

**PREDICTIVE MODELING OF EMERGENCY
DEPARTMENT OPERATIONS:**

**EFFECT OF PATIENTS' LENGTH OF STAY ON ED
DIVERSION**

**ALEXANDER KOLKER, PhD
Froedtert Hospital
Milwaukee, Wisconsin**



Froedtert Hospital, Milwaukee WI

- Primary teaching Hospital for MCW
- Tertiary Referral Center
- Level 1 Trauma Center for SE Wisconsin
- 433 staffed acute care beds
- 23,617 admissions/ 47,176 ED Visits
- 454,780 Outpatient Clinic Visits
- Surgeries - Inpt: 9,034/ Outpt: 5,711

Froedtert HOSPITAL

Froedtert & Community Health

PROBLEM STATEMENT:

- Froedtert Hospital ED ambulance diversion has become unacceptably high due to no ED beds_(average ~21% of time in 2007)
- There are two big groups of patients:
 - (i) admitted to the hospital, and
 - (ii) treated, stabilized and discharged home
- Among factors that affect ED diversion patients' Length of Stay (LOS) in ED is one of most significant one.

The Goal of this work was:

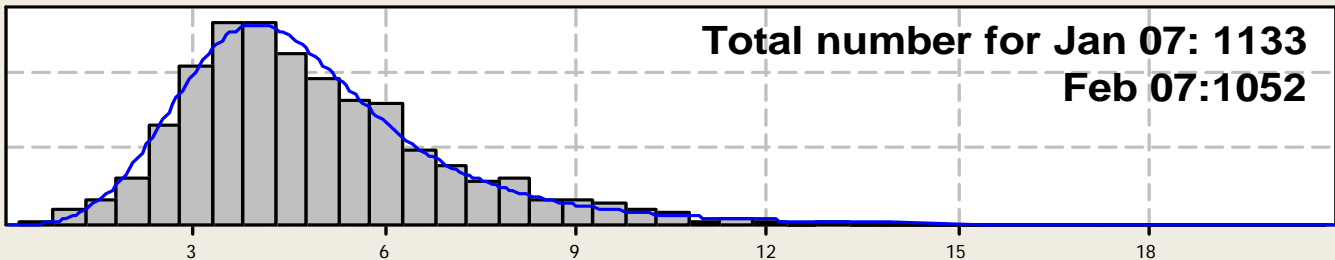
- develop a methodology that could quantitatively analyze and predict an impact of patients' LOS on ED diversion (both for admitted and discharged home patients' groups).
- identify the maximum LOS limit that will result in significant reduction or elimination ED diversion.

STEPS THAT HAVE BEEN PERFORMED:

- Collected data on
 - patients arrival in ED: day of week and time of day
 - mode of transportation: walk-ins or ambulance
 - discharge disposition: home / expired or admitted in the Hospital
- Analyzed LOS distribution and its functional approximation for
 - (i) discharged home, and
 - (ii) admitted patients
- Developed an ED **Process Model** aimed at modeling different scenarios of **LOS upper limits** that will result in significant reduction (or elimination) ED diversion
- Summarized the basics of modeling methodology: What did we learn ?

What is the LOS distribution for admitted and discharged home patients ?

Summary for LOS admitted, hrs

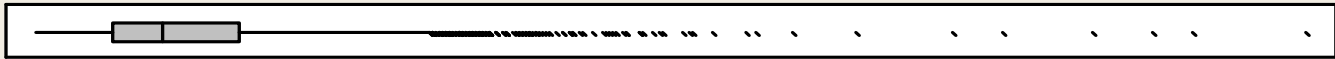
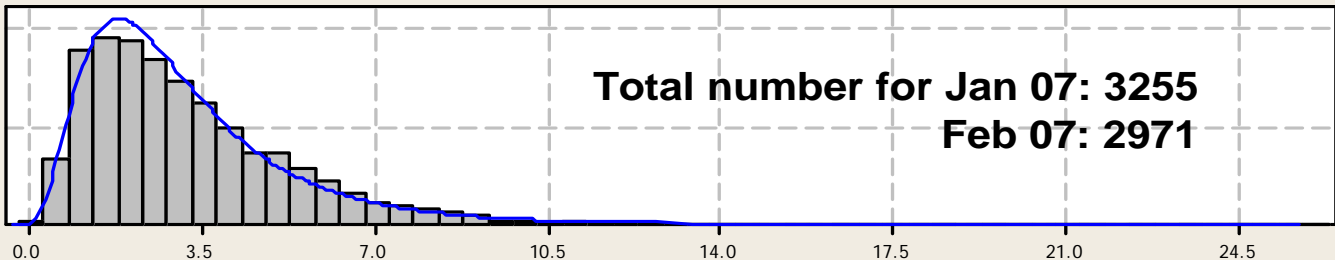


95% Confidence Intervals

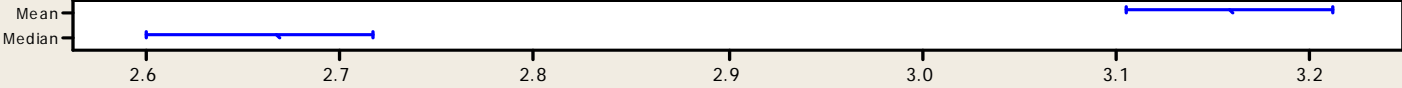


Anderson-Darling Normality Test	
A-Squared	32.74
P-Value <	0.005
Mean	4.8654
StDev	2.1305
Variance	4.5389
Skewness	1.22735
Kurtosis	3.14050
N	2185
Minimum	0.4667
1st Quartile	3.3833
Median	4.4667
3rd Quartile	5.9333
Maximum	20.4333
95% Confidence Interval for Mean	
	4.7761 4.9548
95% Confidence Interval for Median	
	4.3667 4.5667

Summary for LOS home, hrs



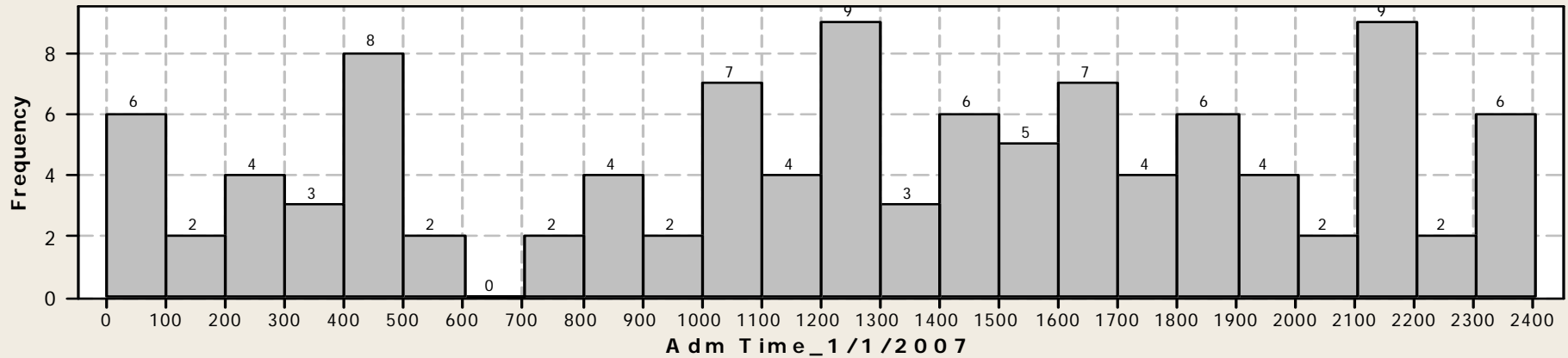
95% Confidence Intervals



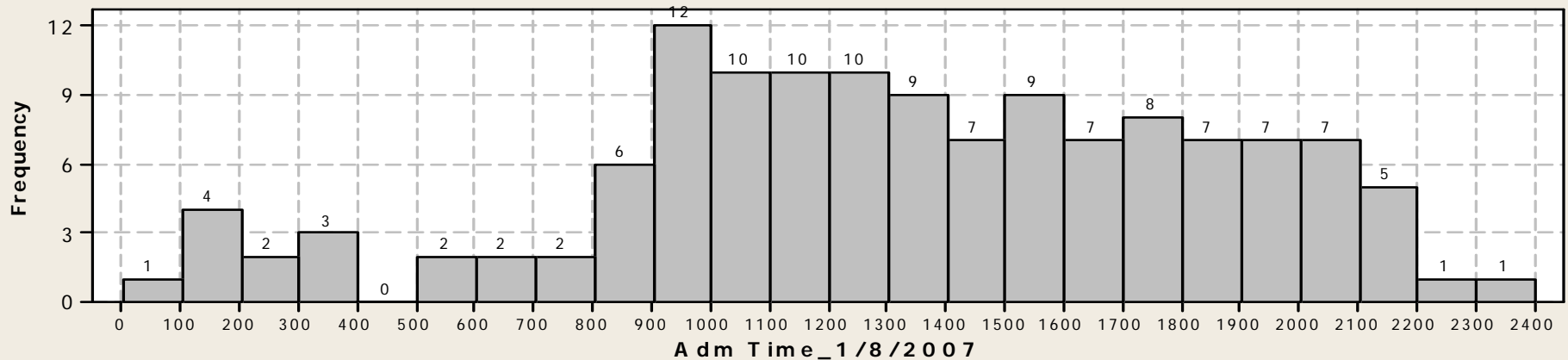
Anderson-Darling Normality Test	
A-Squared	146.60
P-Value <	0.005
Mean	3.1569
StDev	2.1225
Variance	4.5051
Skewness	1.84310
Kurtosis	8.14424
N	6155
Minimum	0.0667
1st Quartile	1.6167
Median	2.6667
3rd Quartile	4.2000
Maximum	25.8167
95% Confidence Interval for Mean	
	3.1039 3.2099
95% Confidence Interval for Median	
	2.6667 2.7167

DO THE SAME DAYS OF WEEK FOR DIFFERENT WEEKS HAVE A SIMILAR ARRIVAL PATTERN ?

