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Analyse visuelle de patterns temporels dans les données des patients

avril 2018

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## Analyse Visuelle de Patterns Temporels dans les Données Patient

**Catherine Plaisant** 



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## Human-Computer Interaction Lab

### **Interdisciplinary Research Community**

Computer Science Information Studies Psychology, Engineering, Business, Education, etc.

www.cs.umd.edu/hcil

#### Designing the User Interface: Strategies for Effective Human-Computer Interaction, 6th Edition

Ben Shneiderman, University of Maryland Catherine Plaisant, University of Maryland Maxine Cohen, Nova Southeastern University Steven Jacobs, Northern Arizona University Niklas Elmqvist, University of Maryland Nicholas Diakopoulos, University of Maryland

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#### SHNEIDERMAN • PLAISANT • COHEN • JACOBS • ELMQVIST



DESIGNING THE USER INTERFACE STRATEGIES FOR EFFECTIVE HUMAN-COMPUTER INTERACTION



# Information visualization

Apply DK Carsoel





Home Finder and Filmfinder prototypes lead to SpotFire



#### Hierarchical Clustering Explorer HCE



Treemap

Lifelines



SpaceTree

### **ANSCOMBE'S QUARTET**

1		( 2	2	3		4	
х	у	Х	у	Х	у	х	у
10.0	8.04	10.0	9.14	10.0	7.46	8.0	6.58
8.0	6.95	8.0	8.14	8.0	6.77	8.0	5.76
13.0	7.58	13.0	8.74	13.0	12.74	8.0	7.71
9.0	8.81	9.0	8.77	9.0	7.11	8.0	8.84
11.0	8.33	11.0	9.26	11.0	7.81	8.0	8.47
14.0	9.96	14.0	8.10	14.0	8.84	8.0	7.04
6.0	7.24	6.0	6.13	6.0	6.08	8.0	5.25
4.0	4.26	4.0	3.10	4.0	5.39	19.0	12.50
12.0	10.84	12.0	9.13	12.0	8.15	8.0	5.56
7.0	4.82	7.0	7.26	7.0	6.42	8.0	7.91
5.0	5.68	5.0	4.74	5.0	5.73	8.0	6.89

### **ANSCOMBE'S QUARTET**

1		( 2	2	3 4		1	
х	у	Х	у	X	у	Х	у
10.0	8.04	10.0	9.14	10.0	7.46	8.0	6.58
8.0	6.95	8.0	8.14	8.0	6.77	8.0	5.76
13.0	7.58	13.0	8.74	13.0	12.74	8.0	7.71
9.0	8.81	9.0	8.77	9.0	7.11	8.0	8.84
11.0	8.33	11.0	9.26	11.0	7.81	8.0	8.47
14.0	9.96	14.0	8.10	14.0	8.84	8.0	7.04
6.0	7.24	6.0	6.13	6.0	6.08	8.0	5.25
4.0	4.26	4.0	3.10	4.0	5.39	19.0	12.50
12.0	10.84	12.0	9.13	12.0	8.15	8.0	5.56
7.0	4.82	7.0	7.26	7.0	6.42	8.0	7.91
5.0	5.68	5.0	4.74	5.0	5.73	8.0	6.89

For ALL 4 datasets

Property	Value
Mean of x	9.0
Variance of x	11.0
Mean of y	7.5
Variance of y	4.12
Correlation	0.816
Linear regression	y = 3 + 0.5x

#### **ANSCOMBE'S QUARTET**



## INFORMATION VISUALIZATION Interactive Compact graphical presentation and User interface

for manipulating large numbers of items  $(10^2 - 10^{6+})$ 

Enables users to make discoveries, decisions, or explanations about patterns or groups of items

• Visual bandwidth is enormous Human perceptual skills are remarkable



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See commercial example online at: https://finviz.com/map.ashx?t=sec



U of Washington - https://vizhub.healthdata.org/gbd-compare/

# **Event Analytics**



## Numerical

## Categorical

#### Patient ID: 12345

04/26/2010 10:0031.0304/26/2010 10:1531.0104/26/2010 10:3031.0204/26/2010 10:4531.0804/26/2010 11:0031.16



#### Patient ID: 45851737

12/02/2008 14:26 Arrival 12/02/2008 14:36 Emergency 12/02/2008 22:44 ICU 12/05/2008 05:07 Floor 12/14/2008 06:19 Exit



# Many application domains



Electronic Health Records: symptoms, treatment, lab test Student records: course, paper, proposal, defense, etc. Web logs, usability logs, security etc. Traffic incident logs: confirmed, unit arrived, lane closed etc.

