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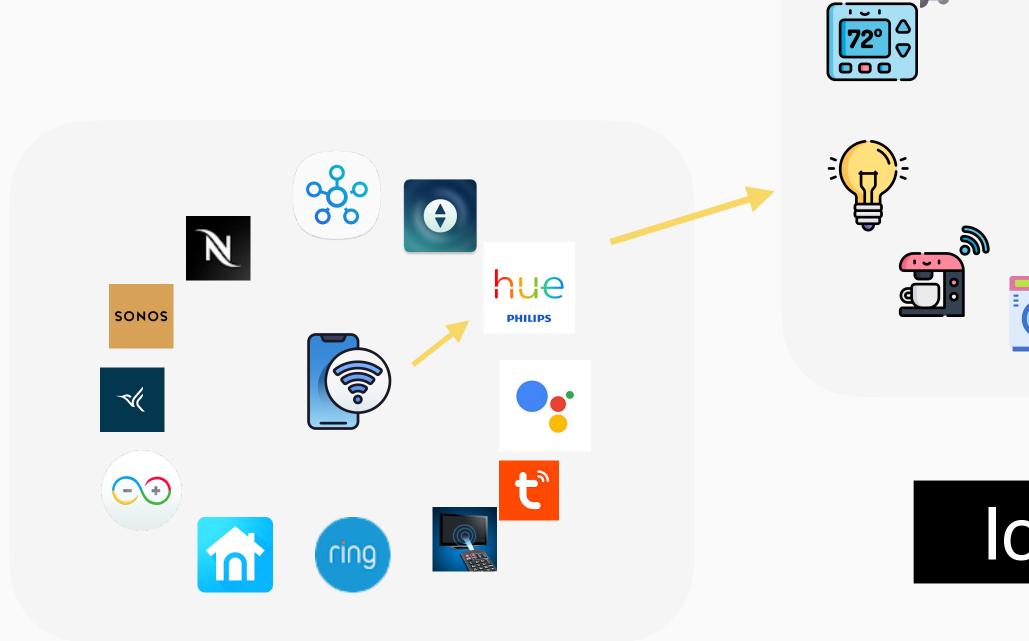
*This work was completed when the author was at William & Mary.

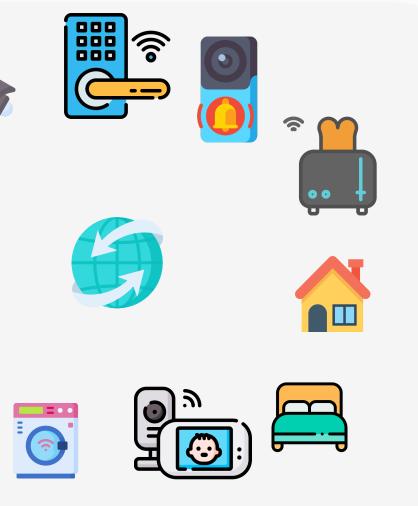
Understanding loT Security from a Market-Scale Perspective



