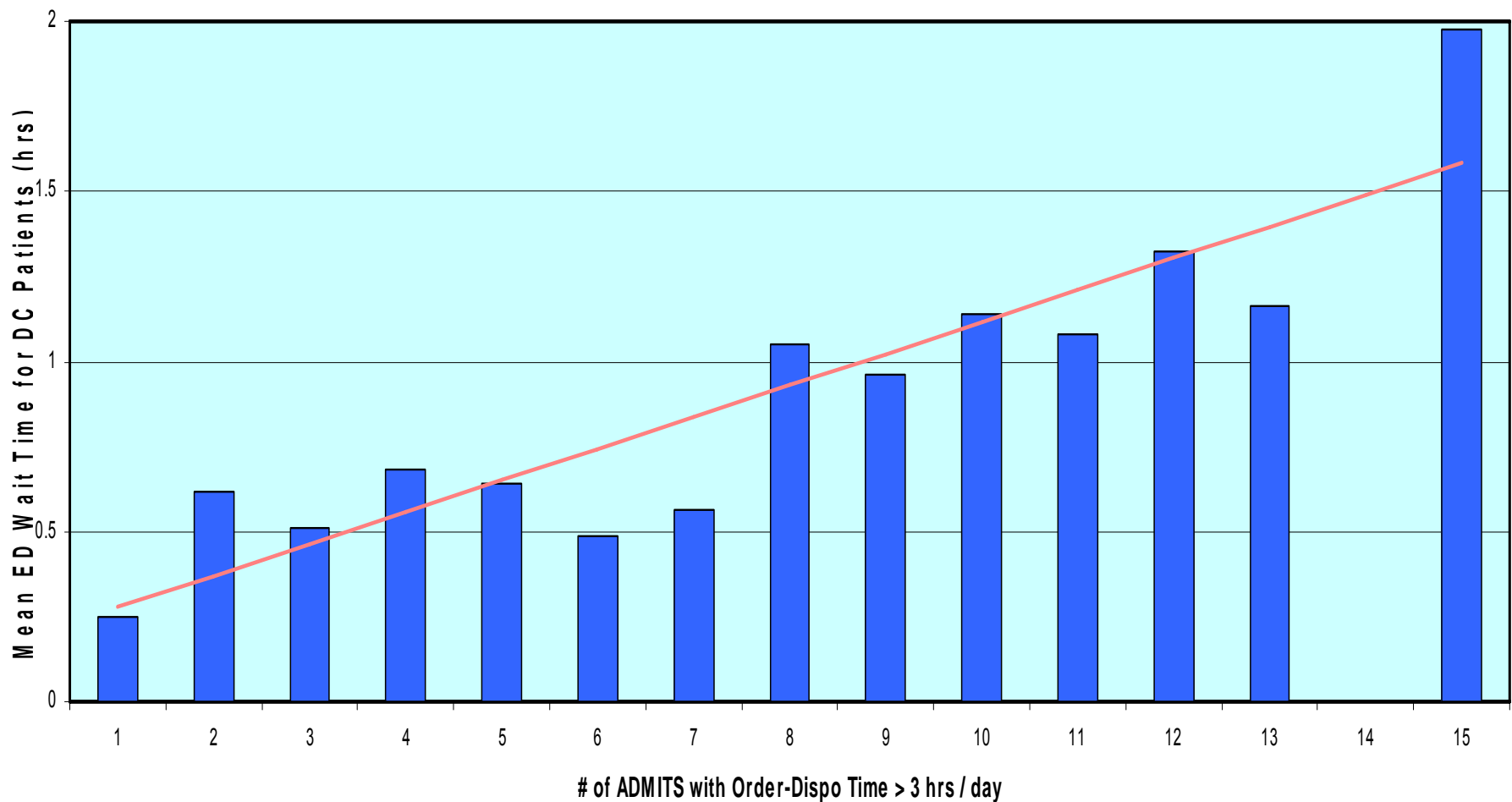
Average Hourly Demand

Triage Time (24 Hour)

