



University of California  
San Francisco



# Precise Patient Monitoring with Few False Alarms - SuperAlarm

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# Disclosures

None

13ICU|16  
XXX,XXX

ECG  
PVC 4

SAVING

RR - II LD II FAIL

NBP 109 / 57 (78)

11:04

SPO2 96% \*\*

ART 1

0

30

CVP 2

0

SPO2

RR II

109/57  
RATE 93 68  
86

\*\*\*  
RATE 94 100

21

CVP 2  
30  
-2

SPO2  
105  
90

RR  
30  
5  
21%

MORE  
MENUS

34.3

T1C

SENSOR

T2C

TP 1  
40.0  
32.0  
37

ESP W 30  
TACHY  
W + 2  
TACHY

10:24  
10:24  
10:19  
10:13

ALARM

# ● Clinical motivations

OPEN ACCESS Freely available online

PLOS ONE

## Insights into the Problem of Alarm Fatigue with Physiologic Monitor Devices: A Comprehensive Observational Study of Consecutive Intensive Care Unit Patients

Barbara J. Drew<sup>1\*</sup>, Patricia Harris<sup>1</sup>, Jessica K. Zègre-Hemsey<sup>2</sup>, Tina Mammone<sup>3</sup>, Daniel Schindler<sup>1</sup>, Rebeca Salas-Boni<sup>1</sup>, Yong Bai<sup>1</sup>, Adelita Tinoco<sup>1</sup>, Quan Ding<sup>1</sup>, Xiao Hu<sup>1</sup>





# 2,500,000+

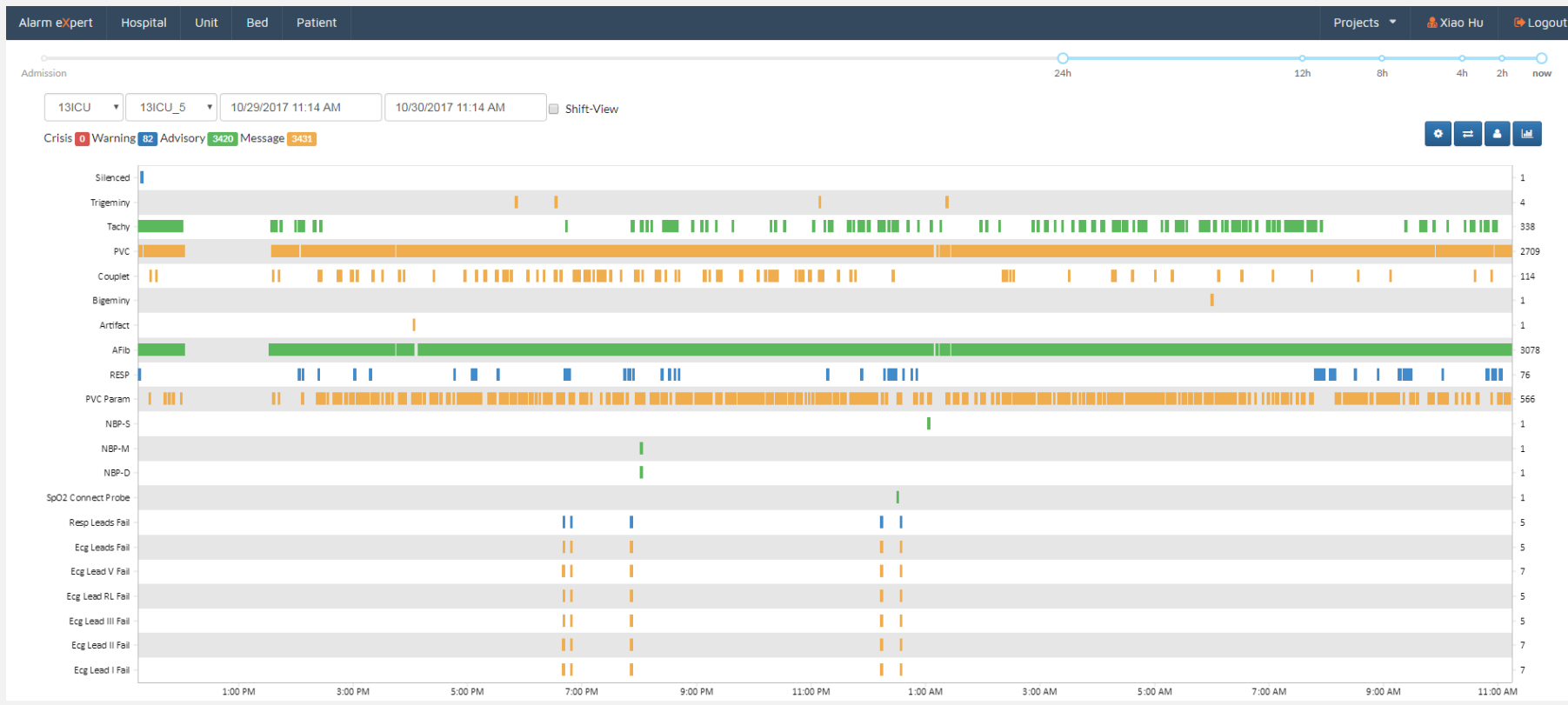
Bedside patient monitor alarms in March 2013 for only 77  
UCSF ICU beds



1 alarm every  
90 seconds



# Alarm Situation at UCSF has not Changed Much!



# Alarm Crisis



**Little boys who cried wolf**



**Desensitized Staff**

- **375,000**/year unplanned ICU transfers
- **> 200,000** cardiac arrests with **80%** mortality
- Many potentially predictable events:
  - Sepsis
  - Bleeding
  - Urgent intubation

