## Symbiosis of Mobile Analytics & Software Testing A practical guide to better testing of Mobile Apps



"Our business branch in 2014 is the 7:01 from Reading to Paddington.

Over 167,000 of our customers use our Mobile Banking app between 7am and 8am on their commute to work every day."



http://www.bbc.co.uk/news/business-your-money-26365616



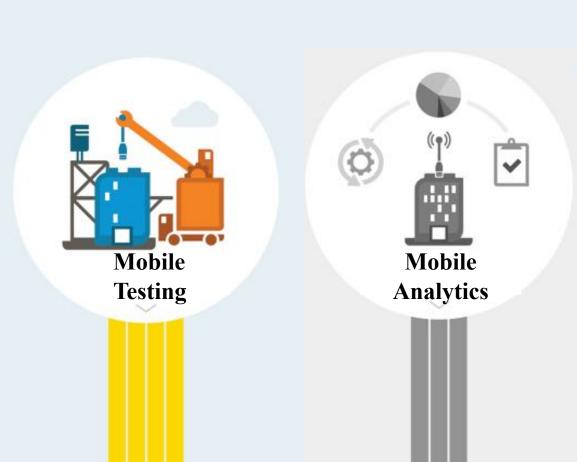


#### Maturing to a broader perspective



How to move from traditional "hands-on" testing to the design, analysis, and application of data generated by mobile apps, the users, etc?

How to better prioritize our development and testing efforts?





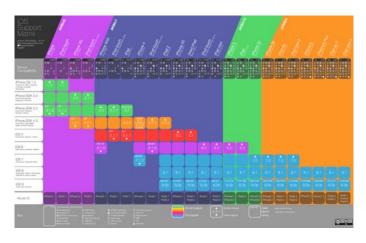
The Confluence

#### FRAGMENTATION, OR THE GODZILLA IN THE ROOM



**Device Characteristics** 

Screenshot from Lirum Info



Platform diversities

iOS Support Matrix from OpenSignal



**Usage Patterns** 

Settings, changing contexts



Fragmentation

Measure Quality Disciplines Boost Testing

Confluence **Analytics** 

#### MEASURING QUALITY SO IT CAN BE IMPROVED



**Defining metrics** ISO/IEC 25010:2011



The virtues of emulators

Screenshot from GenyMobile



Measure **Ouality** 

**Analytics** 

Confluence

#### MOBILE TESTING DISCIPLINES



1 second page load delay results in 7% drop in conversion rate, 11% drop in page views, 16% drop in customer satisfaction

Performance

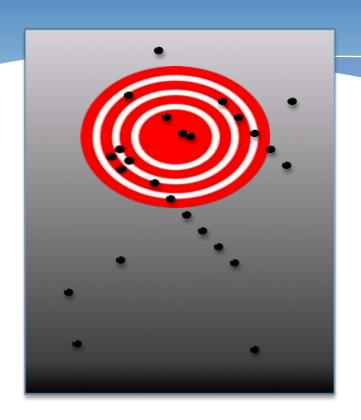
Quote from Getelastic.com



### Are you working too hard, on the wrong things?

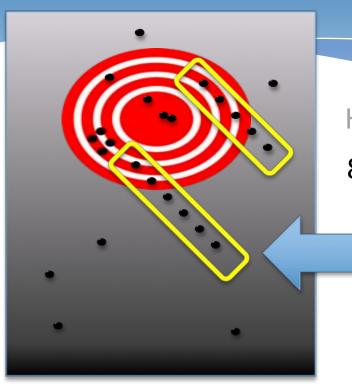
We struggle to decide what to test, how much testing is enough, etc.

### Most of our work is wasted effort



How much of our testing is on target?

### Most of our work is wasted effort



How much of our testing is on target?

80% of reported bugs not addressed

Automated Tests ©



### **Know your users**



Custom drink feature removed[1]