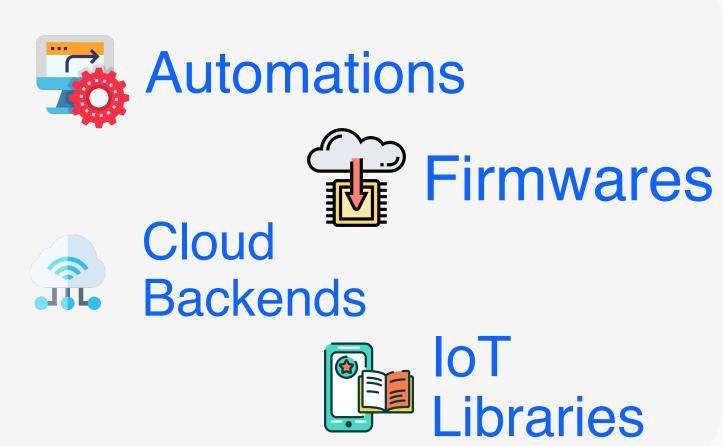
IoT Security



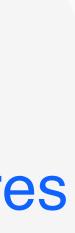






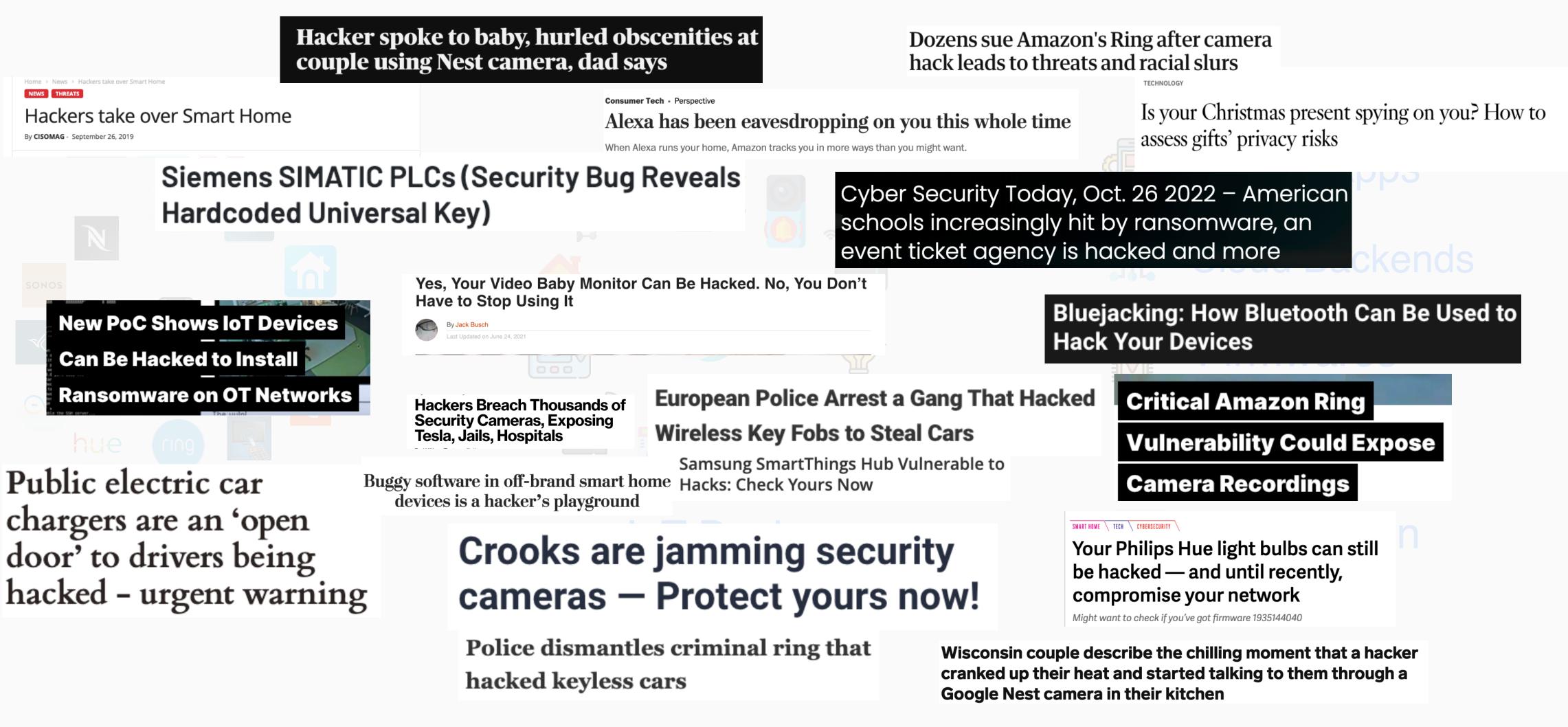


Attack Surface





IoT Security



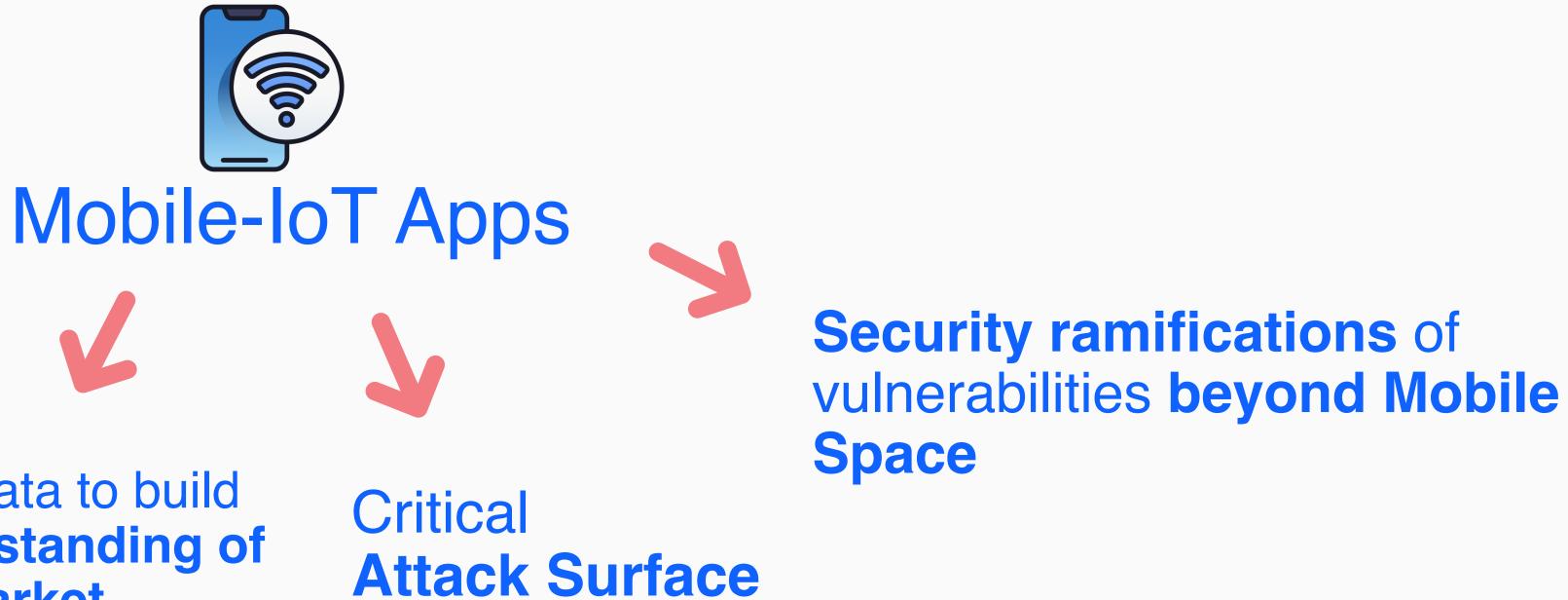


IoT Security Challenges

We do not know what products constitute IoT Ecosystem!



Metadata to build understanding of **IoT Market**