## **HCIL Prototype evolution**

	ΤοοΙ	Event Types	Records	Display	
	LifeLines	Points, Intervals	One	Individual	
	LifeLines2	Points	Many	Individual, Summary	
	Similan	Points	Many	Individual	
	LifeFlow	Points	Many	Individual, Aggregate	
	EventFlow	Points, Intervals	Many	Individual, Aggregate	
(	ORACLE"	Adobe Children	MedStar Sonal Hospital C	Washington Center Nith National Institute of Health	
	JNIVERSITY of MARYLAND School of Pharmacy	Yale School	University ol of Medicine	E IVERSITY OF ITISH LUMBIA	

www.cs.umd.edu/hcil/toolname



#### LifeLines - Single Patient http://www.cs.umd.edu/hcil/lifelines





The Wand Timeline view of a patient record in Allscript's ambulatory EHR iPad application. Used with permission of Allscripts.



http://www.healthtronics.com/physicians/it-solutions/urochartehr-system-product-overview

# A SINGLE RECORD



# MANY RECORDS

Are we following guidelines?

Find patients with Y and Z for my trial

How are opioids prescribed?

Can we improve this workflow?



RECORD

RECORD

RECORD

RECORD







#### Example: Where do patients go after they leave the emergency room?

Data: Transfers within Hospital

Arrival
Emergency
ICU
Die
Floor

🔺 Discharge-Alive

# Summarize all the sequences



## Constructing the EventFlow Overview









































Time

### Eventflow: overview of all sequences





Lien vers la vidéo : EventFlow short demo

https://www.youtube.com/watch?v=ZN1BefRmBMc




#### Interactively manage gaps and overlaps



### Understanding Adherence Using Large Scale Claims Data



Bjarnadottir, M., Malik, S., Onukwugha, E., Gooden, T., Plaisant, C. Understanding Adherence and Prescription Patterns Using Large Scale Claims Data PharmacoEconomics, Volume 34, Issue 2, pp 169-179, 2016.



Data representing men diagnosed with cancer (small sample to explain the design).

Records have been aligned by the cancer diagnosis date (green).

We can see the different sequences of treatment with luteinizing hormone-releasing hormone (LHRH) (purple) and radiation therapy (brown). The most common first treatment in this group is the LHRH. The second most common is radiation therapy and we can see that it occurs earlier on average than LHRH as the distance from green to brown is shorter than the distance from green to purple.







## 18% of adults had the expected sequence (N=541)



# Pediatric unexpected care sequences (n=807)



# Pediatric unexpected care sequences (n=807)



## Pediatric unexpected care sequences (n=807)



## Children's Hospital in DC: Trauma Bay



## **Children's Hospital in DC: Trauma Bay**



### Focusing on the first two events

84% of patients are checked in the correct order.

The most common deviation is that the breathing is checked before the airways (14% of patients) Reversed group takes longer on average than the correct sequence



### Distributions



### Adding third event type (central pulse)



### Combine the 2 pulse types

Adding dista	al pulse	Combine Categories
	Categor	y Name: any_pulse
	Sub-Ca	ategories:
000	EventFlow v2.1-RC2 - DND-ChildrensDemo-06-26-13.txt (DN	I pulse
<u>File Dataset Selection View H</u> elp	🔺 distal p	pulse
Showing: 215/215 patients	Distribution Options 0	
Selected: 0 patient	Zoom: 2	II I None
All Hide selection		9 19:37 1 36m
None Hide others		
Align Filter Merge Window		
No alignment		
1 A ainways		
from the beginning	Other C	Categories:
	∕	'S
Focus:	▲ breath	ing
Betore Both Alter	la aliabil	lity
Advanced Options	- second	dary survey
(no advanced option selected)	- Janke	et
automatic update	•	
✓ ✓ ▲ airways	oxyger	n
✓ ✓ ▲ breathing		
V V A central pulse	-	
V V A distal pulse		
Secondary survey		Apply Cancel survey
✓ ▲ blanket		
	, <u>,</u> ,	▲ inways
		breathing central pulse

### Graphical search & replace to remove duplicates



81% of the patients are treated in the correct order.

The largest deviation is still the airway and breathing being out of order,

but there are also instances where the circulation is checked too early.