=> 1 star feedback ratings



Parallel Kingdom<sup>[2]</sup>

Regular users generate 2.5x daily revenues

[1] example from App Quality book

[2] example from Tale of Two Apps

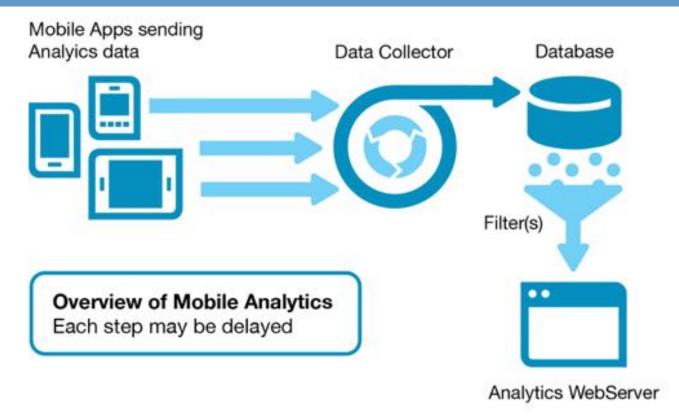
#### **Understand the effects**

- \* Battery drain varied by 3x for similar hardware specifications
- \* Custom code added for **Kindle Fire** to reduce brightness
  - \* 40% less battery drain
- \* Higher network latencies reduced interactivity by 40%
- \* Users preferred Wi-Fi
  - \* 69% for Parallel Kingdom, 58% for StudyBlue
- \* Tablets 2x usage
- \* Pull-out keyboard also increased usage

## Feedback Cycles



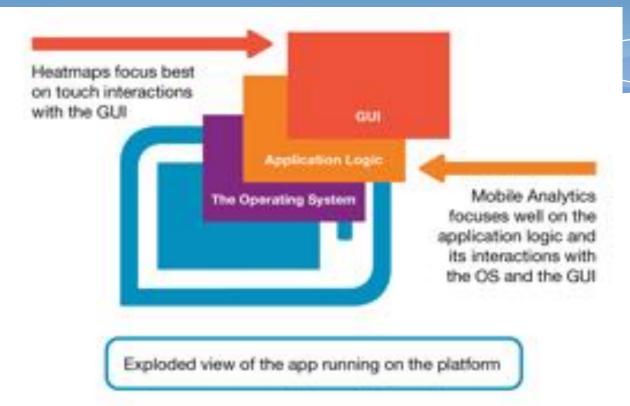
## **TOPOLOGY**



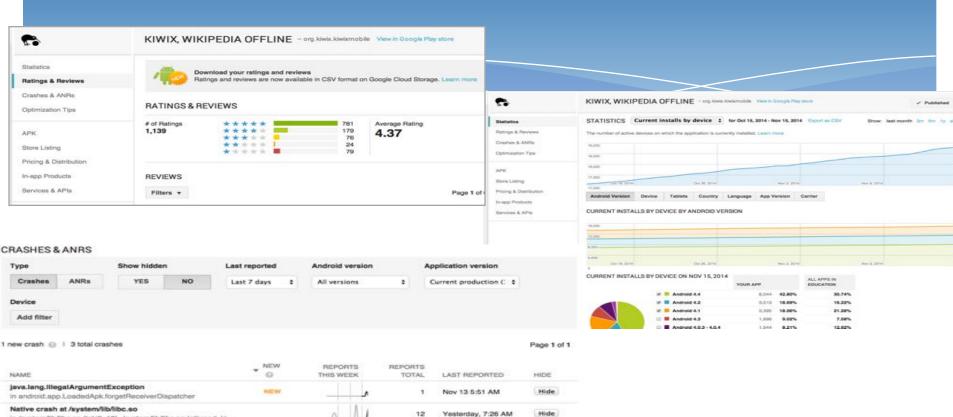
Mobile Analytics

# Some examples

## Layers of an App



## Developer Console (Google Play)

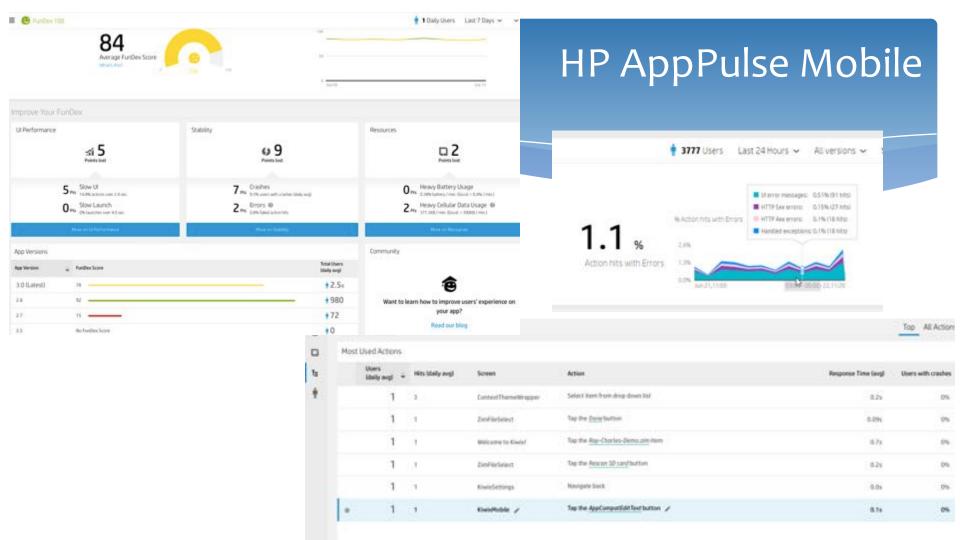


Nov 13 11:55 AM

in /system/lib/libc.so (tgkili+12), /system/lib/libc.so (pthread\_ki...

