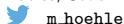


Risk Scoring in Digital Contact Tracing Apps

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09 Dec 2020



Overview

- 1 Introduction
- 2 The Google and Apple Exposure Notification Framework
- 3 Risk Scoring
 - Transmission Risk Level
- 4 Discussion

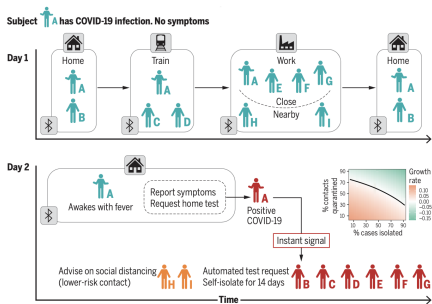
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Contact tracing

- Contact tracing is the process of identifying persons who had contact with an infected person in order to test, isolate and trace onwards
- Contact tracing is an important non-pharmaceutical control measure
- In an influential paper, Ferretti et al. (2020) argued that a digital contact tracing app could speed up contact tracing s.t. $R(t) < 1$ would be possible



Source: Fig. 4 of Ferretti et al. (2020)

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GAENF (1)

- The Google and Apple exposure notification framework (GAENF) enables privacy-preserving contact tracing on mobile devices based on Bluetooth Low Energy
- Each device continuously emits so called Rolling Proximity Identifiers (RPI, changing every ~ 20 min) derived cryptographically from a daily temporary exposure key (TEK)
- Each device regularly (every ~ 5 min) scans all visible RPIs and stores these locally together with their signal strength
- If a user is tested positive for COVID-19, they can volunteer to upload their TEKs of the past 14 days to a central server

GAENF (2)

- An uploaded TEK is called a diagnosis key
- All users download the available diagnosis keys daily and compare the corresponding RPIs to the locally stored RPIs
- The GAENF is used by country specific COVID-19 Apps in several European countries, e.g., Switzerland, Latvia, Denmark, Italy, Poland, Ireland, Germany, UK, ...
- Since the summer the digital contact tracing apps also work between (many of) the countries

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Risk Scoring in the GAENF (v1)

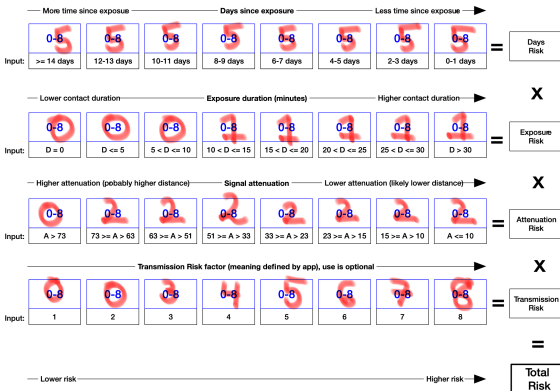
- Consider a specific user. Let I be the user's subset of all stored exposures within the past 14 days which originate from keys known to be COVID-19 test positive
- In the GAENF (v1) we compute the **total risk** for each exposure with a test positive, i.e. for each element $i \in I$, by

$$TR(i) = \text{duration}(i) \times \text{attenuation}(i) \times \text{TRL}(i) \times \text{days}(i),$$

where $\text{TRL}(i)$ is the so called **transmission risk level**.

- The GAENF operationalizes all RHS functions by an integer value between 0 and 8

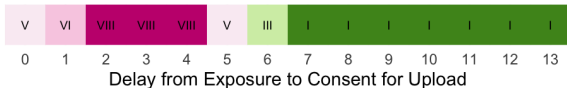
Risk Calculation per Exposure Key



Source: Corona Warn App, Solution Architecture, Fig. 12 (Apache License 2.0) + current config annotated

Transmission Risk Level

- The transmission risk level can be used to enhance the classification with epidemiological information about the potential source



- The following derivations are closely following the document *Epidemiological Motivation of the Transmission Risk Level* (CWA Team, 2020) written as part of the German CWA development¹

¹Reflects COVID-19 knowledge as of May-Jun 2020

Sequence of events for an infected person A

- $T_E = T_{\text{infection}}$: transmission of SARS-CoV-2 to an exposed person A from some unknown source
- $T_I = T_{\text{infectious}}$: start of the infectious period in person A , i.e. A is able to infect others
- $T_S = T_{\text{symptoms}}$: onset of symptoms in person A (also referred to as DSO, day of symptom onset)
- $T_P = T_{\text{sampling}}$: time of sampling of person A
- $T_R = T_{\text{result}}$: time of A obtaining the positive test result
- $T_U = T_{\text{upload}}$: time where person A uploads the positive test result to the system (aka. DU, day of upload)

Information about Symptoms in A at DU

Four cases:

- 1 Knowledge of DSO in A
- 2 Knowledge that A was symptomatic, but DSO not available, only DU
- 3 Knowledge that A was not symptomatic at or before DU (completely asymptomatic or pre-symptomatic)
- 4 Unknown symptom status at DU

Note:

The previously shown TRL corresponds to Case 4, which is computed as a marginalization over Cases 1-3.

