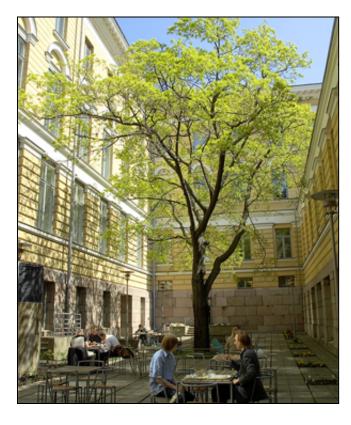
# Overview of the CS Department and NODES: Mobile and 5G Research

Professor Sasu Tarkoma, Head of Department NODES Research Group 22 March 2017

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www.cs.helsinki.fi

## **University of Helsinki**



- The largest and the oldest university in Finland
- Key data for 2015
  - 32 000 students
  - 7 900 employees
  - 300 subjects
  - 6 100 degrees/year
  - 530 PhDs/year
- Founded in Turku 1640
- Moved to Helsinki 1828

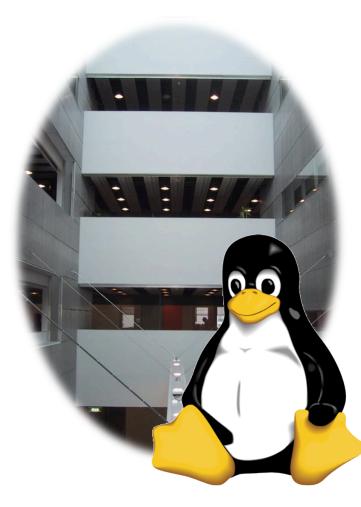
## **Faculty of Science at Kumpula Campus**

#### Departments

- Chemistry
- Computer science
- Geosciences and Geography
- Mathematics and Statistics
- Physics



## **50 Years of Excellence**



- Department of Computer Science
- Leading institution in Computer Science in Finland
  - #1 in Finland in QS Ranking 2017
  - #1 in Nordic Countries and overall #69 in Times Higher Education 2017

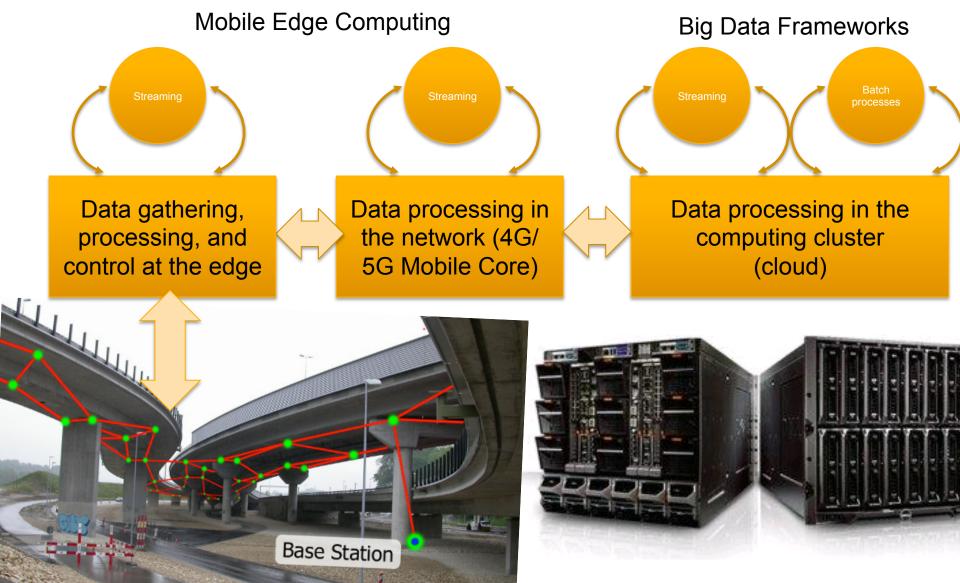
## Core CS and Data Science

- Algorithms, Data Analytics and Machine Learning
- Software Systems
- Networking and Services (NODES)
- Bioinformatics

### **Current research topics include:**

Digital services, IoT security and privacy, software-defined networks, Data Science, ...





# **Mobile Research: Carat**

## Carat team (carat.cs.helsinki.fi)



**Motivation** 

## Battery lifetime?

**Risk level?** 



Many heterogeneous, active devices and many users with different intents. – What kind of behavior is **normal** or **typical**?

# **Introducing Carat**

Carat is the **first system** to use the mobile device community to detect and correct energy problems

Our method for **diagnosing** energy anomalies uses the community to infer a specification (expected energy use), and we call deviation from that inferred specification an anomaly

# Carat

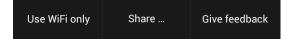
- Originated in UC Berkeley, in collaboration with University of Helsinki
- Mobile app for Android and iOS
- Currently over 850 000 users
- >2.5 TB of data, > 250 million measurements
- Research project with many directions
- http://carat.cs.helsinki.fi



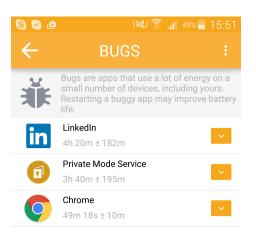
## What is Carat?

