

Behavior Modeling in Physical and Adaptive Intelligent Services

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Project Overview

JST/RISTEX **S³FIRE** (Service Science, Solutions and Foundation Integrated Research Program)

Project title: **Innovation for Service Space Communication by Voice Tweets in Nursing and Caring**

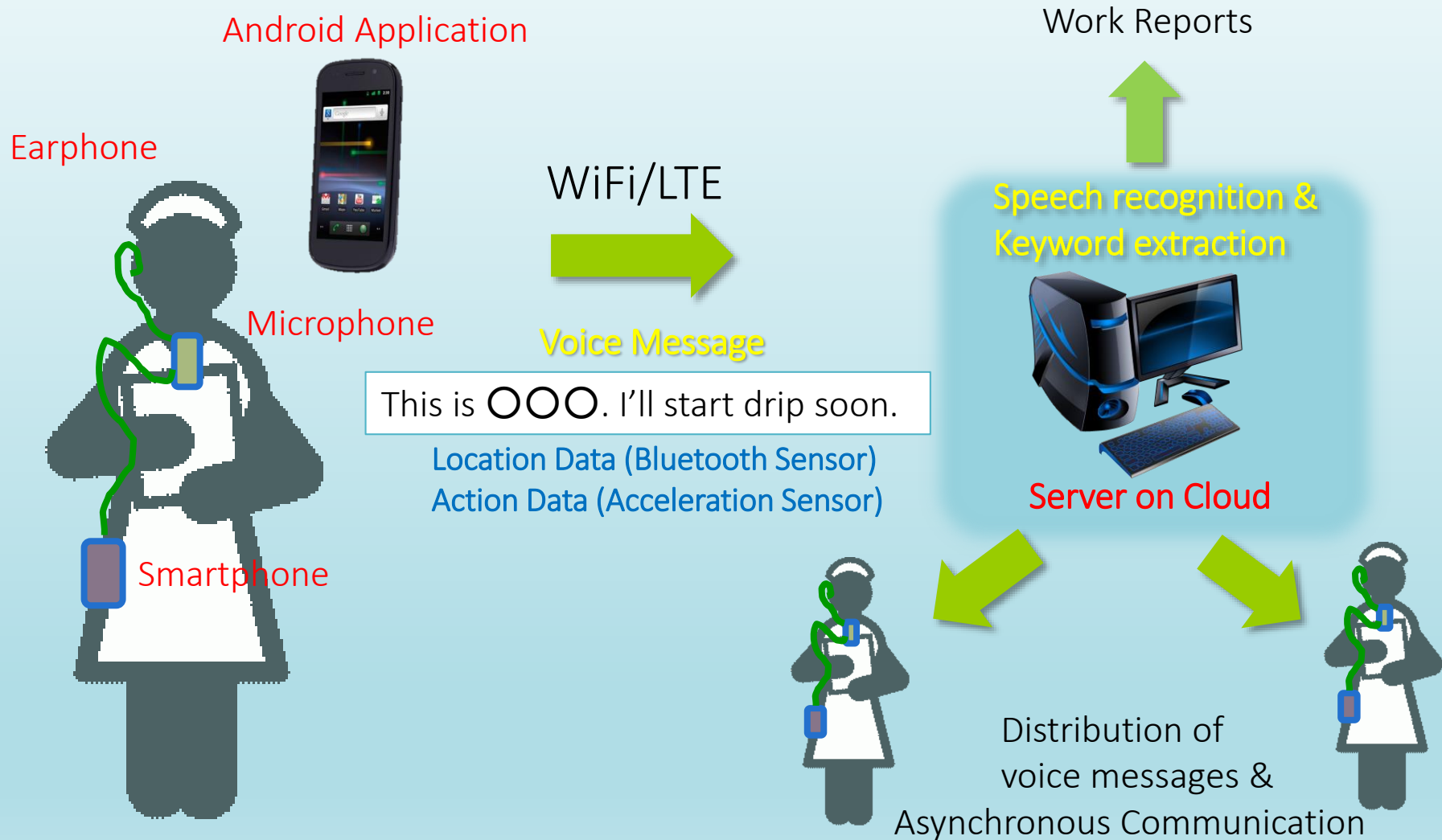
Aim: To improve working environment in hospitals /nursing homes.

Method: Introducing IT Device: smart voice messaging system (“**Voice Tweet Device**”).

Collaboration between



The SVM (Smart Voice Messaging) System



Smart Message Distribution

Doctor



Doctor:
Please be cautious in
XX and follow him up.

Nurse



Nurse M:
I am coming.

Behavior
estimate
engine

Collaboration

Provide voice
messages to right
people at the right
timing

Handover tasks

Real-time risk
management

Reference to
post records

Nursing
records

Nurse N:
We've finished
the check-up.

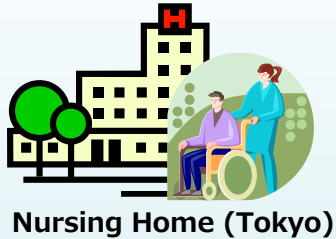
Endoscope
Examination Room

Nurse Station

Four Research Groups

- System development and field experiments
(Toshiba Corp. and all members)
- Visualization/evaluation of space-time communication
(Shimizu Corp.)
- **Modeling and simulation of human behavior**
(JAIST Information Science)
- Evaluation of service quality
(JAIST Knowledge Science)

Modeling and Simulation Group: Method



時刻	5/25/2013	12/30							
日付	5/25/2013	12/30							
入家	17:30:11	17:30:15							
出家	17:30:30	17:30:47							
移動時間	17:30:40	17:31:30							
移動距離	17:32:12	17:33:33							
移動速度	17:33:24	17:35:18							
移動方向	18:00:31	18:00:53							
移動経路	18:00:54	18:06:16							
移動回数	18:06:17	18:06:27							
移動距離	18:06:28	18:12:01							
移動速度	18:12:02	18:12:23							
移動方向	18:12:24	18:12:29							
移動経路	18:12:24	18:12:29							

