



AWARE@ICU

Colin Lea

Automated Workflow and Activity Recognition CIS2 Proposal

Hundreds of micro-tasks

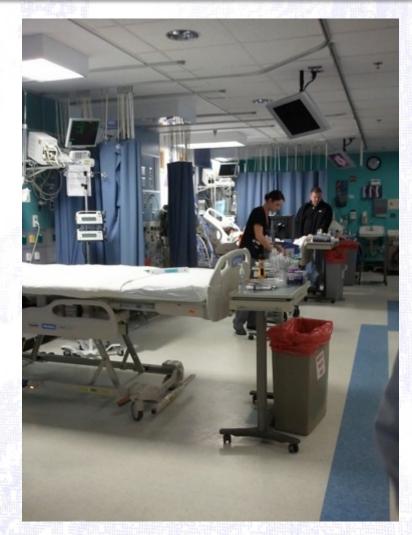
- Drug administration
- Emptying chest tubes
- Documentation

Lack of written direction

- Which nurse is doing what?
- Hierarchy of staff

Sensory overload

- Many alarms
- Sounds become meaningless



Monitors

Infant



Typical Setup

Staff

Movable Equipment

2



Handoffs



Procedures

objectives

Build algorithmic tools for analyzing workflow

- Gather data
- Develop/evaluate techniques for recognition
- Summarize staff activity

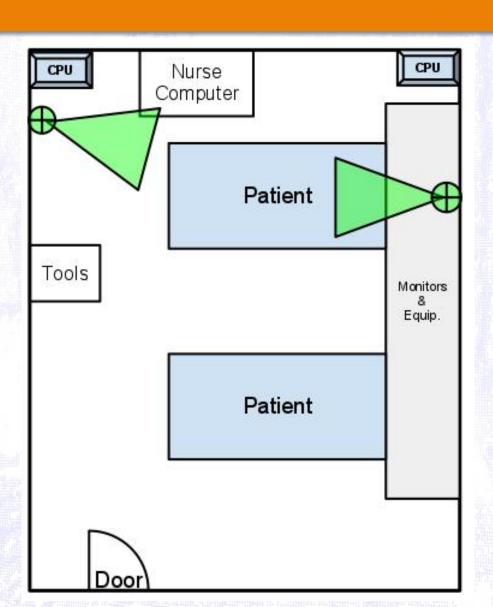
Secondary:

- Understand patient movement
- Analyze microphone data
- Quantify nurse interaction times
- Examine medical device usage

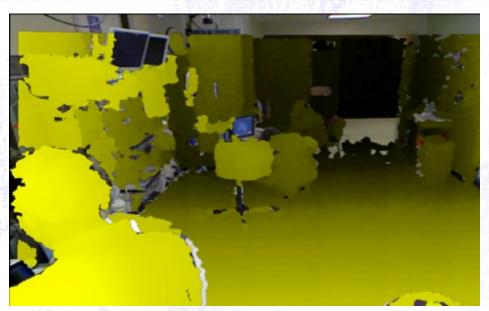
Retroactive + Preventative

hardware





kinect





Spatial cues
What/where are the people and instruments?
Pose detection
What are people doing?

preliminary direction

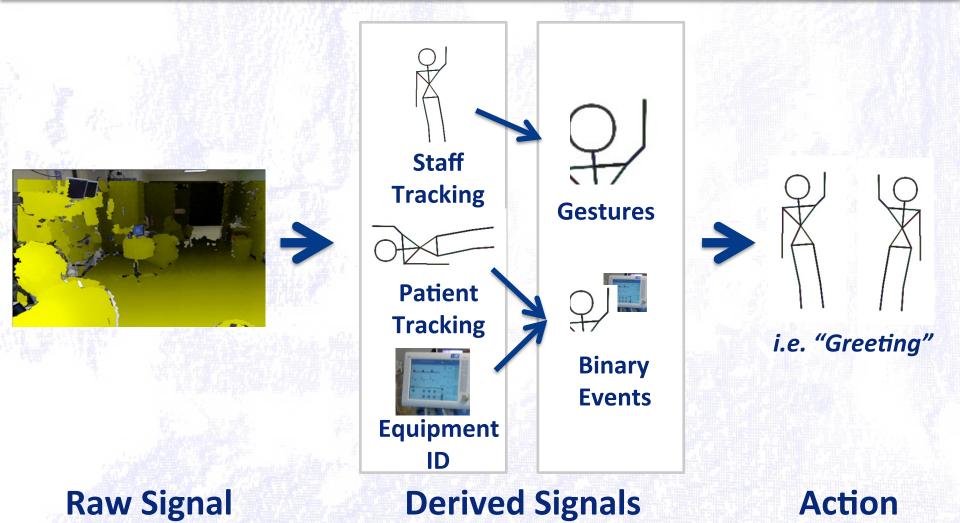
Activity Summarization:

What are the most frequent interactions?

What features are necessary to classify these actions?

With what accuracy can the top actions be recognized?

activity pipeline

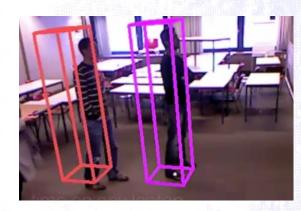


models

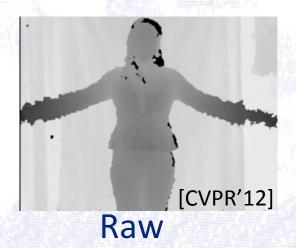
Derived Signals

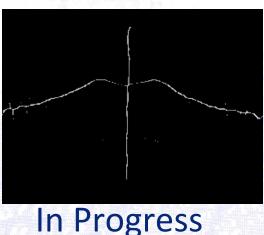
Staff location tracker:

- Bounding box of body from motion Body pose estimation:
- Raw depth -> Joint positions
- Segmentation (PCA-kNN, Watershed)
- Parts model



[EPFL-LTS2 2010]







models

Derived Signals

Determine machine type/location

- Rough depth-based scene segmentation
- Feature-based detectors (i.e. SIFT)

Binary "touch" signals:

- Staff touching patient?
 - Location on patient? [Body tracker]
- Touching Equipment?
 - Type of equipment? [Scene analysis]



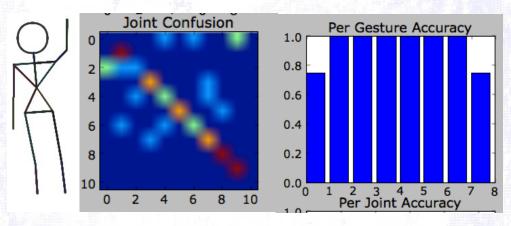


models

Action Recognition

Staff gestures

- **Determine sub-actions**
 - i.e. Inserting tube
- Correlate input w/ each known gesture's PCA basis
- Temporal window?



Current results

- Activity Logger
 Fuse all derived signals and output "action"
- Time-series graphical model (i.e. CRF, S-LDS, HMM)

- Patient Body TrackerEstablish metric for "unsafe" motion
- Track body over time
- Trigger alert (and log) when problematic

Reco	order		Experimentation			
Task	Difficulty	Notes	Task	Difficulty	Notes	
General Purpose Recorder	Easy	In Progress	Get IRB Approval	Easy	Pending	
		In	Record preliminary data	Easy		
Face Tracking	Medium	Progress	Ensure de-identification	Easy		
Blackout stretcher	Easy	The second secon	Hand annotate activities	Medium		
Video tracking failure	Medium	The second secon	Additional recording	Easy		
Derive	d Data	2 1 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2	Activity recognition			
Task Difficulty No		Notes	Task	Difficulty	Notes	
Location tracker	Easy	2000年 - 1000年 - 10000			In	
Face direction tracker	Easy	CSIRO Software	Gesture recognition	Medium	Progress	
Staff body pose			Retrospective analysis	Hard		
estimation	Hard		Patient Tracking	Medium		
Patient tracker	Medium		Multi-camera models	Medium		
Equipment Identification	Hard		Evaluation	Easy	14	