❖ 2015 – 2017 national patient safety goal by Joint Commission:  
**Reduce the harm associated with clinical alarm systems**

❖ 2013 – 2016 top 10 health technology hazards by ECRI  
#1 hazard in 2013, 2014, 2015: **Alarm hazard**  
#2 2016 hazard: **Missed alarms can have fatal consequences**

# ● Current Interventions are Inadequate for Alarm Crisis

Widening  
alarm  
parameters



Frequently  
changing  
ECG  
electrodes



Instituting  
alarm delays



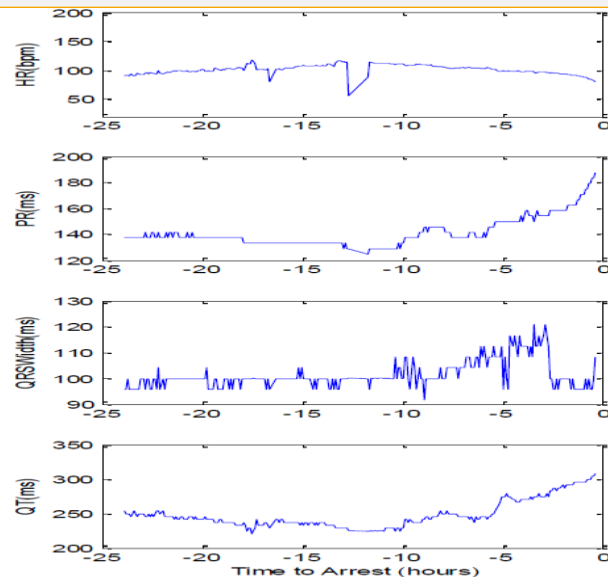
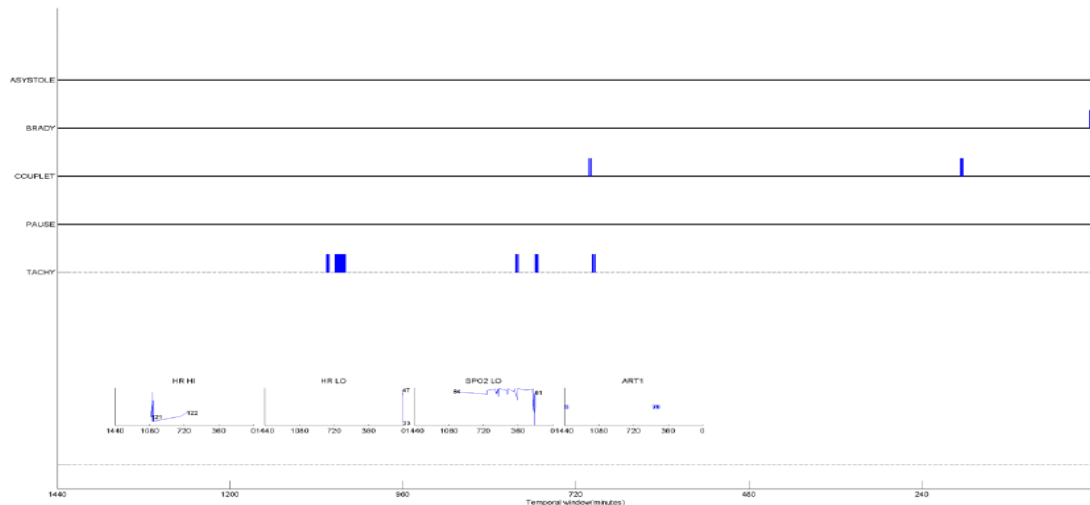
Using  
disposable  
ECG wires







# Alarm Failure: Too Few Right Alarms



## 24-Hour Monitor Alarm Profile

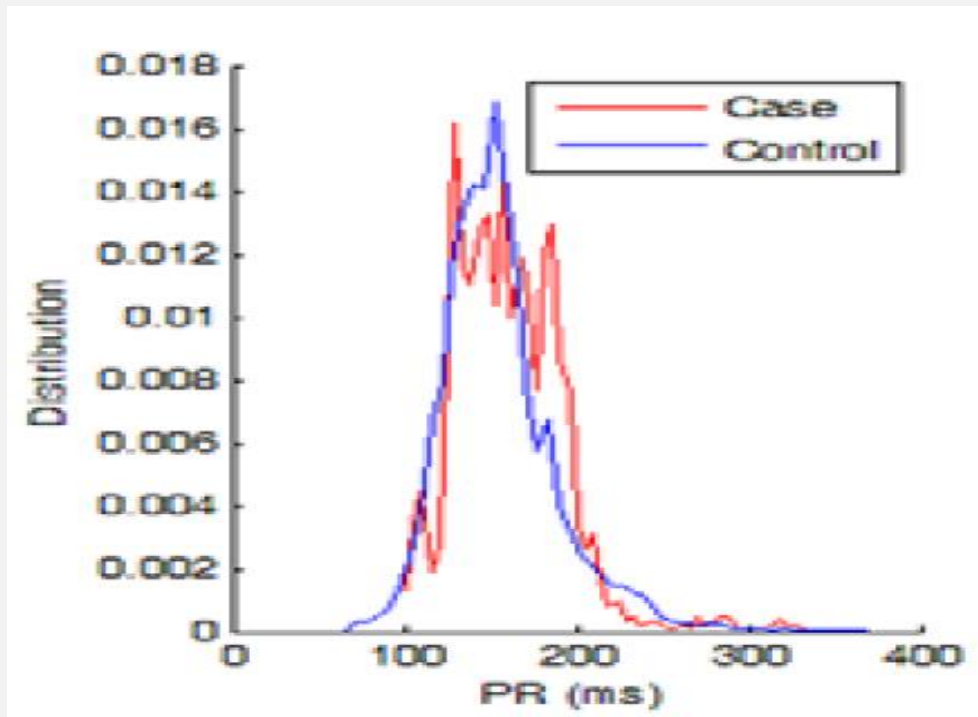
## Trends of ECG Metrics not Monitored

- Potentially useful metrics not monitored
- Failed to consider patterns such as trending