Event Log / Voice-message Log

Reproduction of various situation, Iterative experiments under different conditions, prototyping by alternative devices



JAIST (Ishikawa)
Toshiba (Kanagawa)

Field Experiments

Extraction of unusual behavior, Comparison of individual behavior

Learning behavior models from logs and their analysis



Knowledge toward system improvement

Hypotheses and verification

Virtual Field Experiments

Modeling & Simulation

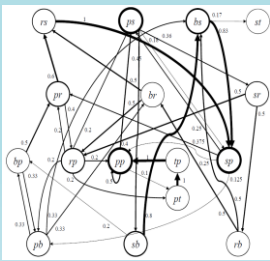


Video

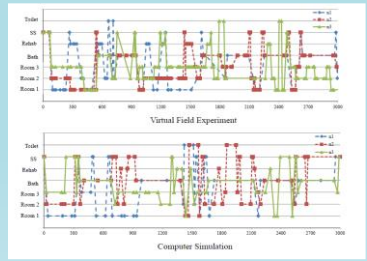
時刻	5/25/2013	12/30							
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Event Log

Building simulation models from process description, Estimation of behavior models from logs



Behavior Model



Traffic Lines

Virtual Field Experiments

Staff station



Wearable camera



Video camera



SVM Terminal

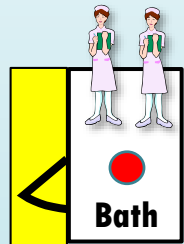


Transceiver
(Replacement of PHS)



Nurse

Observer



Rehabilitation

Commander

Toilet

Room4

Room3

Room2

Experiments

Virtual Field Experiments

20 December 2011 at JAIST

17 January 2012 at JAIST

13 February 2012 at Toshiba Corp.

13 September 2012 at JAIST

22 February 2013 at JAIST

13-15 March 2013 at Toshiba Corp.

28-29 August 2013 at JAIST

Field Experiments (at a nursing home in Tokyo)

5-8 December 2011

21-24 February 2012

27 September 2012

20-24 May 2013

21-23 August 2013

Evaluation of Communication Methods in VFE

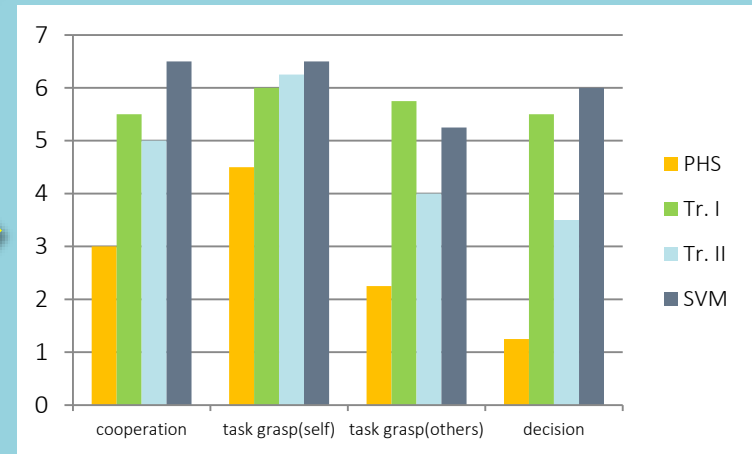
- Traditional Approach -

		PHS	Tr. I	Tr.II	SVM
Type	broadcast / one-to-one / one-to-group	to one	BC	BC	to group
Timing	synchronous / asynchronous	Syn	Syn	Syn	Asyn
Report	status reporting of tasks	-	✓	-	✓
	status reporting of other staffs and resources	-	✓	-	✓
Inquiry	to a staff	✓	-	-	✓
	to all	-	-	✓	✓
	about availability of resources	-	-	✓	✓
Request	assistance of tasks	✓	-	✓	✓
Use System data	event history	-	-	-	✓
Device		PHS	Transceiver	Transceiver	SVM

Communication Methods

	PHS	Tr.I	Tr.II	SVM
Total	3276.4	3286.59	2639.89	2981.24
Nurse1	1062.29	573.46	802.56	864.98
Nurse2	764.69	953.15	725.43	686.65
Nurse3	801.64	922.49	614.24	649.22
Nurse4	647.78	837.5	497.67	780.4