java.lang.lllegalStateException

In android.widget.ListView.layoutChildren



## Heatmaps





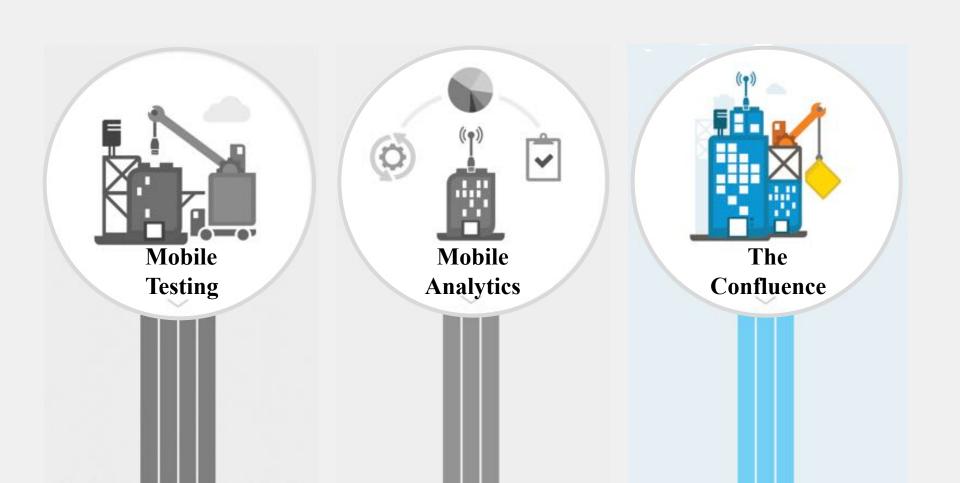
Slide

Zoom/Dezoom

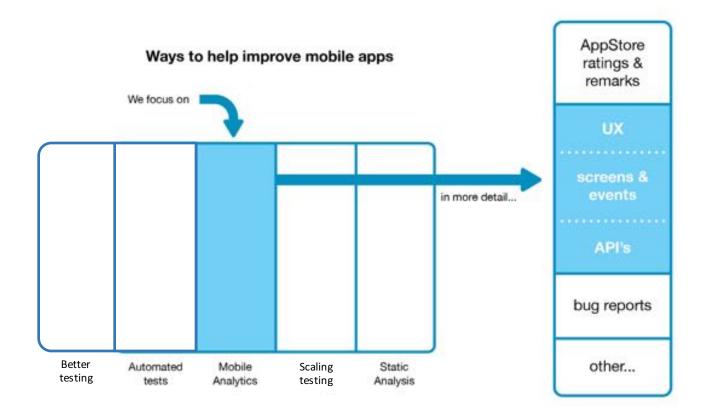
"8 Heatmap Tests that reveal visitor behavior."

- 1. The Link Test Are visitors clicking on dead ends?
- 2. The Distraction Test Are irrelevant elements distracting visitors?
- 3. The Information Test Is there missing information from your page?
- 4. The Device Test Does your page work on different screen sizes?
- 5. The Depth Test Is all your content easily reachable?
- 6. The Engagement Test What are your site visitors really looking for?
- 7. The Fold Test Does your page have what it takes to keep your visitors?
- 8. The Header Test Is your page header helping or hurting your site?

Source: https://www.hotjar.com/heatmaps



# One Way of Many



Confluence

# Tests based on usage

## Some benefits of analytics data

- \* No longer limited to what "we" think "they" need/do/etc. Discover how the app behaves across virtually all the population
- \* Lower cost of operations
- \* Real usage can help drive our testing and analysis; brings realism to our testing

## Testing based on analytics data

#### Testing based on information

- Popularity & volumes
- \* Locales
- \* User-flows, activities, etc.
- \* Crashes

#### Aims include

- \* Fast reproduction
- 1st-hand learning in controlled env.