We do not know what products constitute IoT Ecosystem!

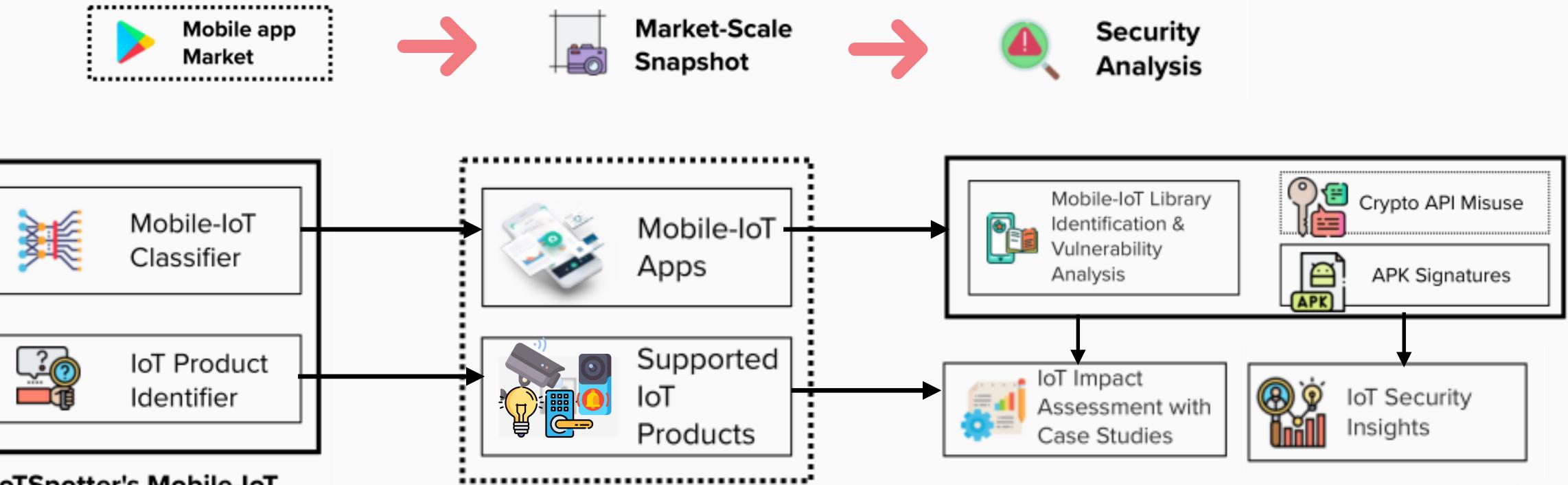
Research Questions



- RQ1: How can we automatically *develop a market-scale snapshot* of mobile-IoT apps from markets containing heterogeneous apps?
- **RQ2:** How can we *make the snapshot useful for security*?