Length of Traffic Line

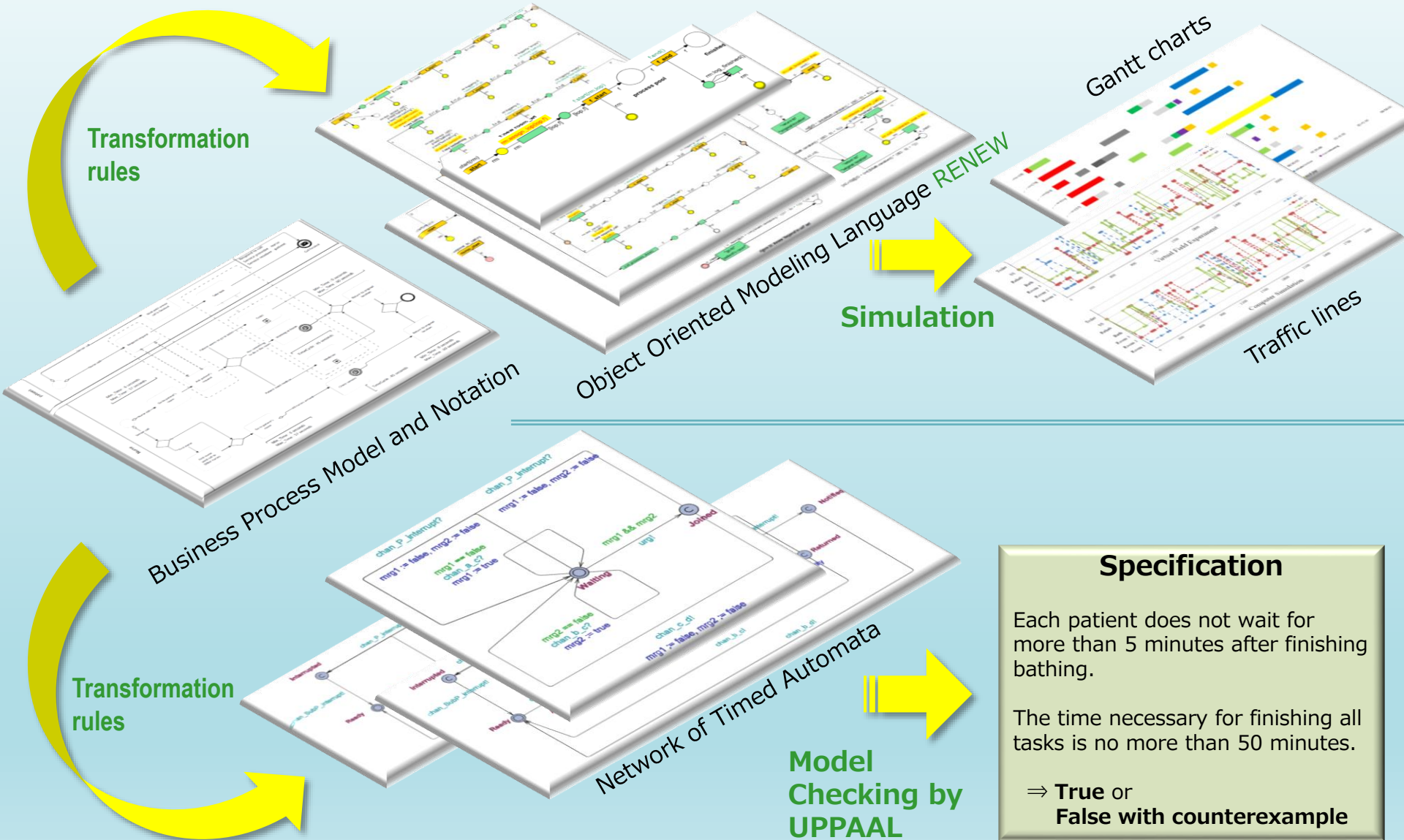


Result of Questionnaire
(Subjective evaluation)

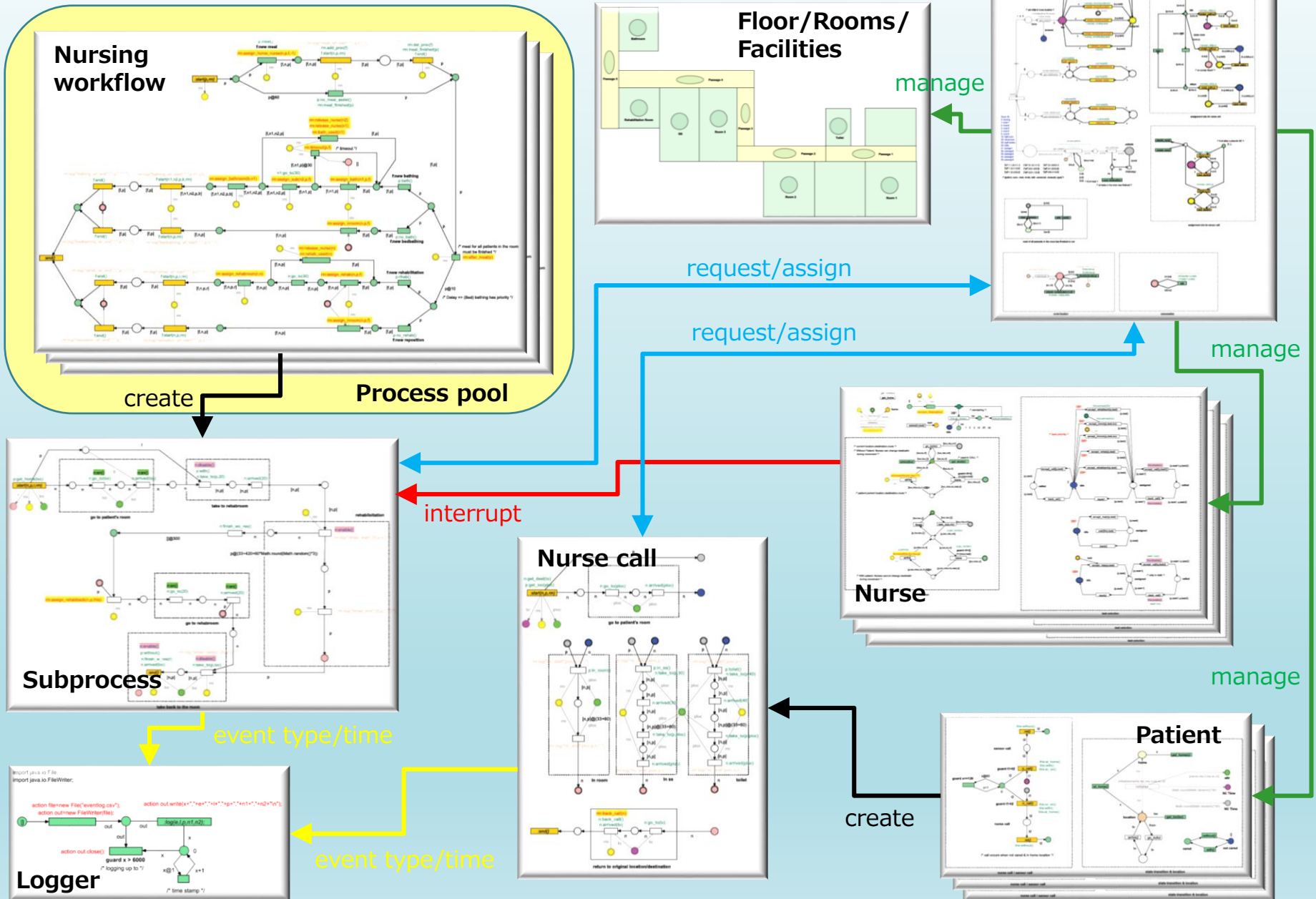
	unfinished tasks (sec.)	waiting time (sec.)	NC response AVG (sec.)
PHS	46	26.9	1.5
Tr.I	42	21.4	1.5
Tr.II	56	35.1	1.7
SVM	34	18.6	1.8