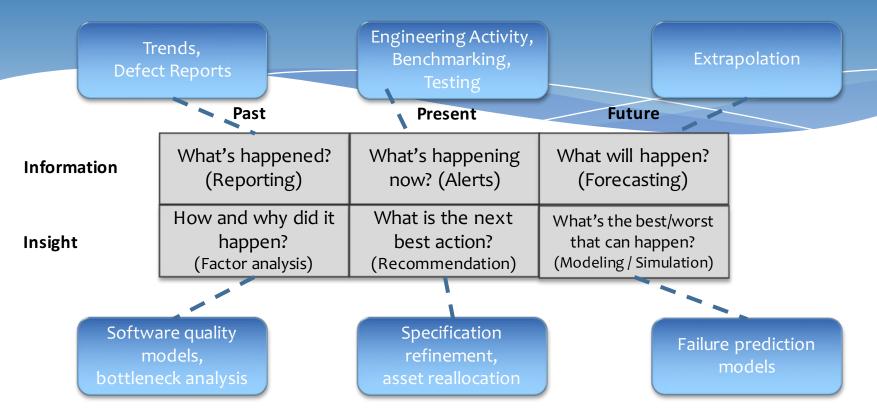
#### Testing based on insights

- Delta's (rate of change)
- \* Crossing thresholds
- \* Anomalies

#### Aims include:

Maximising insights, agility, and ability to adapt & respond

### Analytics for Software Development



## Analytical Questions: Past

Trends,
Defect Reports

**Information** 

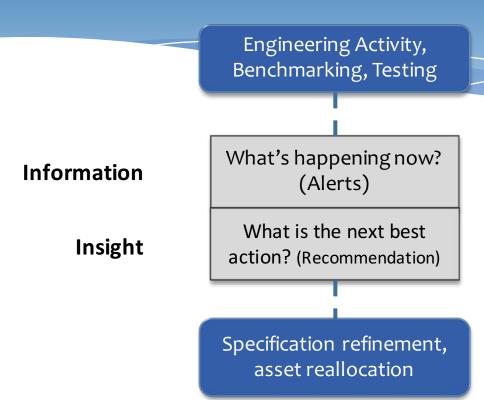
Insight

What's Happened? (Reporting)

How & why did it happen? (Factor analysis)

Software quality models, bottleneck analysis

## Analytical Questions: Present



## Analytical Questions: Future

Extrapolation

What will Happen? (Forecasting)

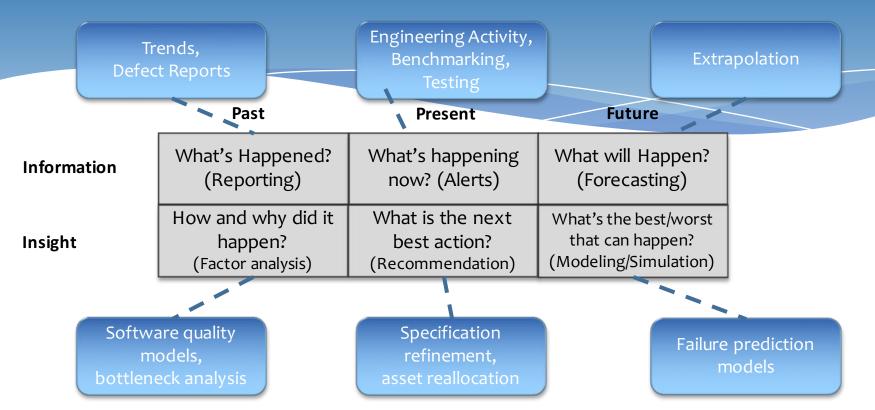
What's the best/worst that can happen? (Modeling / Simulation)