IoTSpotter's Mobile-IoT app Identification

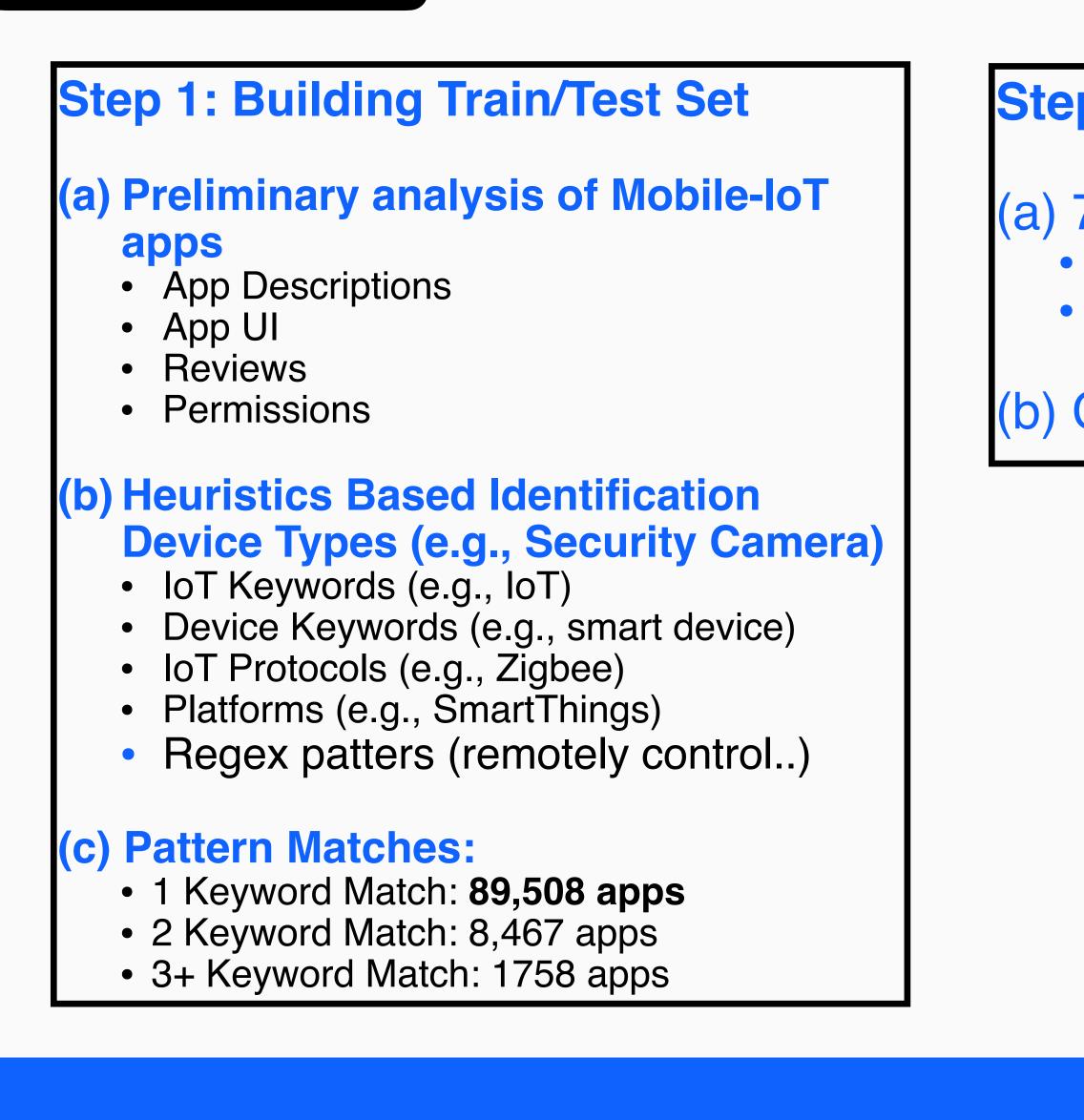
IoTSpotter Framework

Security Analysis of Vulnerability Prevalence and Impact on IoT



Mobile-IoT App Identification

Methodology



Step 2: Manual Labeling

(a) 7196 Labeled Apps
4,123 IoT apps
3,073 non-IoT apps

(b) Cohen Kappa: 0.976

Step 3: Build Classifier Model

(a) Run different Learning algorithms for description:
• Stratified train-test set

(b) Evaluate Performance



iRobot Home

About this app

 \times

The new iRobot Home App is here. With it, enhanced maps, the ability to clean specific objects, custom routines, seasonal suggestions, and intuitive smart home integrations*. Every aspect of the iRobot Home App has been redesigned to give you ultimate control over your clean.





Results

	Description							
Performance	LR	SVM	NB	RF	RNN	LSTM	BiLSTM	BERT
Accuracy Precision Recall F1-Score	0.927 0.932 0.896 0.914	0.916 0.897 0.908 0.902	0.914 0.909 0.889 0.899	0.926 0.929 0.896 0.913	0.947 0.925 0.954 0.939	0.946 0.957 0.915 0.936	0.952 0.962 0.925 0.943	0.957 0.949 0.951 0.950

Ran the model (with hard voting) on the entire market with 2 Million Apps!

Results 2: Manual validation of 2,250 mobile-IoT apps showed 88% are indeed IoT.

Result 1: Identified 37,783 Mobile-IoT apps.