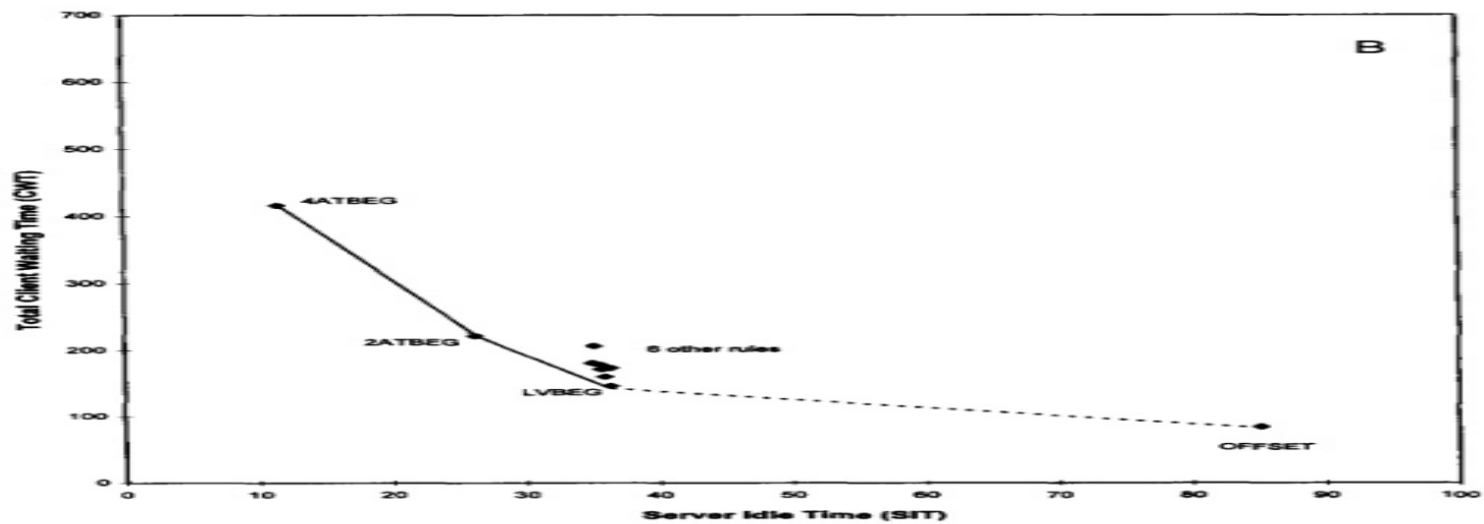
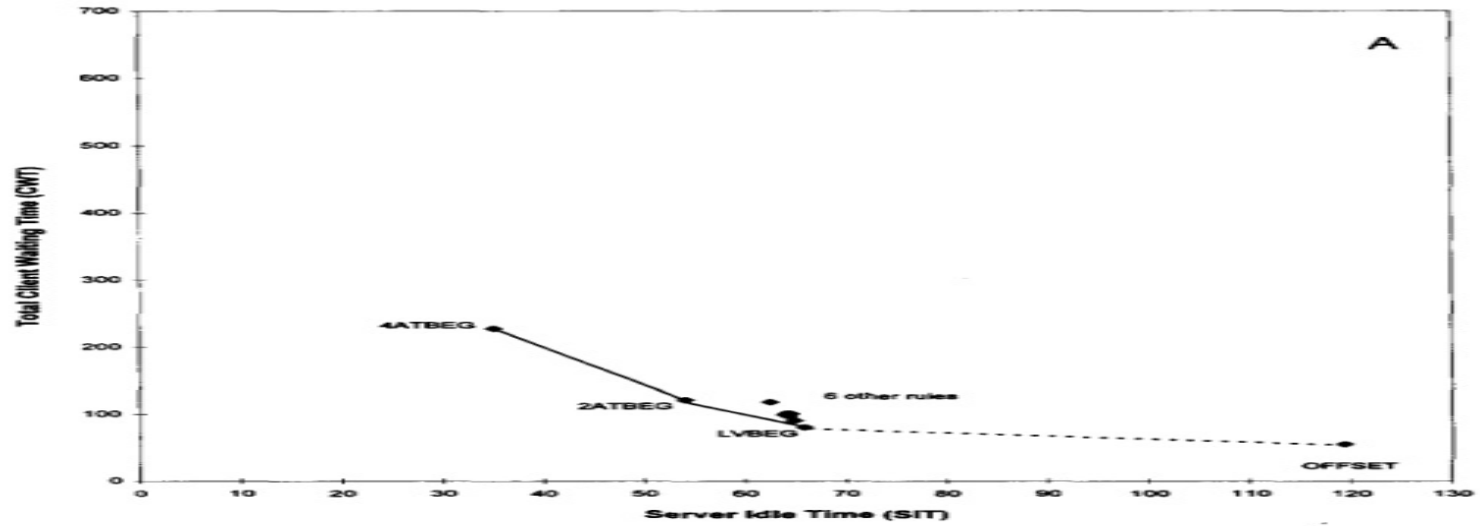