Statistics on Time

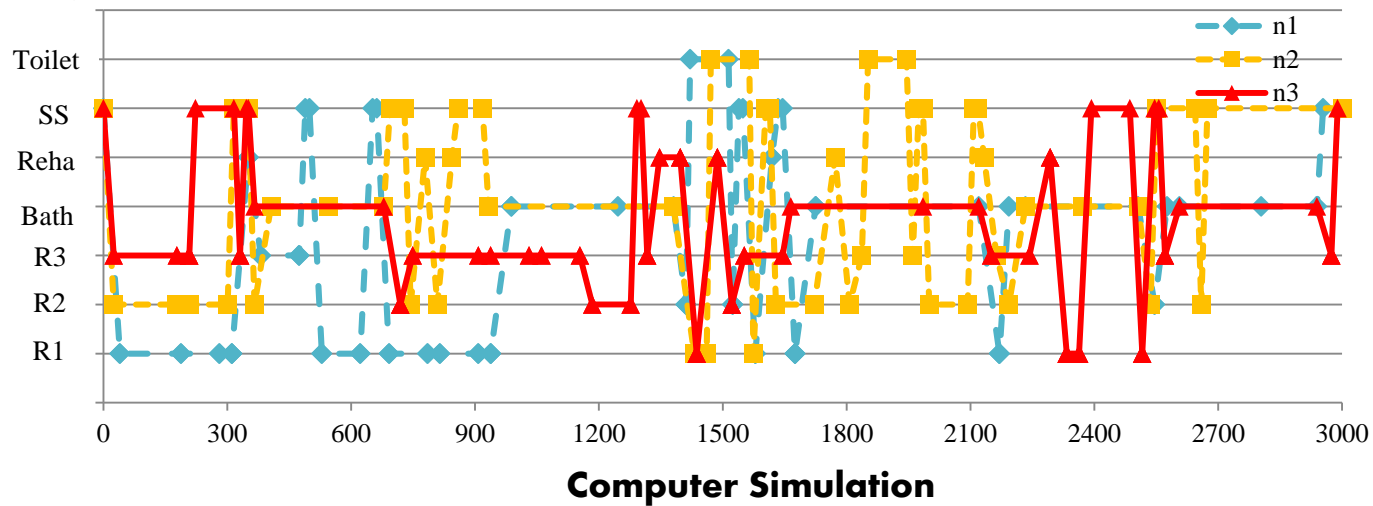
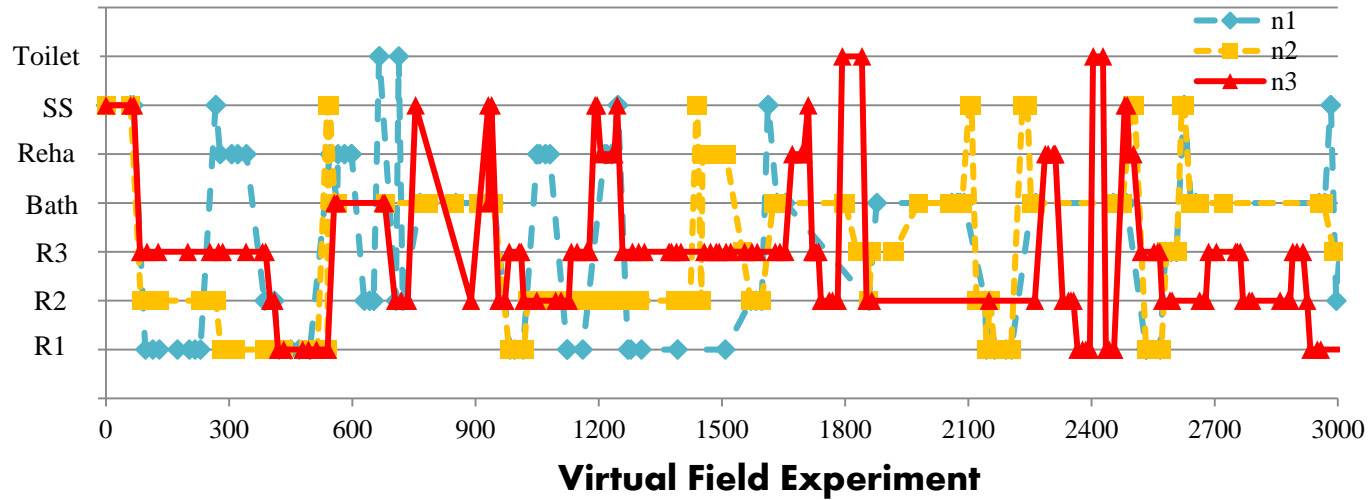
Automatic Synthesis of Executable Models from Process Description



Structure of RENEW Simulation Model



A Result of Simulation – Traffic Line -



Step 1. Event Abstraction

Event format

(date, Staff-ID, type, in-time, out-time, duration)

‘type’ is either the location or “moving”.