**Information** 

Insight

Failure prediction models

### Analytics for Software Development

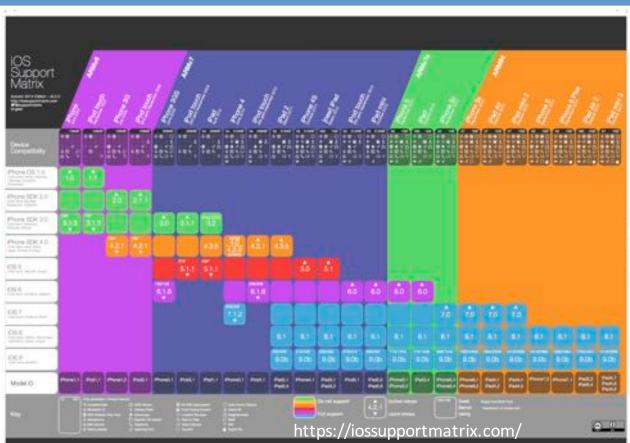


http://research.microsoft.com/pubs/136974/foser-2010-buse.pdf

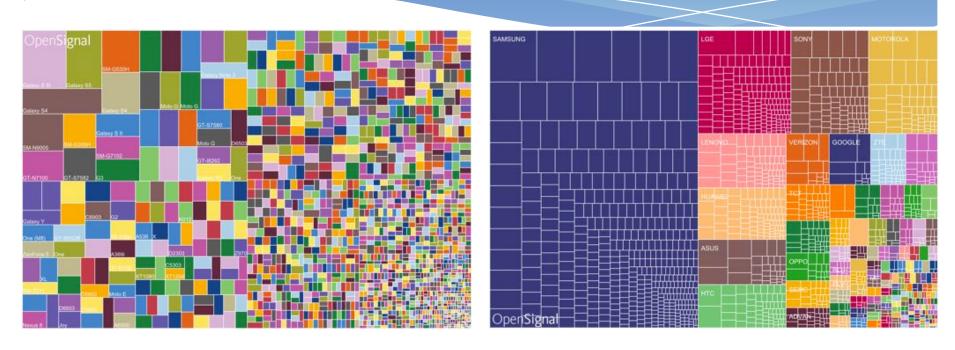
Perennial Question

# How many devices are enough?

## iOS Matrix

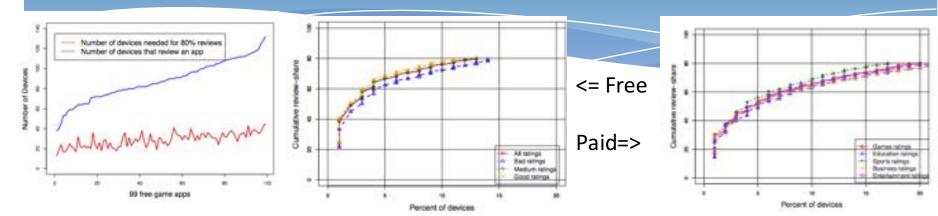


## Android: Devices Matrix



Used with permission from OpenSignal.com

## How many devices are enough?



Number of devices to cover 80% of reviews

"Prioritizing the devices to test your app on : A case study of android game apps"

#### Prioritizing The Devices To Test Your App On: A Case Study Of Android Game Apps

Hammad Khalid', Meiyappan Nagappan', Emad Shihab', Ahmed E. Hassan'
'Softwar Anaysis and Intelligence Lab (SAIL), Quener's University, Kingston, Canada
'Department of Software Engineering, Rochester Institute of Technology, Rochester, USA
'Department of Computer Science and Software Engineering, Concordia University, Montreal, Canada
'hammad@cs.queensu.ca, 'mei@se.rit.edu, 'eshihab@cse.concordia.ca,
'ahmed@cs.queensu.ca

#### ARSTRACT

Sur ratings that are given by the users of mobile apps directly impact the revenue of its developers. At the same time, for popular painforms like Android, these apps must run on hundreds of devices increasing the chance for device-specific problems. Devicespecific problems could impact the rating assigned to an app, given the varying capabilities of devices (e.g., hardware and ordware). To To fix device-specific problems developers must test their apps on a large number of Android devices, which is could want distilligation.

Therefore, to help developers pick which devices to test they appear, our, perspose using the devices that are mentioned in surpers, our, perspose using the Porkies that are mentioned in the super reviews. We mine the user reviews 679 fine game apps and find thus, apprencies user reviews from a large number of devices. However, most of the reviews (80%) originate from a sail subset of devices (on everage, 33%) they therefore, we find that developers of new game apps with no reviews can use the review data of similar game apps to select views can use the review data of similar game apps to select the devices that they should focus on first. Finally, among the set of devices that generate the most reviews for a napp, we find that devices that generate the most review for a napp, we find that for the first of the first original for the same first of the same first of

and large companies are developing an enormous amount of applications (callied mobile app), designed to man on Antroid devices. However, the top-rated or the featured apps in the app markets, are the apps with the most downloads, and hence the most revenue [2, 3]. Also the app market is very competitive, especially for game app developers who have to compete with almost choosing the energy of apps. To compate in this encopier with almost charge or entergery of apps. To compate in this environment, developers need to get (and maintain) good ratings for their apps [2]. This cam be difficult since users are easily annoyed by buggy apps, and that annoyance could lead to bad ratings [4, 5]. Hence, app developers need to text their apps thoroughly not different devices to avoid a

To make matters worse, there exists a large number of Android devices, each with its own nuances. In fact, dealing with device specific issues of (the many) Android devices is considered one of the biggest challenges developers face when creating an Android app [6]. A 2013 survey from Appeclerator, which has aggregated results from similar such survers in the past three vears, shows that Gaining confidence