Monday Adm Time_1/1/2007



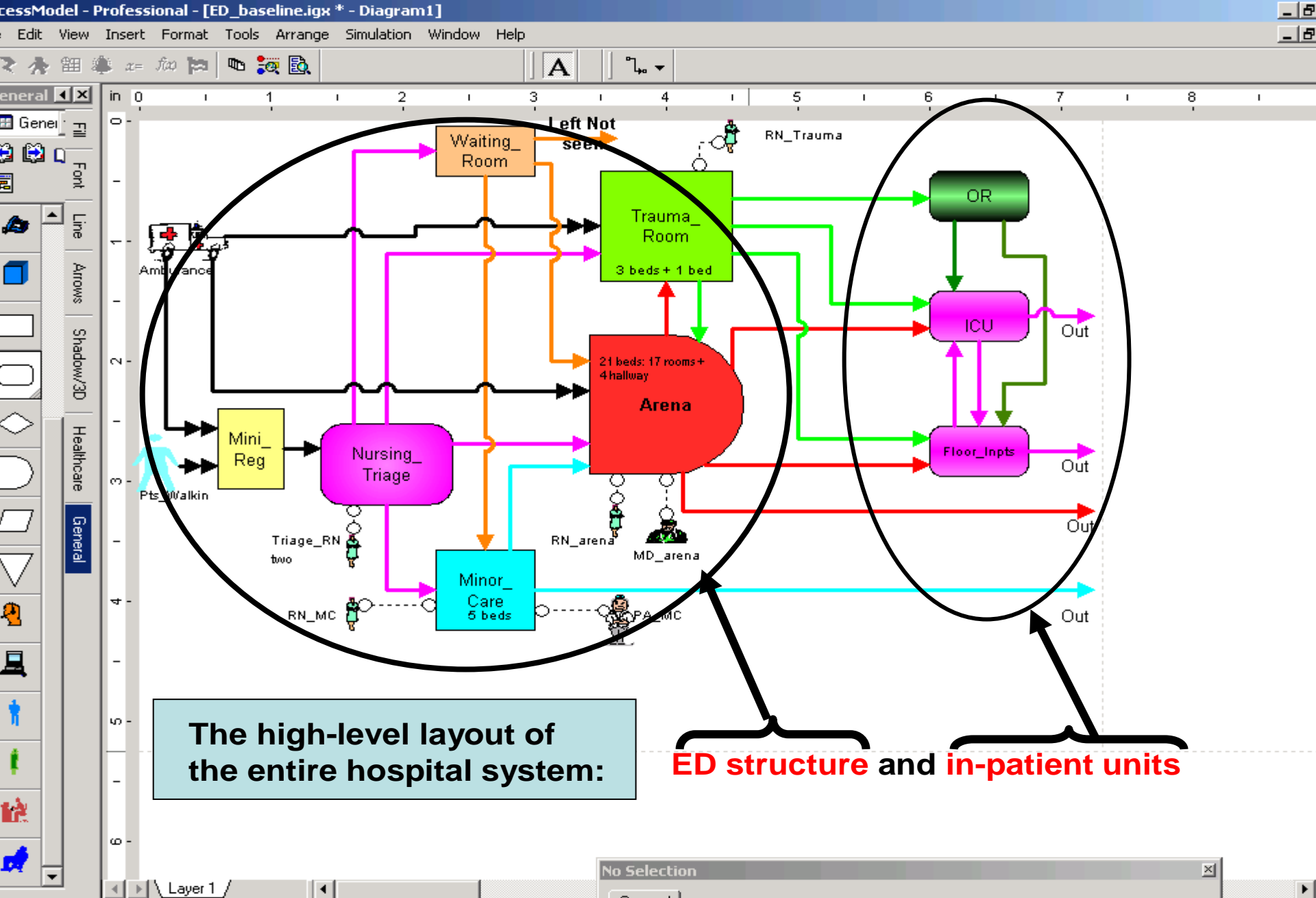
Monday Adm Time_1/8/2007



Take away:

Same days for different weeks have very different patient arrival pattern.

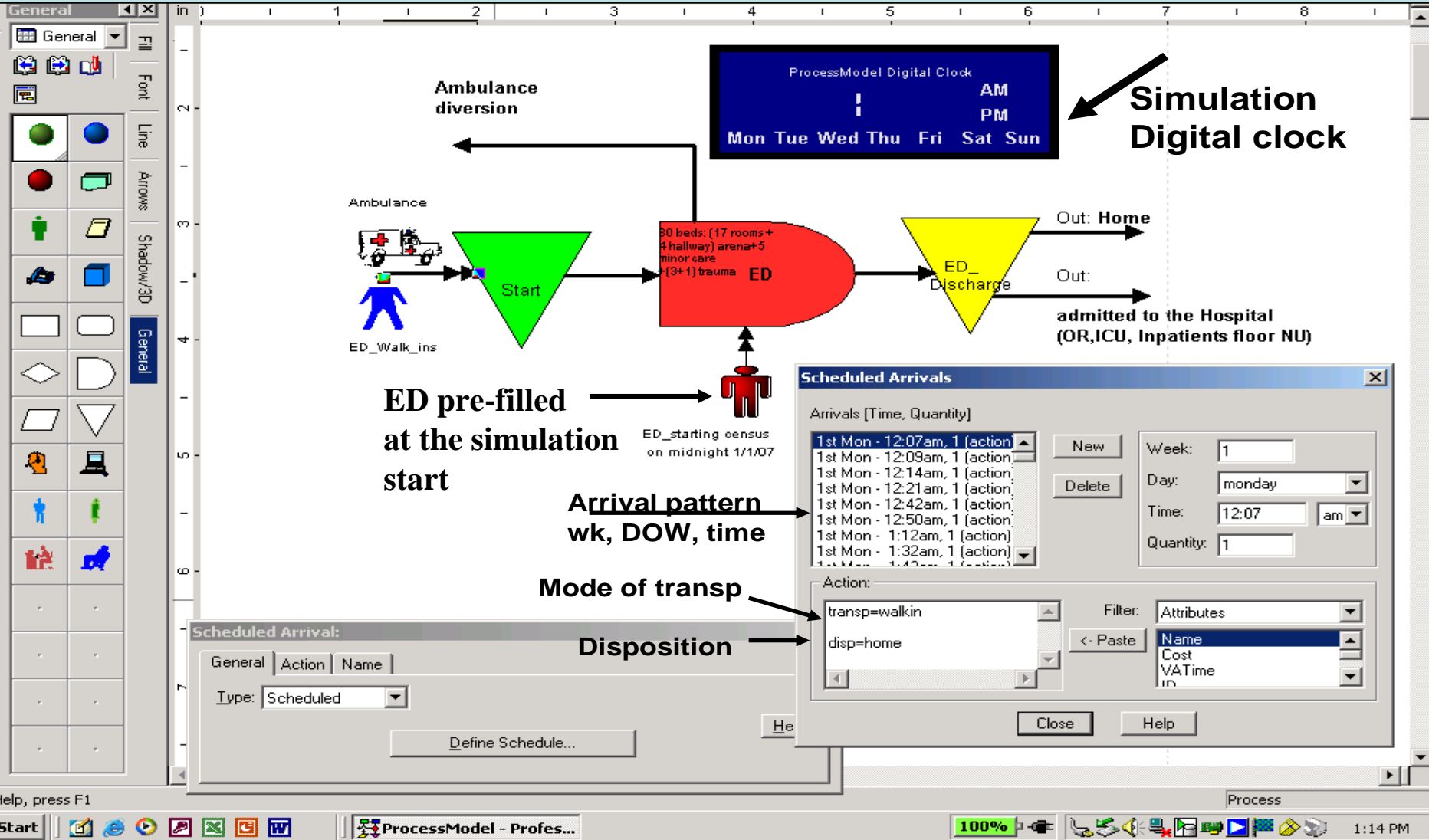
Therefore arrivals for all Mondays, all Tuesdays, and so on should not be combined



The high-level layout of the entire hospital system:

ED structure and in-patient units

ED simulation model layout



MODELING APPROACH

- ED diversion (closure) is declared when ED patients' census reaches ED beds capacity.
- ED stays in diversion **until** some beds become available when patients are moved out of ED (discharged home, expired, or admitted as in-patients).
- % ED diversion = % time ED is at full capacity
- upper LOS limits (simulation parameters) are imposed on the baseline original LOS distributions:
LOS higher than the limiting value is NOT allowed in the simulation run.