Mean ED Wait Times by # of Admits

Mean ED DC Wait Time by # ADMITS > 3 hrs / day

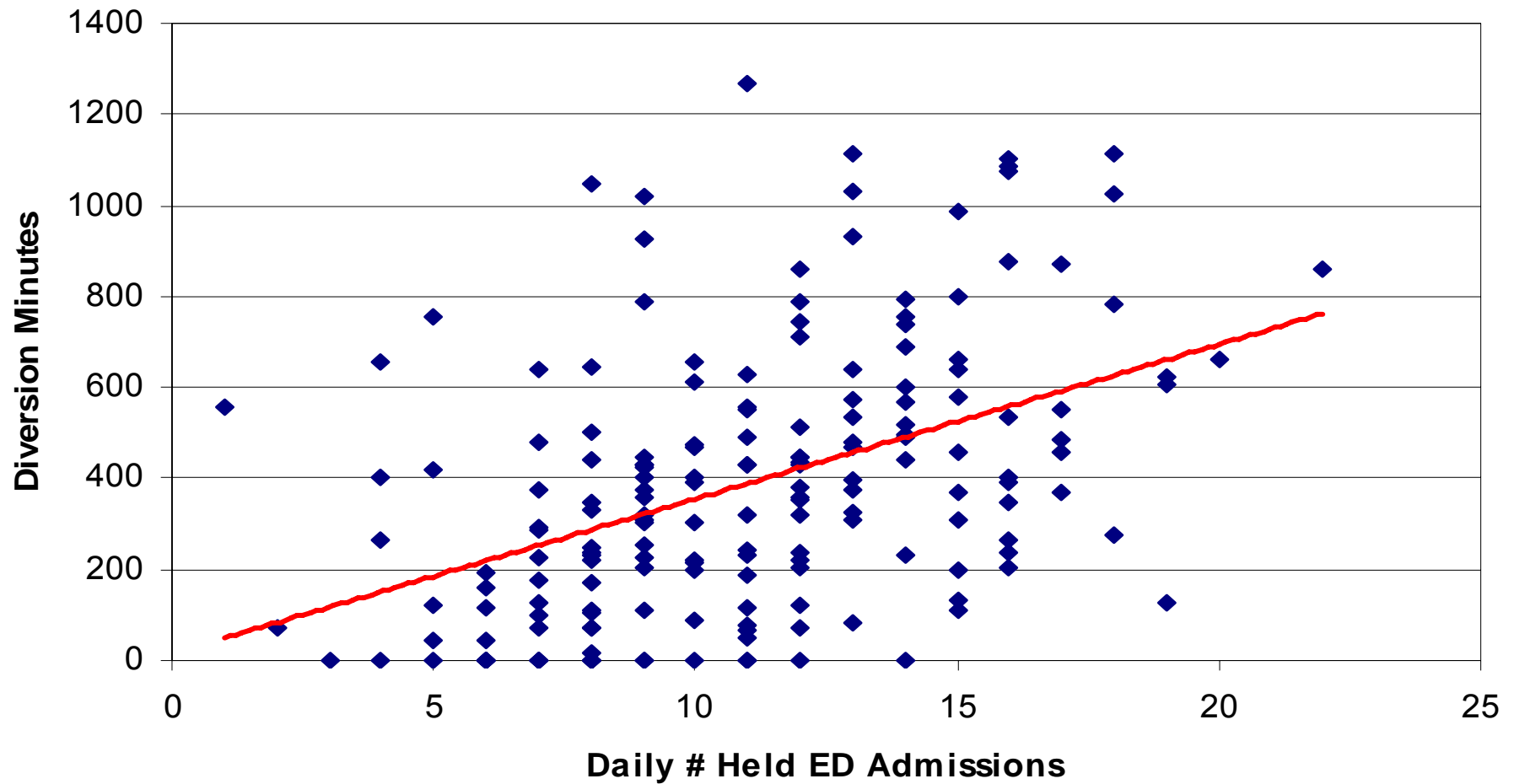


Patient's Waiting Time vs. Doctor's Idle Time



Ambulance Diversion – Impact of Delayed Admits

DAILY AMBULANCE DIVERSION MINUTES



Predictive Modeling

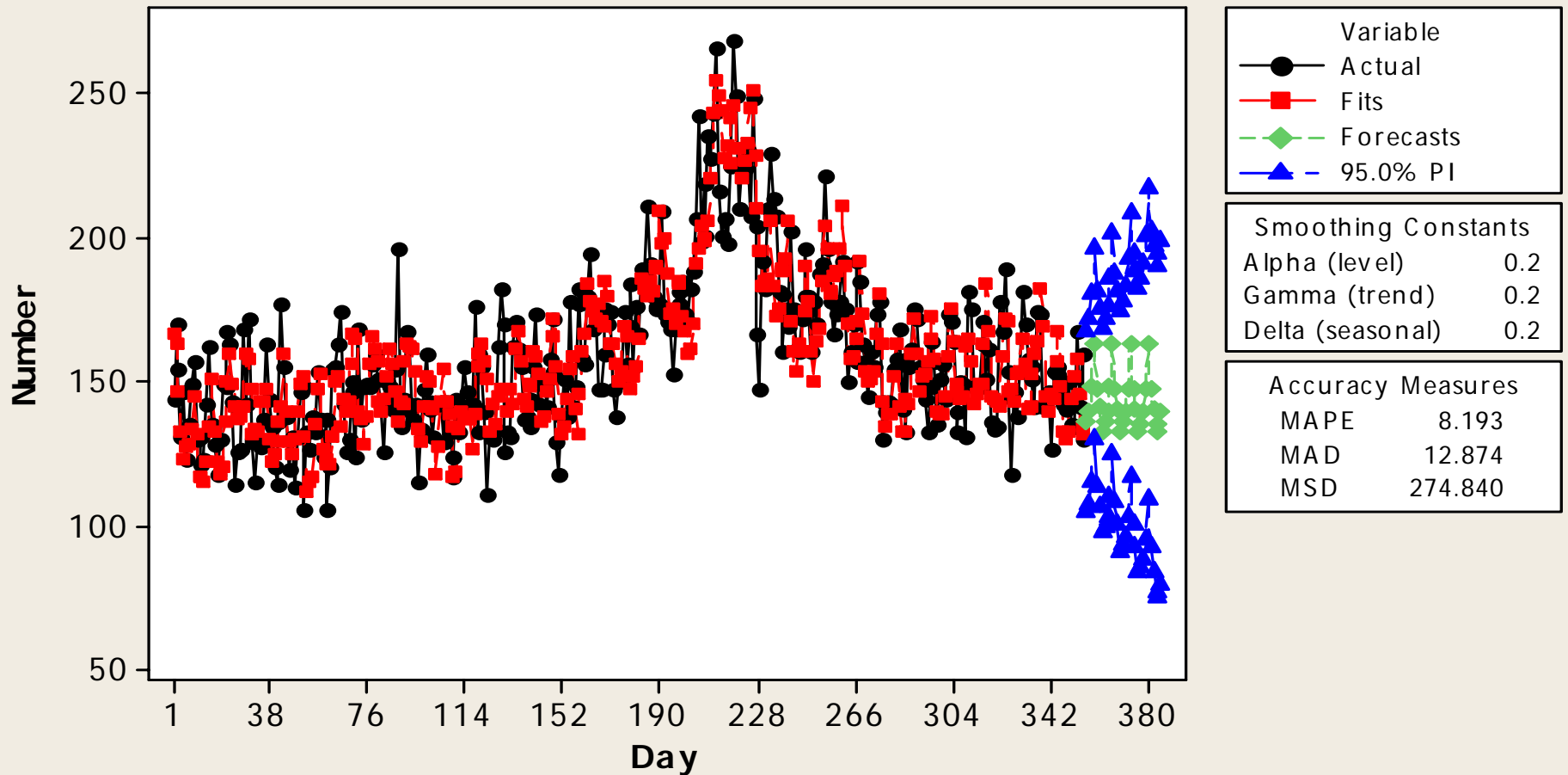
- AutoRegressive Moving Average (ARIMA)
- Exponential Smoothing Models
- Poisson regression Models
 - All assume Gaussian Process
 - But there is obvious interaction in ED Demand