# Design Precise Alarm: Trending

Why trending?

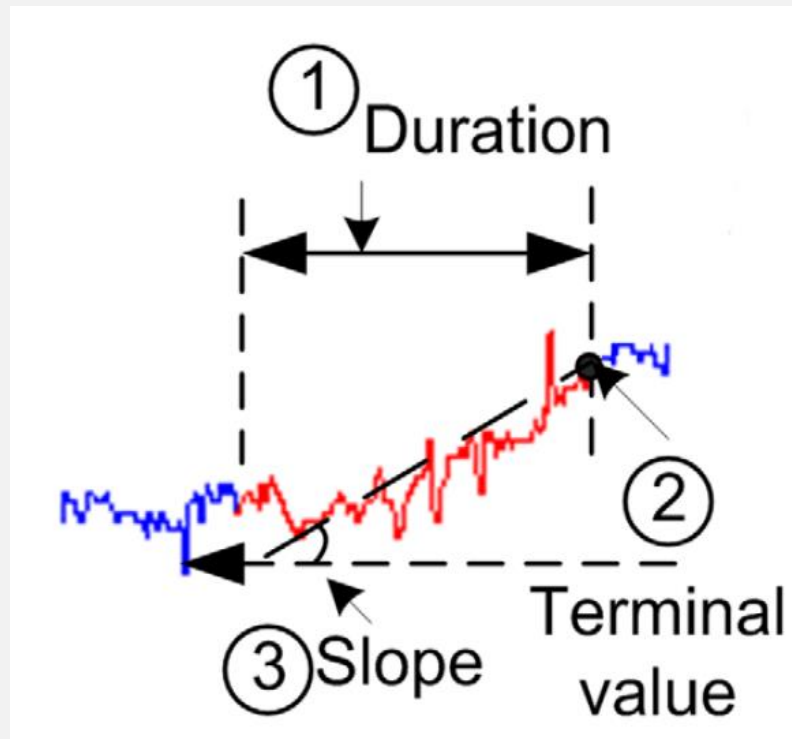
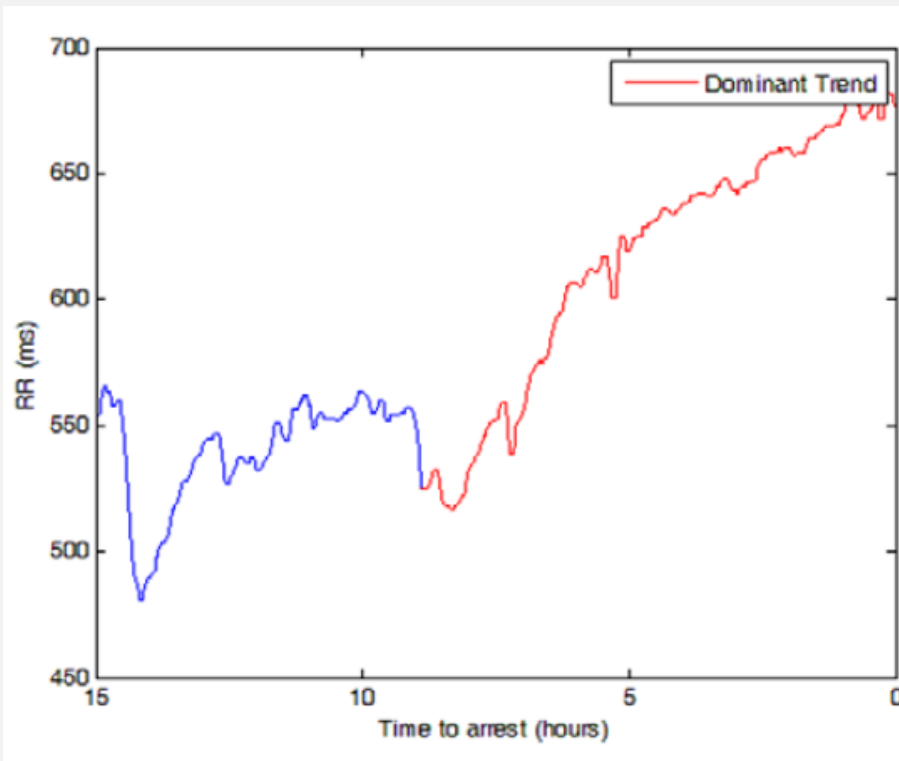
- Monotonic changes are not physiological
- Single threshold crossing alarm is not adequate to differentiate complex physiological states



Ding Q, Bai Y, Tinoco A, Mortara D, Do D, Boyle NG, Pelter MM, **Hu X**. Developing new predictive alarms based on ECG metrics for bradyasystolic cardiac arrest. *Physiol Meas*. 2015 Dec; 36(12):2405-22. PMID: 26502065; PMCID: PMC4838570



