

Detecting inference attacks involving raw sensor data*

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Inference involving raw sensor data: A motivating example

Involving personal data

e.g., Age, Sex $\xrightarrow{\text{Cardiovascular disease}}$ $\text{Pr}(\text{CVD}) < 70\%$ ✓

Involving raw data

e.g., Raw data $\xrightarrow{\text{Activities}}$ Activities ✓

Involving both personal & raw data

Age, Sex + Activities $\xrightarrow{\text{Cardiovascular disease}}$ $\text{Pr}(\text{CVD}) \geq 70\%$ ✗



Yasuhiko Kubota et al. "Physical Activity and Lifetime Risk of Cardiovascular Disease and Cancer". In: *Medicine and science in sports and exercise* 49.8 (Aug. 2017), pp. 1599–1605. PMID: 28350711



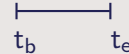
Oresti Banos et al. "Design, Implementation and Validation of a Novel Open Framework for Agile Development of Mobile Health Applications". In: *BioMedical Engineering OnLine* 14.2 (Aug. 13, 2015), S6

The user knowledge model: Metadata knowledge unit obtained by issuing queries

User knowledge

Modeled as $\langle a, c \rangle$, where

- a is an attribute identifier, e.g., ■, ■, ■
- c is a selection condition, e.g., (t_b, t_e) a temporal interval



Example queries

Q₁ SELECT ■, ■, ■
WHERE INTERVAL (t_2, t_3)

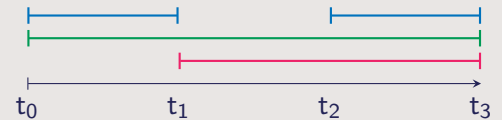
Q₂ SELECT ■, ■
WHERE INTERVAL (t_1, t_2)

Q₃ SELECT ■, ■
WHERE INTERVAL (t_0, t_1)

Query history log (QHL)

Tracks each user knowledge metadata.

After issuing Q₁₋₃, the QHL is:



Inference detection system (InfDS): For inference channels involving raw sensor data

Overall workflow

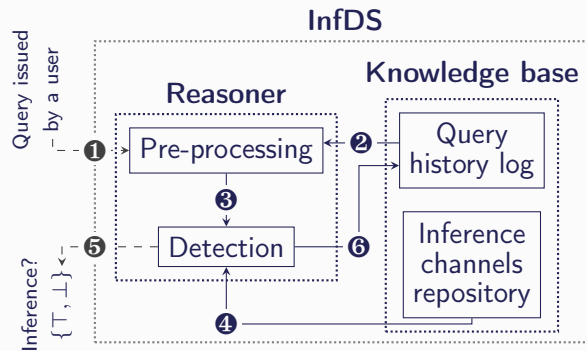
When a new query ① is issued by a user u , the system retrieves the previous knowledge of u ②, i.e., $QHL(u)$, and merges all the metadata into one set ③. The InfDS then checks if ③ can exploit one of the registered inference channels ④ and notifies the detection result ⑤. In case no inference is detected, $QHL(u)$ is updated with the new metadata ⑥

Query based filtering (Qbf)

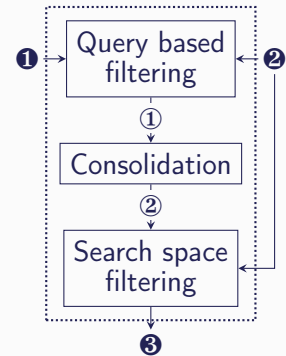
Based on the query knowledge ①, extracts from the QHL ② only the knowledge that can be consolidated ①

Search space filtering (Ssf)

Extracts from the QHL ② the knowledge related to the consolidation results ② that the detection must process ③



Pre-processing



Consolidation

Aims to reduce the QHL size.
Rule example for temporal intervals:

$$\begin{matrix} t_{b_i} & t_{e_i} \\ | & | \\ t_{b_j} & t_{e_j} \end{matrix} \Rightarrow \langle \text{red square}, (t_{b_i}, t_{e_j}) \rangle$$

Hence ■ $\in Q_{1\&2} \Rightarrow \langle \text{red square}, (t_1, t_3) \rangle$

Detection

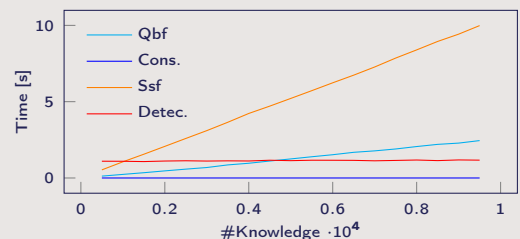
Searches knowledge using patterns:
 $\bigcup \langle \{ \text{red square}, \text{green square}, \text{blue square} \}, (t_b, t_e) \rangle$

Checks if it satisfies the constraints:

$$\exists \Delta t \geq 2s?$$

Performance evaluation

- Linear complexity based on the size of the user knowledge
- Demonstrates the feasibility on the MHEALTH use case
- The InfDS needs further optimization to be scalable



Additional information

Work supported by the DFH-UFA

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