

## Information Exposure From Consumer IoT Devices: A Multidimensional Network-Informed Approach

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# IoT Challenges: Privacy in a World without Walls

20.4 billion IoT devices by 2020 (Gartner Inc.)

- Closed systems and lack of ground truth
  - MITM fails most of the time
- Lack of automation and emulation tools
- Lack of standard testbed and controlled experiments that enable comparisons across IoT deployment sites



# Privacy Concerns

- Personal Information: Stored, Sensor, or Activity data

What information is exposed?

IoT vendor

Cloud/CDN

Other intended destinations

Unintended destinations

- Destination Parties: First, Support, Third, Eavesdroppers

Who receives such information?

?

Non-first party recipients?

?

Traffic going through different privacy jurisdictions?

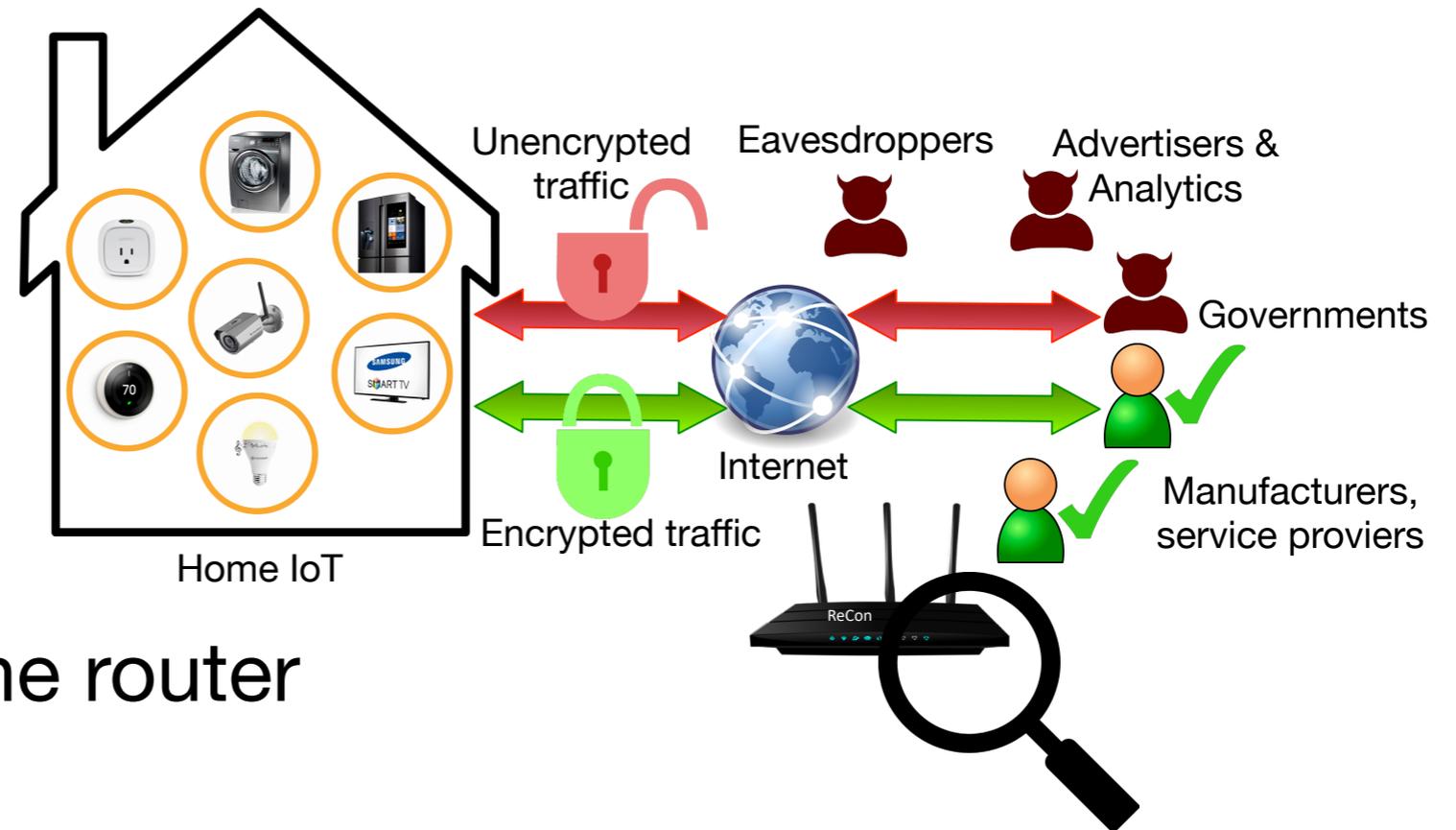
?

Activity data inferred by non-first parties?

# Research Questions

- What is the destination of network traffic?
- To what extent is the traffic encrypted?
- What content is sent?
- Does a device expose information unexpectedly?