Event log

開始	5/25/2013	17:30					
終了	5/25/2013	19:15					
	入室	退室	滞在時間(秒)	移動距離(m)	移動時間(秒)	移動速度(m/秒)	
2Fリビングサロ	17:30:11	17:30:16		5			
↓					12	17	0.7
2Fリビングサロ	17:30:33	17:30:47		14			
↓					0	1	0
2Fリビングサロ	17:30:48	17:31:33		45			
↓					0	39	0
2Fリビングサロ	17:32:12	17:33:23		71			
↓					12	1	12
2Fリビングサロ	17:33:24	17:59:18		1554			
↓					0	73	0
2Fリビングサロ	18:00:31	18:00:53		22			
↓					12	1	12
2Fリビングサロ	18:00:54	18:06:16		322			
↓					35	1	35
208	18:06:17	18:06:27	10				
↓					35	1	35
2Fリビングサロ	18:06:28	18:12:01		333			
↓					0	1	0
2Fリビングサロ	18:12:02	18:12:23		21			
↓					12	1	12
2Fリビングサロ	18:12:24	18:22:19		595			
↓					0	15	0

Abstraction table

Event	Short stay	Long stay
1F Rooms	a	A
2F Rooms	b	B
3F Rooms	c	C
1F Salon	x	X
2F Salon	y	Y
2F Staff Station	s	S
3F Care Station	z	Z
Moving	-	M
Others (stairs/elevator etc.)	e	E

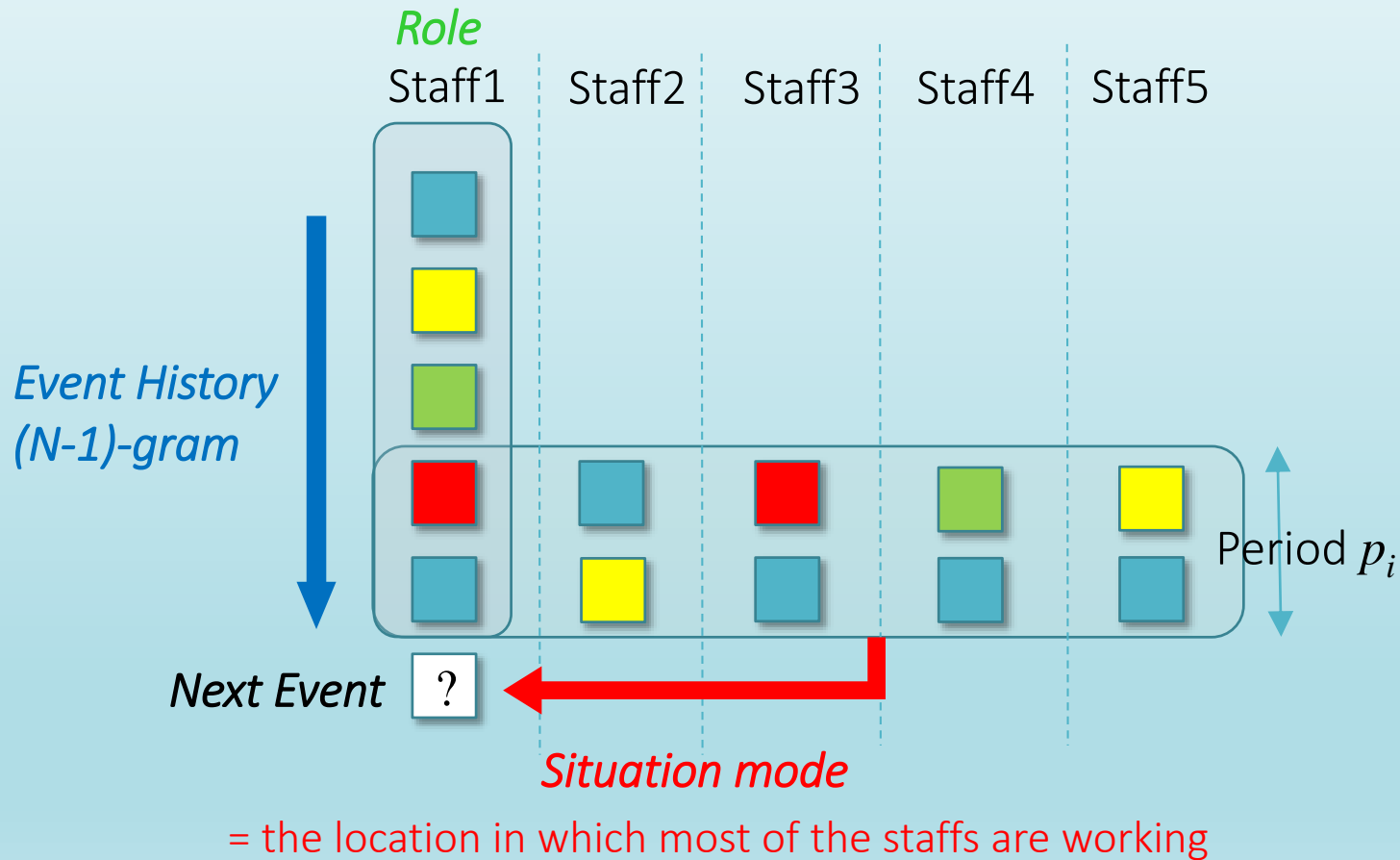
Event sequence

bbByyyyyyYyyyyyyyyyybbbbbbbbbbbbbbyxyyy
 yyyeMyexaXXXyyyeXXxXxXxxexXXxxeeyxyyyxxM
 aaXeyY