Methodology

Step 1: Building Train/Test Set

(a) Select 600 Random Mobile-IoT Apps (b) Manually labeled 3961 statements to identify IoT_Product Entities

Step 2: Train Named Entity Recognition (NER) Model

(a) 82.84% Precision (b) 83.04% Recall

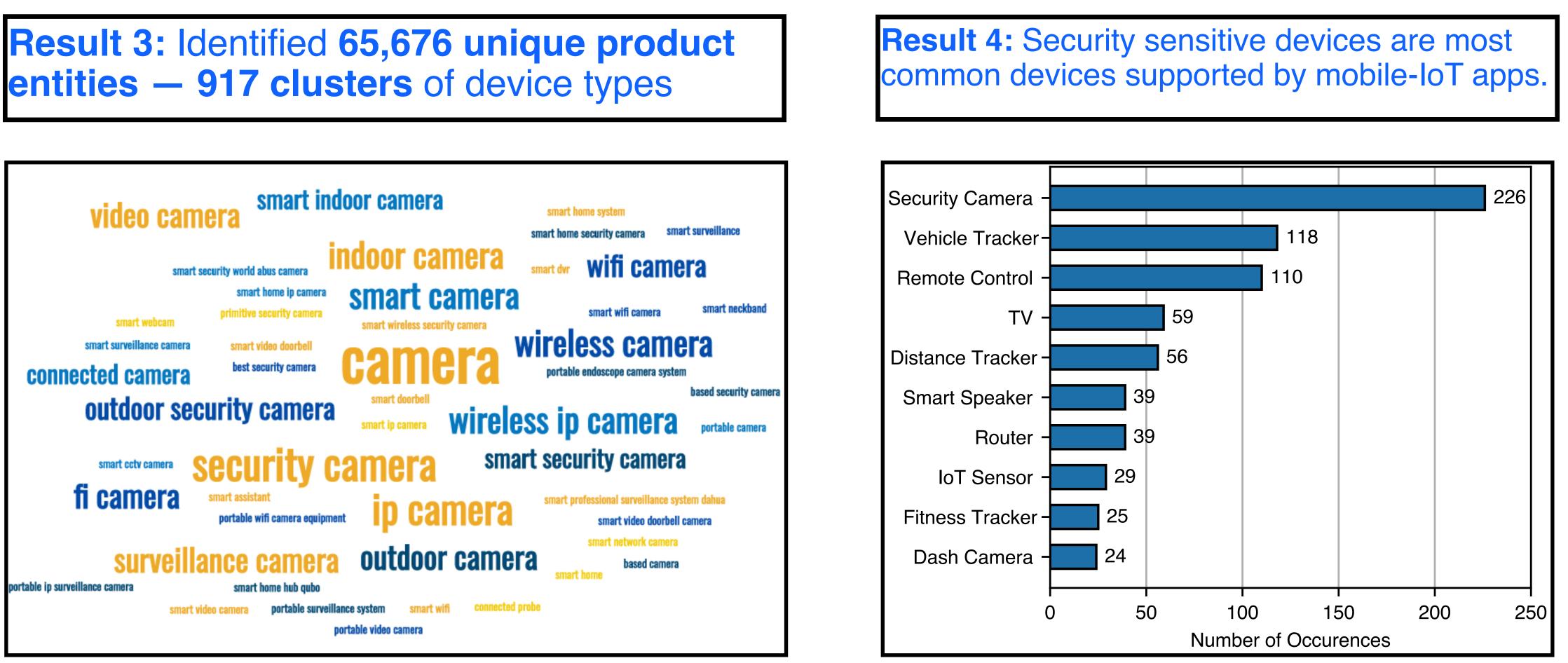
Our first product was the trendsetting Wyze Cam IOT_PRODUCT : indoor smart camera IOT_PRODUCT a multi-purpose that helped our users keep an eye on what matters most without a hefty price tag.





IoT Product Identification

Results





Security Analysis: IoT Library

Methodology

Identifying IoT Libraries

Identified 522,285 third-party library package names from the mobile-IoT snapshot

