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Android SmartTVs Vulnerability Discovery via Log-Guided Fuzzing

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Why is SmartTV Security Important? A *Few Reasons*

Smart TVs

Account for the largest market share of Home IoT devices

Expected to achieve a market value of 253 billion USD by 2023

Plethora of attack vectors:

- **Physical channels: e.g., sending crafted broadcast signals**
- **Malware: SmartTV users can download SmartTV-specific Apps**

Broad Spectrum of Attack Consequences: Cyber + Physical

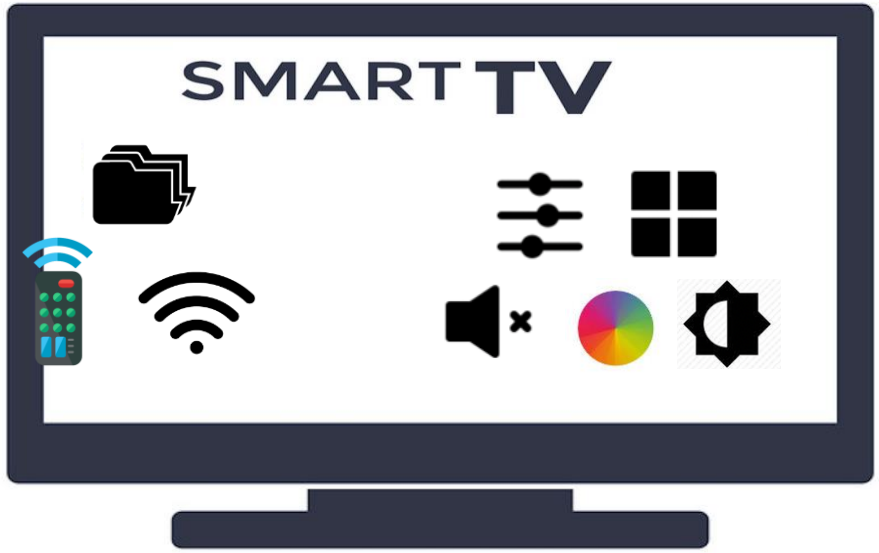
Goal

- Perform a systematic security evaluation of Android SmartTVs.
- Focus on customization aspects, performed to tailor the original OS for the SmartTV functionalities.

Background

Android SmartTVs run a heavily customized version of AOSP:

- Additional hardware, system components.
- Custom Functionalities are exposed to system and app developers through *dedicated APIs.*
- The number of custom APIs is high (up to 101 in H96Pro).



SmartTV APIs can open the door to various damages if not properly protected.

These APIs execute in the context of highly privileged processes.

Motivating Example

- Xiaomi MiBox3 introduces a new native API **SystemControl. setPosition(x, y, w, h)**

SystemControl. setPosition(x, y, w, h)

Motivating Example

- Xiaomi MiBox3 introduces a new native API **SystemControl. setPosition(x, y, w, h)**
  - The API does not enforce any access control and has persistent impact across reboot.
  - With the SmartTV ransomware on the rise, such APIs can be exploited to mount DoS attacks.

SystemControl. setPosition(1000, 1000, 1000, 1000)