# Data Collection Methodology

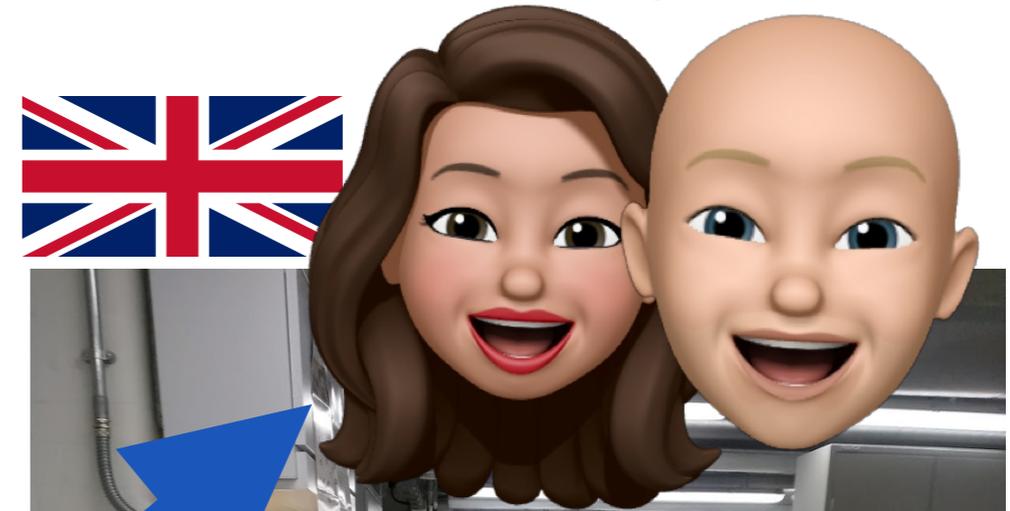
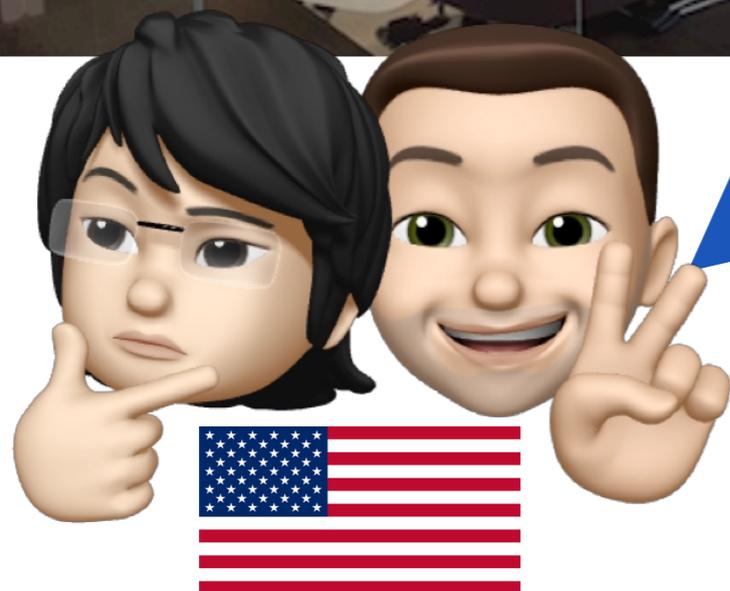
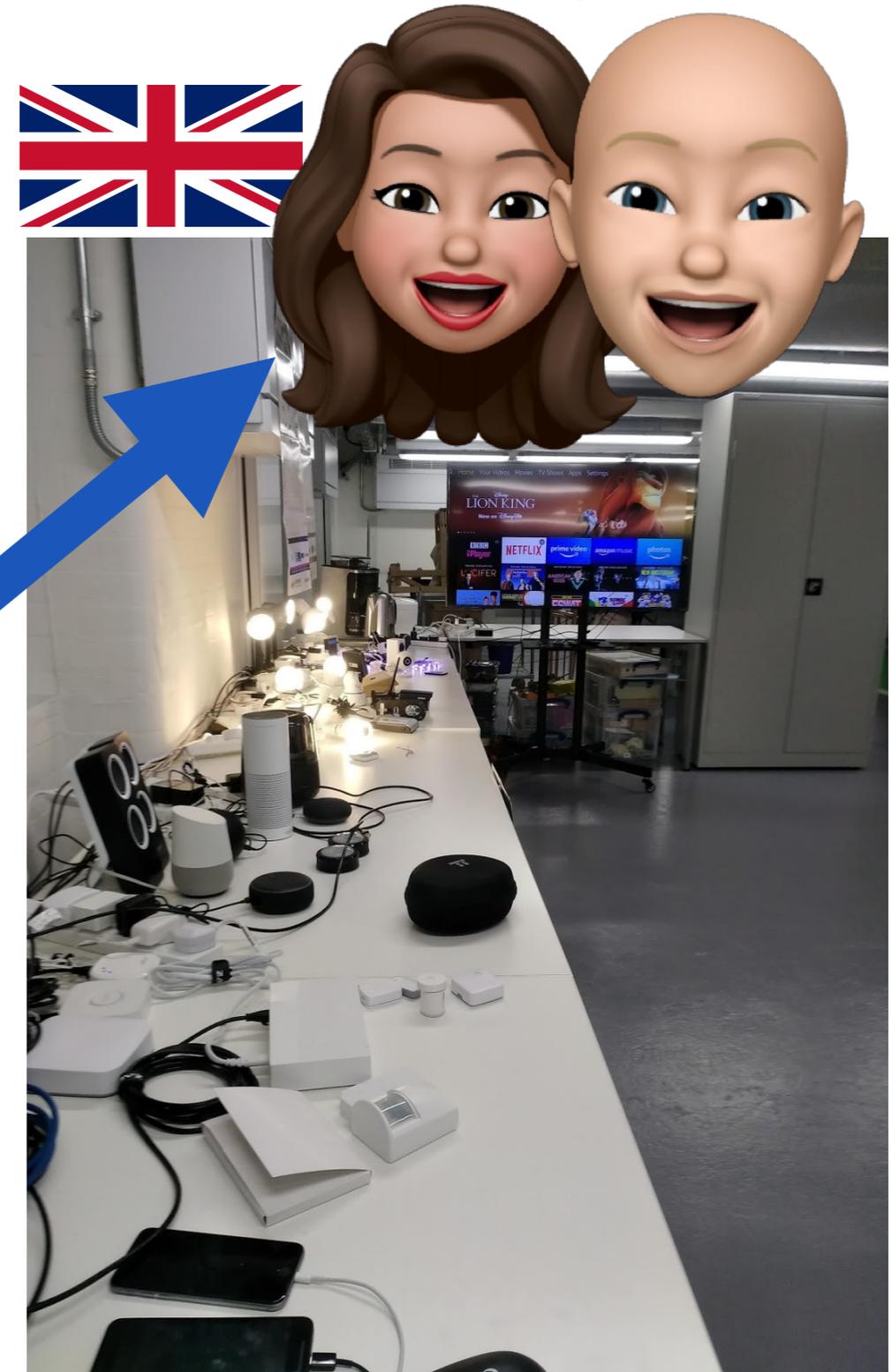
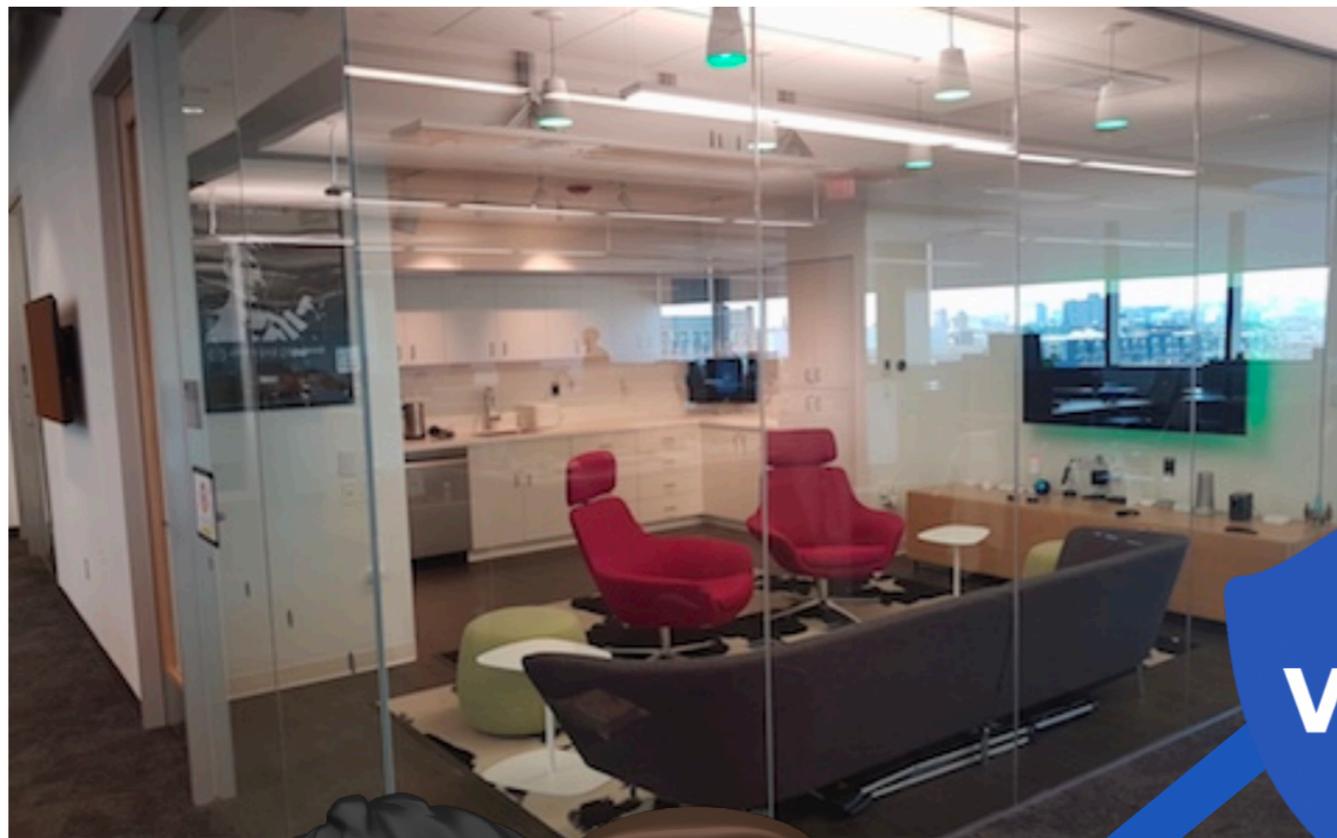