Modeling Inputs

- Data Mining 24 months of patient demand by:
 - DX
 - Patient Characteristics
 - Were patients idled within ED
 - Were staff idled within ED
 - “Feeder” institution data (e.g. County Mental Health discharges)
 - *Readmission within 24 hours of discharge*

Bending Light: Winters Methods Added to an ARIMA Model

Winters' Method Plot for Number
Additive Method



When Staffed to Predicted Demand

- Reduction of 24 hour readmits reduced by 32%
- Non-compliance staffing ratio reduced from 16% to 0%
- Diversions reduced from 16% to 3%
- Non-Emergent patients referred to FDQC's

Field Study 2: Service Demand Modeling and Primary Care Screening Ordering

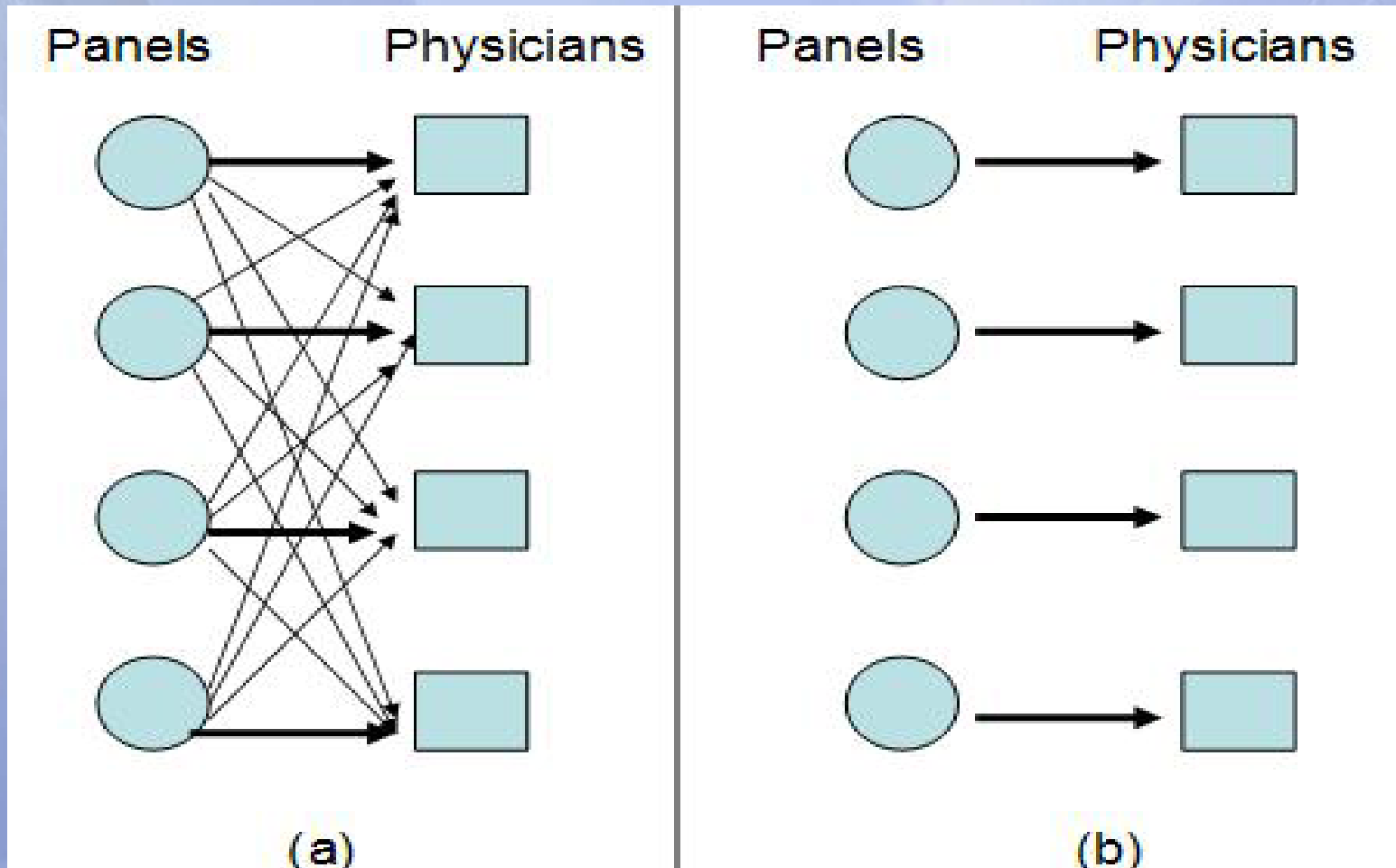
- It is standard of Care to Provide Diabetic Pt's routine, scheduled screening tests
 - Eye Exam
 - Ha1C
 - Lipid Screening
 - Foot Exam
- National Average is about 73% for any 1 item- less then 45% for all items
- Significant impact on Patient Health
- Major Pay-for-Performance Measure

The Moving Parts

Finite appointment slots and fixed individual provider times

- Permitting indiscriminate visits with any GIM physician is expensive
 - takes longer
 - May not identify unique characteristics of patient-e.g. antibiotic response
- Requiring Urgent visits with PCP *only* is not feasible
- GIM exists within a larger Medical Group
 - There is an ED