Take-away:

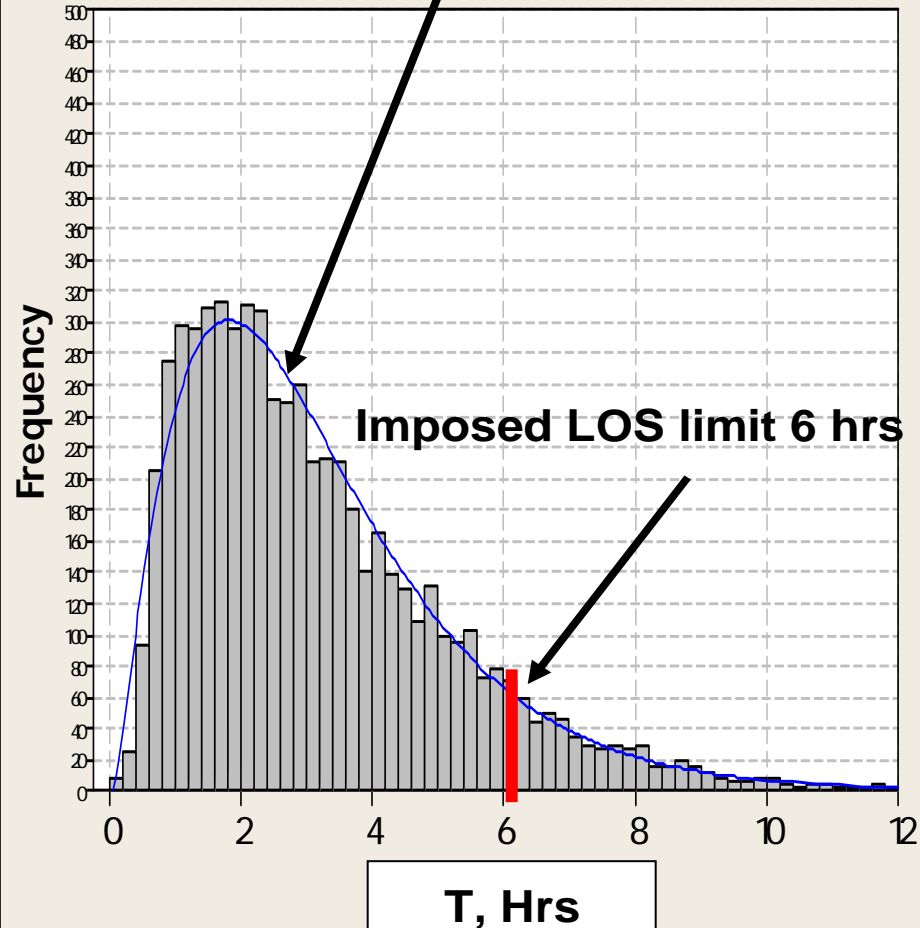
Baseline LOS distributions should be **recalculated** as functions of the upper LOS limits.

MODELING APPROACH (cont.)

Given original distribution density and the limiting value of the random variable T, what is the conditional distribution of the restricted random variable T ?

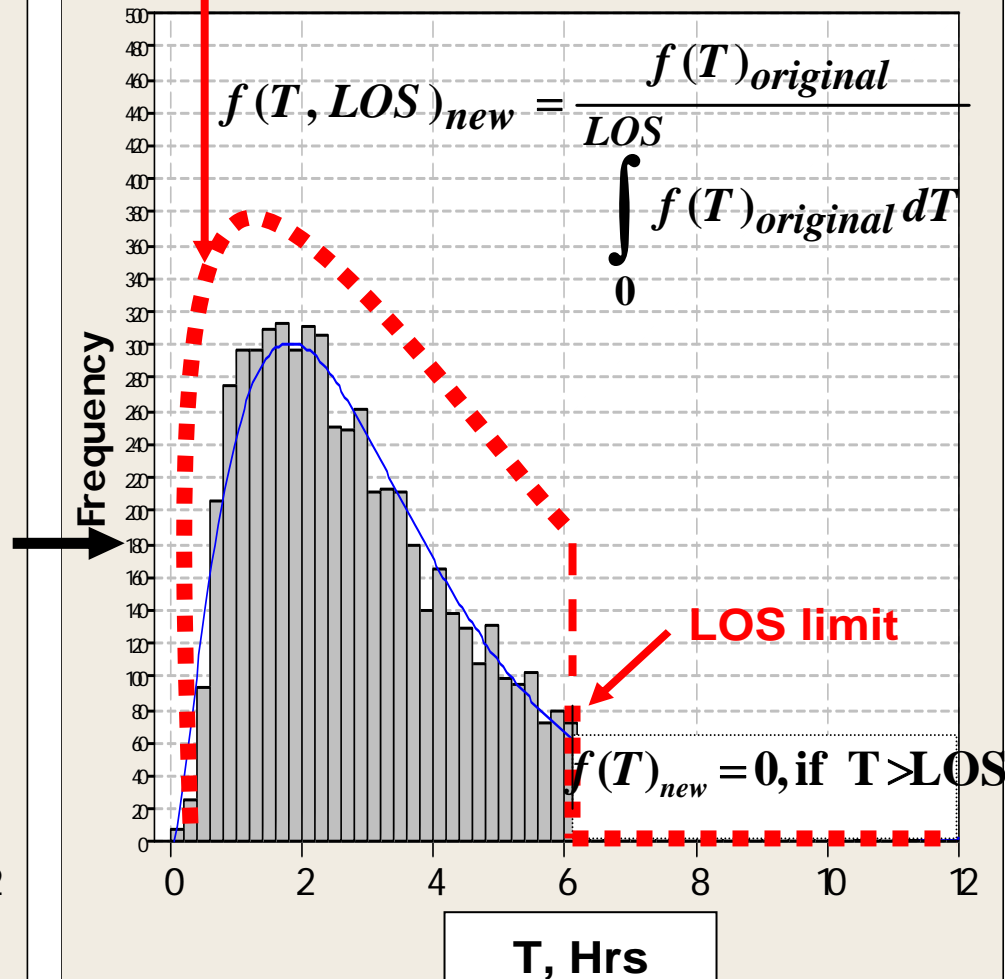
Original unbounded distribution

$f(T)_{orig}$

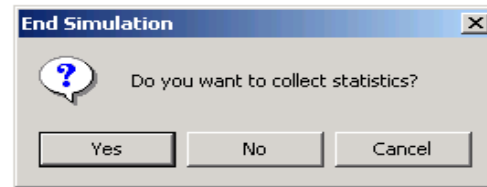
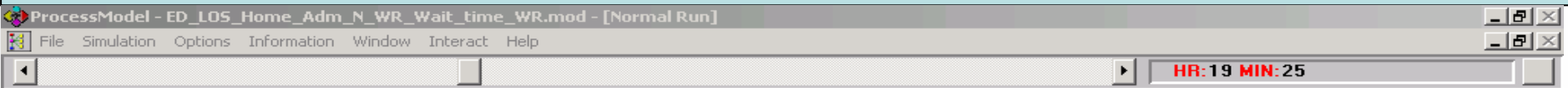