# Design Precise Alarm: Dominant Trend & Alarms



Ding Q, Bai Y, Tinoco A, Mortara D, Do D, Boyle NG, Pelter MM, **Hu X**. Developing new predictive alarms based on ECG metrics for bradyasystolic cardiac arrest. *Physiol Meas*. 2015 Dec; 36(12):2405-22. PMID: 26502065; PMCID: PMC4838570

# Design Precise Alarm: Case-Control Alarm Screening

- Case patients vs Control patients
- Examine the following factors
  - Which variable?
  - Which kind of alarm: duration/value/slope
  - What is the threshold?
- Criteria: FPR↓ and TPR↑



## Design Precise Alarm: a proof of concept study

- 27 patients with brady-asystole arrests
- 304 age/gender/dx matched control patients
- 17 ECG metrics
  - Heart rate
  - ECG morphological metrics
    - PR interval, p wave duration, QRS width
  - ST (J60)
  - HRV (SDNN, LF, HF...)

Ding Q, Bai Y, Tinoco A, Mortara D, Do D, Boyle NG, Pelter MM, **Hu X**. Developing new predictive alarms based on ECG metrics for bradyasystolic cardiac arrest. *Physiol Meas*. 2015 Dec; 36(12):2405-22. PMID: 26502065; PMCID: PMC4838570

		arrests	Control
Number of patients		27	304
Male (%)		18 (66.7%)	220 (72.4%)
Age (mean $\pm$ SD)		61.4 $\pm$ 20.7	63.2 $\pm$ 11.8
Total duration of ECG analyzed (h)		19.4 $\pm$ 7.0	22.8 $\pm$ 4.1
Causes of arrest (%)	Multi-organ failure	12 (44.4%)	Not applicable
	Respiratory failure	7 (25.9%)	
	Cardiac failure	1 (3.7%)	
	Drug-induced	1 (3.7%)	
	Vagally mediated asystolic event, occurring during the turning of a patient	1 (3.7%)	
	Unknown by documentation	5 (18.5%)	
Arrest subtypes (%)	Sinus arrest	10 (37.0%)	Not applicable
	Complete heart block	7 (25.9%)	
	High-degree AV block	4 (14.8%)	

	TPR	Threshold (h)	Lead time (h)
PR	<b>11.1%</b>	<b>4.4</b>	<b>5.0 ± 8.3</b>
Pdur	0.0%	4.1	NA
QRSdur	<b>11.1%</b>	<b>4.1</b>	<b>6.7 ± 6.6</b>
RR	<b>11.1%</b>	<b>6.5</b>	<b>3.6 ± 4.8</b>
QT	0.0%	5.6	NA
SerumK2	7.4%	4.2	4.6 ± 5.3
T Complex	0.0%	4.3	NA
ST I	<b>14.8%</b>	<b>4.2</b>	<b>6.1 ± 5.2</b>
ST II	0.0%	5.3	NA
ST V	3.7%	5.2	0.4 ± 0.0
SDNN	<b>11.1%</b>	<b>4.7</b>	<b>6.3 ± 6.0</b>

- Duration alarms performed much better across different ECG metrics
- Trend duration of PR interval, QRS width, HR, ST-I, and SDNN are top performers at a FPR = 0
- A sensitivity of 33.3% with zero FPR was achieved by combining four metrics
  - HR, QRS width, ST I and estimated K



# First Study



Contents lists available at [SciVerse ScienceDirect](#)

## Journal of Biomedical Informatics

journal homepage: [www.elsevier.com/locate/yjbin](http://www.elsevier.com/locate/yjbin)



### Predictive combinations of monitor alarms preceding in-hospital code blue events

Xiao Hu<sup>a,b,c,\*</sup>, Monica Sapo<sup>c</sup>, Val Nenov<sup>c</sup>, Tod Barry<sup>c</sup>, Sunghan Kim<sup>a</sup>, Duc H. Do<sup>d</sup>, Noel Boyle<sup>d</sup>, Neil Martin<sup>c</sup>

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<sup>d</sup>UCLA Cardiac Arrhythmia Center, David Geffen School of Medicine, University of California, Los Angeles, United States

\*Quality Management Service, UCLA Ronald Reagan University Medical Center, Los Angeles, United States



## ● Introducing SuperAlarm patterns...

...”Combinations of alarms that co-occur *frequently in a time window preceding code blue calls but co-occur **rarely** among control patients*”

- Key algorithm elements

- Discretization of parametric alarm values

Sys ABP Hi 160 > 130 mmHg  $\neq$  Sys ABP Hi 200 > 130 mmHg

- Using *Apriori* algorithm to find frequent combinations of alarms preceding code blue events



## Second Study



Contents lists available at [ScienceDirect](#)

### Journal of Biomedical Informatics

journal homepage: [www.elsevier.com/locate/yjbin](http://www.elsevier.com/locate/yjbin)



## Integrating monitor alarms with laboratory test results to enhance patient deterioration prediction



Yong Bai<sup>a</sup>, Duc H. Do<sup>b</sup>, Patricia Rae Eileen Harris<sup>c</sup>, Daniel Schindler<sup>c</sup>, Noel G. Boyle<sup>b</sup>, Barbara J. Drew<sup>c</sup>, Xiao Hu<sup>c,d,e,f,\*</sup>

<sup>a</sup>Department of Bioengineering, University of California, Los Angeles, CA, United States