The Two Common Models



The Moving Parts: Capacity Allocation for Two Stream Demand

Scheduled and Urgent

- Each has different patient requirements for timeliness and continuity
 - Scheduled is Deterministic
 - Urgent is Stochastic
- Each has different cost-structures and revenue

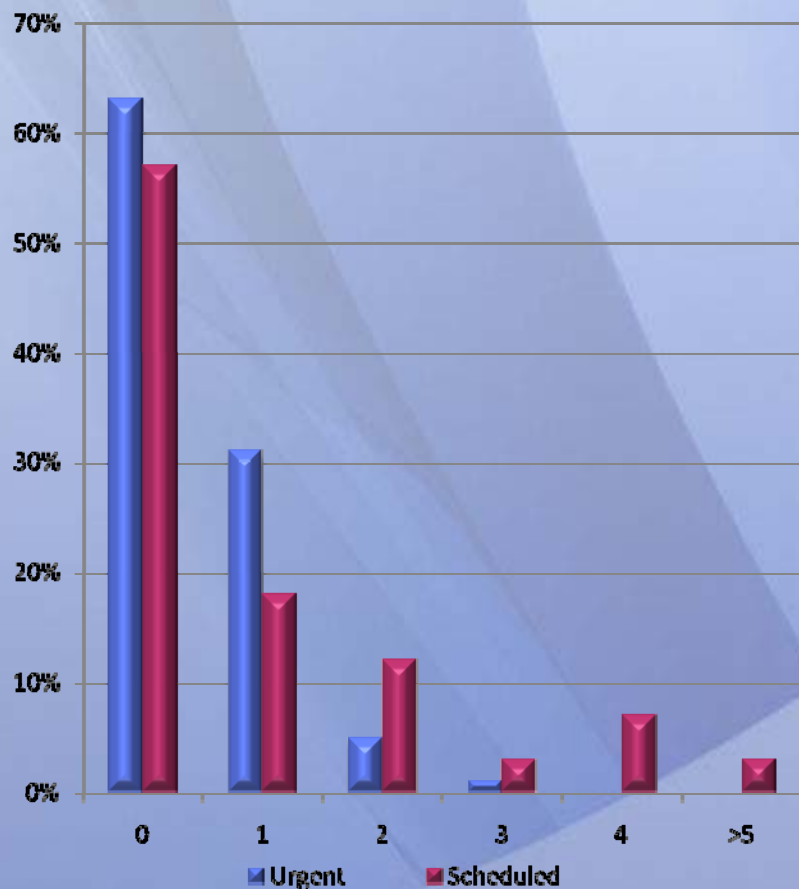
Analyzing Demand-Urgent Care

- Size of provider panel
- Complexity of panel
 - Age
 - Chronic condition
 - Health Expectations
 - Copay structure
- Sub-set Patient demand Characteristics

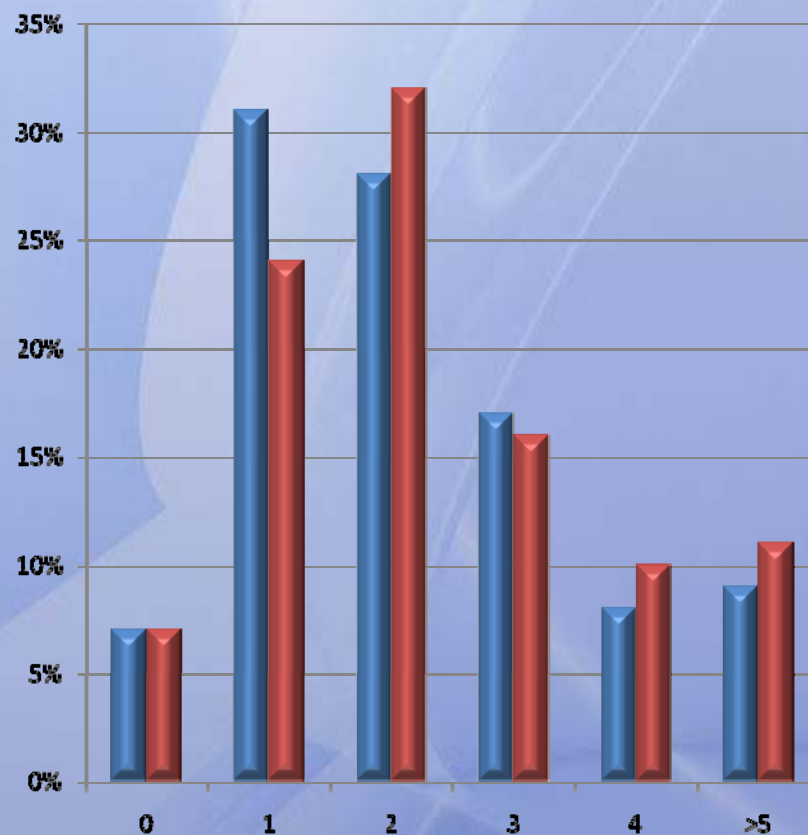
An Example from an Academic Medical Center in Dan Diego

- Group 1: Women < 50 yrs of age with no chronic condition
 - Low demand and mostly for urgent care appointments –
- Group 2: Anyone over 69 with chronic heart disease
 - High demand -mostly for scheduled appointments with a seasonal spike in