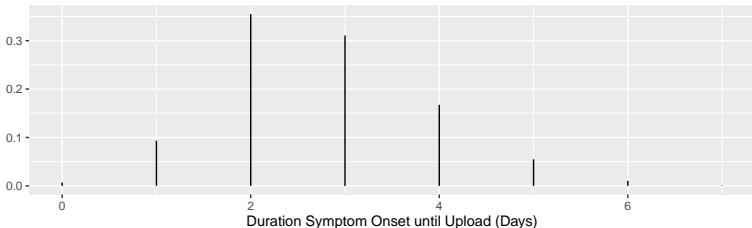
Case 1: Duration between DSO and DU

- Consider the durations $T_U - T_R$, $T_R - T_P$ and $T_P - T_S$
- We find the distribution of

$$T_U - T_S = (T_U - T_R) + (T_R - T_P) + (T_P - T_S)$$

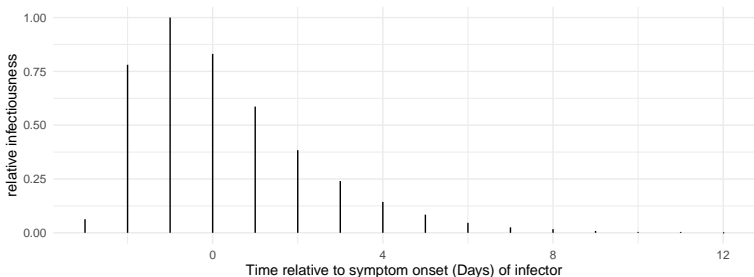
by a convolution of the 3 durations

- Given informed assumptions we obtain the following PMF



Infectiousness (1)

- He et al. (2020a) provides a distribution for time of exposure in B relative to symptom onset in A
- Distribution is used as proxy for the infectiousness of A, $v_A(d)$



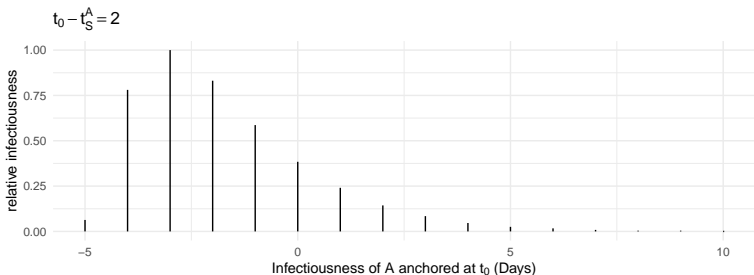
- Recently, a correction appeared moving the graph further to the left (He et al., 2020b).

Infectiousness (2)

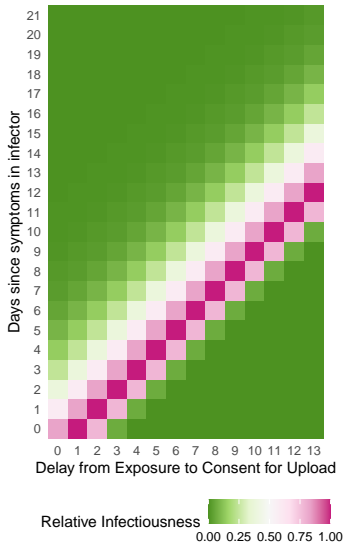
- At time now, t_0 , the onset of symptoms in the primary case A happened $t_0 - t_S^A \geq 0$ days ago.
- Relative infectiousness of A anchored s.t. time zero is at t_0 and a function of how many days d ago the contact between A and B was:

$$\lambda_A^*(d) = v_A(-d + (t_0 - t_S^A))$$

- Example: Plot of the infectiousness profile, if DSO happened 2 days before upload.



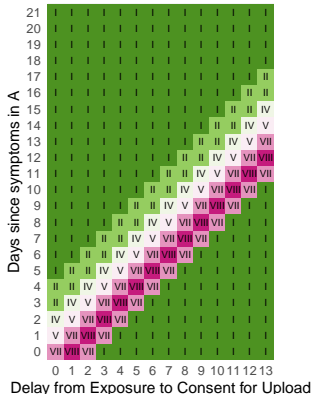
Infectiousness (3)



Infectiousness (4) - Transmission risk level

Discretization of the $[0, 1]$ -scale into 8 equal-sized intervals:

$$\lambda_A(d) = l \Leftrightarrow \frac{l-1}{8} \leq \lambda_A^{(*)}(d) < \frac{l}{8}, \quad \text{where } l \in \{1, \dots, 8\}.$$



Overall Risk (1)

- Exposures with a total risk below the minimum risk score of 11 are set to zero
- A weighted summation of time spent in 3 attenuation buckets (distance: far, intermediate and close) is computed

$$wt_j = \max \left(30, \sum_{i \in I^*} \mathcal{I}(l_j \leq \text{attenuation}(i) < u_j) \times \text{duration}(i) \times w_j \right),$$

where $I^* = \{i \in I : \text{TR}(i) > 0\}$, $w_1 = 0$, $w_2 = 0.5$ and $w_3 = 1$.