- Monitor all traffic at the router
  - per-device
  - per-experiment
- Labs: US and UK (GDPR)

# Testbeds

US: Northeastern University

UK: Imperial College London



# Selecting IoT Devices

- **Criteria:** category; features; popularity; US & UK markets



Flux Bulb  
Xiaomi Strip  
Philips Bulb  
LG TV  
Invoke Speaker  
Behmor Brewer  
GE Microwave  
Samsung Dryer  
Samsung Fridge  
Samsung Washer  
Smarter iKettle  
Xiaomi Rice Cooker

Amazon Cam  
Amcrest Cam  
Lefun Cam  
Luohe Cam  
Micro7 Cam  
ZModo Bell  
Wink2 Hub  
D-Link Sensor

N=46

Blink Cam  
Blink Hub  
Ring Doorbell  
Wansview Cam  
Yi Cam  
Insteon Hub  
Lightify Hub  
Philips Hue Hub  
Sengled Hub  
Smarthings Hub  
Xiaomi Hub  
Magichome Strip  
Nest T-stat

TP-Link Bulb  
TP-Link Plug  
WeMo Plug  
Apple TV  
Fire TV  
Roku TV  
Samsung TV  
Echo Dot  
Echo Spot  
Echo Plus  
Google Home Mini  
Anova Sousvide  
Xiaomi Cleaner

N=26



Bosiwo Cam  
D-Link Cam  
WiMaker Cam  
Xiaomi Cam  
Honeywell T-stat  
Allure Speaker  
Google Home  
Netatmo Weather  
Smarter Brewer

N=35

20 Cameras 13 Smart Hubs 15 Home Automation 9 TVs 11 Speakers 13 Appliances 81 Total



# Design of Experiments

- **Idle: ~112 hours**
- **Controlled interactions**  34,586 experiments (92.6% automated)
  - Manual (repeated 3 times)
  - Automated (repeated 30 times)
    - Text-to-speech to smart assistants (Alexa/Google/Cortana/Bixby)
    - Monkey instrumented control from Android companion apps