## Fools rush in...

## Precision & accuracy



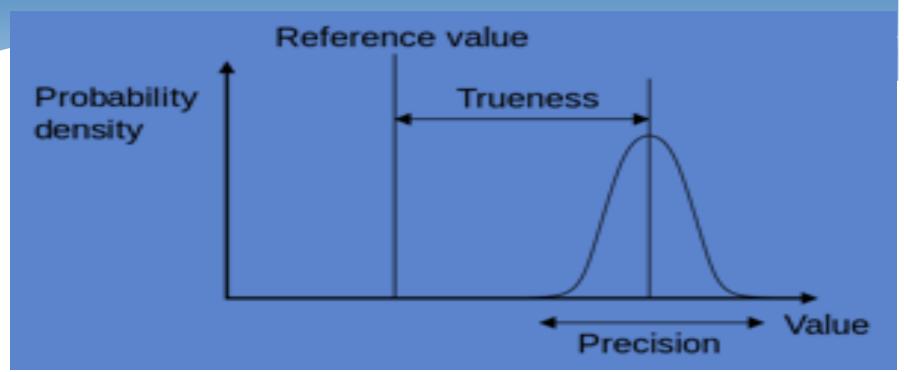
Precision: repeatability

https://en.wikipedia.org/wiki/Accuracy\_and\_precision

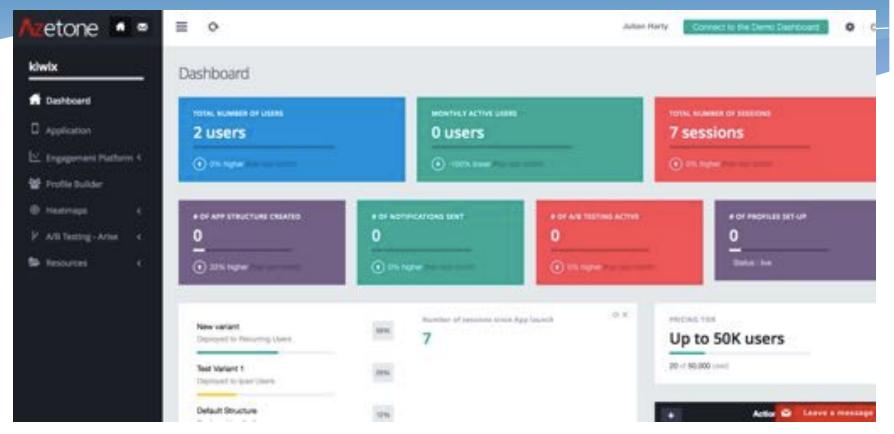


Accuracy: on target

## Precision & accuracy



# A system at rest?



## Divergent answers increase doubt

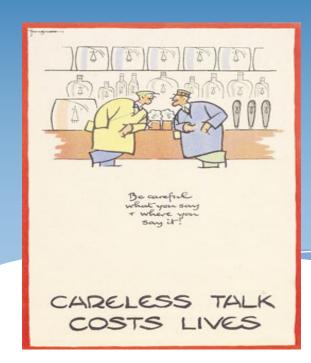
\* A tale of two three mobile analytics libraries (and what happens when bonuses are on the line...)

\* Where were the testers (part one)?

Of Mobile Analytics

## The Dark Side

#### Do no harm to a neighbour



#### THIS POPULAR FLASHLIGHT APP HAS BEEN SECRETLY SHARING YOUR LOCATION AND DEVICE ID

MORE THAN 50 MILLION USERS WERE LEFT IN THE DARK ABOUT HOW BRIGHTEST FLASHLIGHT SHARED THEIR LOCATION AND DEVICE INFORMATION WITH THIRD PARTIES.

sourcedna

Blog

iOS Apps Caught Using Private APIs
October 18, 2015

#### CHINESE TAOMIKE MONETIZATION LIBRARY STEALS SMS MESSAGES paloalto

Palo Alto Networks WildFire has captured over 18,000 Android apps that contain this library. (October 21, 2015)

Where were the testers (part two)?

http://www.vam.ac.uk/users/node/1777

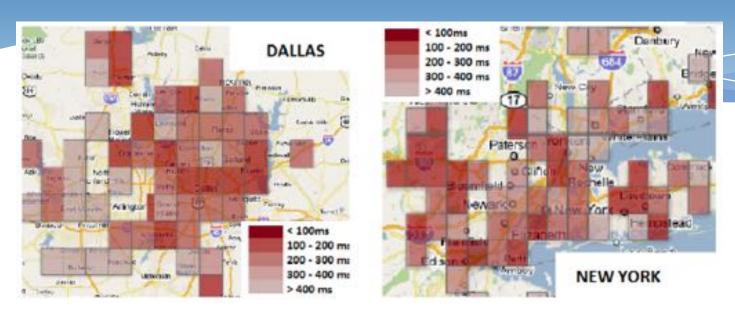
A Study of Third-Party Tracking by Mobile Apps in the Wild ftp://ftp.cs.washington.edu/tr/2012/03/UW-CSE-12-03-01.PDF

https://sourcedna.com/blog/20151018/ios-apps-using-private-apis.html

http://researchcenter.paloaltonetworks.com/2015/10/chinese-taomike-monetization-library-steals-sms-messages/



