Role of Interaction Design & Data Science in Crisis Response

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June 11, 2020

CS-E4002: Human-Centred Research and Design in Crisis

Aalto University

Ketju

Designing the Ketju Contact Tracing App: Interaction Design, Technology and Privacy Implications

11.6.2020

Karri-Pekka Laakso, Reaktor



Reaktor

futurice



FRAKTAL

In cooperation with







ketjusovellus.fi

- Officially started on Mon 23rd March, 2020
- 6 designers
 - 3 from Reaktor (UX, graphic, UX) <= me
 - 3 from Futurice (UX, graphic, UX)



Karri-Pekka Laakso M.Sc. (eng), Helsinki University of Technology

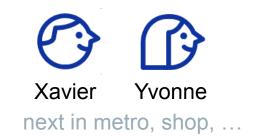
#UX #interaction_design #field_studies #visualizations
#scala #js #java #html_css #node #bash
#public_speaking #teaching
#family #gymnastics #acrobatics #orienteering #piano

Social Distance









Who can Susan recall / name?







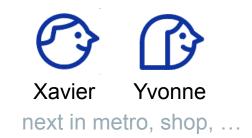


Who would Susan share with?









Who knows Susan by name?

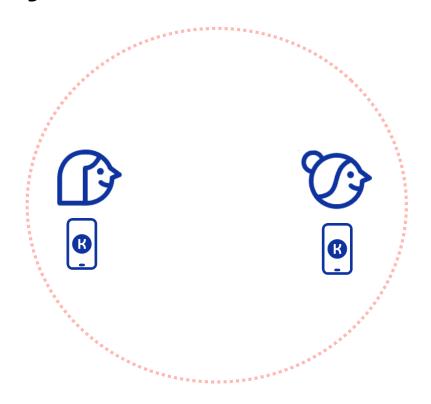




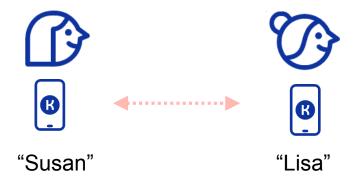




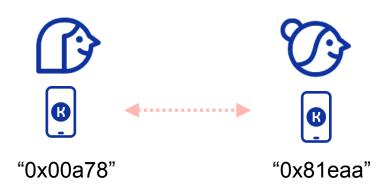
Location



- Location
- Bluetooth contacts



- Location
- Bluetooth contacts
- Anonymous Bluetooth contacts

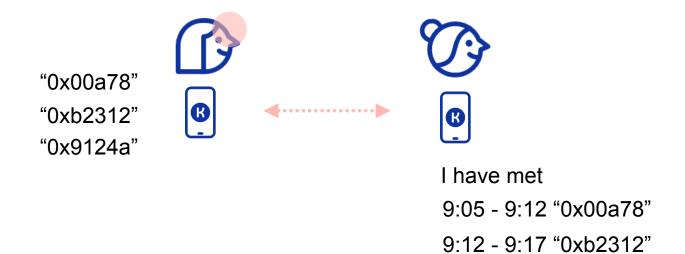


- Location
- Bluetooth contacts
- Anonymous Bluetooth contacts
- Anonymous Bluetooth contacts changing in time (DP^3T, Google Apple Exposure Notification API)



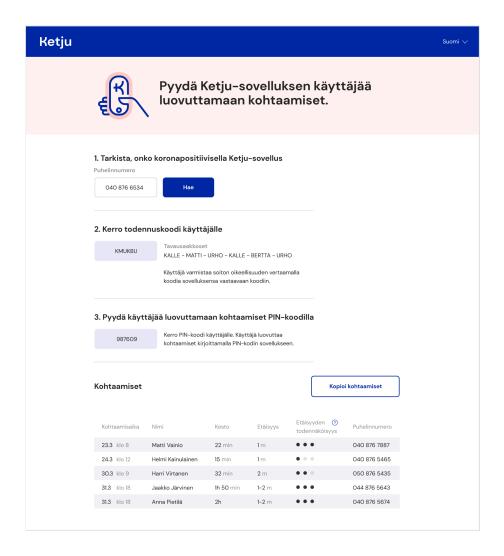


"0x00a78"	"0x81eaa"
"0xb2312"	"0xa292f"
"0x9124a"	"0x93d51"
	"0x00a78" "0xb2312" "0x9124a"



Centralized model

- Authorities know the exposed immediately
- Gather names & phone numbers=> call, interview, quarantine
- Privacy?



Distributed model







1) Citizens load the Ketju app from the App Store and put it on every time they leave home.