New re-calculated distribution

$$f(T, LOS)_{new} = \frac{f(T)_{original}}{\int_0^{LOS} f(T)_{original} dT}$$



ED Simulation Run Example



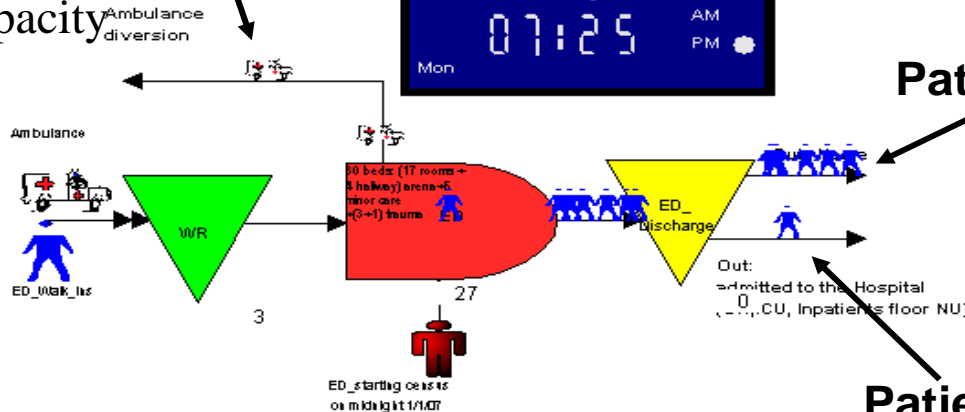
Diversion time & DOW



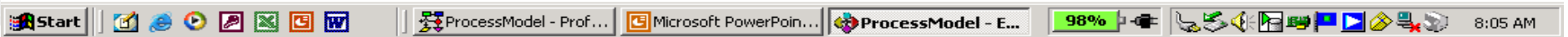
Patients discharged Home

Diverted Ambulance

bounced back when ED is at full capacity



Patient admitted to the Hospital



SIMULATION SUMMARY & MODEL VALIDATION

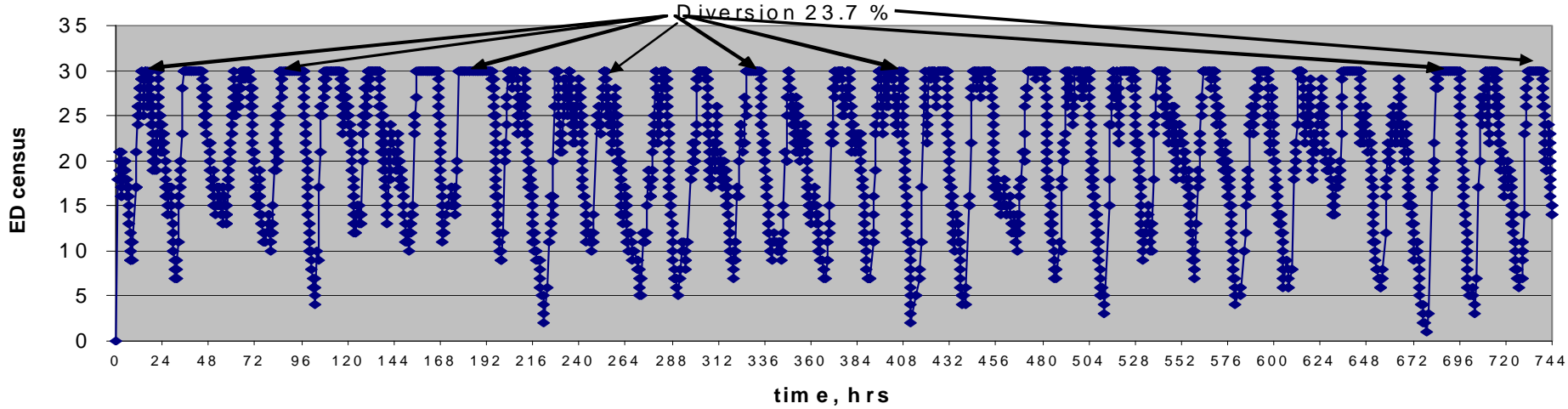
Scenario/option	LOS for discharged home NOT more than	LOS for admitted NOT more than	Predicted ED diversion, %	Note
Current, 07 (Baseline)	24 hrs	24 hrs	23.7%	Actual ED diversion was 21.5%
1	5 hrs Currently <u>17%</u> with LOS more than 5 hrs;	6 hrs Currently <u>24%</u> with LOS more than 6 hrs;	~ 0.5 %	Practically NO diversion
2	6 hrs	6 hrs	~ 2%	Low single digits diversion
3	5 hrs	24 hrs	~4%	Low single digits diversion

Take-away:

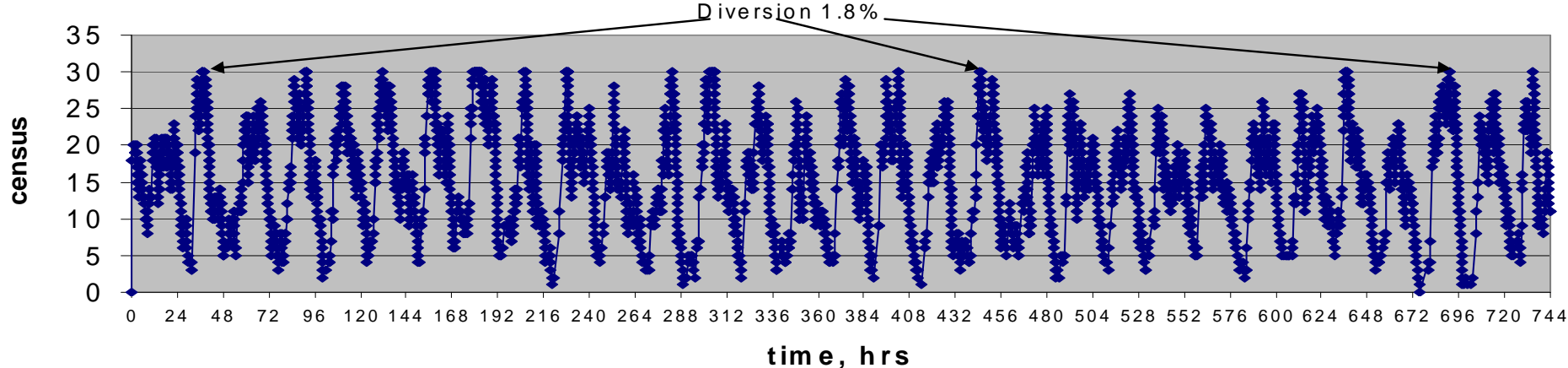
- ED diversion could be negligible (~0.5%) if patients discharged home stay **NOT more than 5 hrs** and admitted patients stay **NOT more than 6 hrs**.
- Relaxing of these LOS limits results in low digits % diversion that still could be acceptable

**Diversion is declared when ED census hits capacity limit.
The longer the census stays at capacity limit the higher is diversion %**

Baseline ED census, January 2007
LOS <= 24 hrs. Capacity 30 beds.



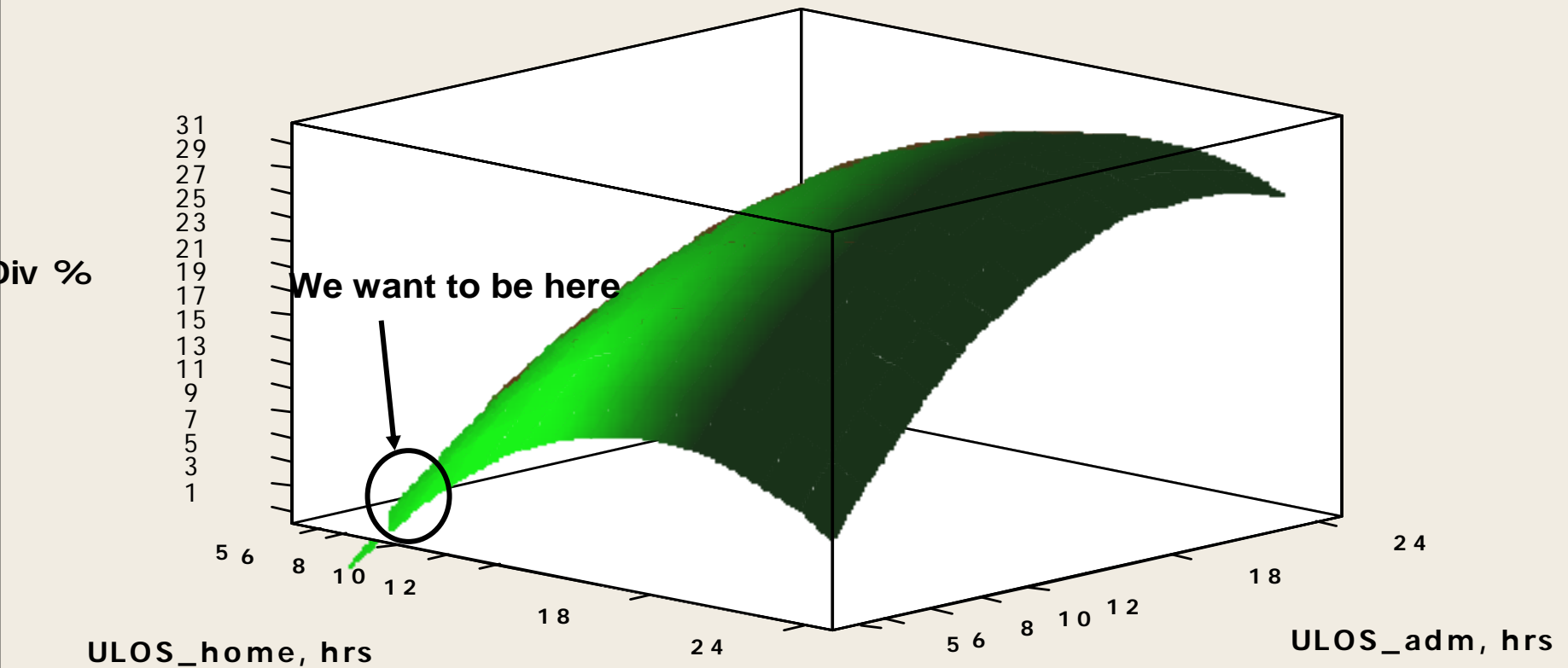
ED census:
LOS_home <= 6 hrs, LOS_adm <= 6 hrs. Capacity 30 beds.



What other combinations of upper limits LOS are possible to get low single digits % ED diversion ?

Performed full factorial DOE with two factors (ULOS_home and ULOS_adm) at 6 levels each using simulated % diversion as a response function.