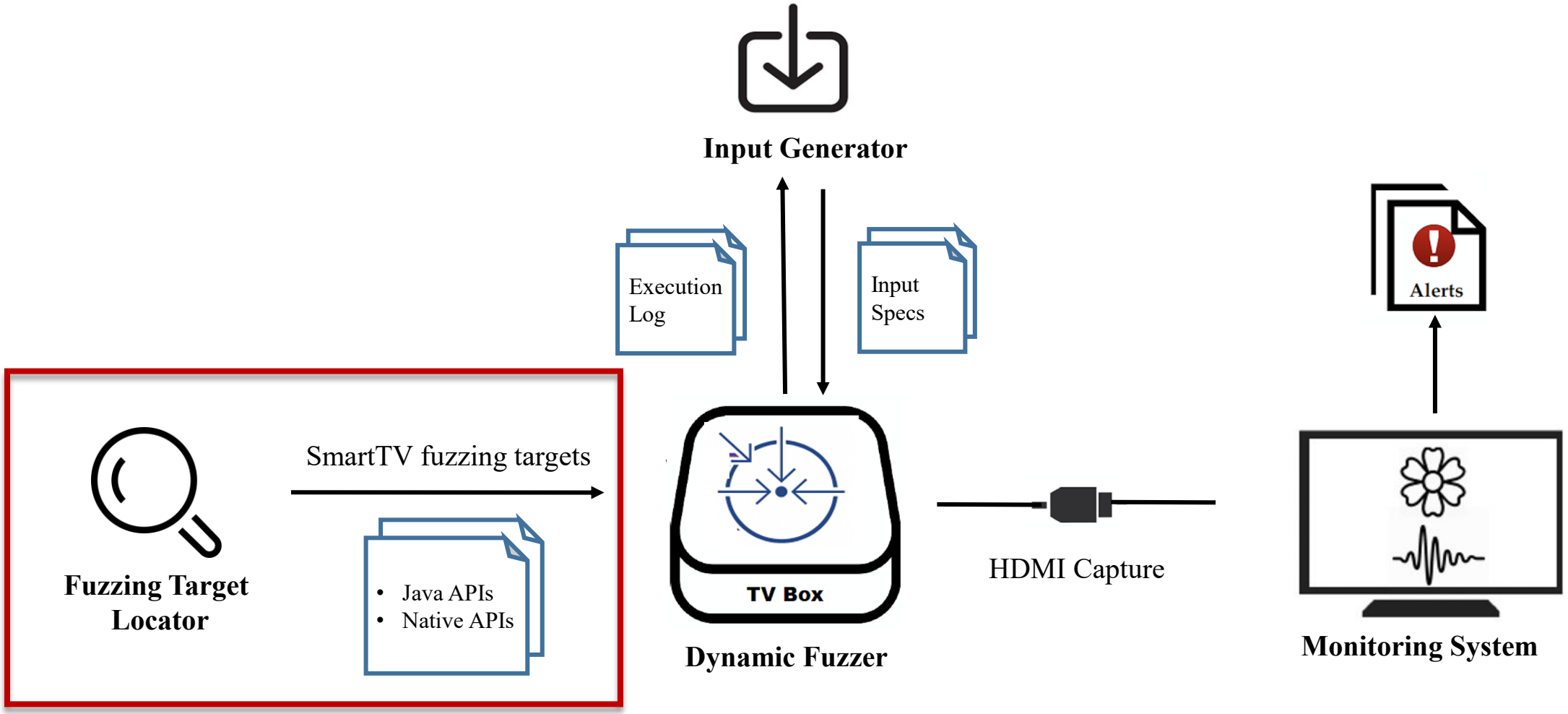
Detecting SmartTV Vulnerabilities

- We develop a specialized analysis framework to uncover hidden flaws, caused by unprotected APIs.

- **Why can’t we directly adopt static analysis tools?**
  - Additions are implemented in C++ and / or Java
- **Why can’t we directly adopt existing testing approaches?**
  - Assessing execution feedback is challenging

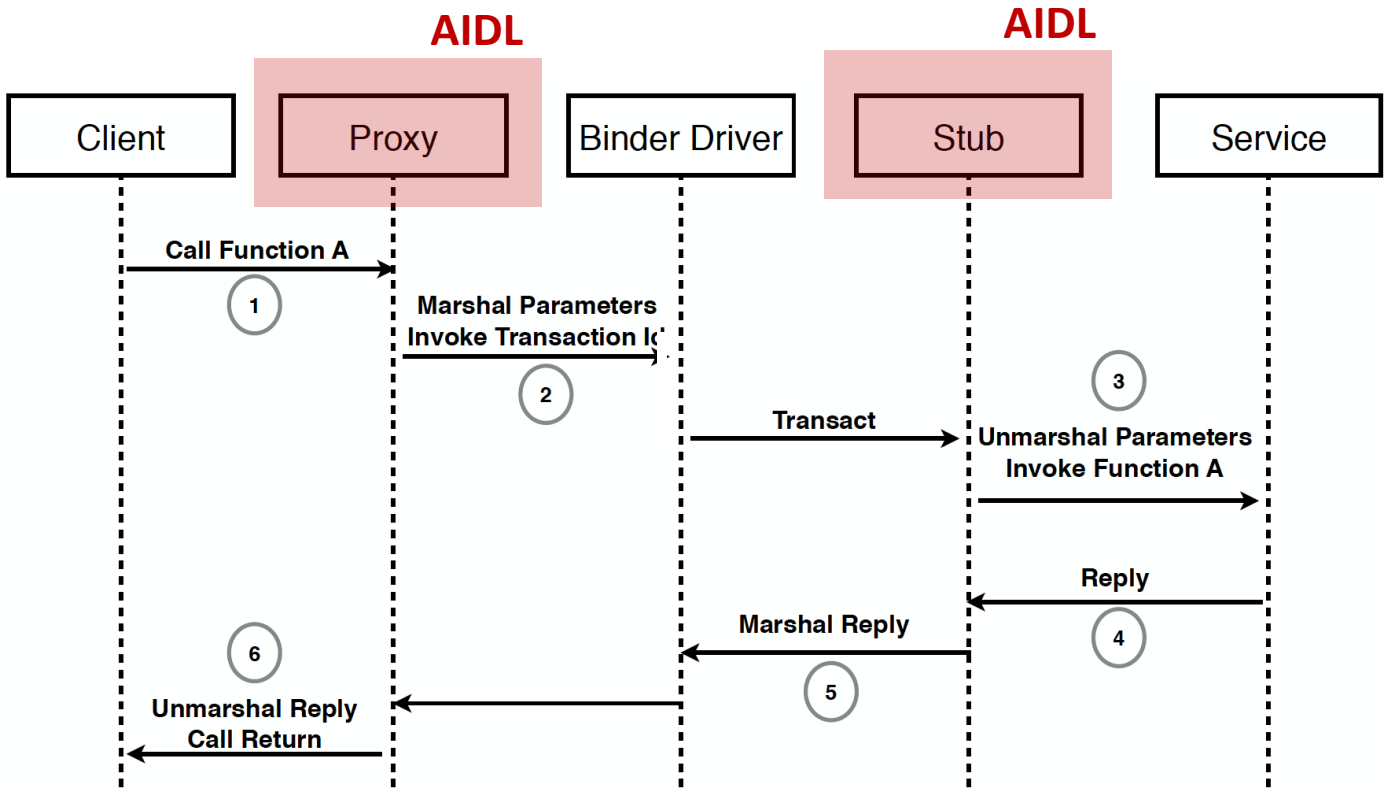
The Audio / Visual behavior is decoupled from the internal states → the system might be functioning correctly when the display / sound is messed up.