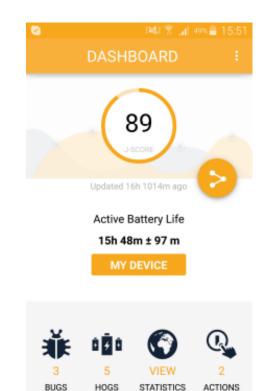
- Mobile app shows users advice: "Kill Facebook for 16m ± 41s battery life"
- Energy hogs and bugs
- Tracks user's battery life average since installation
- Places users within community with a ranking called J-Score





## New UI







in Share < 309 🛛 🕂 +1

171

## Carat: The Brilliant App That Increases Your Battery Life By Showing What Other Apps To Kill



Comment

JOSH CONSTINE ≈

Thursday, June 14th, 2012



**36 Comments** 

"Kill Pandora – Expected Battery Life Improvement: 1 hour 50 minutes" This is what you'll learn from **Carat**, an incredibly useful free **new iOS** and **Android app** that's the first to give you personalized mobile battery life-saving recommendations.

Carat quietly takes measurements from you device, does some math, combines it with other people's anonymized data, and sends back tips on if you should update your OS, kill or restart apps, and how many more minutes of tablet or phone fiddling you'll gain.

As battery tech is expected to improve slowly, some say increasing life just 5% a year, and as we get faster processors, more powerful apps, and brighter screens, everyone could use a Carat in their pocket.

Suddenly...





data, and sends back tips on if you should update your OS, kill or restart apps, and how many more minutes of tablet or phone fiddling you'll gain.

As battery tech is expected to improve slowly, some say increasing life just 5% a year, and as we get faster processors, more powerful apps, and brighter screens, everyone could use a Carat in their pocket.

Suddenly...

#### **Febacke** 11:35 PM Carrier 🛜 By To improve battery life.. 1:45 PM arrier 🔶 **Restart Evernote** Energy Hogs Hog Detail Expected improvement: 4d 20h 11m 25s 'o Kill **Cover Orange** Upgrade the Operating System Expected improvement: 4h 4 Free Carat app finds 'energy hogs,' Help Spread the Word! Expected improvement: +100 'energy bugs' on iOS or Android (Updated 1d 4h 27m 14s devices

**TOP STORIES** 

171

ANDROID | JUNE 14, 2012 | BY: MICHAEL SANTO

<b>₽</b> 34	2	0	£	0	Get Tech Gear alerts!		
~			-		Email	Sign up	
f Like	Tweet	Q +1	1	🔤 email	Eman	Sigirup	



Sι

Carrier 🤤

1:44 PM

running apps: (View Process List) 🔊

Your J-Score:

Average Battery Life:

(Updated 15s ago)

OS version:

device model:

memory used:

memory active:

70 🛛

5.1 🕑

0

11h 7m 32s

Simulator 📀







## iOS and Android app helps you get more from your battery

**Summary:** Carat has been developed by a team of scientists from the UC Berkeley electrical engineering and computer science department's Algorithms, Machines, and People Laboratory (AMP Lab).



By Adrian Kingsley-Hughes for Hardware 2.0 | June 15, 2012 -- Updated 10:21 GMT (03:21 PDT)



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## iOS and Android app helps you get more from your battery

**Summary:** Carat has been developed by a team of scientists from the UC Berkeley electrical engineering and computer science department's Algorithms, Machines, and People Laboratory (AMP Lab).



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# Statistics (October 2016)



471 645 Android and iOS apps 10% energy hogs, 4% energy bugs

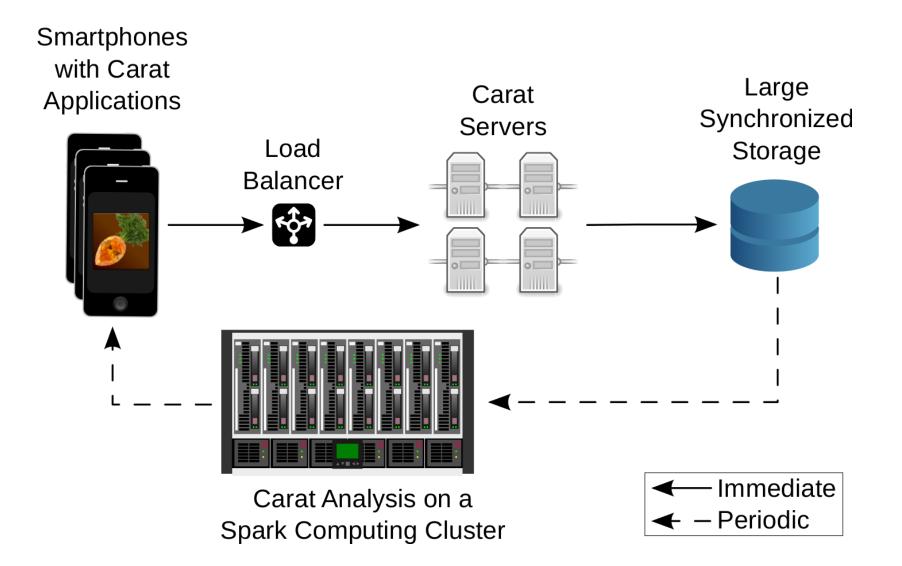
50% of devices have at least one energy bug

Android has a long tail of different device types.