- **Uncontrolled interactions**
  - IRB-approved user study
  - 36 participants, 6 months  
Sep/2018 to Feb/2019

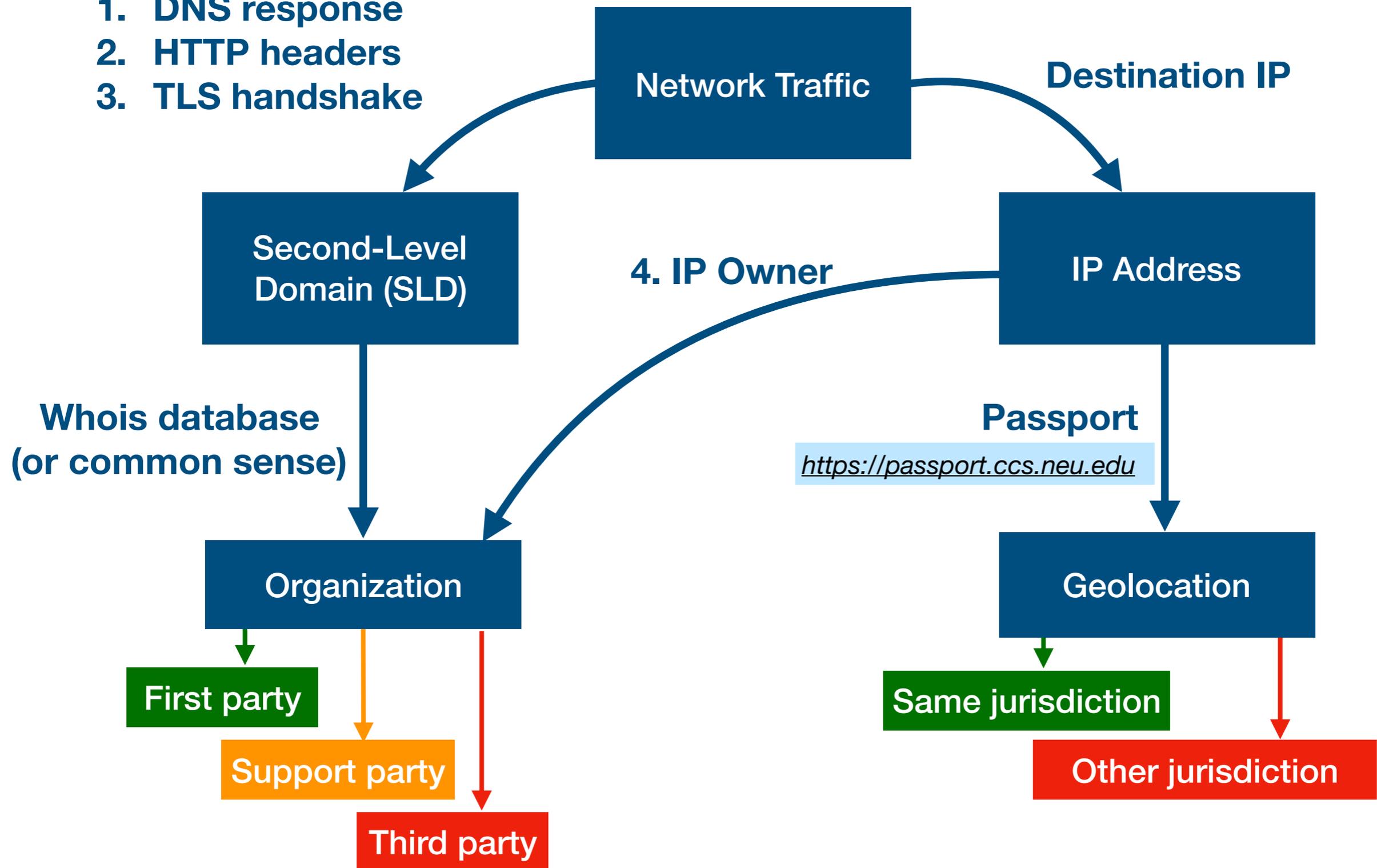
Activity	Description
Power	power on/off the device
Voice	voice commands for speakers
Video	record/watch video
On/Off	turn on/off bulbs/plugs
Motion	move in front of device
Others	change volume, browse menu