Our Approach: Fuzz-testing



Fuzzing Target locator

- We recover native API interfaces at the low-level Binder IPC through binary analysis.
- Recovering Native APIs Interfaces: Binder transaction ids, arguments types and order.



Extracting Native Function Interfaces

```
1  // Binder Proxy
2  class BpSystemControl : public BpInterface< SystemControl >{
3  ...
4    virtual status_t BpSystemControl::abc(...) {
5      ...
6      Parcel data, reply;
7      data.writeInterfaceToken(SystemControl::getInterfaceDescriptor())
8      data.writeStrongBinder(c->asBinder());
9      data.writeInt32(a);
10     data.writeInt32(b);
11     remote()->transact(1, data, &reply);
12     status_t err = reply.readInt32();
```

```
1  // Binder Stub
2  class BnSystemControl: public BnInterface< SystemControl >{
3  ...
4    status_t BnSystemControl::onTransact(...) {
5      switch (code) {
6      case 1: //Transact ID
7        sp<ISystemControlClient> c =
               ISystemControlClient::asInterface(data.readStrongBinder());
8        int32_t a = data.readInt32();
9        int32_t b = data.readInt32();
10       err = mLocal->abc(c, a, b);
11       reply->writeInt32(err);
```

symbol reserved

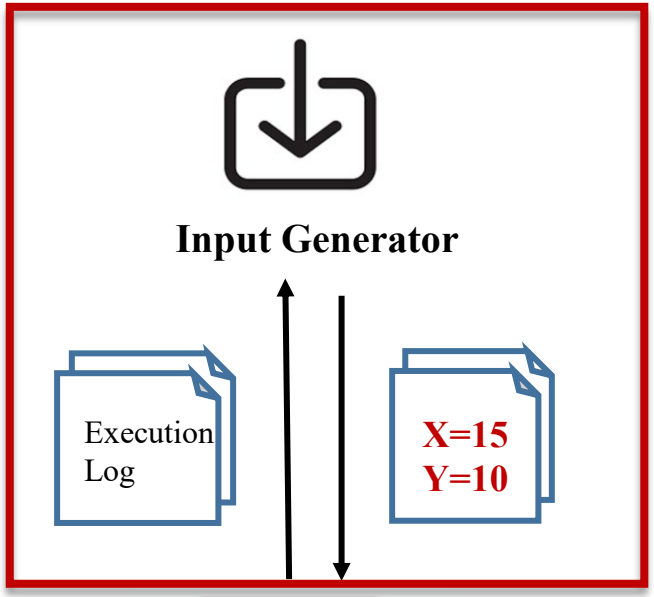
symbol absent

Step:

1. Identify function bodies within the binder proxies
2. Extract traction id and parameter types
   (inferred through the proxy's marshaling methods)
3. Confirm the result by analyzing the binder stubs.