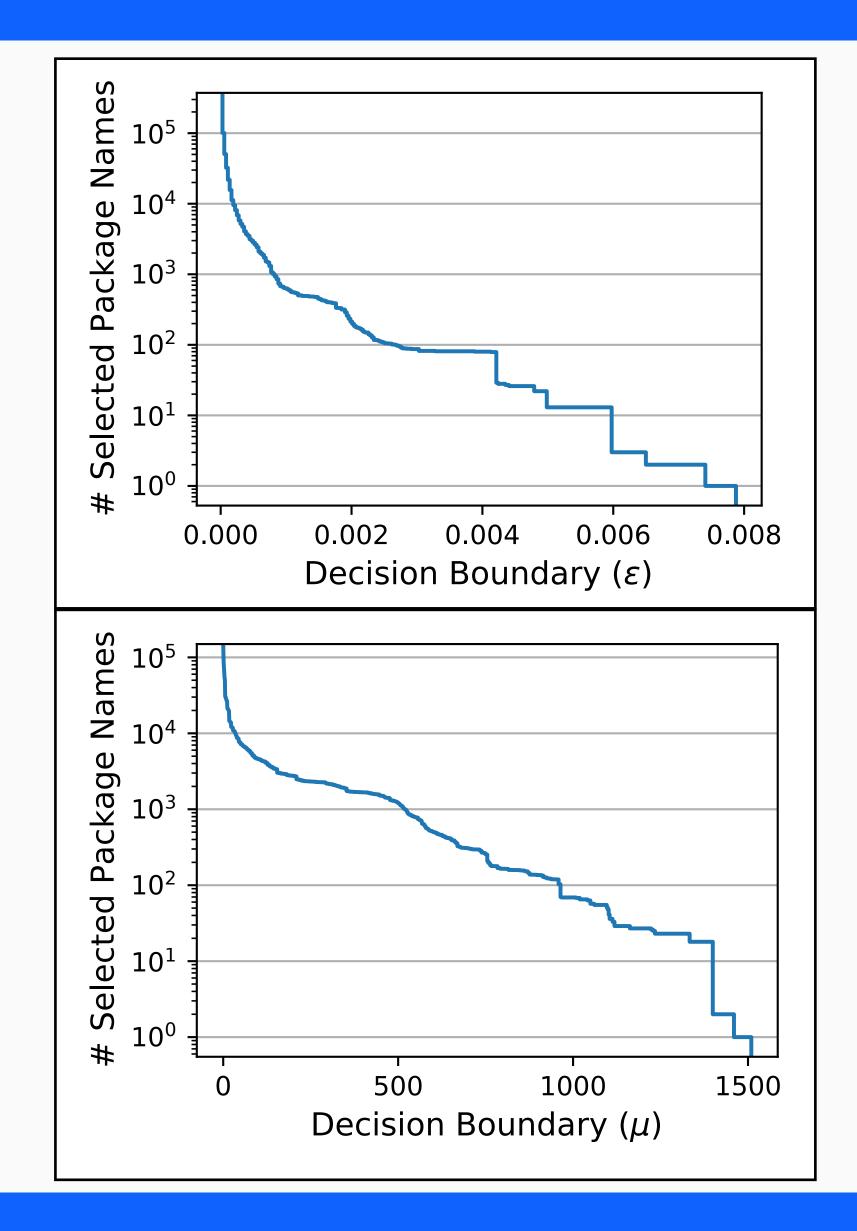
 Popular third-party libraries only found in IoT

 $\epsilon = \underline{\text{no. of apps using the library}}$ total no. of mobile-loT apps

 Third-Party libraries more popular in mobileloT snapshot than non-loT set

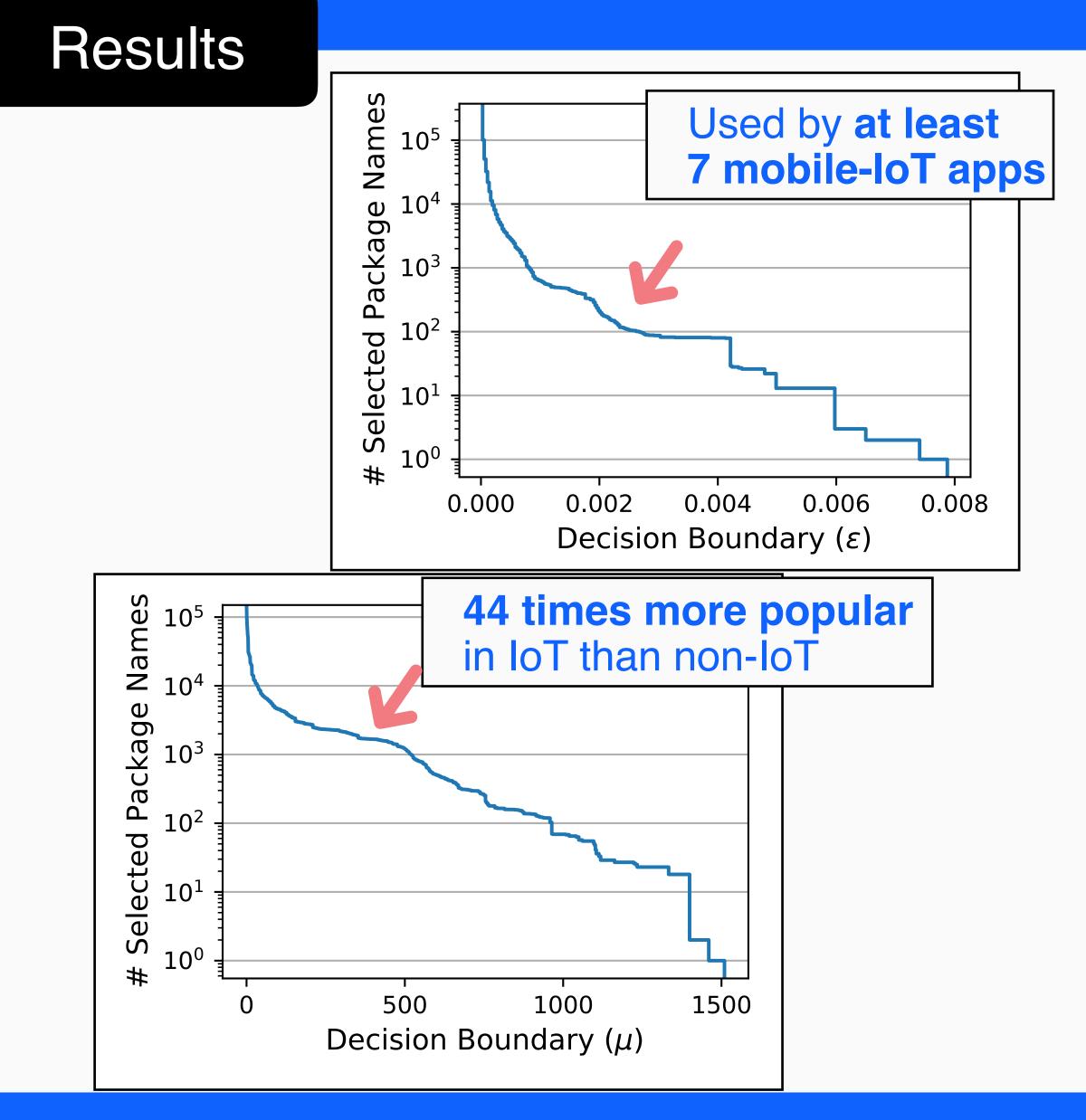
μ = <u>Popularity in Mobile-IoT</u> Popularity in Non-IoT