2) When users A and B meet, their phones register a contact when are close enough for time long enough (for example <5 m, >15 min).

C is too far away to register a contact.



3) The app stores the contact totally anonymously: no personal or location information is stored.



4) A has symptoms and the doctor order a test.



5) The doctor calls to tell that the test result is positive and asks, if A uses the Ketju App.



6) The doctor gives A a PIN code, which A enters to the Ketju app. The app informs the central server that A has been infected.



7) The app of B notices that one of its contacts has reported a positive test result. The app tells B about the exposure and that she should contact the local health center.

C has had no contact with A and thus nothing happens.



8) Only the user knows about the possible infection, since the data at the server and at the phone in anonymous.

Distributed model

- Susan tells the system that she is sick (voluntary)
- Lisa gets a notification
 - => take a test (voluntary)
 - => self-quarantine (voluntary)
- Verification of exposure?
- Granularity of exposure time?
- Contact tracing?



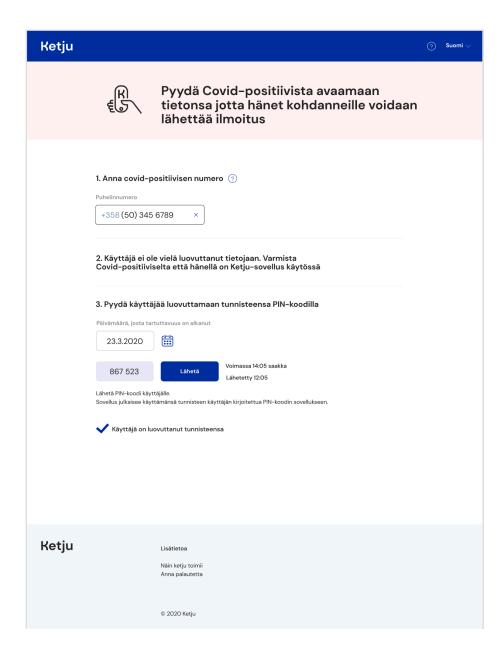


Distributed model

Authorities can only give the release code

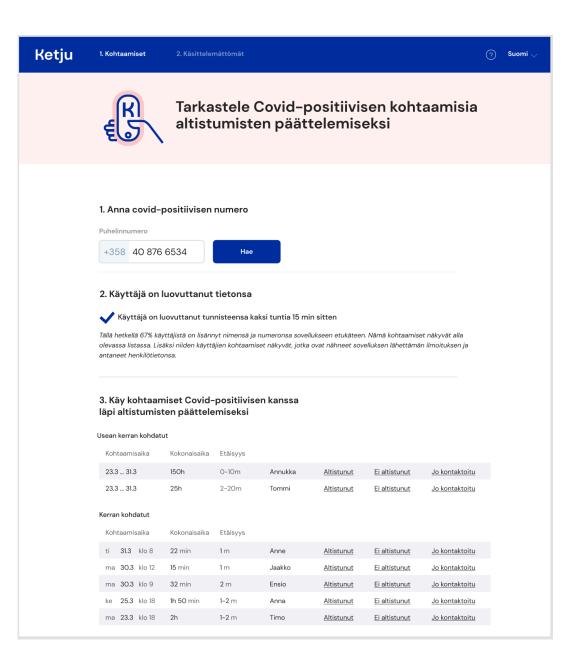
- => no help for contact tracing
- => no official quarantine (≠ self quanrantine)



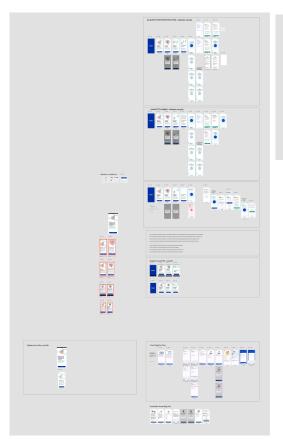


Hybrid models

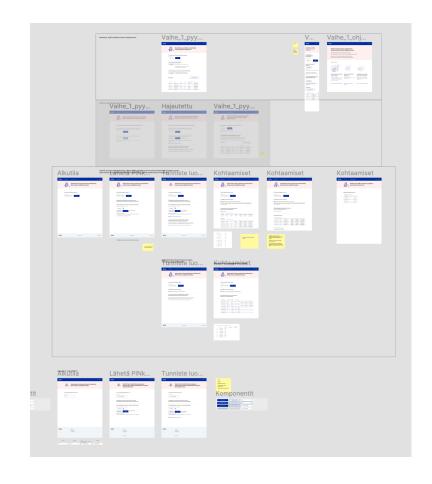
- People voluntarily release information to authorities, e.g.
 - contacts with a sick person
 - exposures of a sick person
 - their contact info
 - ...
- Legal issues
- Privacy issues
- Gapple API issues