carat.cs.helsinki.fi/statistics

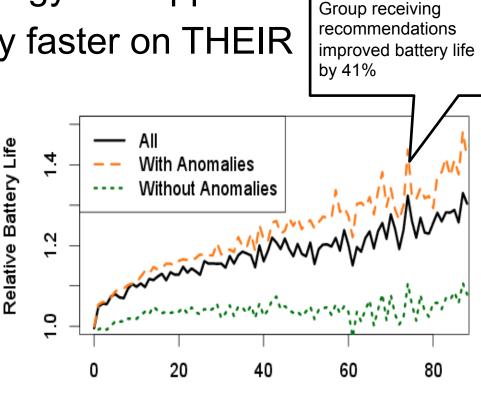


## The Carat project: System



# What is Carat?

- Users see Hogs, high energy use apps
- And Bugs that use energy faster on THEIR device than on others
- Users with these
  issues quickly see
  battery life benefits
  once they are
  addressed



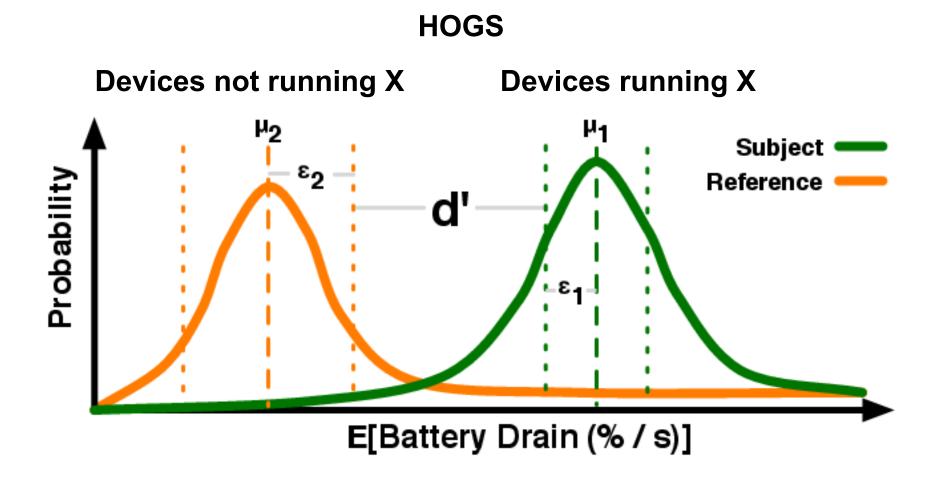
Days Since First Report

- Average improvement 20%
- Those with energy anomalies can improve 41%

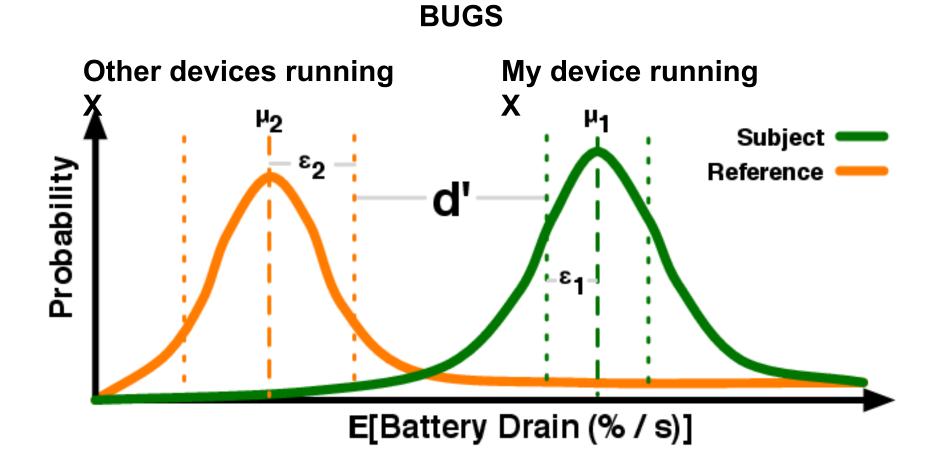
## The Data Analysis

- Samples are combined to obtain energy drain probability distributions (with features)
  - Users, Apps, App and User pairs, OS versions, Device models
- Distributions are compared using the distance between their 95% confidence interval error bars
  - If a distribution has a positive distance from another and a higher mean, it is a:
  - Hog (for an app vs the distribution for other apps)
  - Bug (for app & user combination vs other users of the same app)

## **Hogs and Bugs**



## **Hogs and Bugs**

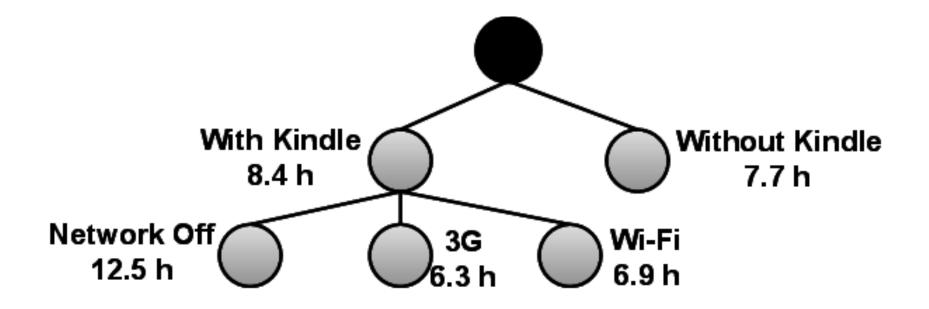


# **Collaborative Data Gathering**

Each device collects: Battery life, timestamp, running apps, context/system settings