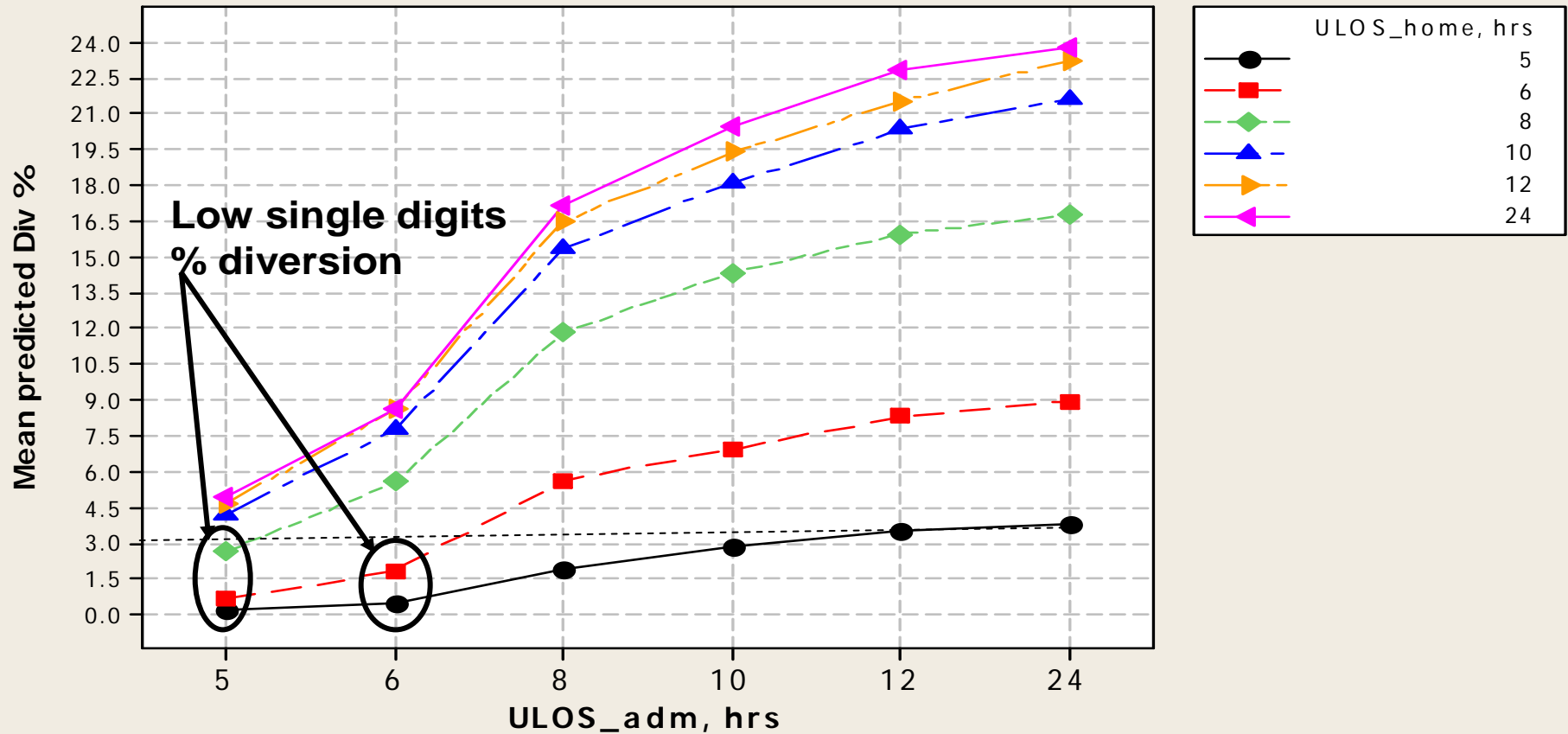
Surface Plot of Div % vs ULOS_adm, ULOS_home



What other combinations of upper limits LOS are possible to get low single digits % ED diversion ?

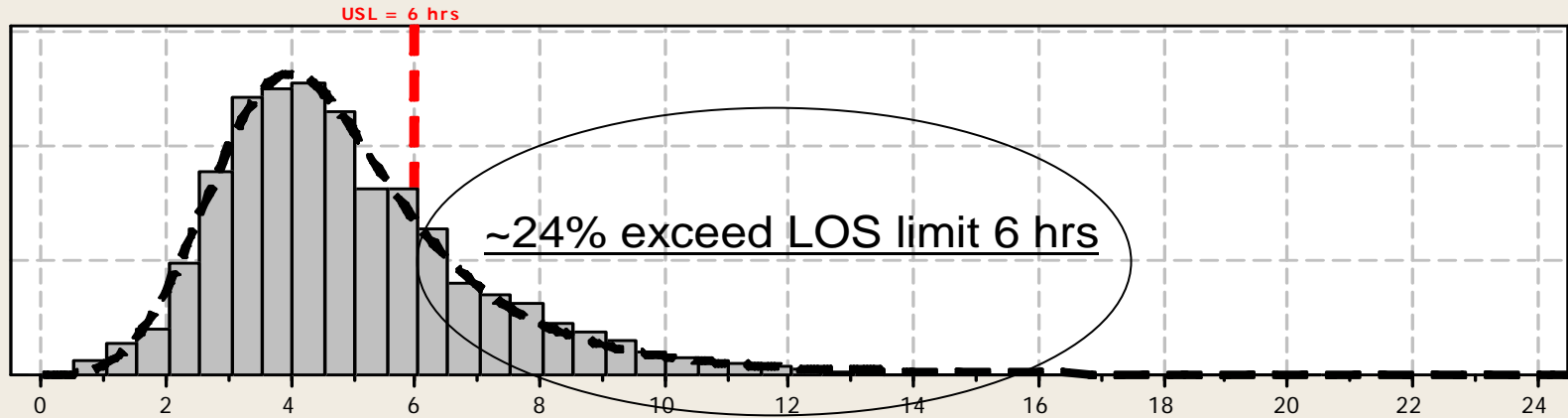
Performed full factorial DOE with two factors (ULOS_home and ULOS_adm) at 6 levels each using simulated % diversion as a response function.