# Research Questions

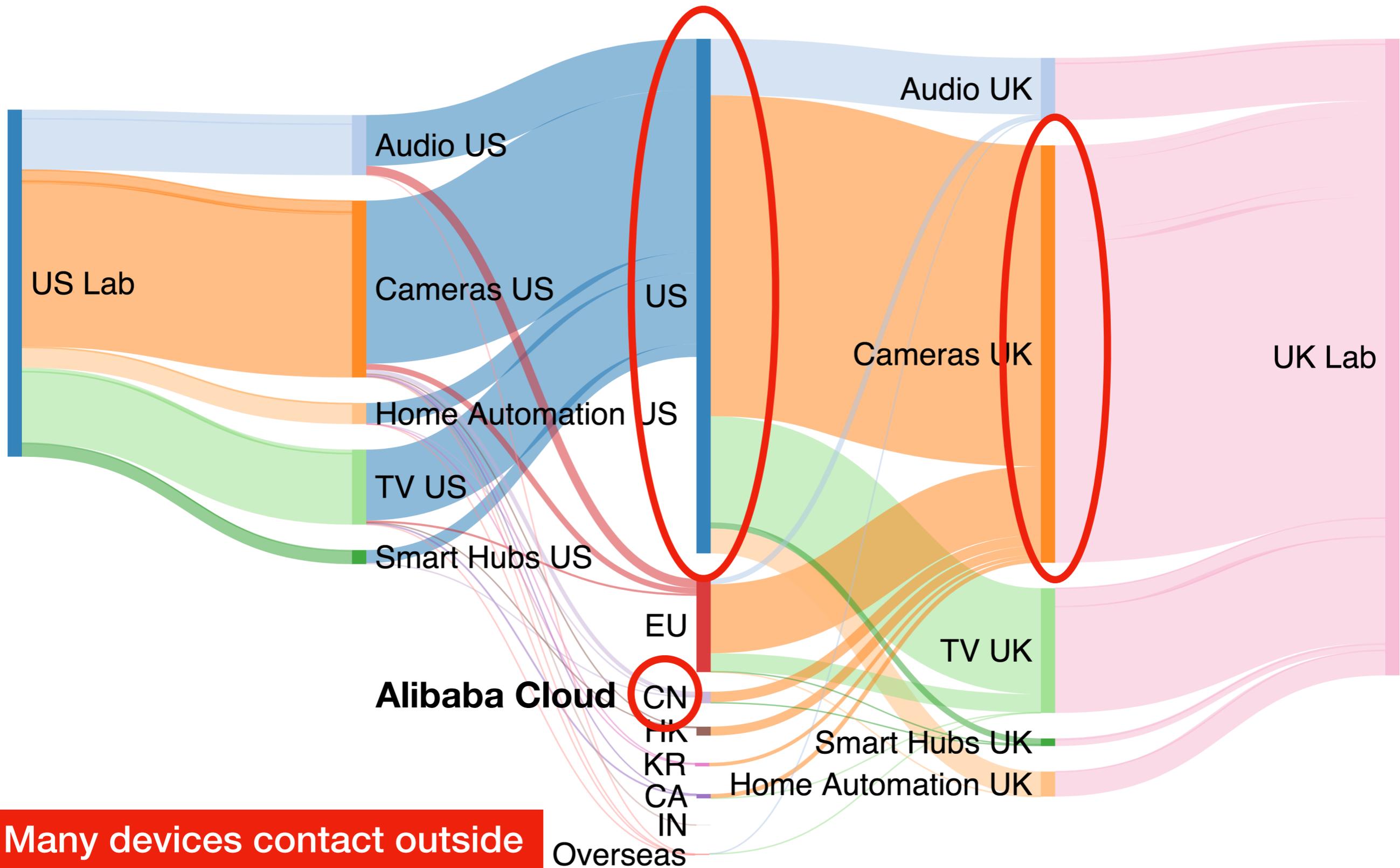
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# What is the Destination?

1. DNS response
2. HTTP headers
3. TLS handshake



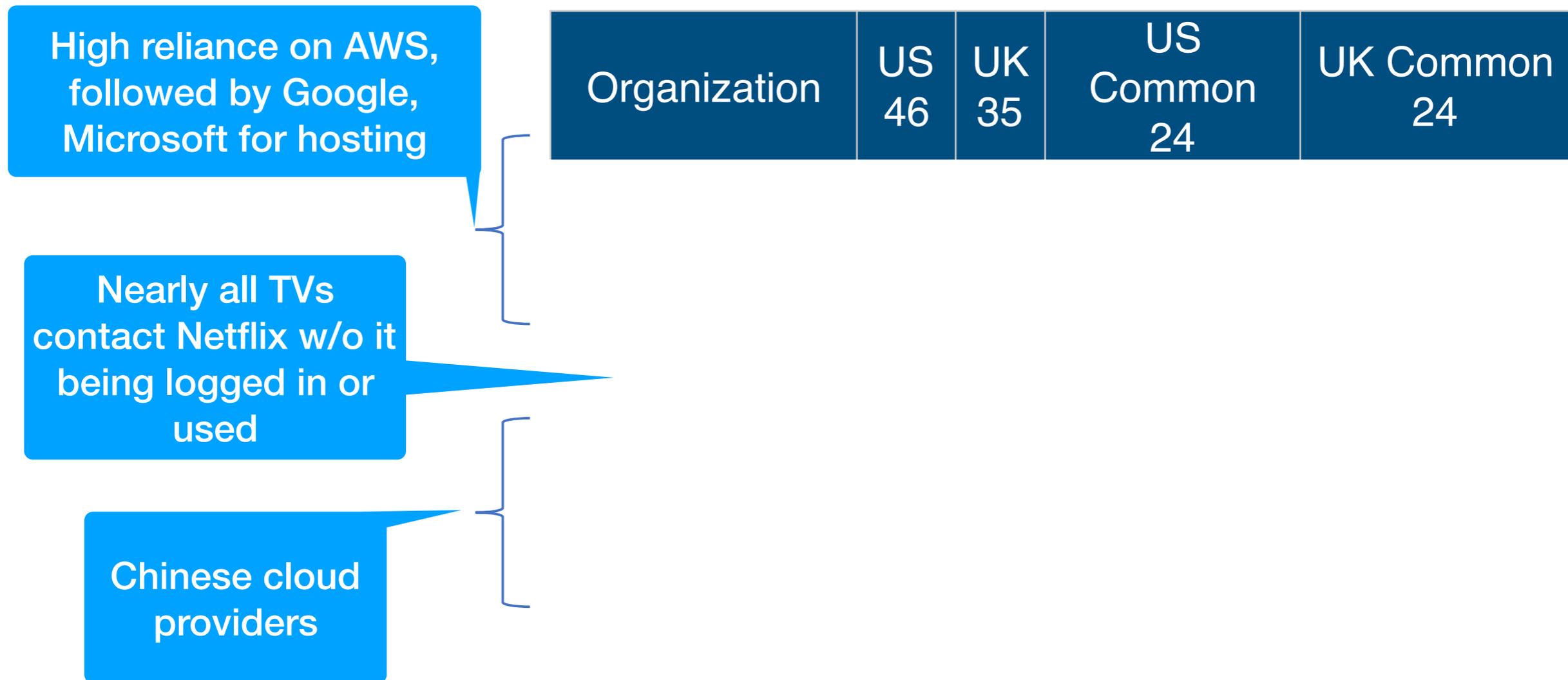
# Destination Characterization



Many devices contact outside testbeds' privacy jurisdictions\*

\*BBC World News: "Who Has My Data? - BBC Click". <https://www.bbc.co.uk/news/technology-48434175>