Our Approach: Fuzz-testing

```
BatteryChangedJob: Running battery changed worker
ImagePlayerService: max x scale up or y scale up is 16
DiskIntentProviderImpl: Successfully read intent from disk
MediaPlayer: not updating
```

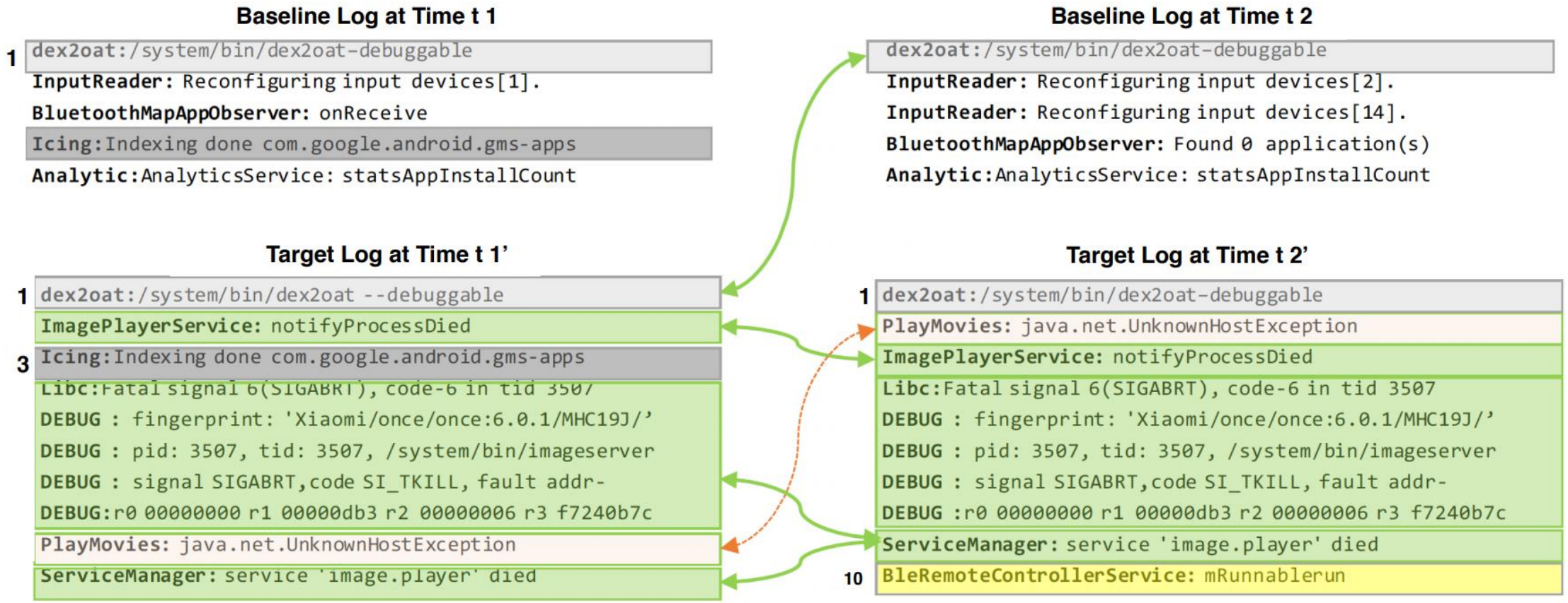


- Challenges to address:

1. Recognizing target logs
2. Recognizing input validations

Recognizing Target Logs

$$score(i) = \begin{cases} 1 & \text{if } \frac{n_i^{target}}{N^{target}} \geq 0.9 \text{ and } \frac{n_i^{baseline}}{N^{baseline}} \leq 0.1 \\ 0, & \text{otherwise} \end{cases}$$



Log Excerpts before and after calling the (native) target API ImagePlayer.XYZ()

Recognizing Input Validations

```
1  public void playSoundEffectVolume(int t, ..) {
2    if (t >= 9||t < 0) Log.w(T, " Value" + t + "out of range"); ...
3  public int dispatchFocusChange(AudioFocusInfo a, ..) {
4    if(a == null) throw IllegalArgumentException("Illegal null Info")
```

Java

```
1  sp<IAudioTrack> AudioFlinger::createTrack(int t, uint32_t r)
2    if (t >= AudioTrack::NUM_STREAM_TYPES) {
3      LOGE("invalid stream type"); goto Exit;}
4    if (r > MAX_SAMPLE_RATE || r > mSampleRate*2) {
5      LOGE("Sample rate out of range: %d", mSampleRate);
```

Native

- Feasibility of learning from log messages in Java to classify log messages from native
- Need of sophisticated NLP techniques as keyword lookup is insufficient.

Deep Learning for Message Classification

```
Intent buildRequestPermissionsIntent(String[] permissions) {
  if (ArrayUtils.isEmpty(permissions)) {
    Log.d("permission cannot be null or empty");
    return;
```

Input Validation

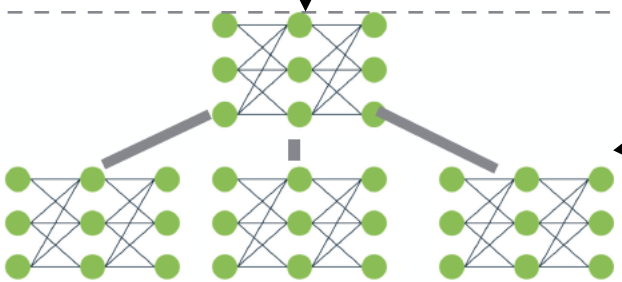
Input Validation

```
void requestBugReportWithDescription(String shareTitle,….){
  if (shareTitle.length() > 50) {
    String err = "shareTitle should be less than " +
                 50 + " characters";
    throw new IllegalArgumentException(err);}
  Slog.d(TAG, "Bugreport notification title" + shareTitle);
```