Iterations







Chaotic surroundings

contract tracing

doctors

politics

infectious diseases

communications

law

media

design

bluetooth

users

protocols

databases

visualizations

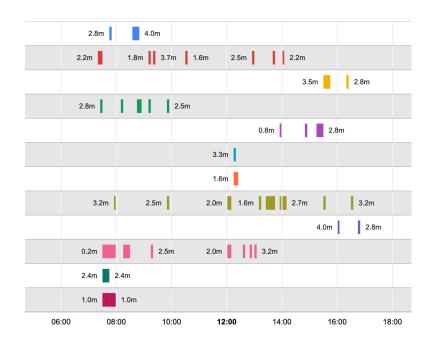
Do what is needed

How to manipulate DBs from pilot users

- 1. Collect new DBs to a directory
- Convert them, so that they don't produce cache files, and tools/convert-dbs.sh Ketju*
- 3. Move the files to the iterations/pilot/all directory
- Rebuild the iterations/pilot/latest directory to contain tests dir) => only the files with the latest timestamps are
- Merge the data into one db: ../../tools/merge-dbs.sh al contacts.sqlite
- 6. (optional step at this point) anonymize the data with .../.
- Create a JS file of all the data and update the html files you
 updated device-data.js and the data updated to the htm
 timelines.html

How to generate screenshots for pilot use

- Create the timelines (in iterations/pilot) ../../tools/c 2020-05-22.html
- 2. Generate the screenshots (in calibration-tests) node t

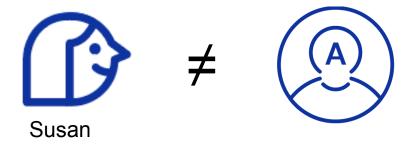


```
<script type= text/javascript >
  google.charts.load('current', {'packages':['timeline']})
      google.charts.setOnLoadCallback(initialize)
//--START-DEVICE-DATA
const deviceData = [...]
//--END-DEVICE-DATA
//--START-MANUAL-DATA
const manualData = [...]
//--ΕΝΩ-ΜΔΝΙΙΔΙ -ΠΔΤΔ
      const data = deviceData.concat(manualData)
      const firstDay = '2020-05-18'
      const lastDay = '2020-06-07
      const pairs = () => Object.values(data
          .map(([a, b]) \Rightarrow ((a < b) ? [a, b] : [b, a]))
          .reduce((acc, [a, b]) => { acc[a+'-'+b] = [a, b]; return acc }
         .sort((a, b) => a.toString().localeCompare(b.toString(), undef
      function initialize() {
        const dayRange = (isoStart, isoEnd) => {
         const start = new Date(isoStart + ' 12:00')
          constrend = new Date(isoEnd + ' 12:00')
          let range = []
          ·let ·day ·= ·start
          while (day <= end) {
            range.push(new Date(day))
            day.setDate(day.getDate() + 1)
         return range
        const daySelector = $('#startDay')
        const fiFmt = Intl.DateTimeFormat('fi', {year: 'numeric', month:
        dayRange(firstDay, lastDay).forEach(day => (
         $(`<option value="${day.toISOString().substring(0,10)}">${fiFm
             appendTo(daySelector)
        daySelector.change(drawCharts)
        drawCharts()
      function drawCharts() {
       $('#timelines').empty()
        const dataTable = new google.visualization.DataTable()
        dataTable.addColumn({ type: 'string', id: 'Origin' })
```

organize 250 database files

produce 330 screenshots for users participate in writing the report

The power of realistic cases



Everything is easier: talking, deducing, ...

Cost / benefit

```
x% install

* y% use

* z% contact noticed

* i% user notices

* j% contacts the authority

* ...
```

vs. cost of implementation & maintenance?

Ketju

Questions?

