

Smart App

instance, opening a door inevitably involves the door's movement, which could be captured by both a contact sensor and

Smart App

an acceleration sensor installed on the door and results in

Channel

two consecutive events. With increasing types of IoT devices

Cyber Part

deployed, physical-channel correlations can be pervasively

observed on many physical properties, such as illuminance,

User

User

Physical

Physical

Activity

Physical Part

Activity

Channel

Channel

Environment

power, sound, and temperature [39].

IoT Device

User Activity Channel. While user activities impose

Figure 3: Correlation channels.

changes on devices, device states also reflect user activities.

Thus, the user activity channel causes correlations between

We assume the IoT platform is not compromised. Like devices. For example, a TV being turned on typically implies other anomaly detection work [35,51,76], we assume there that the user is nearby, which should be captured by the are no or very few anomalies during training. We assume motion sensor. When a user returns home, there should be there are no malicious or conflicting rules in the installed consecutive events, such as "presence on" showing the user's smart apps; how to detect malicious logic [71] and conflicting proximity and "contact-sensor open" for door opening. rules [28,34] are two separate research problems, and there

4.2 Representation of Correlations

are existing solutions to them [28,71], including our prior work [33,34]. Gartner predicts that a typical household could An event reporting that the device A's attribute a should have more than 500 IoT devices by 2022 [72]. Given the be changed to the value a is denoted as $a(A)$ while a state

dense deployment in the near future, we exploit scenarios which indicates that the device B's attribute has the value where an IoT device has one or more other devices nearby b is denoted as $B(B)$ 2 We define two types of correlations.

to interact with, and propose to leverage them to detect a device's anomalous physical behaviors. We discuss the The event-to-event (e2e) correlation. It means that one

case of no interactive devices nearby in Section 8. Jamming

event should be followed by (denoted as another. For that blocks communications reporting IoT events can be example, given a motion sensor A and a light B, the e2e motion(A) easily detected due to session timeout or missing sequence correlation 'active

Eon
.switch(B)

means

the

event

numbers; we thus do not further discuss it.

_motion(A)

'active

should be followed by the event Eon -switch(B)

4 Correlations

The event-to-state (e2s) correlation. It means that

Devices deployed in the same home may correlate in the

one event arising implies (denoted as my) a state

form of co-present or temporally related events [35,39,45,68].

is true. For example, power(plug) switch(heater)

These correlations can be attributed to the execution of smart

means that, when the arises, the state

apps [29], physical interactions [39] or users' activities [45].

switch

As shown in Figure 3, we investigate the causes of these

Son

(heater)

should

be

true.

correlations and categorize them into three channels below.

For the representation of a correlation involving condi-

4.1

Correlation Channels

tions, its anterior event is combined with the conditions

using the "A" symbol. For example, \wedge Presence present

Smart App Channel. Smart apps not only directly cause

Eon

.switch(Light)

means the event Motion active if the condition

correlations between triggers and actions as programmed,

present Presence is true, should be followed by Eon ,switch(Light)

but also imply some extra correlations that should be consid-

We show in Section 5 that the two types of correlations,

ered. For example, the smart App "light follows me" [2] leads

despite their simplicity, are very effective in capturing rich

to the correlation between the motion sensor and the light,

semantic information and modeling the relations of devices

and also implies a possible correlation worth verification,

that correlate via different channels.

that is, "if the light is turned on, then the motion should be in

the active state". The implied correlation is true if the light is

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HAWatcher Design and Implementation

exclusively turned on by the smart app.

We first introduce the workflow of anomaly detection (Section 5.1), and then describe the major modules in HAWatcher, as shown in Figure 4: 1) Semantic Analysis (Section 5.2), 2) Correlation Mining (Section 5.3), 3) Correlation Refining (Section 5.4), and 4) Anomaly Detection (Section 5.5). For simplicity of description, without causing confusion we sometimes omit the device IDs and use the simplified notations E and sB.

different sensor devices can be affected by the same physical event and generate temporally correlated IoT events. For

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