Input Validation

Input Validation

Non-Input Validation



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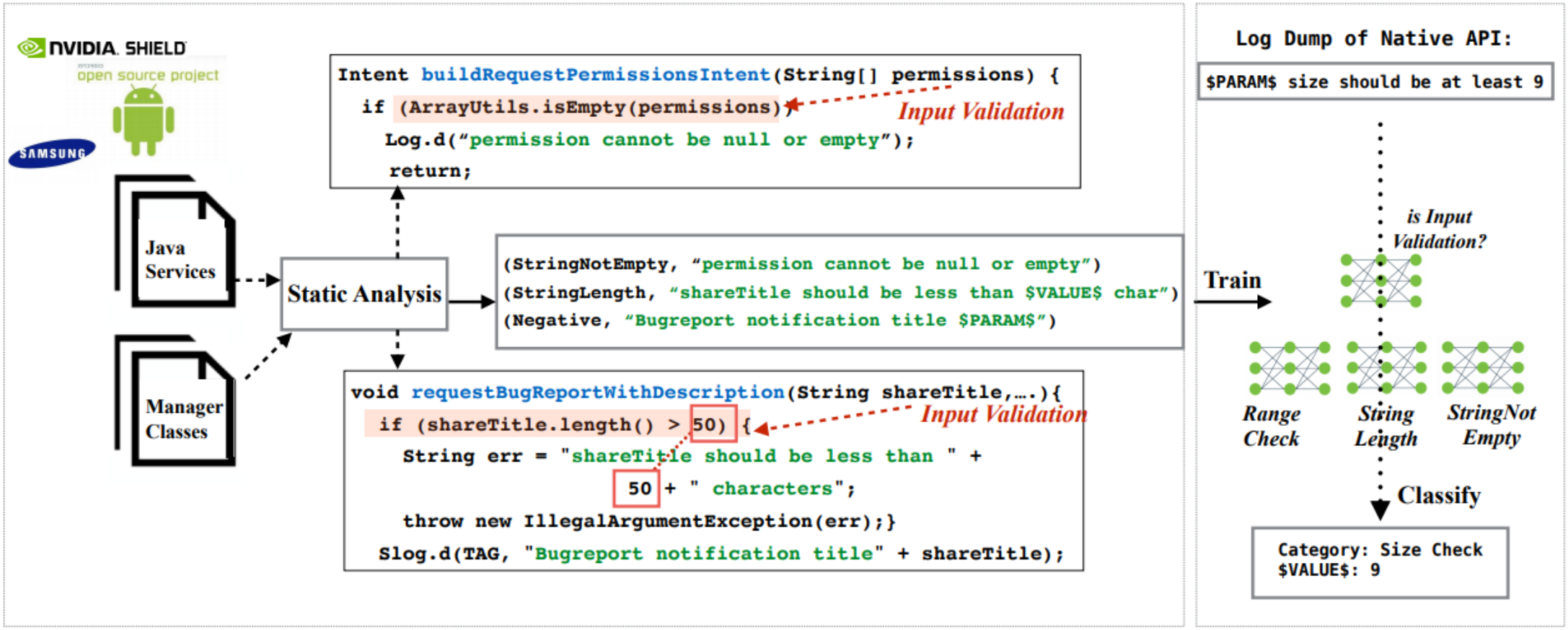
Input Validation Classification