Type and Volume of Visits-2009



Healthy Women < 50



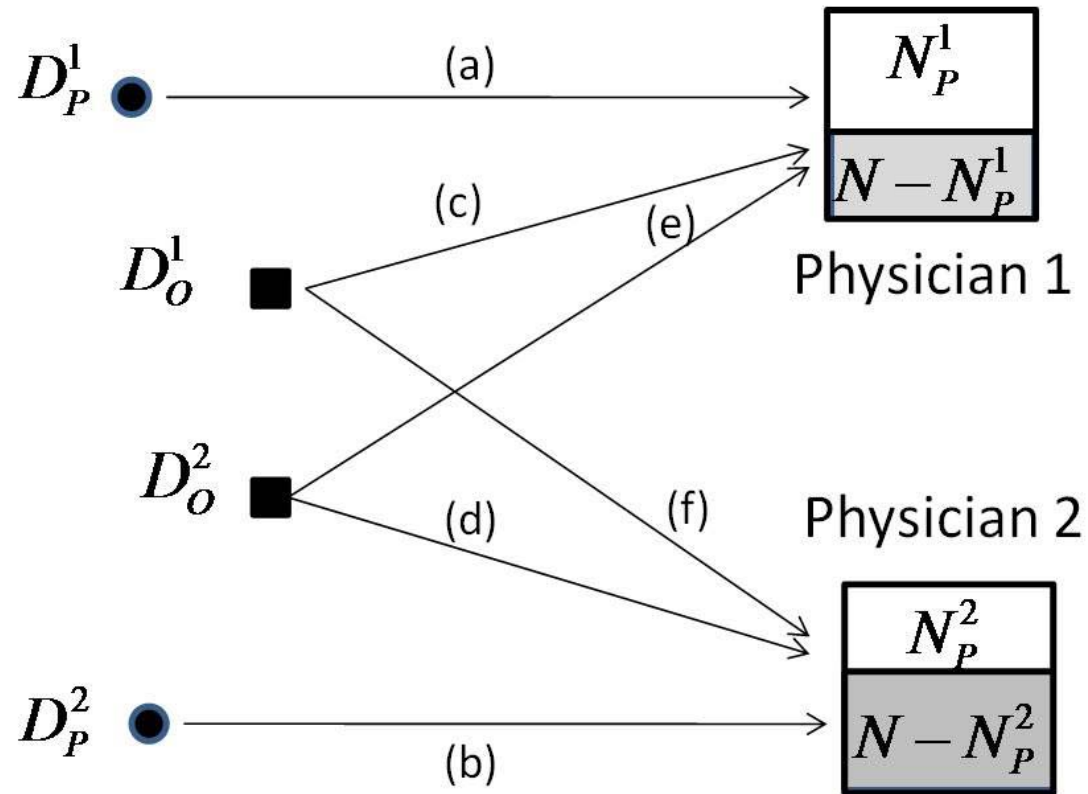
Chronic Illness > 69

■ Urgent ■ Scheduled

Decision Process: Appointment Booking Policies

- 1) Strategic - Build the Patient-Physician relationship through access
- 2) Tactical, reserve XX daily slots for open access patients and anticipate future appointments up to the remaining slots
- 3) Operational level, appointments for daily open access demand given dynamically

Third Model: Partial Flexibility Chained Scheduling



Mathematical Modeling

- N is the total daily capacity
- N_p , decision variables (aka performance and philosophical constraints)
- D_p demand for scheduled
- D_o demand for urgent care