# Who is Contacted by Many Devices?

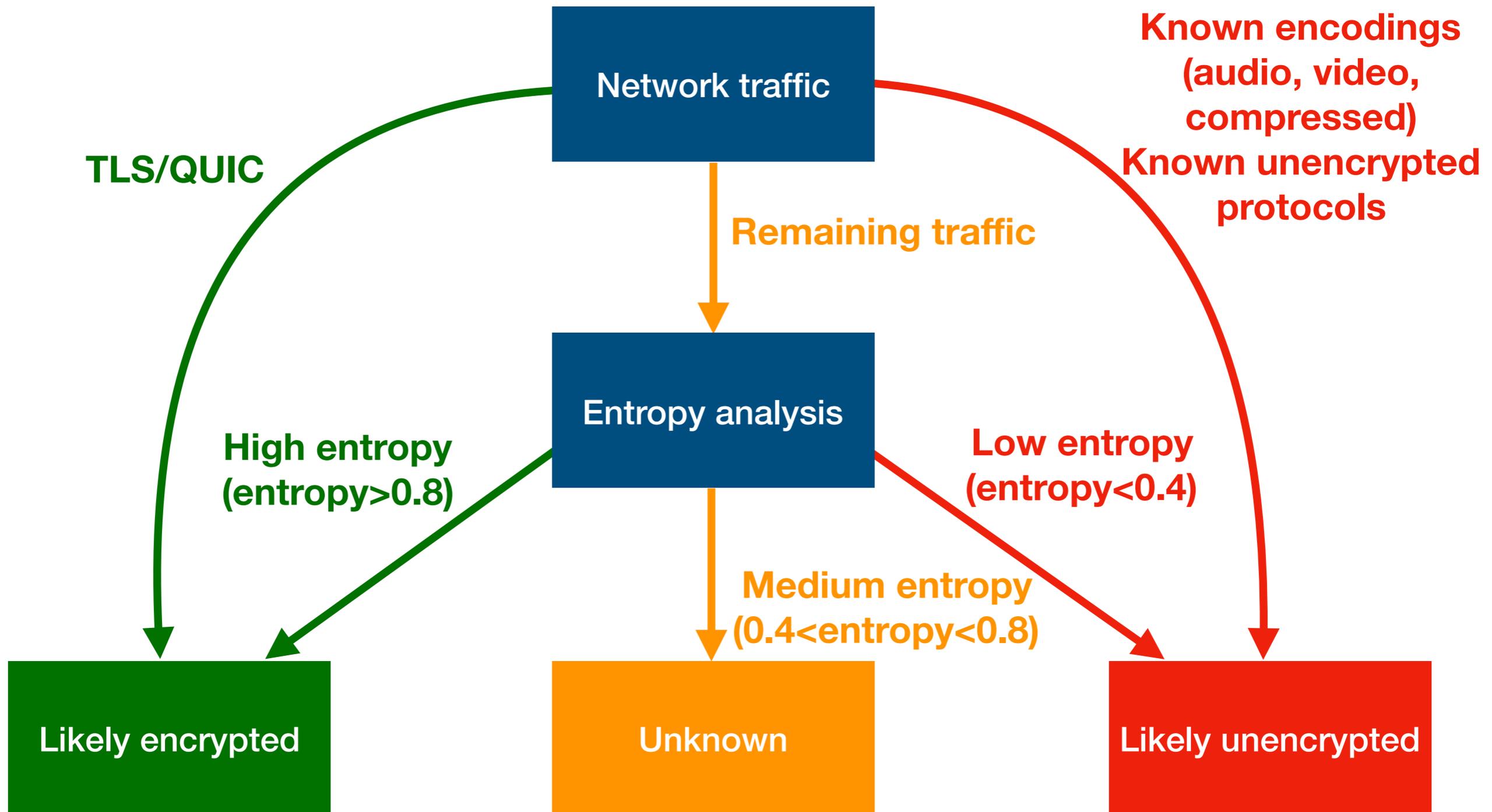


- Non-first party organizations receive information from many IoT devices
- US devices tends to contact more

# Research Questions

- What is the destination of network traffic?
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# Is the Traffic Encrypted?



Entropy thresholds calculated using min and max over 5311 randomly sampled IoT traffic traces

# How Many Devices Do Encrypt Their Traffic?

Only 2/81 devices have most traffic unencrypted

	Range (%)	US 46	UK 35	US Common 24	UK Common 24
Unencrypted	>75				
	50-75				
	25-50				
	<25				
Encrypted	>75				
	50-75				
	25-50				
	<25				
Unknown	>75				
	50-75				
	25-50				
	<25				

26/81 devices have most traffic encrypted

43/81 devices have most traffic unknown

# How Much Traffic is Sent Unencrypted?

	Device Type	US 46	UK 35	US Common 24	UK Common 24
Unencrypted	Appliances				
	Speakers				
	Hubs				
	Automation				
	Cameras				
	TVs				
Encrypted	Appliances				
	Speakers				
	Hub				
	Automation				
	Cameras				
	TVs				
Unknown	Appliances				
	Speakers				
	Hubs				
	Automation				
	Cameras				
	TVs				