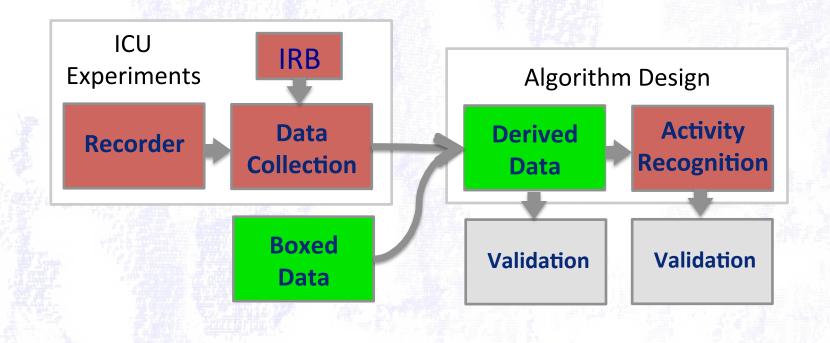
Reco	order	2.1	Experimentation			
Task Difficulty		Notes	Task	Difficulty	Notes	
General Purpose Recorder	rpose Easy		Get IRB Approval	Easy	Pending	
Face Tracking		In Progress	Record preliminary data	Easy		
Face Tracking	Medium		Ensure de-identification	Easy		
Blackout stretcher	Easy	The second secon	Hand annotate activities	Medium		
Video tracking failure Medium			Additional recording			
Derived Data Minimum						
Task Difficulty		Notes	Task	Difficulty	Notes	
Location tracker	Easy				In	
Face direction tracker	Easy	CSIRO Software	Gesture recognition	Medium	Progress	
Staff body pose			Retrospective analysis	Hard		
estimation	Hard		Patient Tracking	Medium		
Patient tracker	Medium		Multi-camera models	Medium		
Equipment Identification	Hard		Evaluation	Easy	15	

Reco	order		Experimentation				
Task Difficulty		Notes	Task	Difficulty	Notes		
General Purpose Recorder	Easy	In Progress	Get IRB Approval	Easy	Pending		
Face Tracking	Medium	In Progress	Record preliminary data Tensure de-identification	Easy			
Blackout stretcher	Easy	The state of the s	Hand annotate activities	Medium			
Video tracking failure 7	Medium		Additional recording	Easy			
Derived Data <u>Expected</u> Activity recognition							
Task Difficulty		Notes	Task	Difficulty	Notes		
Location tracker	Easy	AND THE RESERVE			In		
Face direction tracker	Easy	CSIRO Software	Gesture recognition	Medium	Progress		
Staff body pose			Retrospective analysis	Hard			
estimation	Hard		Patient Tracking	Medium			
Patient tracker	Medium		Multi-camera models	Medium			
Equipment Identification	Hard		Evaluation	Easy	16		

Reco	order	2.8	Experimentation			
Task Difficulty		Notes	Task	Difficulty	Notes	
General Purpose Recorder	Easy	In Progress	Get IRB Approval	Easy	Pending	
Face Tracking	Medium	In Progress	Record preliminary data Tensure de-identification	Easy		
Blackout stretcher	Easy		Hand annotate activities	Medium		
Video tracking failure 🦙	Medium		Additional recording	Easy		
Derived Data Maximum						
Task Difficulty		Notes	Task	Difficulty	Notes	
Location tracker	Easy	A41.			In	
Face direction tracker	Easy	CSIRO Software	Gesture recognition	Medium	Progress	
Staff body pose		(1) (1) (1) (1) (1) (1) (1) (1) (1) (1)	Retrospective analysis	Hard		
estimation	Hard		Patient Tracking	Medium		
Patient tracker	Medium		Multi-camera models	Medium		
Equipment Identification	Hard		Evaluation	Easy	17	

Task	Difficulty
CVPR One Shot Gesture Competition	Medium
Multi-view reconstruction	Easy
Semi-supervised activity learning	Medium

supplemental tasks



dependencies

milestones

Generic recorder

(depth + video + skeleton)

Experimental Data

- Pending IRB Approval
- Preliminary + additional datasets

Activity Logger

Record actions

Derived Data Modules

- Gesture based
- Scene analysis
- Touch sensors

Recognition Modules

- Time-series recognition
- Patient body tracker

literature

Key Directions:

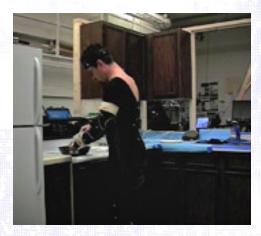
- Spatiotemporal techniques
- Sequential methods (le. HMMs)
- Context free grammars
- Hierarchical Graphical Models



Georgia Tech Aware Home



TUM Operating Room



CMU Kitchen

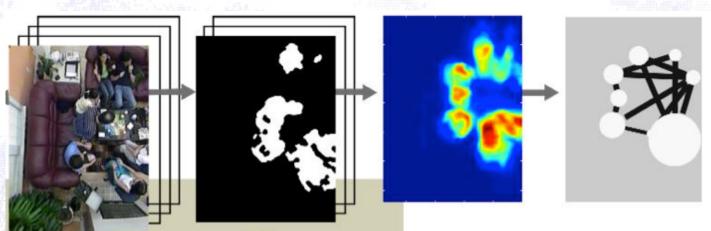
Aware Home

Goal: Capture & aggregate motion data. Visualize.

- Unsupervised techniques
- Suffix trees (2008)
 - Discrete-data subsequences
 - Reduce redundant info
 - Finds & merges new activities

	Suffix Tree		V	SM	3-grams		5-grams	
	P	R	P	R	P	R	P	R
Aloo Dam	55	60	55	50	50	60	55	60
Babka	50	40	-	-	56	50	38	30
Cereal	63	50	60	60	57	40	33	30
Fruit Salad	50	60	-	-	-	-	33	40
Omelet	63	70	-	-	-	-	-	-
Raita	47	70	-	-	18	70	33	70
Chicken	70	70	16	100	44	40	42	50
Setup Table	88	70	60	60	50	50	45	50
Green Salad	64	70	-	-	40	20	38	30
Wash Dishes	60	30	50	50	44	40	28	20
Average	61	59	24	32	36	37	35	38
% Discovery	100			50	8	0	9	0

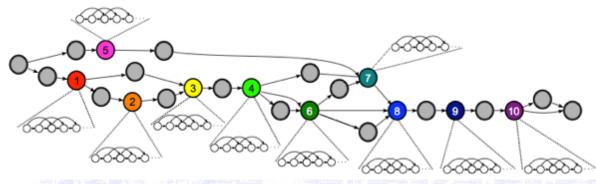
Activity Discovery

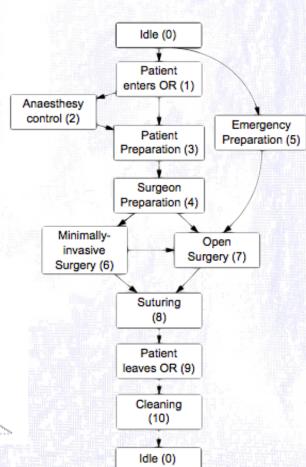


Operating Room

Goal: Analyze surgical workflow

- Supervised techniques
- Workflow HMMs
 - Prob. of going in and out of state
 - Sub-HMM per phase
- Dynamic Time Warping

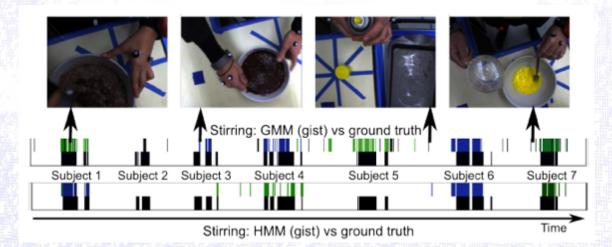




QoLT Kitchen

Goal: Recognize actions in food preparation

- Video + IMU + Motion + RFID
- Unsupervised temporal segmentation
 - PCA + GMM
- Supervised activity classification
 - Scene Gist (le. counter vs fridge)
 - k-NN beat HMM with multimodal data
 - Less data can beat more



LCSR:

- Dr. Russ Taylor
- Dr. Greg Hager
- Dr. Suchi Saria

JHMI:

Dr. Jim Fackler

collaborators