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@kplaakso

Privacy-preserving contact statistics collection using COVID-19 contact tracing apps

Antti Honkela¹ and Tejas Kulkarni²

Finnish Center for Artificial Intelligence FCAI

1 University of Helsinki

2 Aalto University

Human-Centred Research and Design in Crisis
11 June 2020



Outline

A cartoon of an epidemic

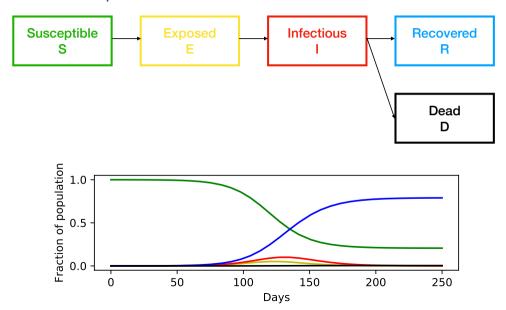
Contact tracing apps

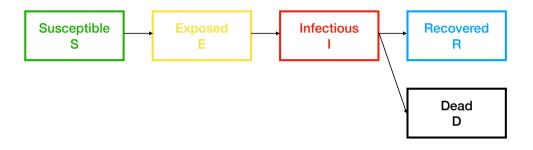
Privacy-preserving collection of contact statistics

Disclaimer: while I talk about mathematical epidemic models, I am not an epidemiologist. Listeners beware.

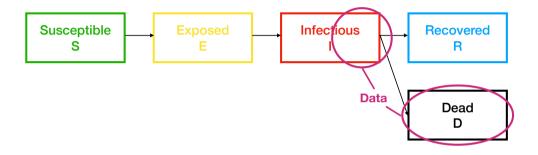


- ▶ Population divided to classes of individuals based on infection status
- ▶ Infection spreads when S has sufficiently strong contact with I
- Current growth rate of the epidemic measured by reproductive number $R_e \approx$ ratio of new infections over recoveries





- Modelling by fitting the curve to observed confirmed cases / hospitalisations / deaths / ...
- ► Estimation delay: changes in infection rate only show in tests and hospitalisations after a week or more



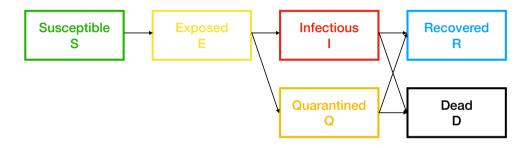
- Modelling by fitting the curve to observed confirmed cases / hospitalisations / deaths / ...
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Managing an epidemic



- ▶ Epidemic management activities aim at limiting contacts between S and I
 - Social distancing limits all contacts
 - ► Contact tracing aims to quarantine exposed individuals before they become infective

Managing an epidemic

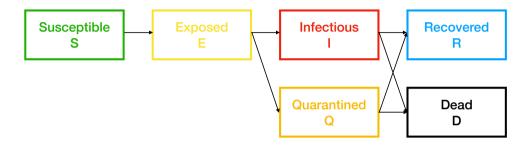


- ▶ Epidemic management activities aim at limiting contacts between S and I
 - Social distancing limits all contacts
 - ▶ Contact tracing aims to quarantine exposed individuals before they become infective

Contact tracing and mobile apps

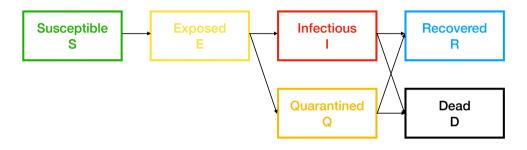
- ► Classical contact tracing works remarkably well, but is very labour intensive
- Many countries are developing mobile apps to assist the activity
- Privacy is an absolute requirement
- ► Typical mode of operation (e.g. DP-3T, Google/Apple):
 - No location data gathered (low utility, bad privacy)
 - Use Bluetooth to record random identifiers broadcast by nearby devices
 - Identifiers change relatively frequently
 - Recorded recent contacts shared only when user is diagnosed positive
- ▶ Discussion to decide between decentralised vs. centralised mode of operation

Contact tracing app theory



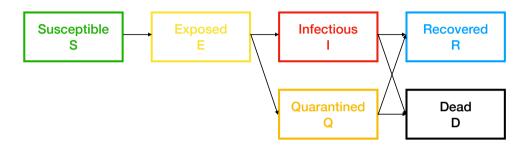
▶ Assuming 50% of population use the app and all detected exposed individuals are perfectly quarantined, we prevent 25% of potential future infections

Contact tracing app theory



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Contact tracing app theory



- ▶ Assuming 50% of population use the app and all detected exposed individuals are perfectly quarantined, we prevent 25% of potential future infections
 - E.g. reduce $R_e = 1.2 \rightarrow R_e = 0.9$
- ▶ Over 70% users needed for theoretical 50% efficiency
 - E.g. reduce $R_e = 1.8 \rightarrow R_e = 0.9$