<sup>b</sup>UCLA Cardiac Arrhythmia Center, David Geffen School of Medicine, University of California, Los Angeles, CA, United States

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<sup>f</sup>UCB/UCSF Graduate Group in Bioengineering, University of California, San Francisco, CA, United States

# Extending the definition of SuperAlarm patterns

...“**Combination of alarms and abnormal laboratory test results**”

- **Key algorithm elements**

- **Two approaches to representing abnormal lab test results as events:**

- ✓ **Abnormal lab: most recent results with flag [H, HH, L, and LL]**



- ✓ **Delta lab: abnormality flag combinations of two results (e.g.,  $N \rightarrow H$ )**

- **Use of a more efficient algorithm for mining maximal frequent alarms**

# ● Useful Properties

## Filter individual false alarms

V TACH &  $19.5 < \text{BP MEAN LO} < 52.5$  &  $20.5 < \text{BP DIA LO} < 37.5$  &  $27.5 < \text{BP SYS LO} < 75.5$

Ventricular Tachycardia alarm unlikely false when co-occurring with Low blood pressure Alarms!

## Capture trend of physiologic variables

$\text{SpO}_2 \text{ LO} < 57.5$  &  $57.5 < \text{SpO}_2 \text{ LO} < 82.5$  &  $82.5 < \text{SpO}_2 \text{ LO} < 93.5$

A likely SpO2 trending pattern!

## Inherently multivariate

$57.5 < \text{SPO}_2 \text{ LO} < 82.5$  & BNP H→H & PT H→H & Hematocrit L→L & Hemoglobin L→L

## ● Third Study

Proceedings — AMIA Joint Summits  
on Translational Science



[AMIA Jt Summits Transl Sci Proc.](#) 2015; 2015: 162–167.

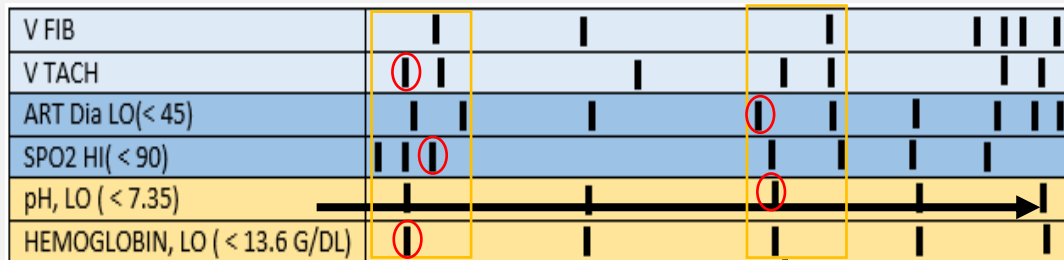
PMCID: PMC4525278

Published online 2015 Mar 25.

### **Cumulative Time Series Representation for Code Blue prediction in the Intensive Care Unit.**

[Rebeca Salas-Boni](#), PhD,<sup>1</sup> [Yong Bai](#), MS,<sup>2</sup> and [Xiao Hu](#), PhD<sup>1,3,4,5</sup>

# Event Sequence Representation of Heterogeneous Temporal Clinical Data

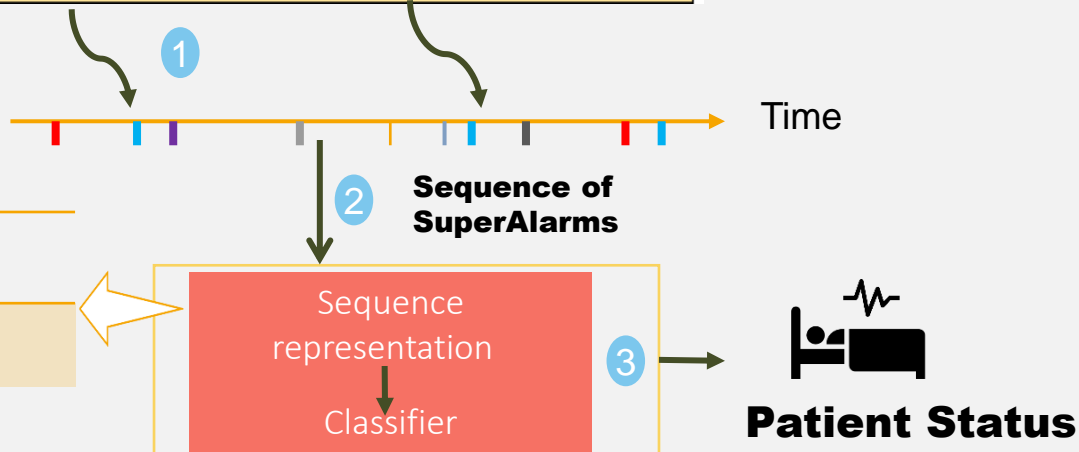


- Key algorithm elements

A novel sequence representation technique



SuperAlarm Sequence	Fixed-dimensional feature space
$\{a, b, c, c, e, f\}$	[0.1 0.5 0.9 0.8]
$\{e, f, g, c\}$	[0.2 0.05 0.1 0.2]