Cameras and TVs have the most recognizable unencrypted traffic

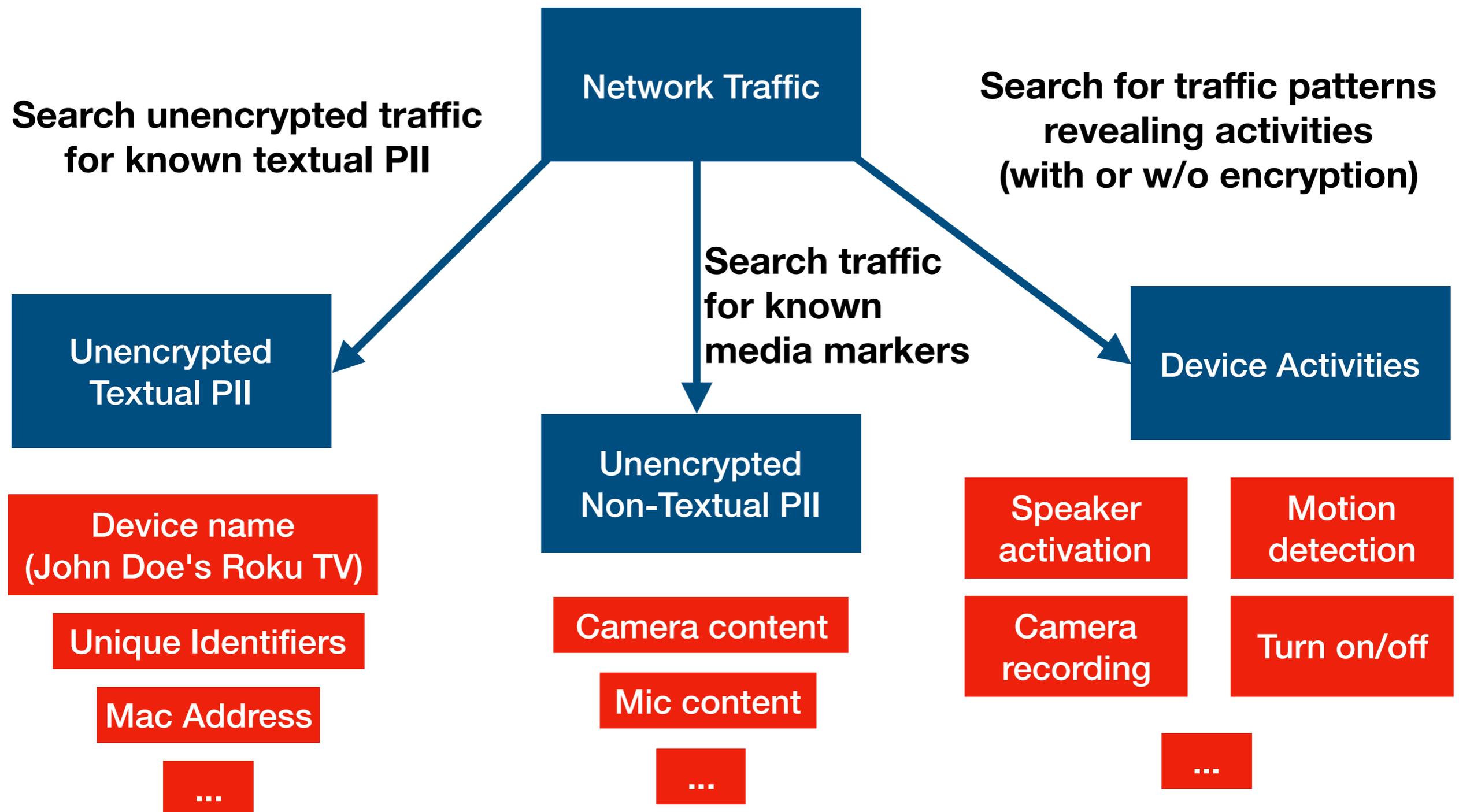
Speakers and TVs have the most recognizable encrypted traffic

26/81 devices have most traffic unrecognizable

# Research Questions

- What is the destination of network traffic?
- To what extent is the traffic encrypted?
- **What content is sent?**
- Does a device expose information unexpectedly?

# What Information is Sent?



# Unencrypted Content Leakage

**MagicHome LED**



**Samsung Fridge**



**Insteon Hub**

**PII: MAC Address unencrypted!**

**PII: MAC Address and Timestamped Video unencrypted each time a motion is detected!**



**Xiaomi Camera**

## Other unencrypted content

- Device toggle actions (e.g., on-off)
- Firmware updates
- Metadata pertaining to initial device set up

# Can we Infer an Activity from Network Traffic?

## Hypothesis:

Eavesdroppers may infer **activity information** even from encrypted traffic

Interaction method  
(local, app, or voice?)

Functionality  
(e.g., toggling a light)

**Idea:** Given the traffic patterns of an activity, look for similar patterns

**Solution:** use supervised machine learning

### ML APPROACH

- Random Forest Tree Classifier
- Features (*assuming encrypted*):
  - packet size, inter-arrival times
  - min, max, mean, deciles, ...

### ML VALIDATION

- Cross validation:
  - 7/3 split, averaged across 10 times
- F1 score (val=[0,1]):
  - 0 is the worst, 1 is the best

# Device Activity Inference

- An activity is predictable when F1-score is  $>0.75$

Number of predictable devices by activity

Activity	Total devices	US	UK	US common	UK Common
Power					
Voice					
Video					
On/Off					
Movement					
Others					