- The data is combined and the results for your apps and your device are sent back to you
- Context feature analysis: how various context features affect the energy consumption of the device
- Collaborative aspect: We observe trends in the community, as well as how your device is different
- The method can be used for phones, sensors, houses, base stations, servers, laptops, ... anything that generates measurements

## Example: The Kindle WhisperSync bug



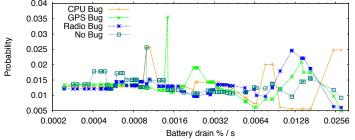
The decision tree allows "what-if" analysis and the generation of recommendations

## **Project Infrastructure**

- Data Analysis: Amazon EC2
  - -10 x X-Large VM (4 cores, 15G memory)
- Server facing mobile devices: Amazon EC2
  - -4 x medium VM (1 core, 4G memory)
  - Load balancer, independent DNS name for easy changing of infrastructure when required
- Amazon S3
  - Storage of data (incoming 0.5-1.0 GB / week)

## **Lessons learned**

- Research prototype != product
- It is not easy to scale
  - 100 000 users in one day when we launched
  - Scaling will cost, cloud is not free
  - Managing clusters is not easy
- Design system so that it can evolve (no hardcoding, extensible formats)
- Validation is not easy
  - Ground truth
  - Injected bugs, validated bugs