Opportunity

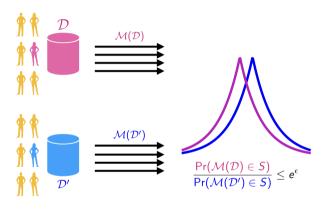
- ► Contact tracing apps collect data of close contacts in the population
- ► Collecting statistical information of contact frequencies would provide data for epidemic modelling as well as monitoring and planning other interventions (e.g. school and business closures)
- ▶ While such statistical information is not very sensitive, strong privacy protection is still necessary; collection should be opt-in
- ▶ Such data likely to be useful with fewer users: even at 30% use we would expect to see 30% of contacts of each user and can correct that from known rate of users

Asking sensitive questions: Randomised response (Warner, 1965)

Assume respondents are instructed to answer a potentially sensitive query (e.g. were you in contact with more than 10 individuals yesterday?) as follows:

- 1. Flip a coin in secret.
- 2. If tails, then respond truthfully.
- 3. If **heads**, then flip a second coin and respond "Yes" if heads and "No" if tails.
- ▶ Outcome: the answer is flipped with probability $\frac{1}{4}$
- Everyone gets plausible deniability: "It was just the coins"
- Statistics can be estimated from population responses by compensating for the noise
- ► Probabilistic loss of privacy: "Yes" response makes it more likely your true response was "Yes"

Differential privacy (DP; Dwork et al., 2006)

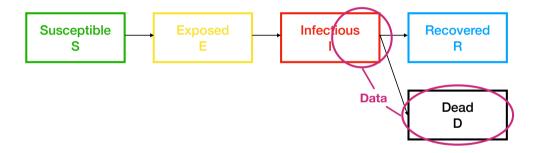


- ▶ Provides protection against adversaries with side information
- Degrades gracefully under repeated use
- Invariant to post-processing

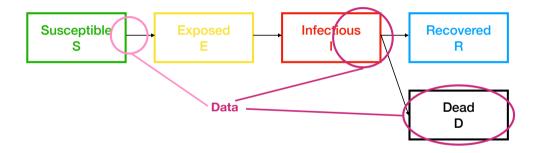
Proposal: Locally DP collection of contact statistics

- Using local DP (LDP) to collect population histogram of number of contacts
- Data are anonymised before they leave the user device
- Privacy guarantee: even if you see a user's report knowing it came from her, you can only guess what she answered
- ► Rapidly changing identifiers make collection of full daily statistics difficult
 - ► Collect e.g. maximum number of contacts over 30 min period each day instead
- ► Observed data fed to a Bayesian model for denoising, integrating responses from multiple days in a probabilistic model
 - Produces real-time data on behaviour changes and effects of changes to interventions and guidelines

Extended data collection



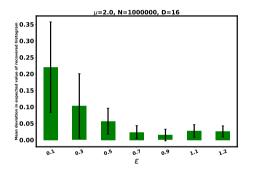
Extended data collection

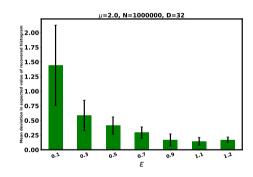


More about privacy

- Caveat: theoretical privacy guarantees will degrade over repeated data collection
 - Using fresh data for each report mitigates the impact
 - Ethical review of costs vs. benefits still needed
- Introducing secure shuffling to eliminate linking reports to individuals tightens the privacy guarantee by approximately factor \sqrt{n} for n users
 - ▶ Basically eliminates privacy concerns even after repeated collection
 - ▶ NB: even breaking the shuffler would not completely compromise the privacy

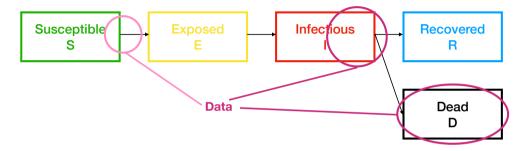
Preliminary simulation results





Error in estimating the mean of a simulated population using LDP histogram estimation with D bins. Simulated using geometric distribution with true mean 2.0. Bars show MAE of 10 repeats, error bars show std over the repeats.

Conclusion



- ▶ COVID-19 pandemic has just started, still long way to vaccine or herd immunity
- Daily contact statistics observed by contact tracing apps could provide direct measures of the most important spreading mechanism, enabling more timely modelling
- ▶ Differential privacy provides the means to collect these data under strong privacy guarantees to the users

Next Tuesday:

Why Contact Tracing is Neither Secure or Privacy-Preserving

Janne Lindqvist,

Prof. at the Department of Computer Science, Aalto University