#### Reducing precision to protect privacy



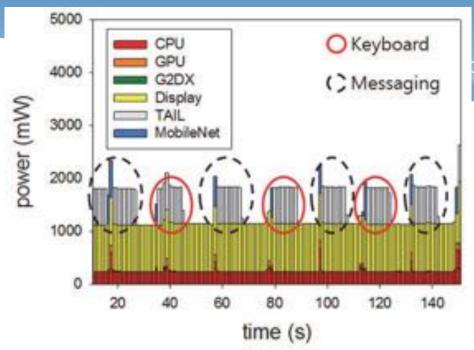
10km x 10km squares

From: Capturing Mobile Experience in the Wild: A Tale of Two Apps Figure © ACM

Listening is a means to an end

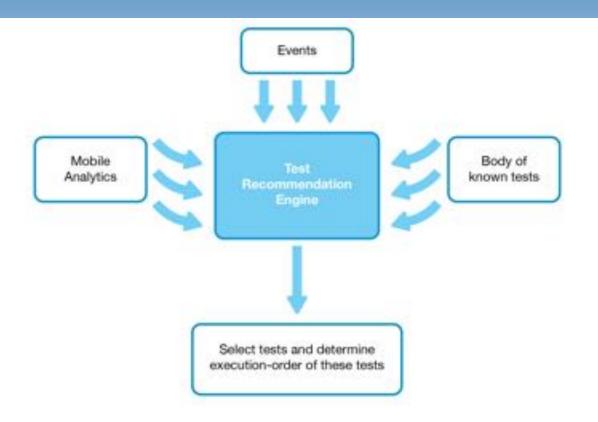
## Necessary but not sufficient

#### **Correlation? Causation? Appropriate?**



From: User Interaction-based Profiling System for Android Application Tuning Figure © ACM

## Test Recommendation Engine

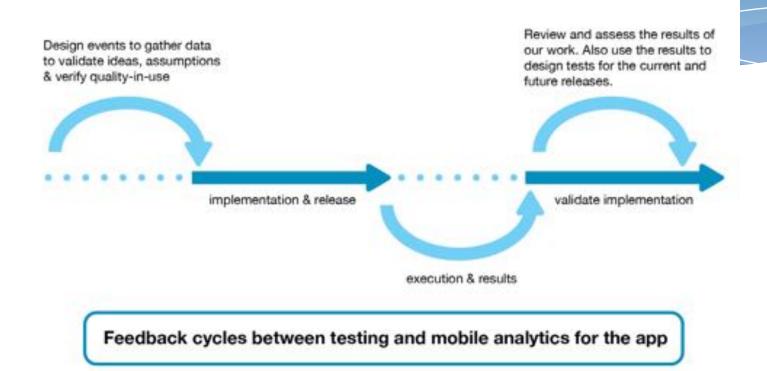


#### Beware the automation bias

"When presented with an automated solution 40% of pilots reasoned less or none at all"

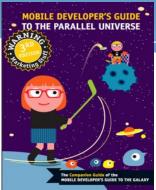
Automation Bias in Intelligent Time Critical Decision Support Systems http://citeseerx.ist.psu.edu/viewdoc/download?doi=10.1.1.91.2634&rep=rep1&type=pdf

# Getting Involved

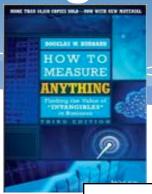


Further reading





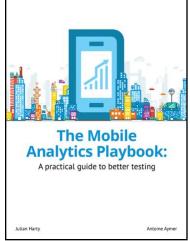






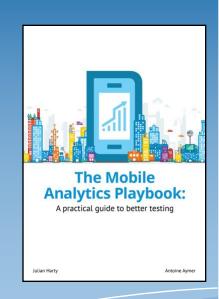
Email me for academic references
These books available at:

- [1] http://enough.de
- [2] http://wip.org/
- [3] http://www.appqualitybook.com/
- [4] http://www.howtomeasureanything.com/
- [5] https://www.nostarch.com/silence.htm



Later: julianharty@gmail.com

Q&A Now?