Simulated Div % as a function of upper LOS limits, hrs

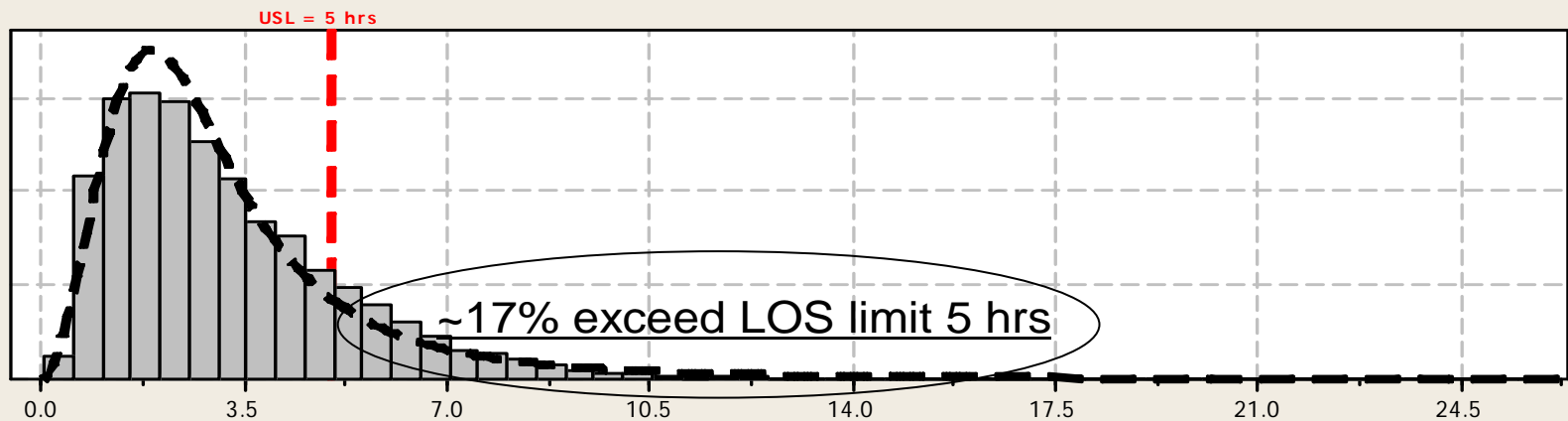


% of patients that stay longer than the upper LOS limit

LOS_adm Jan and Feb, hrs
 Calculations Based on Loglogistic Distribution Model



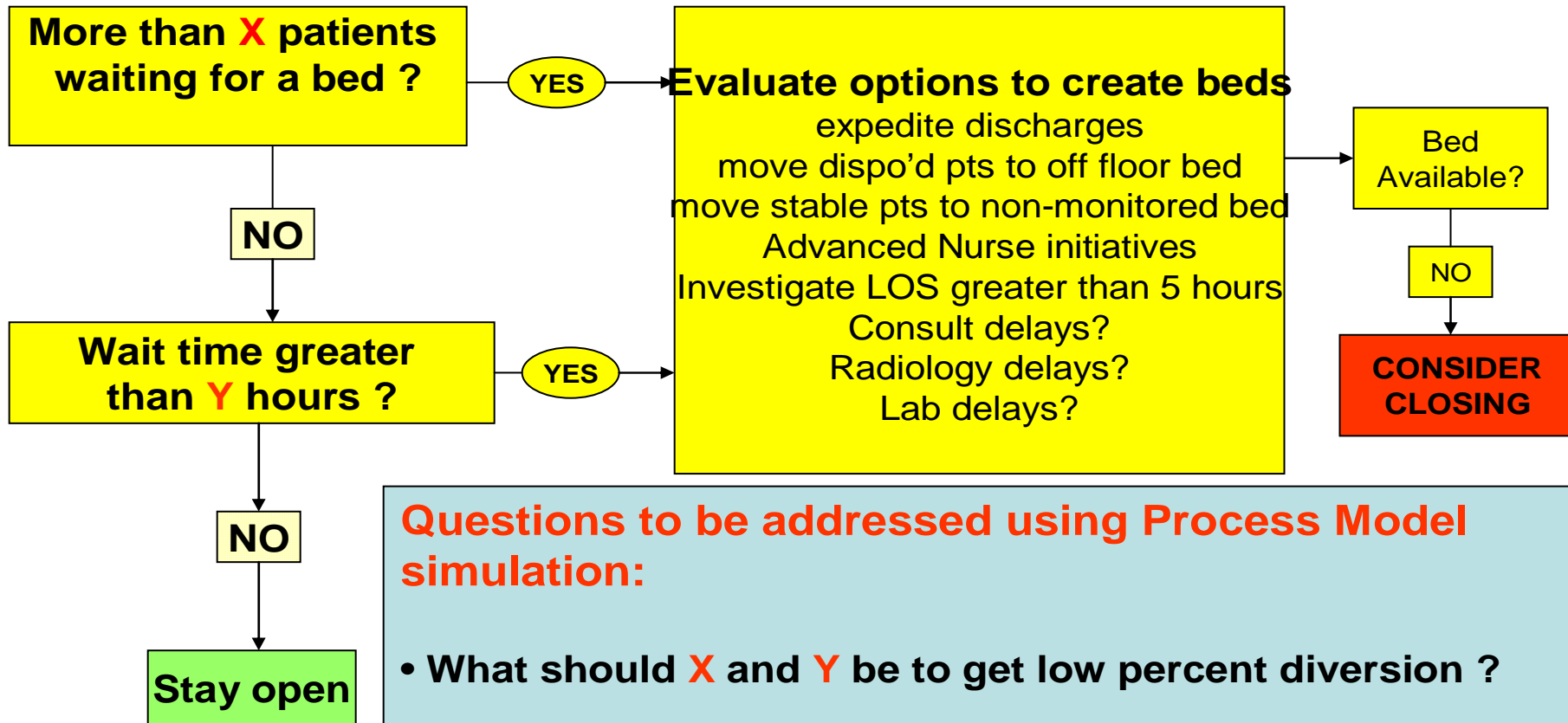
LOS_home Jan and Feb, hrs
 Calculations Based on Loglogistic Distribution Model



ALTERNATIVE CLOSURE CRITERIA

Objectives

1. Maintain Safe Patient Care
2. Decrease ED Diversions
3. Decrease Length of Stay
4. Decrease Left Not Seens



Questions to be addressed using Process Model simulation:

- What should **X** and **Y** be to get low percent diversion ?
- Are **X** and **Y** correlated ?

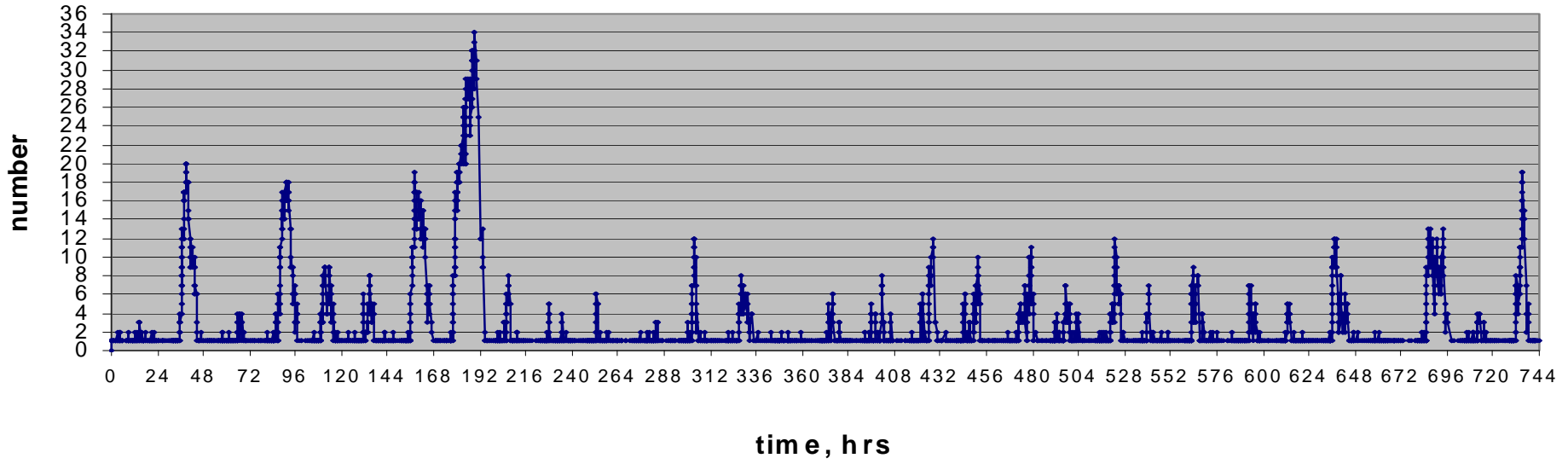
SIMULATION OF ALTERNATIVE CLOSURE CRITERIA

Scenario/option	LOS for discharged home NOT more than	LOS for admitted NOT more than	Predicted ED diversion, %	Max number of patients in Waiting Room	Waiting time for admitted patients in Waiting Room
Current state, (Baseline)	24 hrs	24 hrs	23.7%	31	3 hr 30 min
1	5 hrs	6 hrs	~ 0.5 %	7	35 min
2	6 hrs	6 hrs	~ 2%	10	43 min
3	5 hrs	24 hrs	~4%	12	1 hr

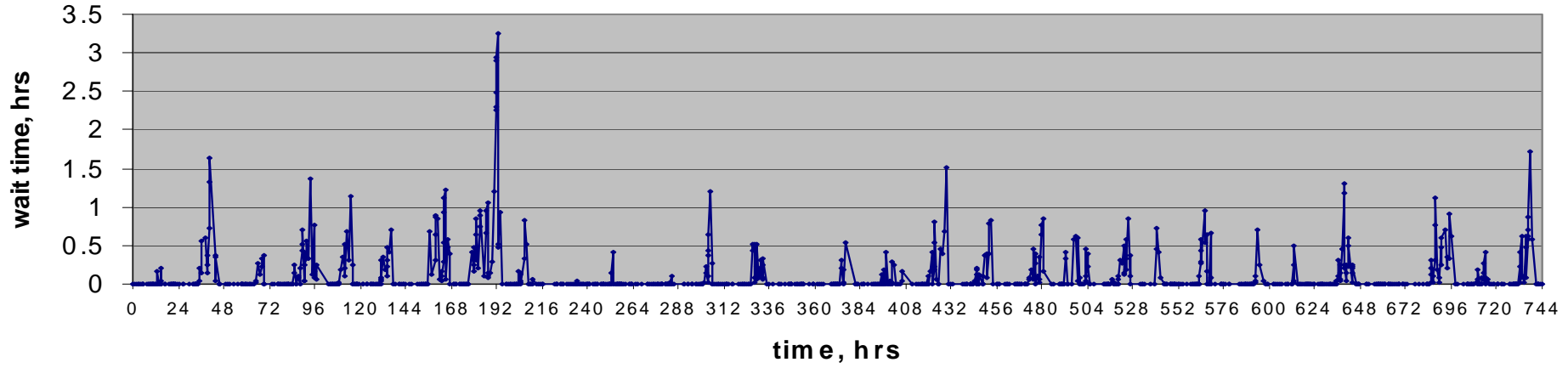
Take-away:

- The lower max number of patients in the waiting room and max waiting time the lower is ED diversion
- Locations of peaks of the max number of patients in Waiting Room is strongly correlated to locations of peaks of the max waiting time (see next slides)

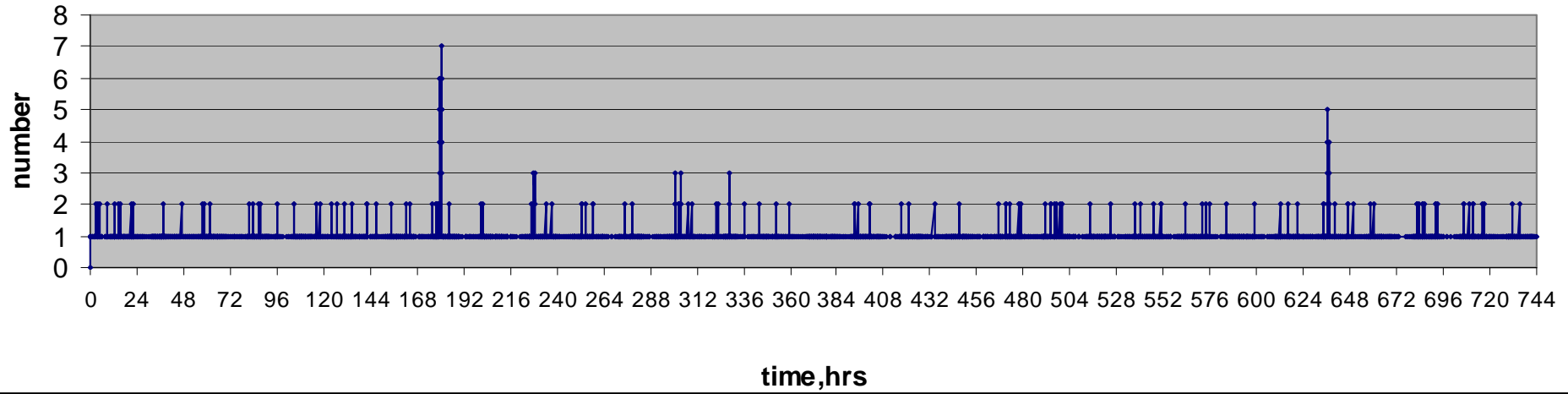
Number of patients in Waiting Room
ULOS_home=24 hrs, ULOS_adm=24 hrs