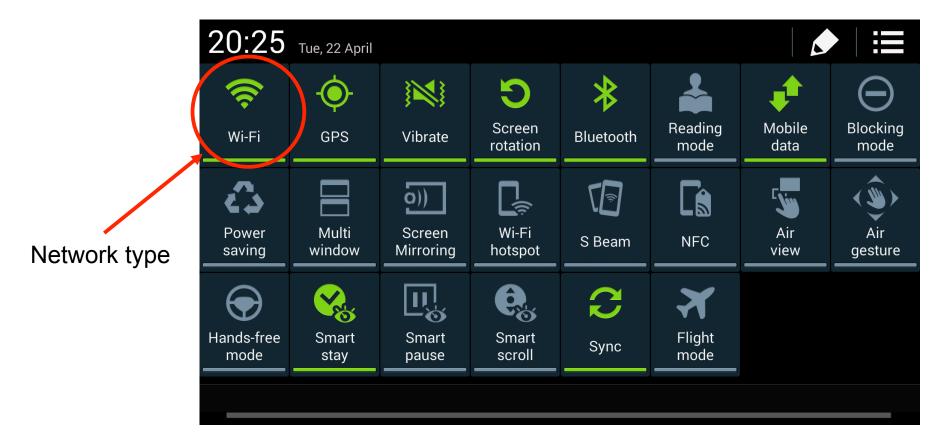




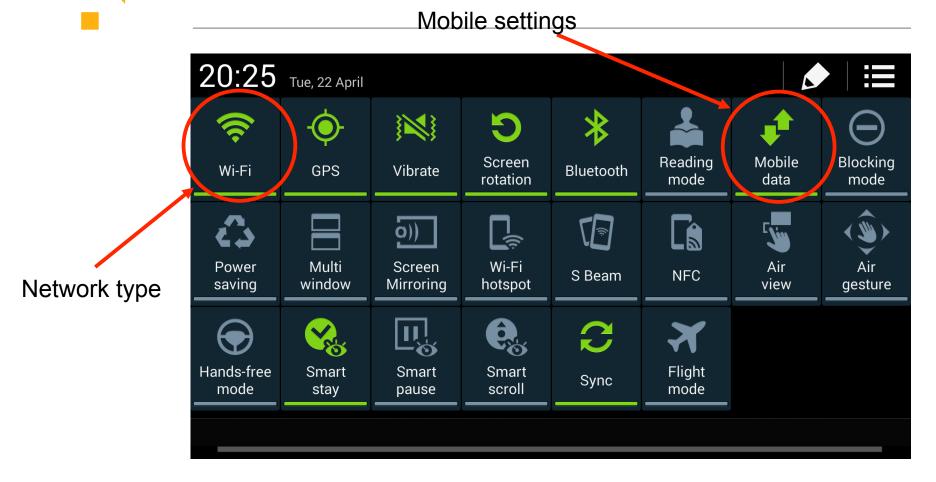
## **Energy efficient configuration?**

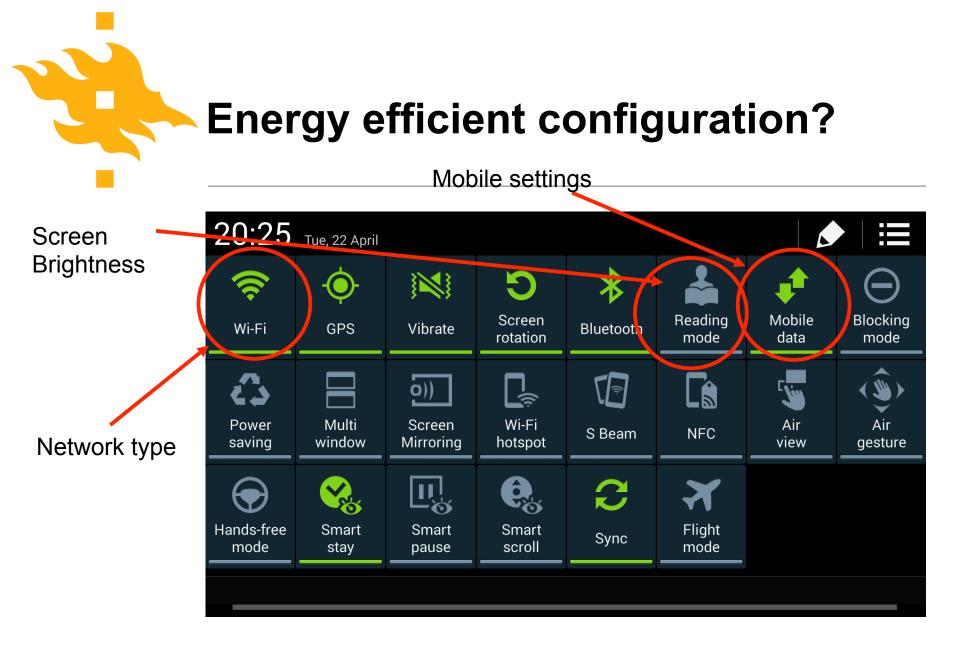
20:25 Tue, 22 April							
((r-	Ó	Ŵ	ß	*		4	$\bigcirc$
Wi-Fi	GPS	Vibrate	Screen rotation	Bluetooth	Reading mode	Mobile data	Blocking mode
Power saving	Multi window	<b>o</b> )) Screen Mirroring	Wi-Fi hotspot	S Beam		Air view	Air gesture
Hands-free mode	Smart stay	Smart pause	Smart scroll	<b>C</b> Sync	Flight mode		

## **Energy efficient configuration?**

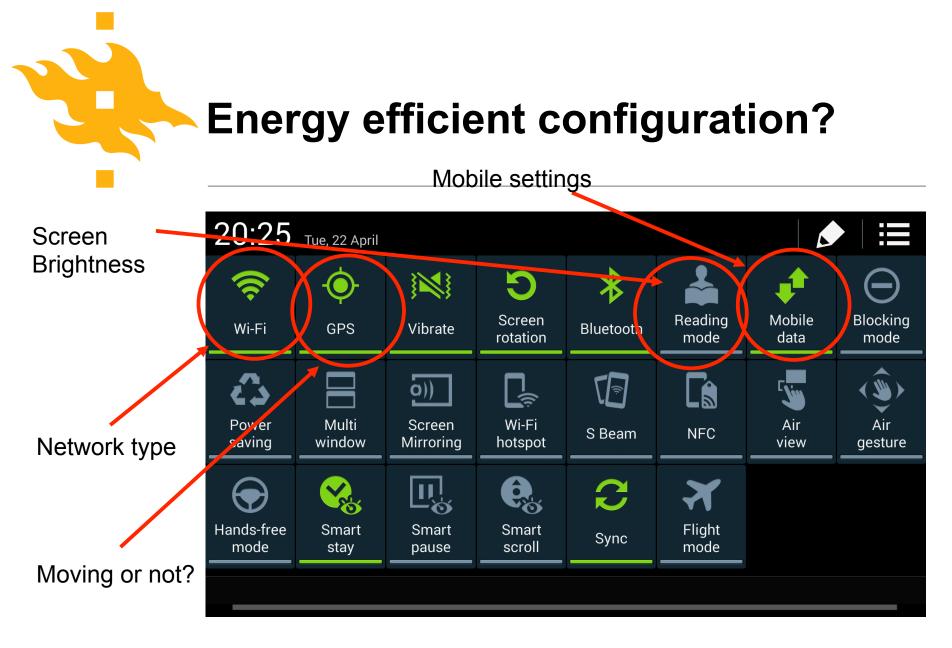


## **Energy efficient configuration?**





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## **Selected findings**

- Wi-Fi signal strength dropping one bar can result in over 13% battery loss
- High temperature can cause 50% battery loss, and high temperature is not always related to high CPU load
- Automatic screen brightness is, in the most cases, better than manual setting
- In addition to CPU, battery temperature and distance traveled are useful in predicting battery lifetime

## **Battery lifetime – an example**

Battery Temperature	Distance Traveled	CPU Use Level	Screen Brightness	Estimated Battery Life (h)
Under 30°C	>0	Low	Automatic	8.83 - 9.12
Under 30°C	>0	Low	Manual	8.49 - 8.82
Under 30°C	>0	High	Automatic	8.09 - 8.24
Under 30°C	>0	Medium	Automatic	7.65 - 7.89
Under 30°C	>0	Medium	Manual	7.34 - 7.60
Under 30°C	>0	High	Manual	7.27 - 7.41
Under 30°C	None	Medium	Automatic	6.57 - 6.64
Under 30°C	Just wan	t to play	Automatic 9	6.28 - 6.35
Under 30°C	Noneust wall	Medium	a game.	6.13 - 6.20
Under 30°C	Netich CD	Low of	Manual	5.88 - 5.96
Under 30°C	<b>High CP</b>	Uliguse:	Automatic	5.78 - 5.82
Over 30°C	>0	Low	Automatic	5.08 - 5.22
Under 30°C	None	High	Manual	5.00 - 5.04
Over 30°C	>0	Low	Manual	4.73 - 4.88
Over 30°C	>0	High	Automatic	4.62 - 4.69
Over 30°C	>0	Medium	Automatic	4.59 - 4.70
Over 30°C	>0	Medium	Manual	4.28 - 4.39
Over 30°C	None	Medium	Automatic	4.25 - 4.29
Over 30°C	>0	High	Manual	4.08 - 4.14