Security Analysis: IoT Library



Result 5: Identified 19,939 IoT library package names.

Analysis of 50 library package names.

Library Family	Functionality	# Apps
com.tuya	IoT framework	1,362
no.nordicsemi.android	BLE & firmware services	1,097
javax.jmdns	DNS services	852
com.amazonaws.mobileconnectors	IoT cloud services	751
com.connectsdk	Device control	378
com.inuker.bluetooth	BLE services	358
com.clj.fastble	BLE services	333
com.hiflying	Device control	285
com.telink	Device control	250
com.hikvision	Device control	191
org.fourthline.cling	Device control	187

Results 6: Identified 11 library families; 10 provide functionalities associated with IoT







Security Analysis: IoT Library

Findings

Vulnerabilities in IoT libraries

Finding 1: 65 IoT Libraries (481 unique versions) are subject to **79 CVEs**

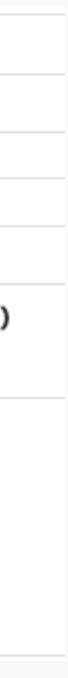
Finding 2: IoT libraries are less vulnerable relative to non-loT libraries; Out of 2500 samples: Non-IoT: 193 CVEs, 63 libraries, 7,105 versions **IoT:** 7CVEs, 10 libraries, 98 versions

Use of Vulnerable IoT Libraries

Finding 3: 40 popular mobile-loT apps are vulnerable because of vulnerable **IoT library usage.**

Finding 4: Vulnerable library usage in non-loT is 12.7X (507/40) more than in IoT.

	N REPOSITORY		
Categories	Android Packages		
Tags	panel aar android		
Date	Jan 08, 2021		
Files	aar (90 KB) View All		
Repositories	JCenter		
Ranking	#405161 in MvnRepository (See Top Artifacts) #56265 in Android Packages		
	Vulnerabilities from dependencies:		
	CVE-2022-25845		
Vulnerabilities	CVE-2022-24329		
	CVE-2021-36090		
	View 5 more		





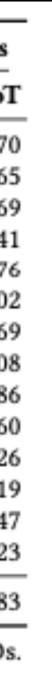
Security Analysis: Crypto APIs

Findings

Flaws detected by CryptoGuard

Finding 5: 94.11% apps contain at least 1 Crypto-API misuse according to CryptoGuard out of 917 apps with 1M+ installs (96.29% non-loT). Finding 6: 82.5% high severity violations detected by CryptoGuard is true positive.

	CryptoGuard's Rules (IDs as per [55])	# Vulnerable Apps		
п	Rule Name	Mobile-IoT	Non-IoT	
9	Insecure PRNGs (e.g., java.util.Random) [M]	842	870	
16	Insecure cryptographic hash (e.g., SHA1, MD5) [H]	825	865	
1	Predictable/constant cryptographic keys [H]	577	669	
7	Occasional use of HTTP [H]	438	441	
14,11	*64-bit block ciphers (c.g., DES, RC4), ECB mode [M]	406	376	
5	Custom TrustManager to trust all certificates [H]	380	302	
4	Custom Hostname verifiers to accept all hosts [H]	293	269	
12	Static IVs in CBC mode symmetric ciphers [M]	239	208	
6	SSLSocketFactory w/o hostname verification [H]	186	86	
3	Predictable/constant passwords for KeyStore [H]	142	60	
13	Fewer than 1,000 iterations for PBE	70	26	
15	Insecure asymmetric cipher use	66	19	
2,10	*Predictable passwords, static salts in for PBE [H/M]	63	47	
8	Predictable/constant PRNG seeds [M]	50	23	
-	Number of apps that violated at least one rule	863	883	