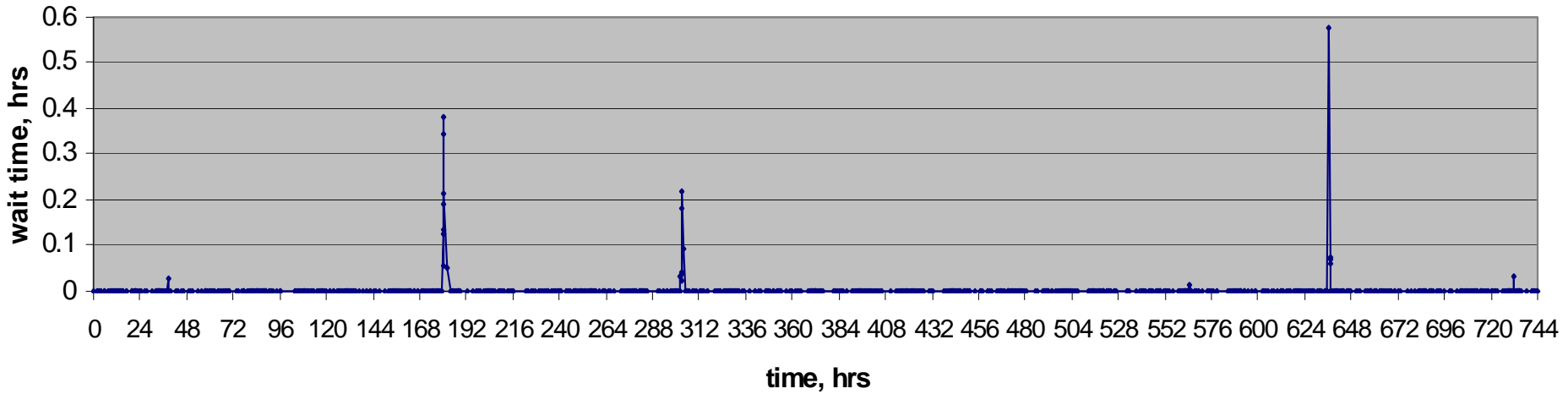
Wait_time_adm in Waiting Room, hrs
ULOS_home=24 hrs, ULOS_adm=24 hrs



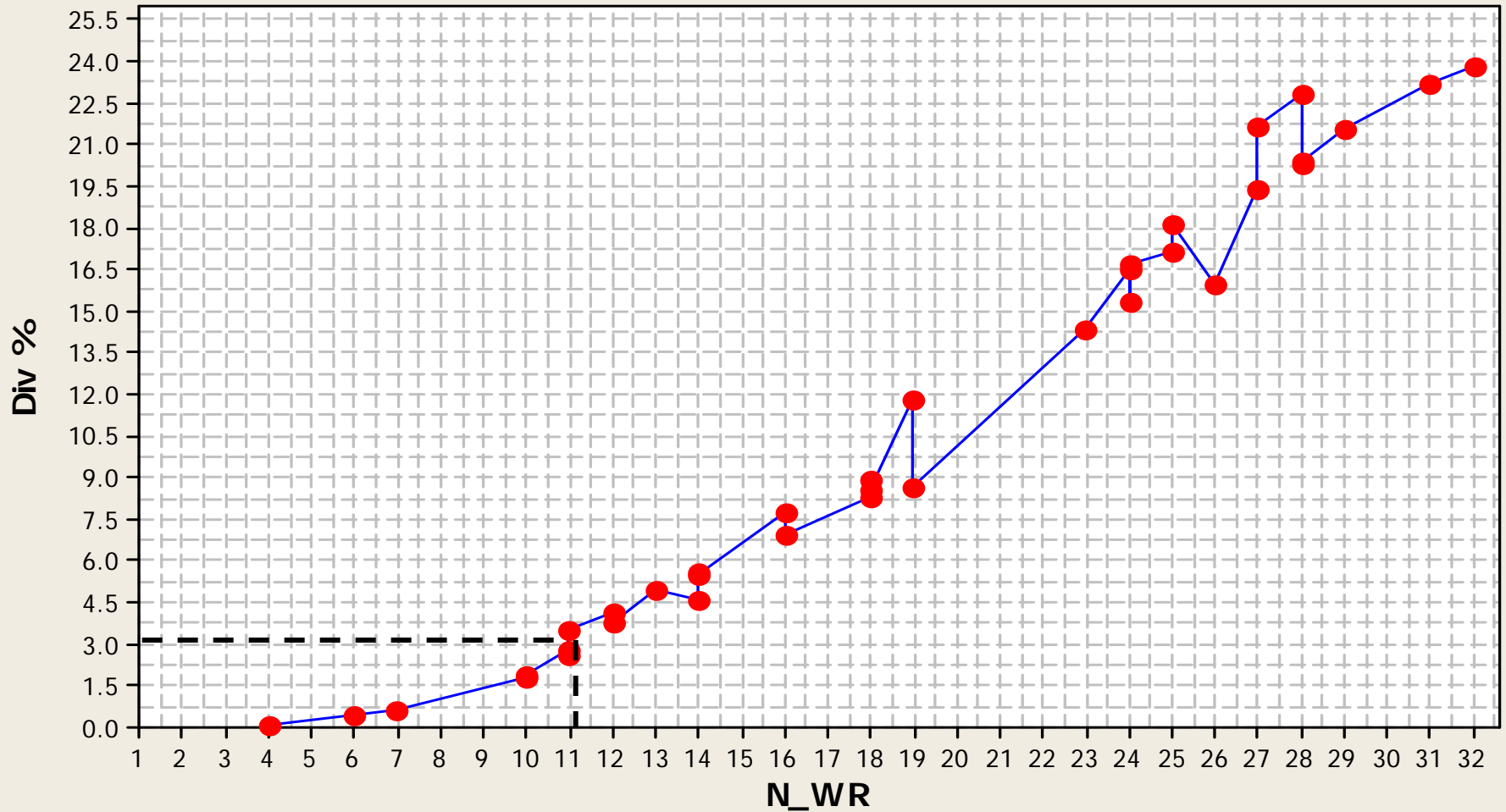
Number of patients in Waiting Room
ULOS_home=5 hrs, ULOS_adm=6 hrs