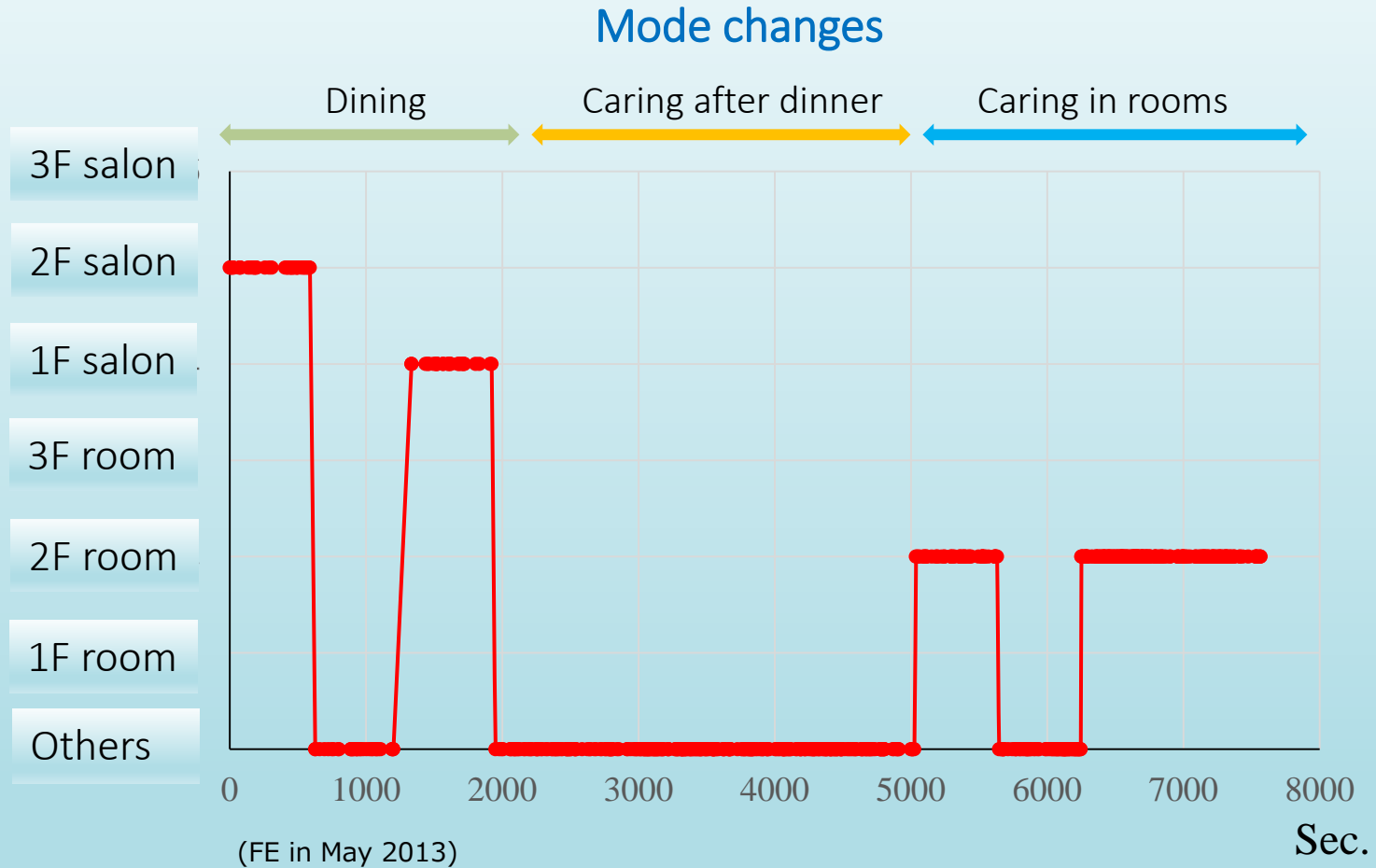
Step 2. Modeling: Attributed N-Gram Model

Conditional Probability

$Pr(\text{Next Event} \mid [\text{Role}, \text{Situation Mode}] : \text{Event History})$



Step 2. **Modeling:** Situation mode



Step 2. **Modeling: Learning Models from Event Logs**

Maximum likelihood estimation

$$Pr(\sigma|y) = \frac{O_{y\sigma}(w)}{\sum_{\sigma' \in \Sigma} O_{y\sigma'}(w)}$$

w : the given event sequence.

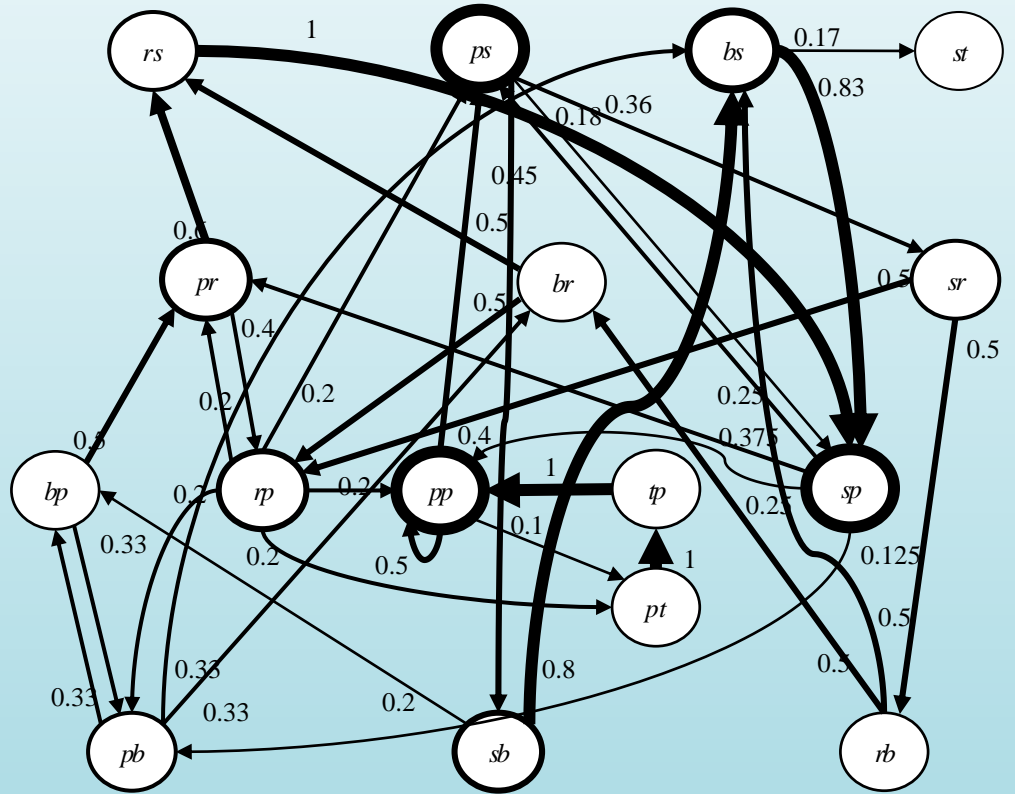
$O_s(w)$: the number of times string s occurs in w .