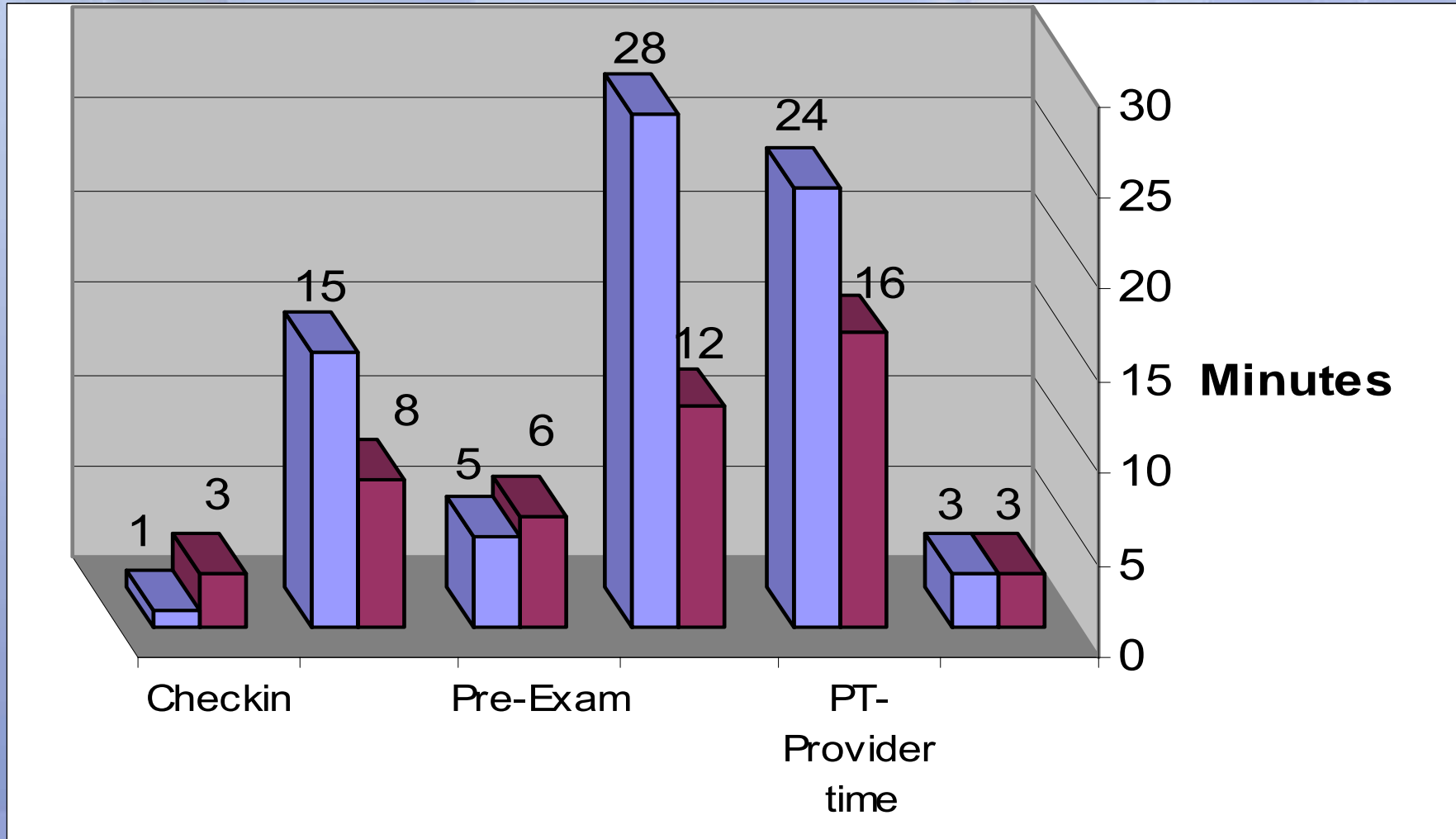
Creates

- $\text{Min} N_p \text{ } c_p E[D_p - N_p]^+ + c_o E[D_o - (N - \min(N_p, D_p))]^+$

Results of Improved Scheduling: Adherence to Policy

	Baseline	Follow-up
Activity	Frequency	Frequency
Patient reports having been reminded of appt	31%	85%
Contact information verified at check-in	74%	92%
Patient visit terminated before seeing a Provider	3%	0%
Patient chart “pulled” by check-in/pre-exam	60%	86%
Health history taken	26%	86%
Influenza vaccination rate	43%	68%

Results of Improved Scheduling : Patient Throughput



Results of Improved Scheduling: Rate of Screenings

	Baseline	Follow-up
Ha1c	67%	93%
Lipid screening	63%	91%
Eye Exam	43%	84%
Foot Exam	37%	89%

Results of Improved Scheduling: Patient Satisfaction

	Baseline	Follow-up
Registration	3.0	4.2
Waiting Room	3.0	4.4
Waiting for Provider	3.0	4.2
Doc	4.0	5.0

Field Study 3: Operational Conditions Affecting Patient Understandings of Treatment

- 13% of Pediatric Transplant “fail” annually
 - failure to comply with treatment recommendations
- Traditional Studies of Non-Compliance Focus on difficult to change factors
 - Patient SES
 - Provider practice type
 - Patient-Provider communication style
- Little analysis of the structural determinants of “understanding”

Input: Uncontrolled Patient Arrival Patterns Determine Wait Times

- Traditionally assume patients arrive randomly (bell shaped curve peaking at appointment time)
 - Each patient’s “timeliness” is independent of the other’s
 - “Timeliness” is within patient control
 - Consistent arrival patterns throughout the day