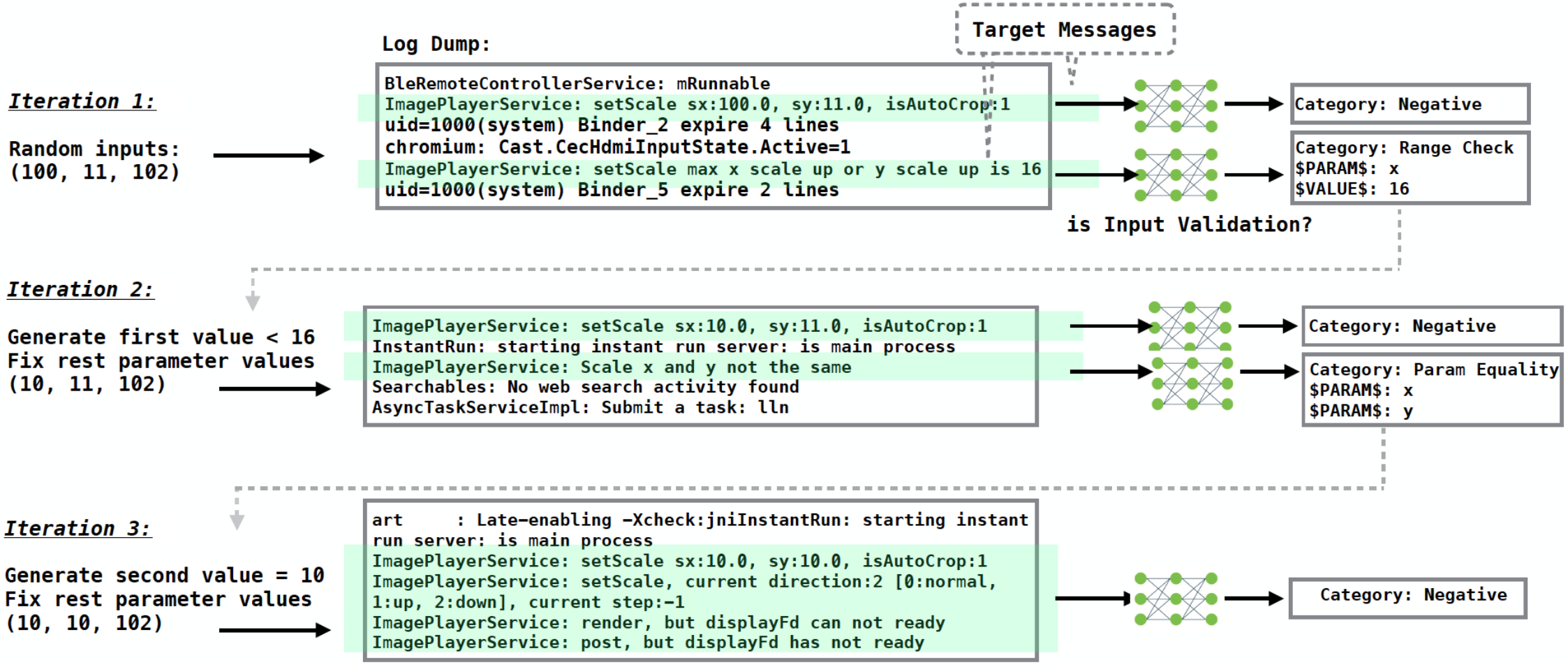


(A) Training Data Collection

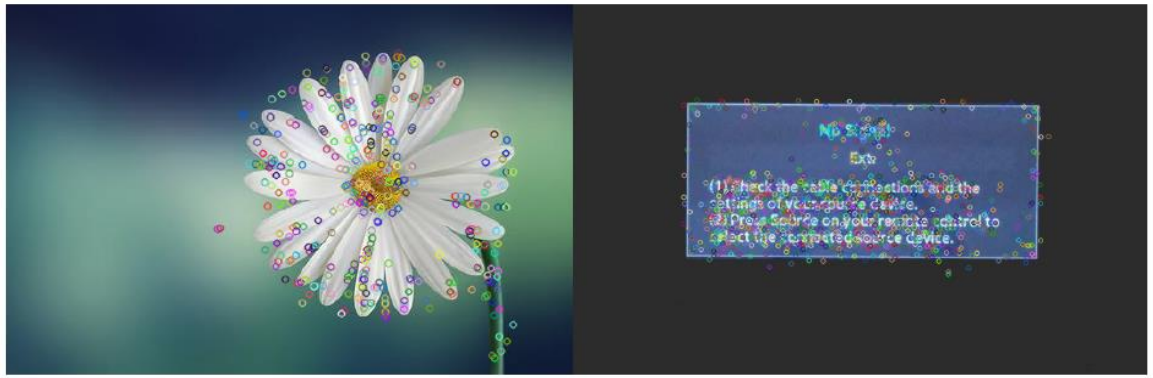
(B) Log Classifier Training

Log-Guided Fuzzing

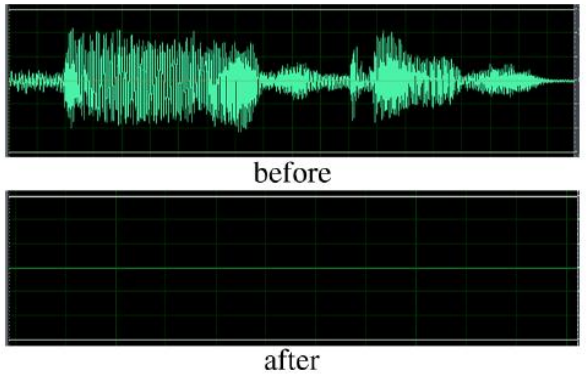
Example: fuzzing ABC(int, int, float)