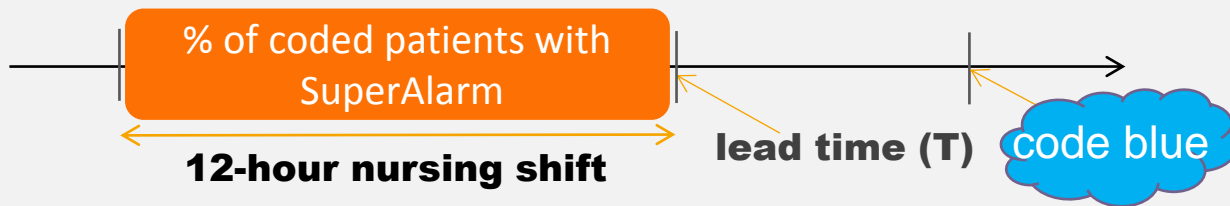


# ● Performance Metrics

- Alarm Frequency Reduction Rate (AFRR)

$$1 - \frac{[\text{hourly \# of SuperAlarms}]}{[\text{hourly \# of Monitor Alarms}]}$$

- Sensitivity<sup>L</sup>@T



- Work-up to Detection Ratio (WDR)

$$\frac{[\text{patients with SuperAlarm triggers}]}{[\text{coded patients with SuperAlarm triggers}]}$$

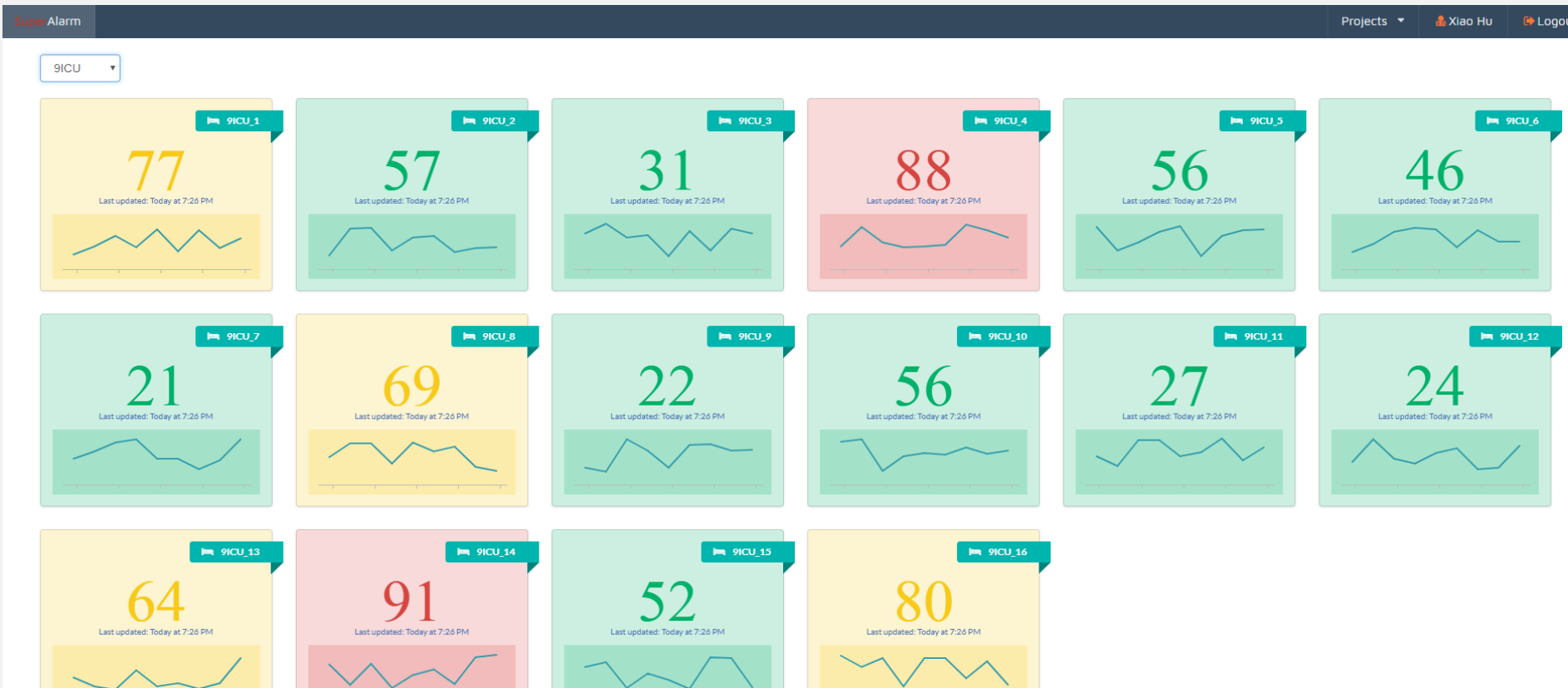
# Performance Results

Based on SuperAlarm Dataset 1.0

	Sensitivity at Different Lead Time					Alarm Frequency Reduction Rate	Work-up to Detection Ratio
	0.5 Hour	1.0 Hour	2.0 Hour	6.0 Hour	12.0 Hour		
<b>Paper #1</b>	86.7 %	86.7 %	80.0 %	73.3 %	60.0 %	85.3±10.5%	7.8±0.3
<b>Paper #2</b>	93.3 %	90.0 %	90.0 %	86.7 %	80.0 %	85.3±9.8%	6.5±0.2
<b>Paper #3</b>	90.0 %	90.0 %	90.0 %	86.7 %	70.0 %	88.5±21.4%	4.8±0.1