wait_time_adm in Waiting Room, hrs
ULOS_home=5 hrs, ULOS_adm=6 hrs

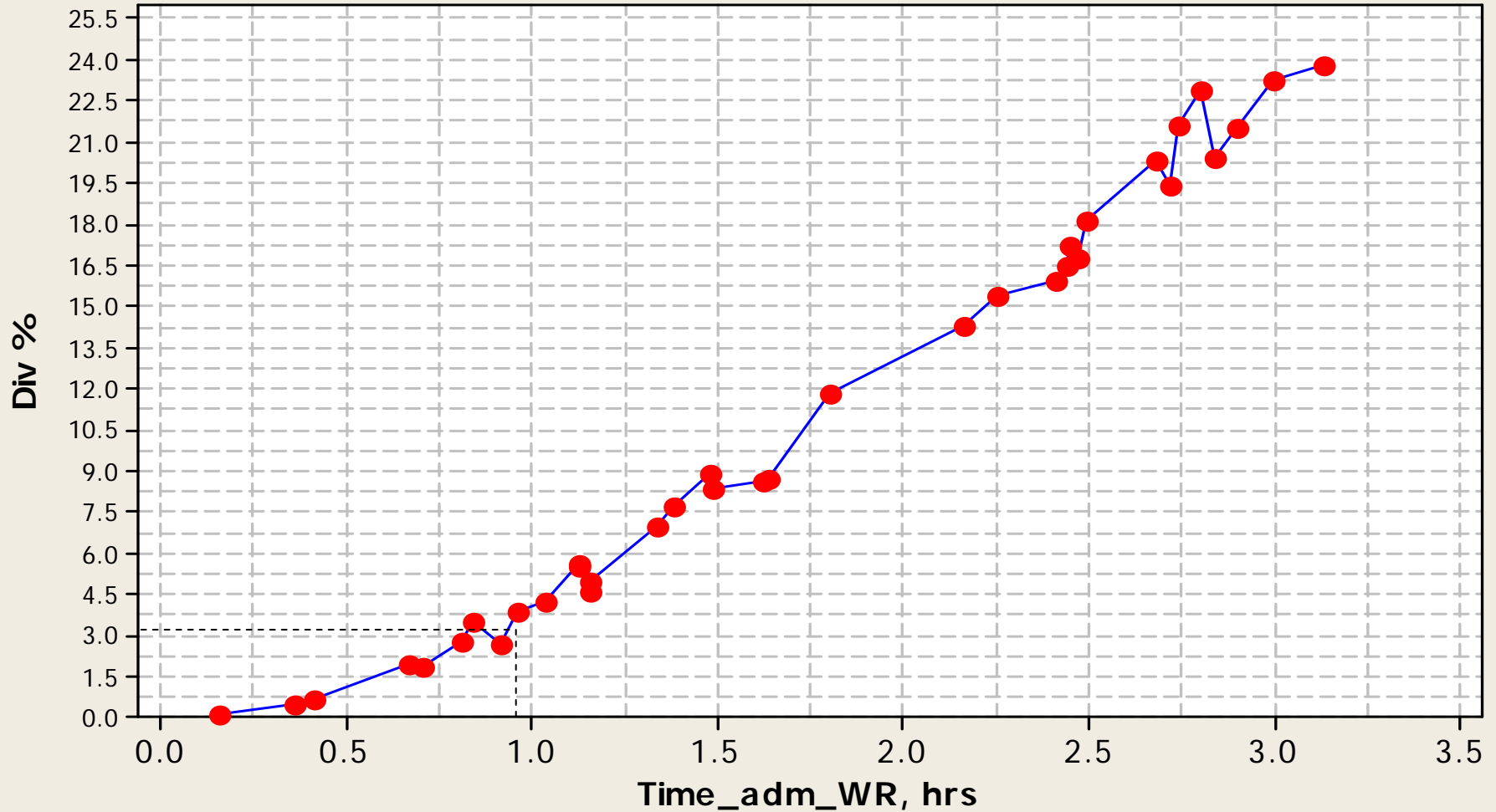


Scatterplot of Div % vs N_WR



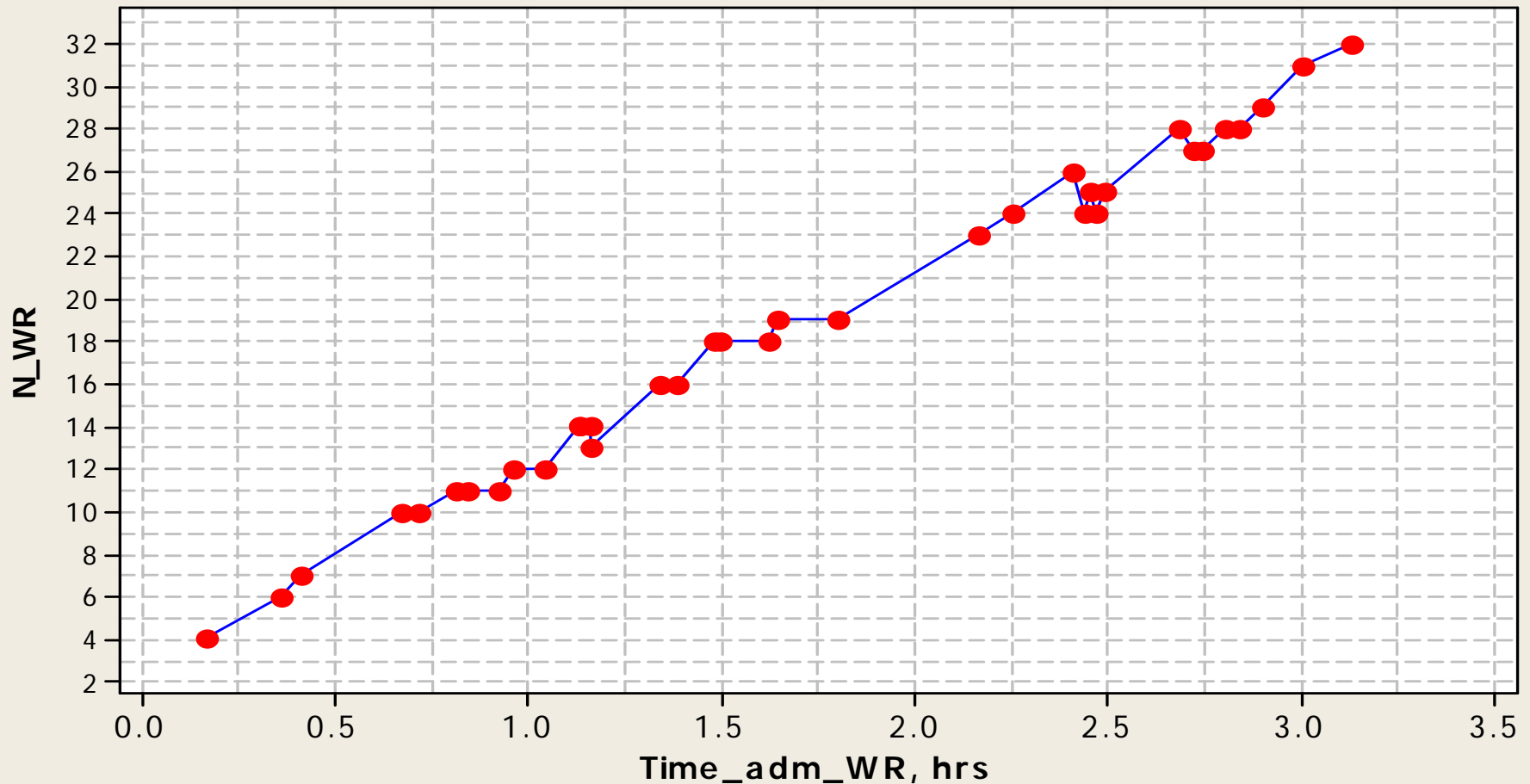
Take-away:
The number of patients in waiting room 11 or less corresponds to single digits diversion less than 3%

Scatterplot of Div % vs Time_adm_WR, hrs



Take-away:
Waiting time for admitted patients in waiting room 1 hr or less corresponds to single digits diversion less than 3%

Scatterplot of N_WR vs Time_adm_WR, hrs



Take-away:

- The number of waiting patients and the waiting time for admitted patients is strongly correlated to each other. Pearson linear correlation coefficient is 0.996 !!.
- Strong correlation indicates that either one or another criteria should be enough, **not both**.

Conclusions

- ED diversion could likely be negligible (less than 1 %) if patients discharged home stay **NOT more than 5 hrs** and admitted patients stay **NOT more than 6 hrs**.

Currently:

-17% of patients discharged home stay above this limit up to 24 hrs;

-24 % of admitted patients stay above this limit up to 20 hrs.

This long LOS for large % of patients results in ED closure/diversion

- Some relaxing of these LOS limits will result in low single digits % ED diversion that still could be acceptable
- Other combinations of LOS upper limits that result in low single digits % diversion have been determined using full factorial DOE with two factors.
- An alternative diversion criteria could be used: the number of patients in waiting room. The number of patients **11 or less** corresponds to single digits diversion, less than ~3%

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What did we learn about simulation methodology?

- Patient Throughput flow is an example of the general Dynamic Supply & Demand problem .
Dynamic means that the system's behavior depends on time (not a one-time snapshot)
- There are three basic components that should be accounted for in this type of problems:
 - The number of patients (or, generally, any items) entering the system at any point of time
 - The number of patients (any items) leaving the system at any point of time
 - Limited Capacity of the system which limits the flow of patients through the system
- All three components affect the flow of patients that the system can handle.
- A lack of the proper balance between these components results in the system's over-flow and closure/diversion
- Process Model Simulation methodology provides the only means of analyzing and managing the proper balance

APPENDIX

WHAT IS THE PROCESS MODEL ?

- It is a computer model that mimics the dynamic behavior of a real process over the time in order to visualize and quantitatively analyze its performance in terms of:
 - Cycle times
 - Throughput capacity
 - Resources utilization
 - Activities utilization
- It is a tool to perform ‘WHAT-IF’ analysis and play different scenarios of the model behavior as conditions and process parameters change.

This allows to make experiments on the computer model, and test different solutions (changes) for their effectiveness before going to the floor for the actual implementation.

WHAT ARE THE BASIC ELEMENTS OF THE PROCESS MODEL?

- **Flow chart of the process:** Diagram that depicts logical flow of a process from its inception to its completion
- **Entities:** Items to be processed: patients, documents, customers, etc.
- **Activities:** Tasks performed on entities: medical procedures, document approval, customer check out, etc
- **Resources:** Agents used to perform activities and move entities: service personnel, operators, equipment, nurses, physicians.
- **Connections:**
- **Entity arrivals:** Define process entry points, time, and quantities of the entities that enter the system to begin processing
- **Entity routings:** Define directions and logical conditions flow for entities

WHAT INFORMATION (DATA) IS REQUIRED TO FEED THE MODEL ?

- **Entities quantities and arrival time:** periodic, random, scheduled, daily pattern, etc
- **The time that the entities spend in the activities.** This is usually not a fixed time but a statistical distribution. The wider the time distribution the higher the variability of the system behavior.
- **The capacity of each activity,** i.e. the max number of entities that can be processed concurrently in the activity.
- **The size of input and output queues** for the activities
- **The routing type or the logical conditions** to take a specified routing.
- **Resource Assignments:** their number and availability, and/or resources shift schedule