[2] Ella Peltonen, Eemil Lagerspetz, Petteri Nurmi, and Sasu Tarkoma. Energy Modeling of System Settings: A Crowdsourced Approach. PerCom '15. Best Paper Award.

## **Battery lifetime**

## → 98% better expected battery life

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## **The PADS Project**

The PADS project will develop new privacy enhancing algorithms and methods for **Data Science** 

## Two methodological goals

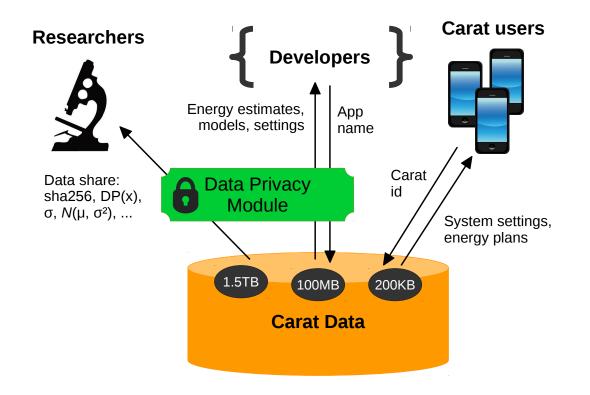
- Develop the algorithmic framework for privacy-aware predictive modelling, and
- Scalable implementation of the framework

## Two key use cases

- Large-scale genome data processing and sharing for personalized medicine, and
- Mobile and environmental sensing

#### **Research status**

Mobile sensing data analysis in progress Full paper prepared based on questionnaire study Privacy SDK outlined in the recent Big Data workshop paper.



#### **Malware Infection Rates**

We studied malware based on the dataset McAfee, Mobile Sandbox, MalGenome, ...

Malware infection rates are higher than conservative estimates (0.26% of devices)

Google says 0.12% of manually installed packages are malware, not very far from this number Our infection estimate is

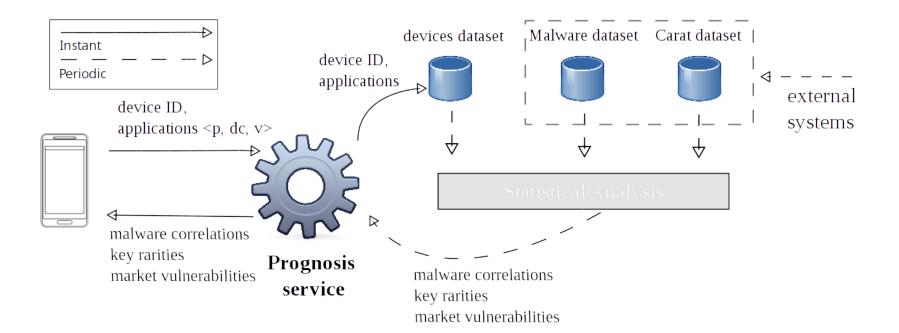
Lookout Antivirus predicts >1%

higher than previous research, but lower than some AV vendors. 4.3%

## An Early Warning System for Malware

A lightweight technique for identifying devices at risk By looking at applications that occur with malware, it is possible to predict infection 5x better than choosing devices at random

Useful for administrators, organisations (Bring Your Own Device scenario)