https://www.surveymonkey.com/r/853LC3K

Sources of data

## For Mobile Devices

# Ways to collect data

What	Technique	Tool(s)
Device details	Research, App Store, Mobile Analytics	API calls, published
GUI elements and layout	Static Analysis, Dynamic Analysis	Code Quality utilities Test Automation Tools
Run-time environment	Queries, Mobile Analytics	Shell commands
Run-time behaviours	Instrumentation, Logging, Mobile Analytics	Development Tools, Test Automation Tools
Network traffic	Observation, Logging	Network Analyzer
Test results	Explicit collection	Data collection utilities, Test Automation Tools
App Store Data	Developer account access	Provided by App Store
Ratings & Reviews	Online access, data collection	App Store, other sources
In-app analytics	Integration of Mobile Analytics Library	Development tools

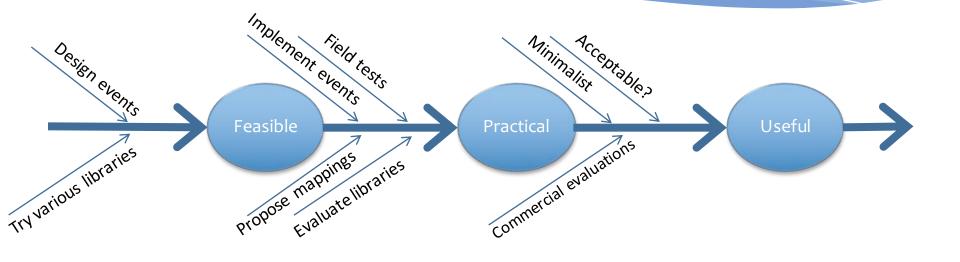
# Data sources for mobile apps

What	On device	Off device
Device details	Υ	
GUI elements and layout	Υ	
Run-time environment	Υ	
Run-time behaviours	Υ	
Network traffic	Υ	Υ
Test results	?	Υ
App Store Data		Υ
Ratings & Reviews		Υ
In-app analytics		Υ

## Sources of the data

What	Static	Runtime	Human
Device details	Υ		
GUI elements and layout	Υ	Υ	
Run-time environment		Υ	
Run-time behaviours		Υ	
Network traffic		Υ	
Test results	Υ	Υ	Υ
App Store data		Υ	
Ratings & Reviews			Υ
In-app analytics		Υ	

## Implementation Flow



Questionnaire

# Results

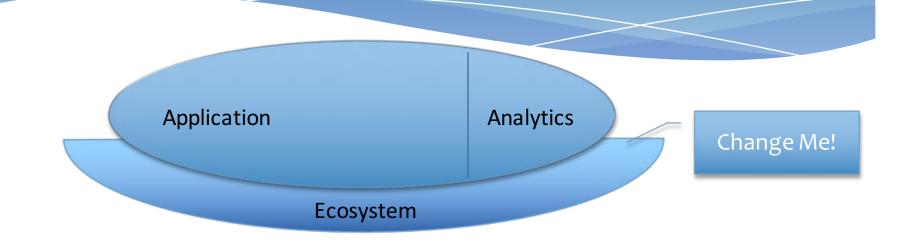
# Aims: to glean 2 perspectives

- As a potential beneficiary
- As someone being observed

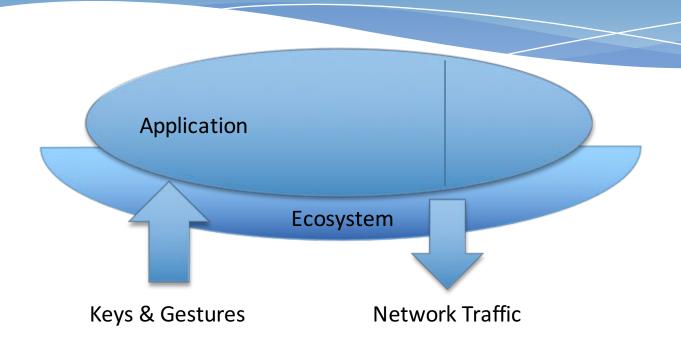
#### **Emerging Patterns**

- 1.Takers 3/10
- 2.Balanced 7/10
- 3. Dislike 2/10
- 4. Lagging what's happening

### **Instrument the Ecosystem**

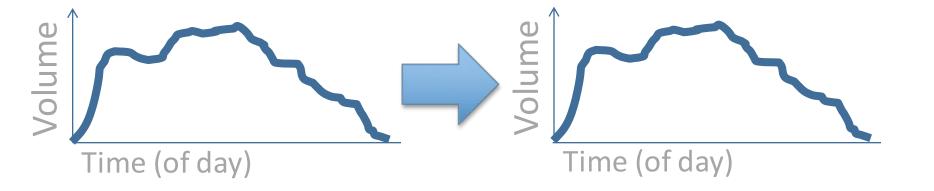


#### **Observe the Behaviours**



#### **Network profiling**

Discover



Use

### **Network profiling**