- The limits (l_j, u_j) of the three classes are chosen at $[63, \infty)$, $[55, 63)$ and $[0, 55)$ dB.
- The **Exposure Score** is then calculated as $\sum_{j=1}^3 wt_j$.

Overall Risk (2)

- Next, a normalized total risk score defined as

$$\text{Normalized Total Risk Score} := \frac{\max_{i \in I} \text{TR}(i)}{50}.$$

- The combined risk score is now given as

$$\text{Combined Risk Score} = \text{Exposure Score} \times \text{Normalized Total Risk Score}$$

- A warning is given by the app, if

$$\text{Combined Risk Score} > 15.$$

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Summary...

- Mathematical parametrization of the GAENF risk classifier in an app downloaded 23.5 mio times²
- Contact tracing apps are not the silver bullet for COVID-19 control. However, they provide an important supplement to manual contact tracing and speed up the turn-around time for test results
- The success of the app heavily depends on the widespread voluntary use in the population
- One component of winning as many users as possible is **transparency**
 - Source code and documentation of the CWA app and server back-end is available from GitHub
 - Description of the risk scoring is available at several audience levels (user, epidemiologist, mathematician)

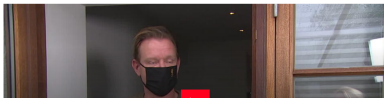
²As of 2020-12-04

A not so good example... (from Denmark)

 PENGE

Jens fik corona, men smittestop-app gav ikke hans kone besked

Problem med appen Smittestop er et tillidsbrud, mener De Konservative.



TEKNOLOGI

Myndigheder erkender fejl i Smittestop-app

De allernærmeste kontakter har ikke i alle tilfælde fået besked om smittefare. Uvist, om fejlen har været der fra starten af.



 PENGE

Fejl i Smittestop vokser i omfang: Kollegaer og venner kan have misset beskeder

Der er en masse smittekæder, der ikke bliver brudt på grund af fejl i appen, vurderer epidemiolog.



... it turns out that NetCompany, which created the app for the Ministry of Health and the Elderly, has either misunderstood or been misinformed about a key element of the technology developed by Apple and Google - an element that determines who is notified of the corona infection.

Source: DR Nyheter Sep 2020

Opinions...

- There is a trade-off to make between complexity and robustness in the risk scoring model
- Living up to scientific standards is more than code on GitHub:
 - transparency is needed both at the high-level as well as at for detailed mathematical descriptions
 - evidence is continuously evolving and has to be updated
 - engagement with the user community is an opportunity for peer-review
 - scientific peer-review publication assures scientific standard
 - a plan is needed how to collect data about the use and impact of the app (data protection and privacy preservation is IMO not an excuse to do nothing)
- Details and an example of the GAENF risk calculation can be found in a recent blog post (Höhle, 2020)

References I

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He, X., Lau, E. H. Y., Wu, P., Deng, X., Wang, J., Hao, X., Lau, Y. C., Wong, J. Y., Guan, Y., Tan, X., Mo, X., Chen, Y., Liao, B., Chen, W., Hu, F., Zhang, Q., Zhong, M., Wu, Y., Zhao, L., Zhang, F., Cowling, B. J., Li, F., and Leung, G. M. (2020b). Author correction: Temporal dynamics in viral shedding and transmissibility of covid-19. Nature Medicine.

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Höhle, M. (2020). https://staff.math.su.se/hoehle/blog/2020/09/17/gaen_riskscoring.html. Blog post.

Appendix

Appendix

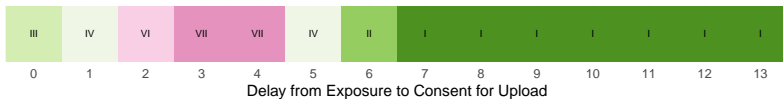
Case 2 (1)

- We know that A is symptomatic at time of upload, but do not know the exact DSO
- We can infer a probability distribution for the DSO T_S^A from knowledge about the duration between DSO and upload given by and knowing the date of upload t_0
- Marginalizing over the DSO

$$\begin{aligned}\lambda_A^{(*)}(d, T_S^A \leq t_0) &= \mathbb{E}_{T_S^A} \left[\lambda_A^{(*)}(d, T_S^A = t_S^A) \mid T_S^A \leq t_0 \right] \\ &= \sum_{t_S^A = t_0 - 13}^{t_0} \lambda_A^{(*)}(d, T_S^A = t_S^A) \cdot f_{T_U - T_S}(t_0 - t_S^A)\end{aligned}$$

Case 2 (2)

Resulting risk levels:



Case 3 and 4

See the details in CWA Team (2020).