Monitoring System

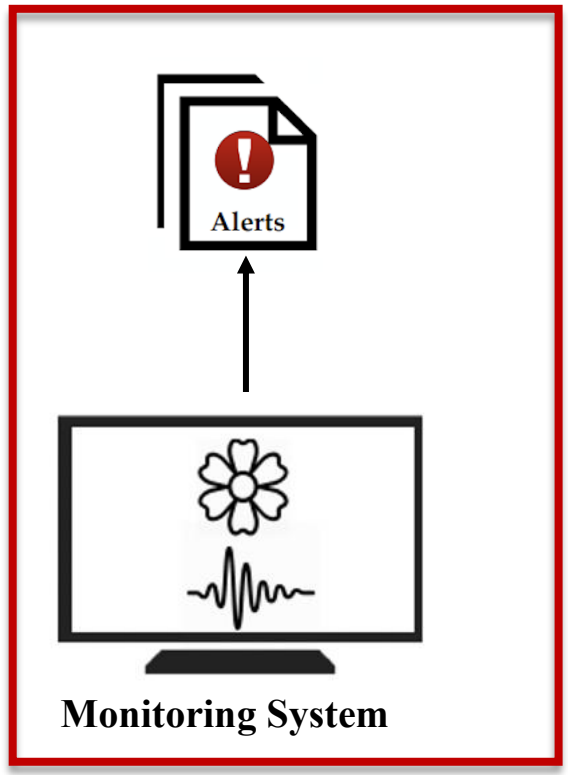


(a) Display before and after invoking `DisplayManager.enableInterface`



(b) Audio comparison

HDMI Capture



Evaluation

- 11 Android TVBoxes evaluated
  - including Nvidia Shield, MIBOX 3, etc.
  - each analyzed device contained 1 to 9 vulnerabilities
- 37 flaws discovered
  - 11 high-impact cyber threats
  - 10 new memory corruptions
  - 16 visual/auditory anomalies
- confirmed and fixed by the vendors