Step 2. Modeling: Probabilistic Automata

Event	Symbol
Move to Room1~4	p
Move to Bathroom	b
Move to Rehab. room	r
Move to Staff Station	s
Move to Toilet	t

Conditional probabilities

Prev.	Next	PHS	Tr.I	Tr.II	SVM
br	b	0	0	0	0
	p	0.40	0.67	0.67	0.25
	r	0	0	0	0
	s	0.60	0.333	0.333	0.75
	t	0	0	0	0
pr	b	1.0	0	0.25	0
	p	0	0.67	0.25	0
	r	0	0	0	0
	s	0	0.33	0.05	1.0
	t	0	0	0	0
sb	b	0	0	0	0
	p	1.0	0.20	0.20	0.25
	r	0	0.20	0.40	0.25
	s	0	0.60	0.40	0.50
	t	0	0	0	0



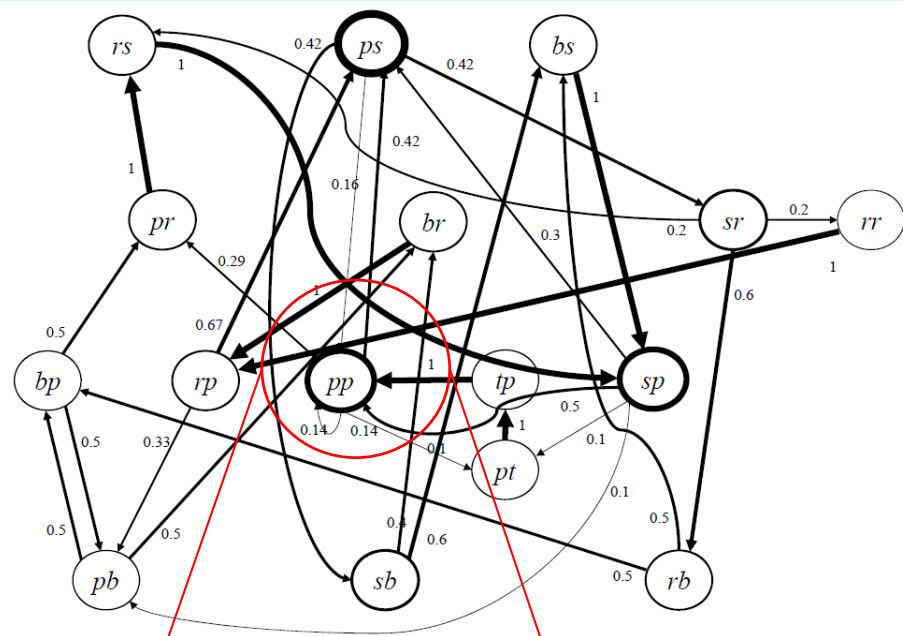
Graphical Representation by Probabilistic Automaton

(VFE in Sept. 2012)

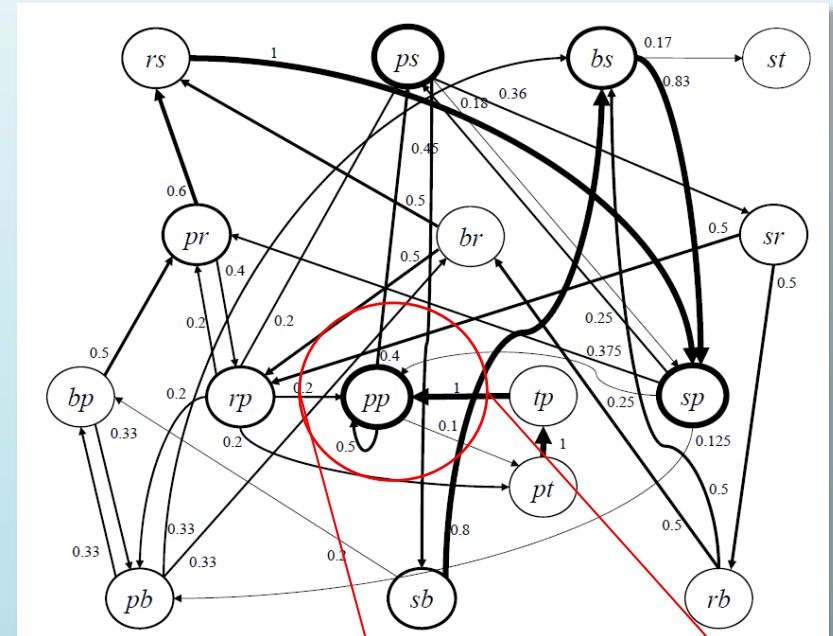
Step 3. Analysis:

Comparison of Models in different Situations

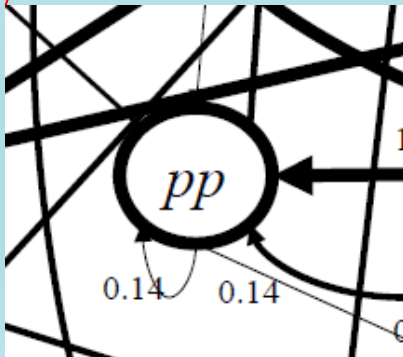
Transceiver II



SVM

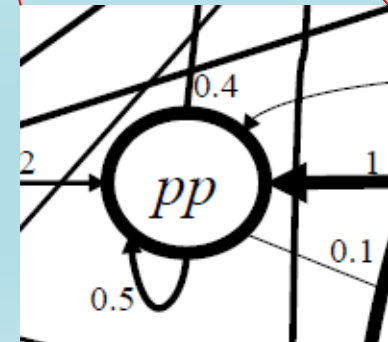


(VFE in Sept. 2012)



$$\Pr(\text{Room} \mid \text{Room} \rightarrow \text{Room}) \\ = 0.14 \text{ (Tr. II)}, 0.5 \text{ (SVM)}.$$

In-room tasks were interrupted often in experiments other than SVM.