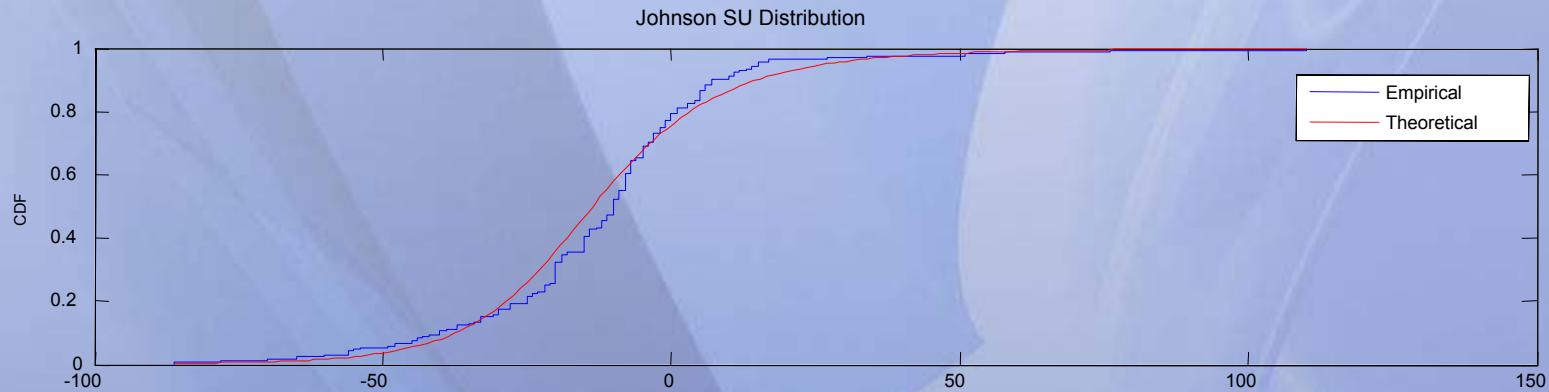
Tardiness(GM): Johnson S_U fits well

Method: Moment matching

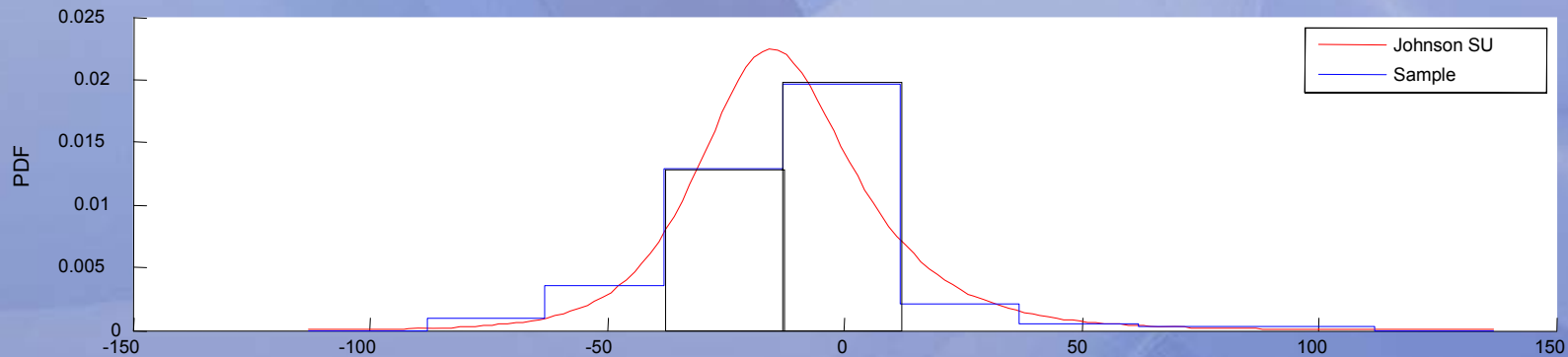
$$\begin{aligned} \gamma &= -0.3099 & \delta &= 1.432 \\ \lambda &= 24.5789 & \xi &= -19.02 \\ &= 25.03 & & \end{aligned}$$

Goodness-of-fit tests

K-S statistic	0.1110	χ^2_{GOF} statistic	
Significance probability	0.0296	Significance probability	0.0009



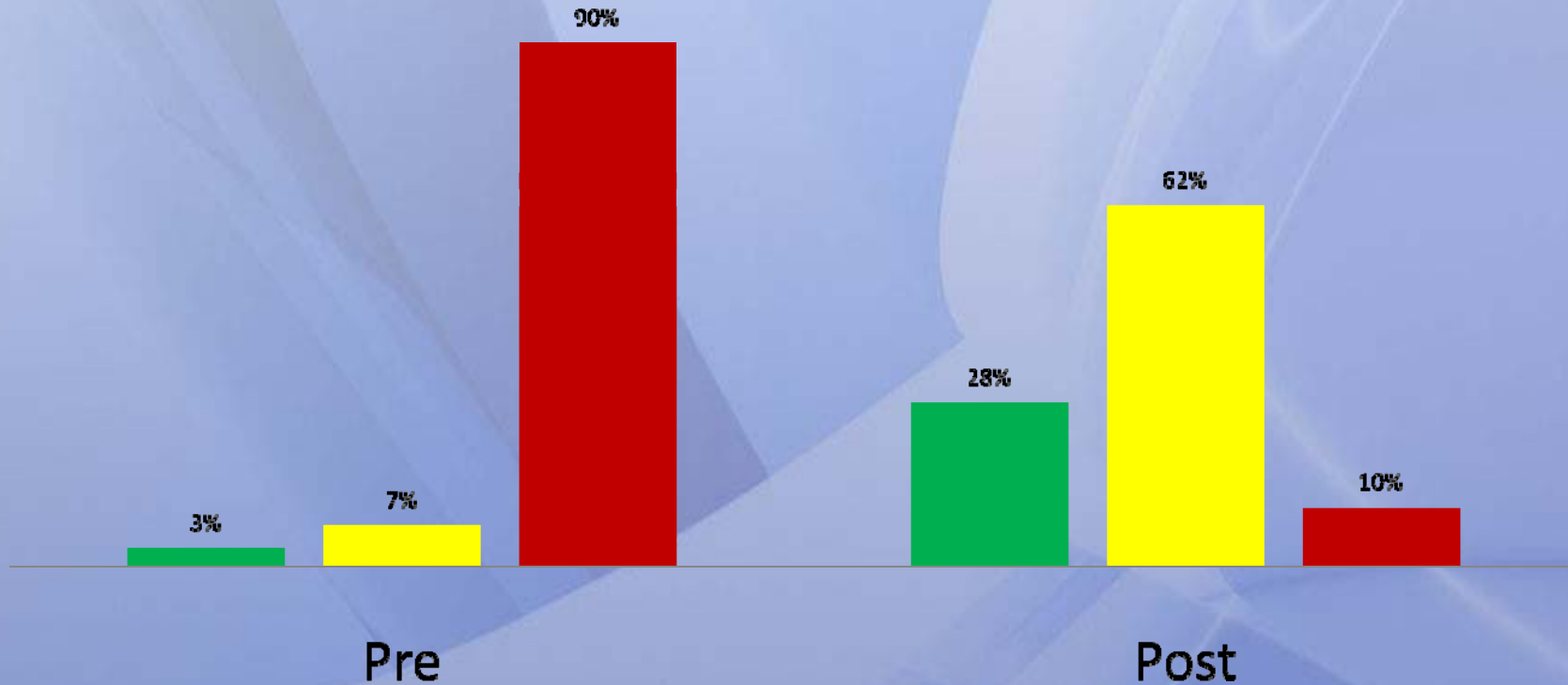
Patient Tardiness



Patient Tardiness

Change in Patient Understanding of Provider Recommendations

■ 100% ■ <50% ■ >50



Scheduling Optimization goals

1. Revenue
2. Patient throughput
3. Patient access
4. *Fairness*

Results of Naïve Modeling

- Schedule using *Easy* First Five rule gives 10.8% decrease in total cost over present scheduling system
- If classification of patients into easy and hard is implemented, it can reduce total cost by more than 10% on average.