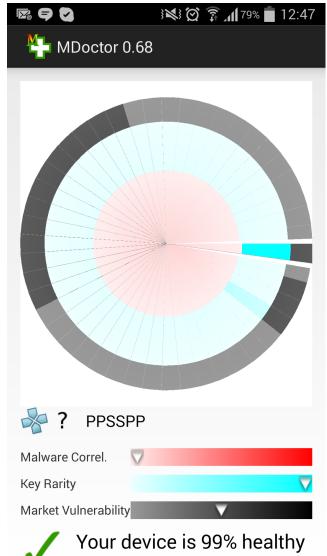


### MDoctor: Increasing Awareness of Infection Vulnerability

MDoctor shows status of applications according to a malware dataset

Infection vulnerability can be seen from device health

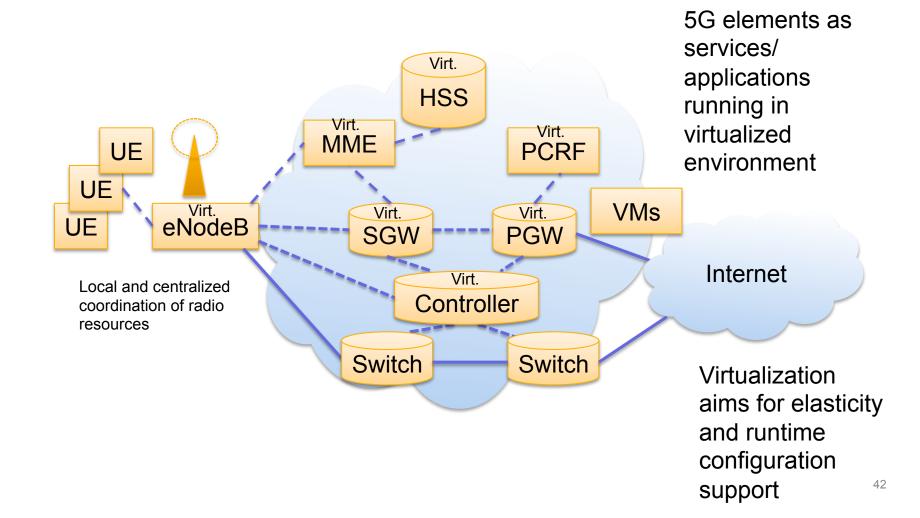
Three metrics for application analysis: malware correlation, key rarity, and market vulnerability



## **5G Research**

Te

### Starting point in 2014: LTE RAN and EPC with SDN and Cloud



#### NOKIA CENTER FOR ADVANCED RESEARCH (NCAR)



NCAR was launched in April 2016 and is a joint research center with University of Helsinki, Aalto University, and Nokia.

To foster **wider cooperation** between the universities and Nokia to enable **cross-unit research** delivering high quality results: thesis, publications, holistic concepts and demos.

# ncar.cs.helsinki.fi

### **Highlights**

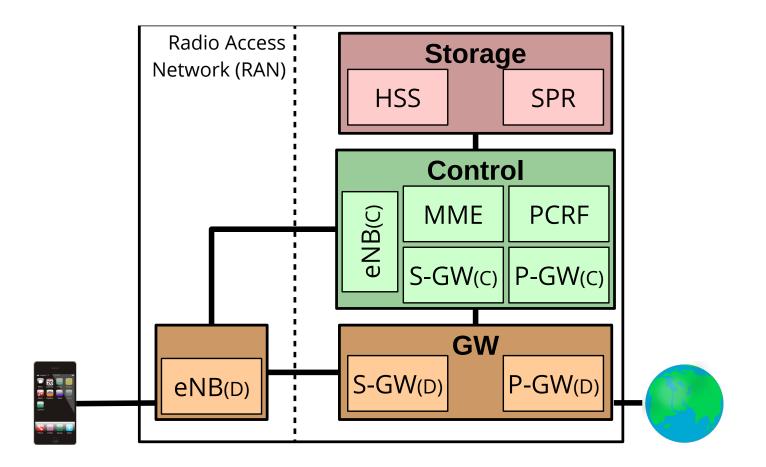
Refactoring Approach for Optimizing Mobile Networks

Coreless Mobile Networks: A state management perspective

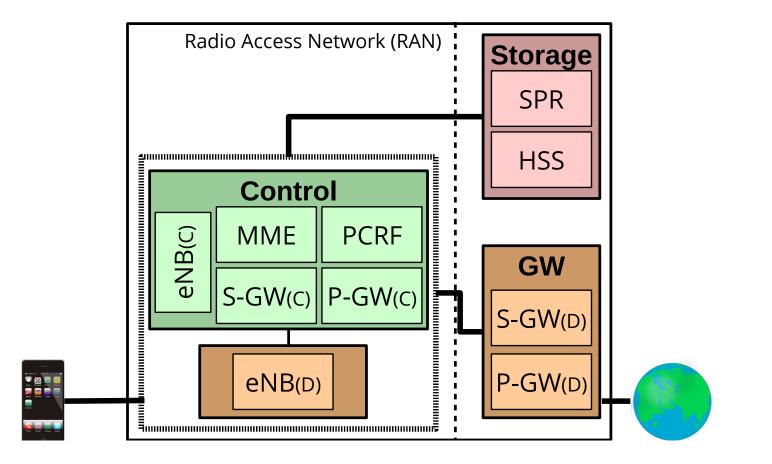
Off-the-Shelf Software-defined Wireless Networks

Service Function Chaining: Enabling Technologies and Protocols

#### **Refactoring: Thin Edge**



#### **Refactoring: Intelligent Edge**

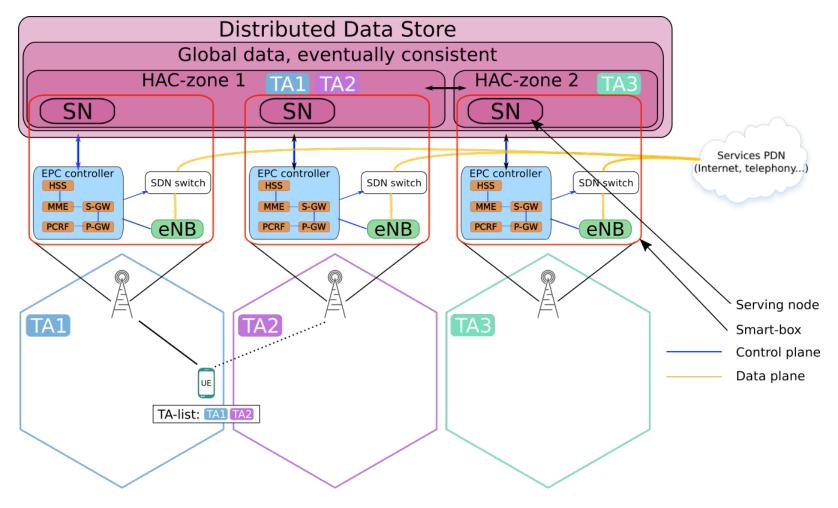


#### Refactoring Approach for Optimizing Mobile Networks

	Total number of signals per event				
	Initial Attach	Active to Idle	Idle to Active (UE)	Idle to Active (Net)	Handover (S1H)
Implementation			(	(1100)	
LTE (Baseline)	35	6	13	17	22
Thin Edge	24	6	13	16	16
Intelligent Edge	17	3	10	12	12