Power is the most predictable activity

Number of predictable devices by category

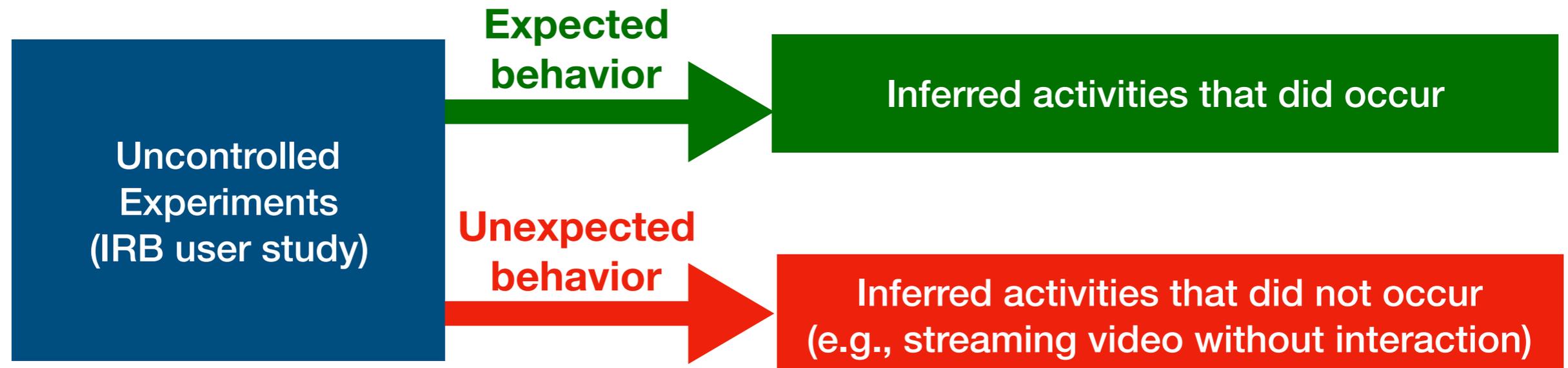
Category	Total devices	US	UK	US common	UK Common
Appliances					
Speakers					
Cameras					
Home Automation					
Smart Hubs					
TVs					

An activity/device is more predictable when it generates more traffic

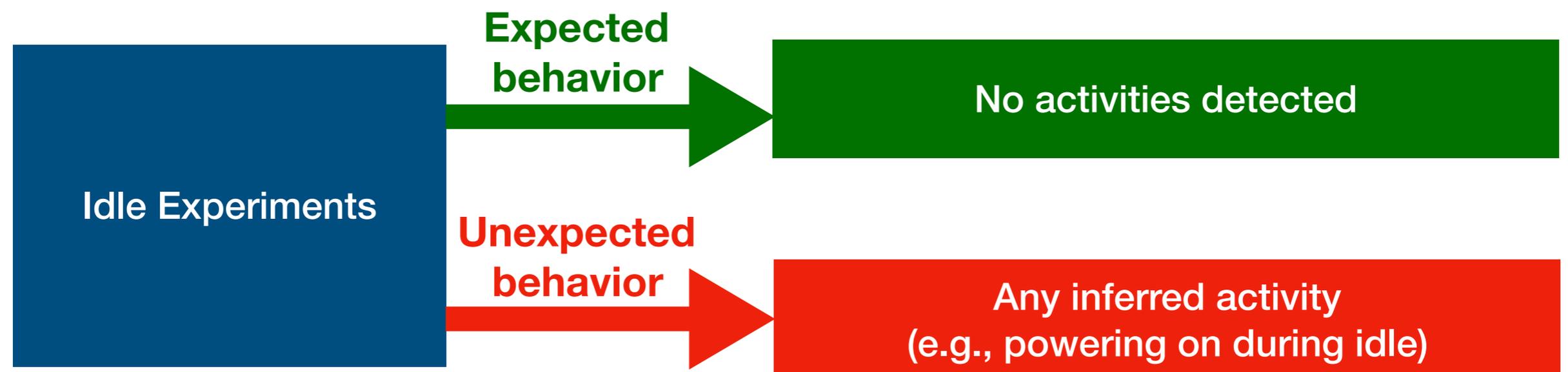
# Research Questions

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# Does a Device Expose Information Unexpectedly?



**Ground truth provided by camera and user interviews.**  
**Activity inference models provided by the controlled experiments.**



# Cases of Unexpected Behavior



**Popular doorbells**

Video recording on detected motion (cannot be disabled)



**Popular smart TVs**

Contact Netflix, Google, and Facebook unexpectedly



**Alexa-enabled devices**

Frequently falsely triggered (e.g. "I like Star Trek")

*Financial Times: "Smart TVs sending private data to Netflix and Facebook".  
<https://www.ft.com/content/23ab2f68-d957-11e9-8f9b-77216ebe1f17>*

- Other notable cases of activities detected when idle
  - **local move**: cameras triggered "falsely"
  - **power**: devices frequently (dis)connect from WiFi

# Conclusion

- First step towards more large-scale IoT measurements
- Non-first parties are contacted by many devices
- Some (24/81) devices are vulnerable to activity inference
- Inference models to identify *unexpected* activities
- **Testbed framework, data, and analysis scripts are publicly available at:**

<https://moniotrlab.ccis.neu.edu/imc19/>



# Fostering Further IoT Privacy Research

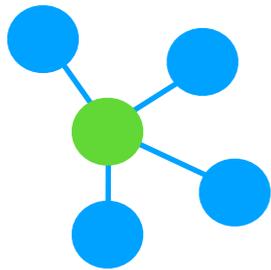


Testbed and automation code

Build your own IoT testbed

Repeat our experiments

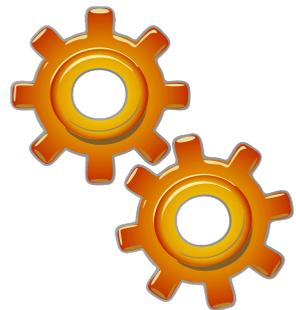
Design new experiments



Network traffic traces for  
81 IoT devices!

Idle Traffic: 112 hours!

Controlled experiments:  
34,586 tagged PCAPs!



Analysis scripts

Destination Analysis

Encryption/Entropy Analysis

Activity inference ML models

<https://moniotrlab.ccis.neu.edu/imc19/>



# Entropy Threshold

