HAWatcher: Semantics-Aware Anomaly Detection for Applified Smart Homes
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Abstract
As IoT devices are integrated via automation and coupled
(a)
Command
Valve Not
Room
with the physical environment, anomalies in an appified
Intercepted
Closed
Flooded
smart home, whether due to attacks or device malfunctions,
may lead to severe consequences. Prior works that utilize data
mm
mining techniques to detect anomalies suffer from high false
(b)
alarm rates and missing many real anomalies. Our observa-

Broken
Heater
Fire
Relay
Overheating
Hazard
tion is that data mining-based approaches miss a large chunk
of information about automation programs (also called smart
apps) and devices. We propose Home Automation Watcher
(HAWatcher), a semantics-aware anomaly detection system
(c)
Spurious
Presence
Unlock
for appified smart homes. HAWatcher models a smart home's
Presence On
Unlock App
Front Door
normal behaviors based on both event logs and semantics.
Given a home, HAWatcher generates hypothetical correla-
Figure 1: Examples of anomalies in a smart home.
tions according to semantic information, such as apps, device
types, relations and installation locations, and verifies them
Despite advances in appified smart home, there are grow-
with event logs. The mined correlations are refined using
ing concerns about its safety and security [41]. First, IoT de-

correlations extracted from the installed smart apps. The vices make it possible for cyber-space attacks to be extended refined correlations are used by a Shadow Execution engine to the physical world. As shown in Figure 1(a), the command to simulate the smart home's normal behaviors. During runof "close the valve" is maliciously intercepted, which may time, inconsistencies between devices' real-world states and cause room flooding. Second, very often a device malfuncsimulated states are reported as anomalies. We evaluate our tion is hardly noticeable until certain consequences arise. As prototype on the SmartThings platform in four real-world shown in Figure 1(b), an electronic heater controlled by a testbeds and test it against totally 62 different anomaly cases. smart app "It's too cold" [15] could result in fires because of a The results show that HAWatcher achieves high accuracy, broken relay (an electronically operated switch), which presignificantly outperforming prior approaches. vents the plug from shutting the power for the heater. Third, as IoT devices are chained together via automation [28,29,39], abnormal behaviors of one device might trigger undesired 1 Introduction

actions of another, which further exaggerates the impact of anomalies. As shown in Figure 1(c), a smart lock that auto-With the rapid growth of Internet of Things (IoT), smart matically unlocks upon the resident's presence is unlocked homes gain booming popularity. As predicted by Gartner,

due to a fake event of the presence sensor.

there will be more than 500 IoT devices deployed in a typical To address these concerns, many anomaly detection syshousehold by 2022 [72]. IoT devices become increasingly intems [30,35,54,56,60,68,76] utilize data mining techniques to tegrated, thanks to IoT platforms such as \$martThings [21], profile the system's normal behaviors and report events that Homekit [47], and OpenHAB [55]. These platforms provide deviate from profiles as anomalies. However, these works interoperability among home IoT devices by different venusually take event logs as inputs without fully considering dors, and allow them to work according to user-specified each event's semantics, which actually may be acquired from automation programs (also called smart apps).

smart apps, device types, and device functionalities. The lim-USENIX Association

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