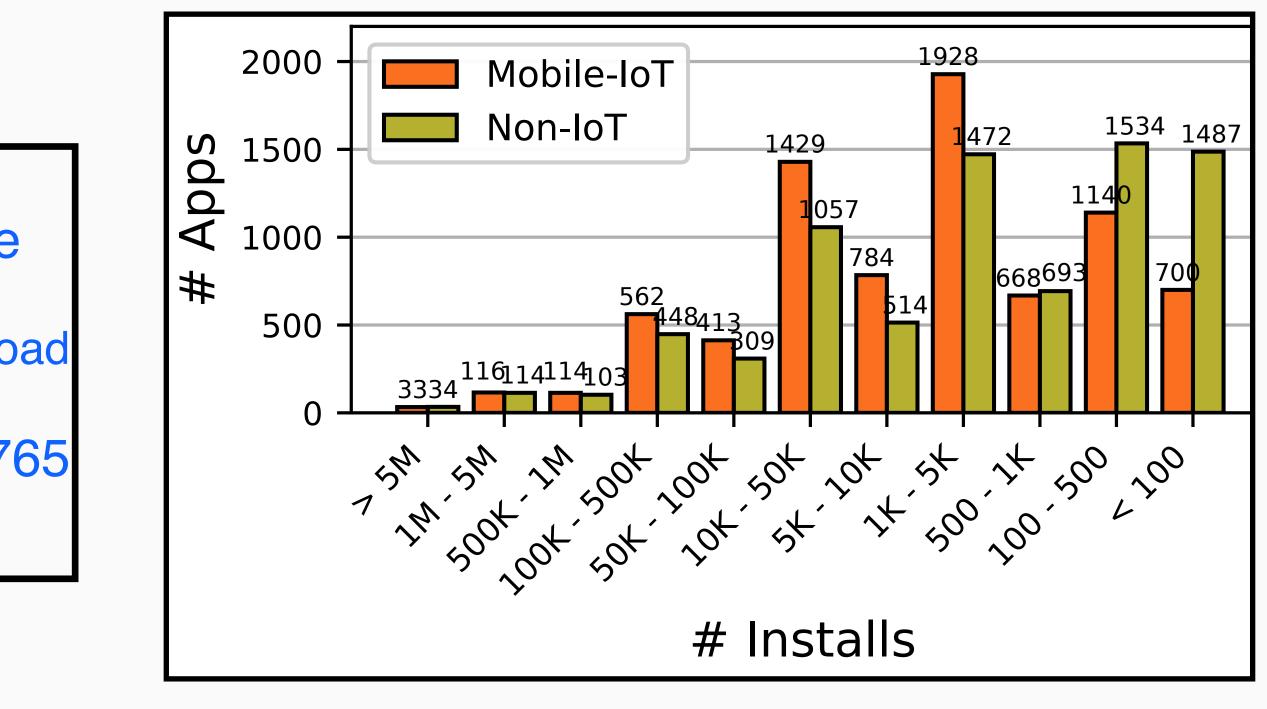


Janus Vulnerability in Apps

Finding 7: 7,887 (20.87%) mobile-IoT apps are susceptible to Janus Vulnerability

- 263 with 1M+ download and 33 have 50M+ download

Finding 8: Non-IoT are similarly vulnerable (7765 apps).





Case Study: Contextual Analysis

Findings

Contextualization

Finding 9: Every class of vulnerability impacts critical IoT functions.

Finding 10: Vulnerable IoT apps support security/privacy critical devices.

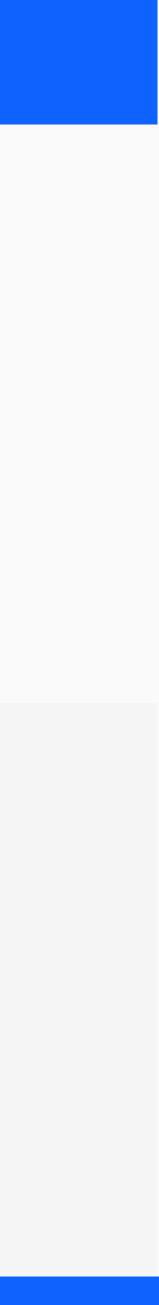
IoT Impact	Vulnerabilities	Devices Affected
Firmware (Malicious	Crypto (HTTP,	Camera
Modification)	no integrity	JBL Speaker
	checks)	IP Camera
App/Device	IoT Libraries	PTZ Camera
Functions (hijack,	(multiple CVEs)	Smart TV
code execution)		Vestel Smart TV
		IP Camera
		Wi-Fi Routers
User Credentials,	Crypto (MD5,	IP Camera, NVR
Authentication	TrustManager,	TVs, Chromecast
	HTTP)	Camera
Admin Password	Crypto (constant	IP Camera
Leakage	password, HTTP)	
App Integrity	Janus	Activity trackers
(Malware)		Echo Devices
General Data	Crypto (DES,	Smart TV
Security	MD5, ECB mode)	Washer, AC, TV
-		Lights, Blinds, TV





- Focused Effort on Mobile-IoT Apps
- Precise Exploration of Mobile-IoT Security
- Contextualized, Automated, Security Analysis for Mobile-IoT

Lessons



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