A Refactoring Approach for Optimizing Mobile Networks. Matteo Pozza, Ashwin Rao, Armir Abujari, Claudio Pallazi, Hannu Flinck, and Sasu Tarkoma. *Paper in IEEE ICC 2017* 

#### **Coreless Mobile Networks: A state management perspective**



Frans Ojala, 2016

## Implications

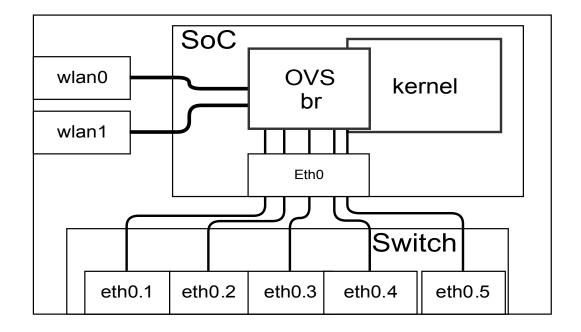
In theory, if the data store is the bottleneck, our results indicate the following numbers for a simulation of 15 eNB:

Current deployments are seeing a maximum of 1000 UE / eNB UE per area increases depending on configuration: ~84 - 740 x

5G prospects 100 - 1000 x

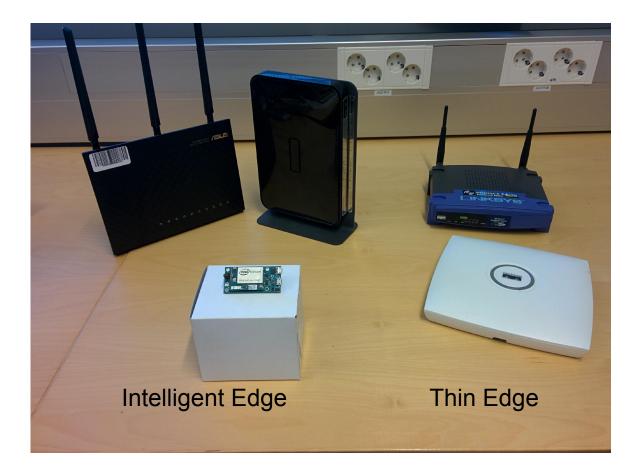
## Off-the-Shelf Software-defined Wireless Networks

- OVS in base station
- Use Wireless Isolation to force flows to OVS.
  - Either all packets or jus ARP queries depening on implementation.
- Can be used with OpenWrt capable APs or with (at leas some) enterprise APs
- Two approaches, Intelligent and Thin Edge
- Thin edge:
  - Traffic is forced to flow through external host.

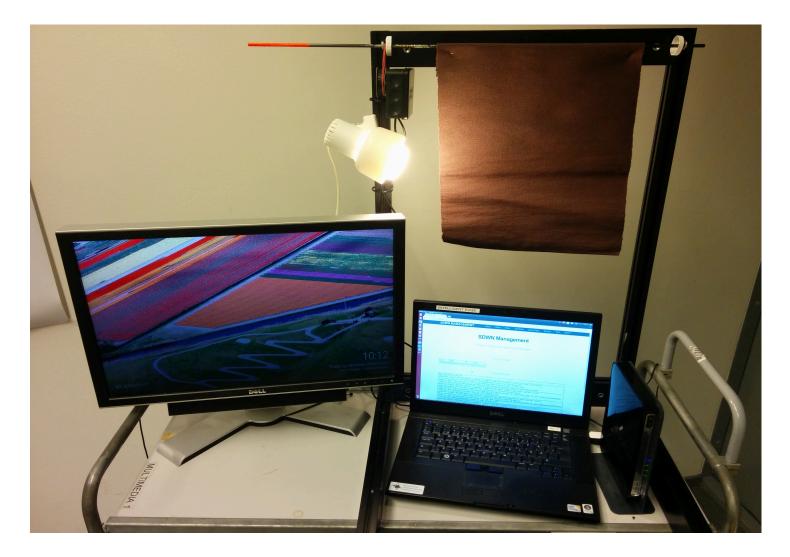


Seppo Hätönen, Petri Savolainen, Ashwin Rao, Hannu Flinck, and Sasu Tarkoma. SIGCOMM 2016 demo.

#### **Deployable on Off-the-Shelf Devices**



#### **Unified Mobile Edge for IoT Devices**



#### IoT hub running at the edge as an SFC service



# **Thank You!**

#### www.cs.helsinki.fi carat.cs.helsinki.fi ncar.cs.helsinki.fi

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