Step 3. Analysis:

Detection of Unusual/Suspicious Behavior I

Event log

開始	5/25/2013	17:30							
終了	5/25/2013	19:15							
	入室	退室	滞在時間(秒)	移動距離(m)	移動時間(秒)	移動速度(m/秒)			
2Fエレベーター	17:30:11	17:30:16		5	12	17	0.7		
2Fエレベーター	17:30:33	17:30:47		14				0	0
2Fエレベーター	17:30:48	17:31:33		45	0			39	0
2Fエレベーター	17:32:12	17:33:23		71					
2Fエレベーター	17:33:24	17:59:18		1584	12			1	12
2Fエレベーター	18:00:31	18:00:53		22				73	0
2Fエレベーター	18:00:54	18:00:16		322	12			1	12
1	2008	18:06:17	18:06:27	10				35	35
2Fエレベーター	18:06:28	18:12:01		333				1	35
2Fエレベーター	18:12:02	18:12:23		21				0	0
2Fエレベーター	18:12:24	18:22:19		585				12	12
2Fエレベーター				0				15	0

Fragments

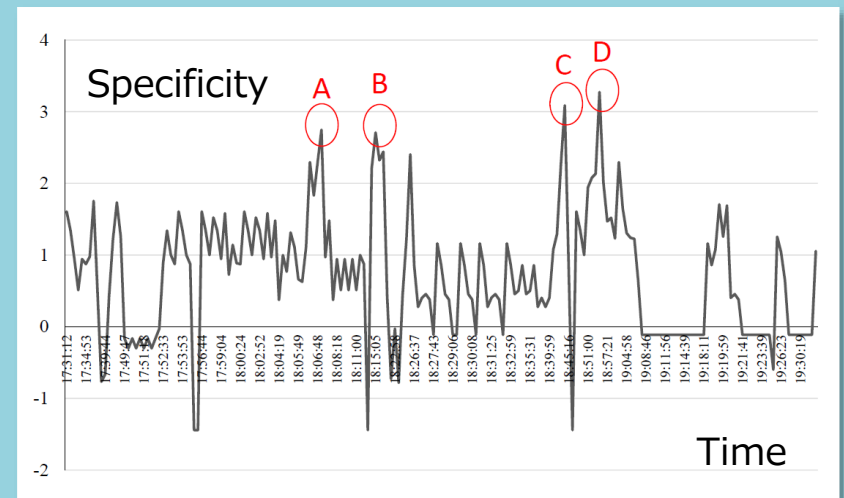
a	b	c	d	e	f	g
a	b	c	d			
b	c	d	e			
c	d	e	f			
d	e	f	g			

degree of specificity

$$= \log \left(\frac{\text{actual occurrences}}{\text{expected number}} \right)$$

Behavior model of
average behavior

- A. Frequent movements between different floors:
1F salon → 2F salon → 3F room → 2F room
- B. Long movement and long task at the same location.
- C. The following voice message was sent just before the point: "Ms. XXX has returned to her room by herself. I will go to see her now." (Usually Ms. XXX needs assistance on her movement.)
- D. Frequent movements using an elevator.



(FE in May 2013)

Time series of specificity

Step 3. Analysis:

Detection of Unusual/Suspicious Behavior II

Event log of each staff

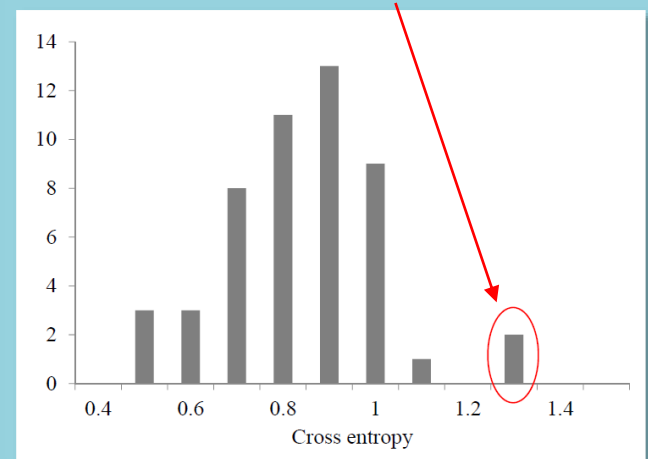
[illegible]

Cross entropy analysis (Discrepancy between log and model)

$$H(w) = - \sum_{j=1, |w|} \frac{1}{|w|} \log_2 \Pr \left(s(w_{[j]}) | a(w_{[j-N+1, j-1]}): s(w_{[j-N+1, j-1]}) \right)$$

Behavior model of average behavior

Different behavior:
Most of time is spent in 1F salon.



(FE in May 2013)

Histogram

Conclusion

Behavior modeling for physical and adaptive intelligent services:

- Detailed process description,
- Automatic generation of executable models for simulation and verification,
- Learning probabilistic models from event logs,
- Diagnosis of human behavior: detection of unusual/suspicious activities, discrepancy between individual log and average behavior

Ongoing/Future work

- How to utilize the results for improving service quality
- Modeling collaboration of staffs
- More experiments on other fields