## Correct procedure only 48% + description of the 27 variations



Carter, E., Burd, R., Monroe, M., Plaisant, C., Shneiderman, B. Using EventFlow to Analyze Task Performance During Trauma Resuscitation In Proceedings of the Workshop on Interactive Systems in Healthcare (WISH), 2013



## Goal= How are asthma meds used?

## 15M US Army records

 $\rightarrow$  only 182,000 records with asthma diagnostic (~1%)

### ightarrow only the asthma medication events

Further reduction # of events and unique complete sequences by 33%

Extract from a database first Interactively analyze sample within VA tool → strategy Repeat with larger dataset

#### Another example Analysis of prescription patterns of asthma medication in collaboration with Army PharmacoVigilance Center



This case study is described in: http://hcil2.cs.umd.edu/trs/2014-26/2014-26.pdf

Catherine Plaisant, Megan Monroe, Tamra Meyer, Ben Shneiderman

Interactive Visualization (a book chapter in Big Data and Health Analytics, Marconi, K. and Lehman, H. (Eds), CRC Press - Taylor and Francis, pp 243-262, 2014.



# Asthma Therapy Example



### PATIENT X

#### Acronyms

ICS: Inhaled Corticosteroid LABA: Long-acting Beta Agonist LABA + ICS: LABA combined with ICS

## Search and Replace: Identifying the Index LABA



### Limit to the 3 Months Surrounding the Index LABA



## Find Patients with ICS Before and After LABA



From: Gigi Lipori Date: Friday, April 24, 2015 at 11:45 AM

Got everything to work. Worth its weight in gold (again).

All are er/la opioid scripts

That would be a pain in the neck to code up.







## 20 case studies



### 20 case studies



Du et al. (2017) Coping with Volume and Variety in Temporal Event Sequences: Strategies for Sharpening Analytic Focus *IEEE Transactions Visualization and Computer Graphics* 

## **Event Analytics**

to:

review the data from individual records search for temporal patterns of interest summarize all the event sequences perform data transformations select cohorts of interest for further studies

### Many other projects in event analytics



# Thank you

#### Thank you to U. of Maryland colleagues

Ben Shneiderman Fan Du, Sana Malik Megan Monroe, Krist Wongsuphasawat, David Wang

#### + all case study partners



# hcil.umd.edu/eventflow

## plaisant@cs.umd.edu



#### Tableau Desktop: Free to download, easy to use



#### Analysis at the speed of thought

See the drag-and-drop analytics solution built for speed and ease of use. Start building the analytical skills employers are looking for in today's datadriven workplace.



#### EventFlow: Visual Analysis of Temporal Event Sequences and Advanced Strategies for Healthcare Discovery

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#### **CoCo: A Visual Analytics Tool for Comparing Cohorts of Event Sequences**

vs. lived died CoCo | Cohort Comparison Version 0.5 About (HCIL) Choose Files 2 files Human-Computer Interaction Lab, University of Maryland, College Park NUMBER OF RECORDS CONTAINING SEQUENCES (30) COHORT OVERVIEWS **P-VALUE & METRIC FILTERS** <sup>(C)</sup> Time C Frequency % Prevalence 0.0 Legend 1,000 died (1178) **[**--] l-I consecutive non-consecutive ■≤ 0.01 2/26/27 14/14 4/4 2/2800 -0/0 **■**≤ 0.05 1/1 3/6 4/4 0/0 - 000 pi □> 0.05 2/4 0/81 6/6 3/3 1/1 400 -SORT Default (p-value, difference) P-value Difference ∆ died 200 lived (1178) ∆ lived % died % lived Sequence Length 0 400 600 lived 800 1,000 200 Ι. P-VALUE CORRECTION SEQUENCE LENGTH Default Bonferroni 2 3 5 6 7 8 9 10 4 **RESULTS (48/155)** SEQUENCE DETAILS | NORMAL FLOOR BED -> ICU Event Types (7) died lived This sequence takes 1.9 times longer in lived than died (p=0.0013). ICU Emergency •• lived lived 400 7 died died Exit Normal Floor Bed ••• 2300· 323 806 Aspirin B200 3d 1h 5d 18h ••---Average 止 100 Min 56m 3h 17m Max 2w 2d 6w 3d 0 3d 0 5d Northal Root Red 3w 2d Duration 0 1w 4d 4w 6d 1h 18h St. Dev. 4d 4h 1w 3d 0 OTHER RESULTS 2x •• **...** 

Jump to: Participants | News | Software | Publications | Related Projects

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#### hcil.umd.edu/coco

#### **EventAction: Visual Analytics for Temporal Event Sequence Recommendation**



#### SUMMARY

**Human-Computer** 

Interaction Lab University of Maryland

HCIL

EventAction is a prescriptive analytics interface designed to **present and explain recommendations of temporal event sequences**. EventAction provides a visual analytics approach to (1) identify similar records, (2) explore potential outcomes, (3) review recommended temporal event sequences that might help achieve the users' goals, and (4) interactively assist users as they define a personalized action plan associated with a probability of success. EventAction's usage scenarios include **student advising**, **treatment formulating**, **customer retention**, and **sports coaching**.

#### hcil.umd.edu/eventaction

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