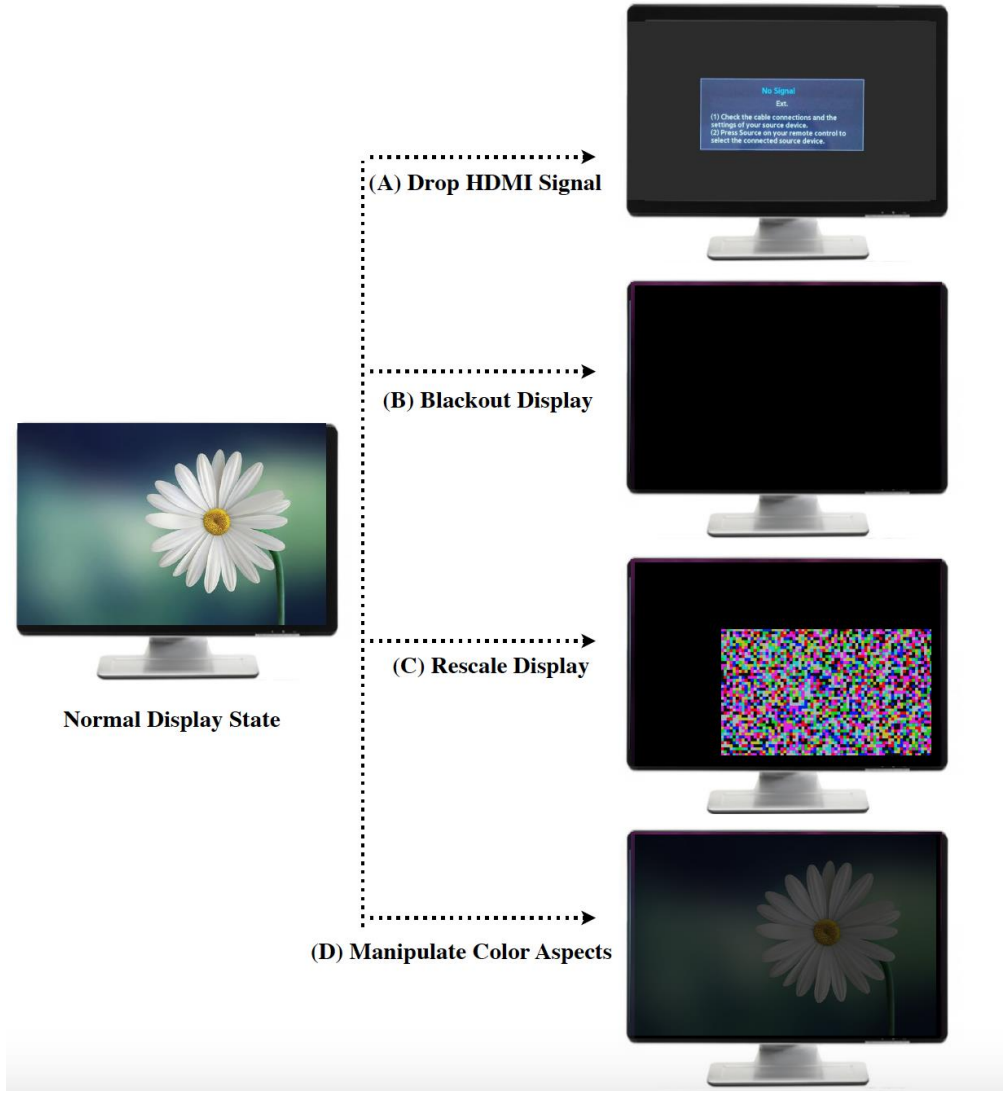
Evaluation

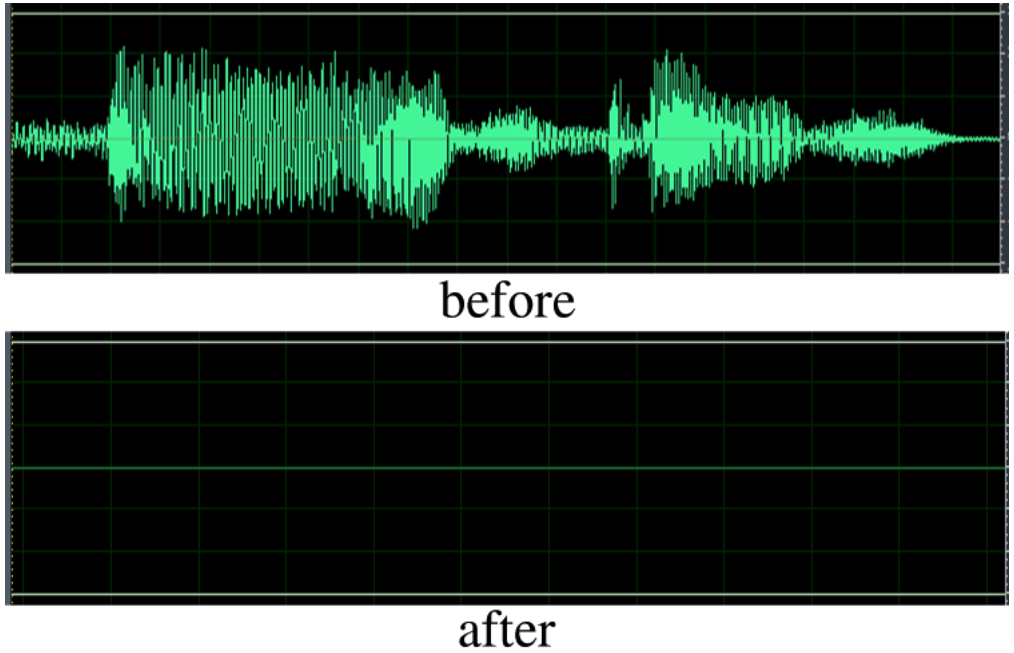
Cyber threats and Memory Corruptions

Description	Victim Devices (s)	Log-Guided Input Inference	Log-Guided Feedback Inference	External Feedback	Exposing Time	
					Random	Guided
Corrupt boot environment variables	H96 Pro	✓	✓	✓	Timed out	0.11h
Overwrite System Directories	Nvidia Shield	✓	✓	✓	Timed out	4.71h
Delete Files in internal memory	Nvidia Shield	✓	✓	✓	Timed out	2.14h
inject mouse coordinates	V88, Max	✗	✗	✓	0.03h	0.04h
inject mouse coordinates	V88, Max	✗	✗	✓	0.03h	0.03h
Change persistent system properties	Q+	✓	✓	✗	Timed out	0.14h
read highly-sensitive data	Q+	✓	✓	✗	Timed out	0.14h
overwrite certain system files	Q+	✓	✓	✗	Timed out	0.19h
read highly-sensitive data	Q+	✓	✓	✗	Timed out	0.15h
create hidden files under /sdcard/	GT King	✓	✓	✗	Time out	0.05h
reboot device into recovery mode	MIBOX4	✗	✓	✓	0.03h	0.03h

Evaluation

Physical Vulnerabilities





Related Work

- **IOT-Fuzzer: Discovering Memory Corruptions in IOT through App-Based Fuzzing.** In Proceedings of NDSS 2018.

- **FIRM-AFL: High-Throughput Greybox Fuzzing of IoT Firmware via Augmented Process Emulation.** In Proceedings of Usenix Security 2019.

Conclusion

- New technique
  - integrate static analysis and log-guided dynamic fuzzing
  - automatically detect cyber and physical anomalies
  - provide a solution when instrumentation and execution feedback is not feasible

- New findings
  - reveal security-critical threats of Android SmartTV API additions
  - cyber threats, memory corruptions and physical anomalies

Thank you!

Q & A

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