# ICU Dashboard of Patient Scores – by SuperAlarm





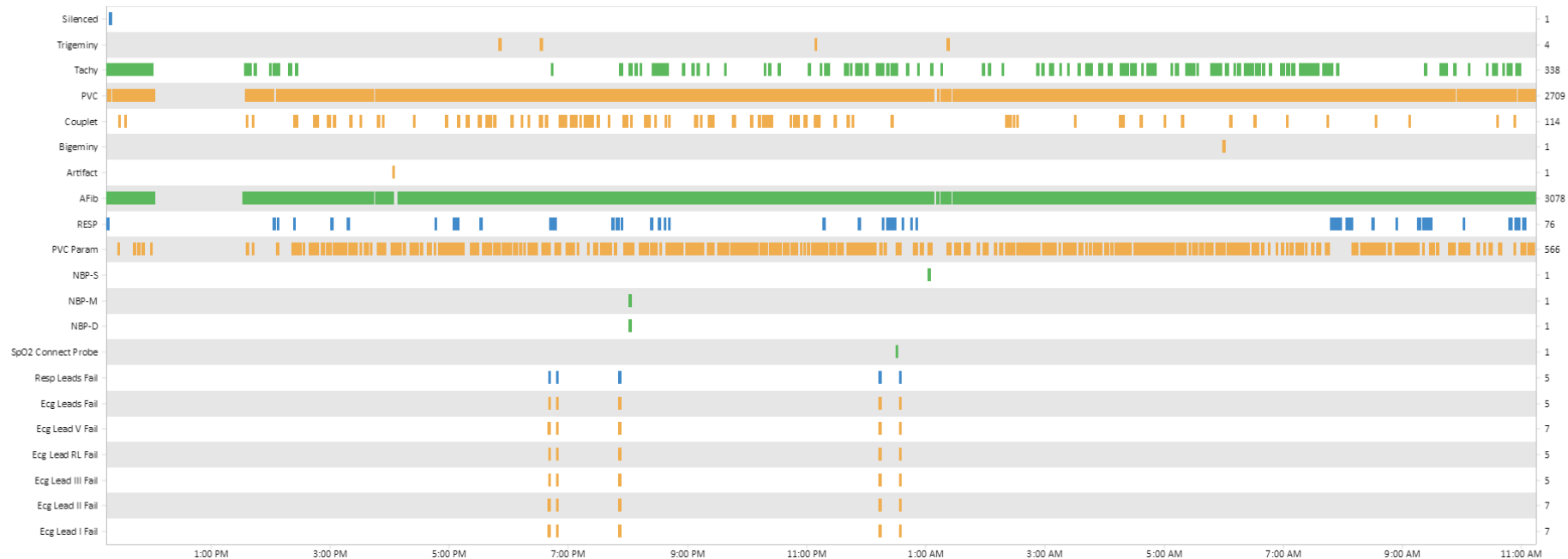
# Drilldown of Dashboard: Alarms

Alarm eXpert Hospital Unit Bed Patient Projects Xiao Hu Logout

Admission 24h 12h 8h 4h 2h now

13ICU 13ICU\_5 10/29/2017 11:14 AM 10/30/2017 11:14 AM Shift-View

Crisis 0 Warning 82 Advisory 3420 Message 3431





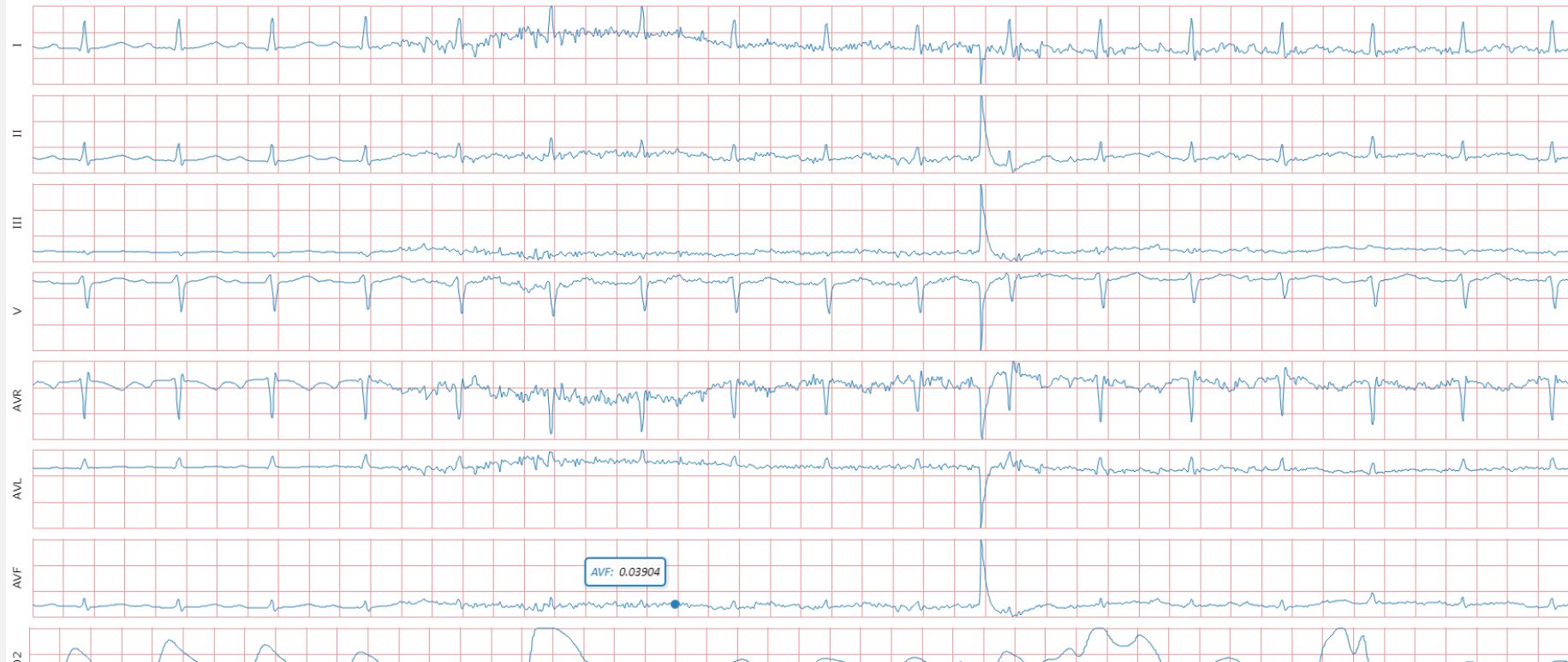
# Drilldown of Dashboard: Signal Waveforms

PVC



HR 100 PVC 0 RESP 25 SPO2-% 94 SPO2-R 95 NBP-S 134 NBP-D 79 NBP-M 99

November 2nd 2017, 10:09:07 pm





# Research History Built on Data



## SuperAlarm 1.0



### Dates

- Start: March 2010
- End: June 2012



### Identified Cohorts

- Code Blue patients
- Control patients



### Data Sources

- Monitor Data (BedMaster)
- Electronic Health Records (Epic)



### Institutions

- UCLA Medical Center

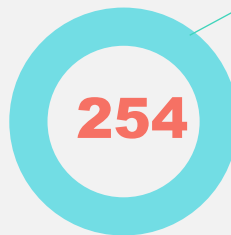
### ✓ EHR

- Patient Demographics
- ADT
- Labs

### ✓ Monitors

- Alarms
- ECG

## Code Blue case patients at UCLA



- $61.6 \pm 18.2$  average age (year $\pm$ SD)
- 54% male
- 46% female
- $10.4 \pm 16.9$  average monitoring duration (days, median $\pm$ SD)
- 2,213 total control patients



# SuperAlarm Dataset 2.0: Code Blue Patients



## Dates

- Start: March 2010
- End: December 2016



## Identified Cohorts

- Code Blue patients
- Control patients



## Data Sources

- Monitor Data (BedMaster)
- Electronic Health Records (Epic)



## Institutions

- UCLA Medical Center
- UCSF Medical Center

## UCSF\*(325)

From 2013

Average **age** (year)  $61 \pm 15$  ☐

Male 54.61% ☐

Female 45.39% ☐

Average **monitoring duration**  
(days, median  $\pm$  SD) ☐

$7.56 \pm 15.37$

**Control** patients ☐

2,234

## UCLA(777)

From 2010

☐ Average **age** (year)  $62 \pm 17$

☐ Male 57.82%

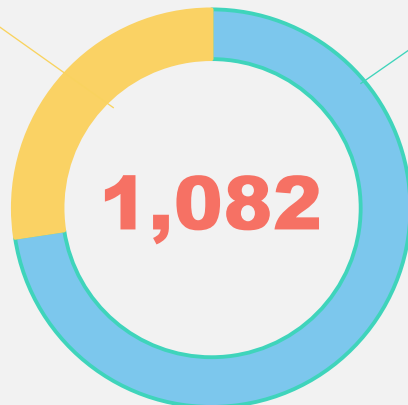
☐ Female 42.18%

☐ Average **monitoring duration**  
(days, median  $\pm$  SD)

$12.96 \pm 39.21$

☐ **Control** patients

3,734



\* Data collection started in 2013



# Conclusions



Current patient monitoring practice is imprecise

- ✓ Alarm fatigue – too many false and nuisance alarms
- ✓ Alarm failure – too few right alarms, patient deterioration missed



Precise patient monitoring is the future

- ✓ Systematic design of precise single alarm
- ✓ Multimodality data integration (monitor + EHR)
- ✓ Pattern recognition across modalities and temporal dimension

# ● Team Members

## Hu Lab

Xiao Hu, PhD, Director

## Algorithm Researchers

Kais Gadhomi, PhD

Ran Xiao, PhD

Mary Liu, PhD

Nate Tran, PhD Student

## Data master

Andrea Villaroman, MTM

## Lead software engineer

Peter Li, MS

## Software engineers

Del Bold, MS

Jacob Abba, BS



## Clinicians - UCSF

Rich Fidler, CRNA, PhD

Daphne Stannard, CNS, PhD

Michele Pelter, RN, PhD

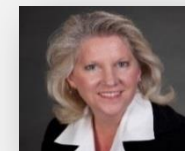
Matt Aldrich, MD

## Clinicians - UCLA

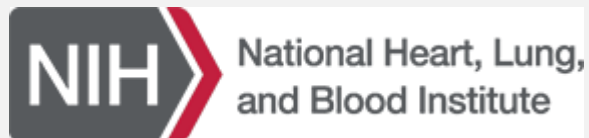
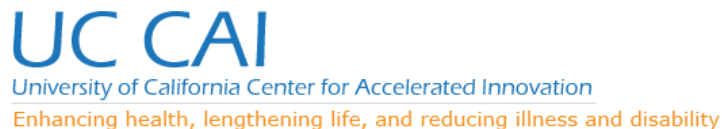
Paul Vespa, MD

Noel Boyle, MD

Duc Do, MD



# Acknowledgement







THANKS !