

RUTGERS THE STATE UNIVERSITY OF NEW JERSEY

Mining Smartphone Mobility Data 1

## MINING SMARTPHONE MOBILITY DATA

Introduction

*Spiros Papadimitriou*, Tina Eliassi-Rad, Katharina Morik  
Rutgers University Northeastern TU Dortmund



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## Tutorial plan

8:30 – 9:25 Spiros

- Mobile technology overview
- Mobile sensing: localization

9:25 – 10:05 Tina

- Local-based social networks
- Mobile advertising

10:05 – 10:30 Katharina

- Resource-constrained Graphical Models for App Usage Mining

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## Mobile devices



Smartphones

IoT

Network  
(Cellular, WiFi, Bluetooth, ZigBee, ...)

Sensors

Medical

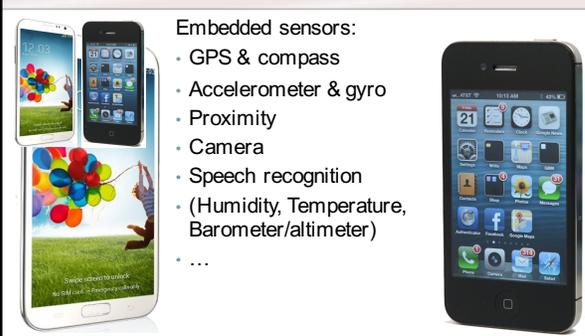
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## Mobile devices: smartphones

Embedded sensors:

- GPS & compass
- Accelerometer & gyro
- Proximity
- Camera
- Speech recognition
- (Humidity, Temperature, Barometer/altimeter)
- ...

(more later)

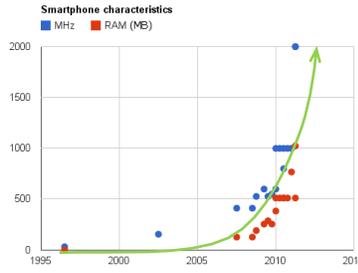


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## So what?

Smartphone characteristics

■ MHz ■ RAM (MB)



Year

...you have a pretty powerful computer in your pocket!  
...and it's connected!

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## So what?

It's what I and many others have worked towards our entire careers. It's just happening *now*.

– Eric Schmidt (on cloud computing)

- The same could be said about mobile sensing and mining
  - Sensing & sensor networks
  - Ubiquitous computing
  - Mobility tracking
  - ...
- But all are becoming mainstream now!

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## Mobile “vs” web

Modern applications:

```

graph LR
    subgraph Cloud [Identity / Authentication]
        IA[Identity / Authentication]
    end
    Frontend[Frontend] <--> API[API]
    API <--> Backend[Backend]
  
```

- Browser (e.g., HTML5 + Javascript)
- Smartphone
- ...
- Typically JSON (increasingly, authenticated)
- Amazon EC2
- Microsoft Azure
- Google GCE
- ...

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## Mobile “vs” web

Identity / authentication (e.g., OAuth):

- Users
- Applications & developers
- Mobile APIs for managing identity/accounts & content

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## Mobile “vs” web

E.g.: what is the difference between Facebook in your web-browser, vs Facebook on your smartphone

Not much:

- It's the same backend & API, just running a different frontend

A lot:

- Access to content and data only on the device (e.g., photos, location, accelerometer, etc...)

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## “App”

- So... “app” vs “non-app” is maybe a better distinction...
- App has well-defined:
  - API (w/ semantics)
  - Entry points (controlled)
  - User identity (controlled)
- No longer entirely free (cf. web crawler vs Twitter firehose)
- Provide better UX and integration (cf... vs FB OpenGraph)
- Trade-off / balance: distributed and centralized (in organizational sense)

“Killer app for privacy/identity is ‘social’” ...

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## Example applications

- Geo-location
- Urban computing
- Quantified self
- Healthcare
- Security
- ...many more!

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## Geo-location

### Example applications

What most people think (mainstream applications):

Google Maps      Waze      Yelp      Foursquare

- Maps
- Navigation
- Local search (+ social)

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## Geo-location

### Example applications

Locale / Tasker      Google Now      Glympe

- Context-based:
  - Locale: e.g., "if I'm within 0.5mi of work address and I have a meeting on my calendar, then set my phone to silent"
  - Google Now: "if I have a dentist appointment on my calendar, notify me when I need to leave, based on current traffic conditions, to be on time" or "if my email contains records of a booked flight, show flight status"
- Location reporting and sharing: Glympe, Google Latitude, etc.

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## Urban computing

Use broadly collected data for urban planning and analytics:

- Zoning and planning
- Traffic monitoring and management
- Public transportation planning
- Crisis detection and management
- Energy consumption sensing
- Air quality monitoring
- ...

Much of this data comes from traces of mobile activity!

[ ICWSM 2016 Tutorial: "The Web of Cities and Mobility" ]

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## Quantified self

### Example applications

Withings devices      Sleep Cycle      Instant HR

- Measure "self", visualize, and correlate
- Idea dates back to 70s; term coined ~2007 by Kevin Kelly
- Both peripheral sensors as well as just apps; e.g.
  - Heart rate, Sleep quality
  - Weight, Activity
  - ...

<http://quantifiedself.com/>

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## Healthcare

Quantified self: log everything

Medical applications: glucose, asthma, ECG, ...

- Related to quantified self
  - Many of these services can send data to your doctor
- Distinction: specific goal vs. "log everything" approach
- Micro-level (personal) and macro-level (population)

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## Privacy

### Examples

... Sonntag, 11. Oktober 2009

- Vast data that allows quite accurate activity tracking or inferences
- Clearly raises privacy concerns
- Policy ( & technology ?)

"Tell-all telephone" – Die Zeit & Malte Spitz

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## Security & Malware

### Examples

- Mobile malware: 6x [Juniper]
- E.g., BadNews: malware on Google Play (30+ apps, 2M downloads, fake app update prompts, mobile "pickpocketing")

Some challenges:

- Role mining: characterize groups of permissions more meaningfully
- Unusual activity detection

Better: iOS-style permissions (now also on Android)

- User asked when permission needed
- Can grant/deny individual permissions

## Mobile mining

- The mobile “revolution” (like the “PC revolution”) brings together many disciplines and touches many areas
- So, we had to draw some (occasionally arbitrary) divisions, and leave several things out

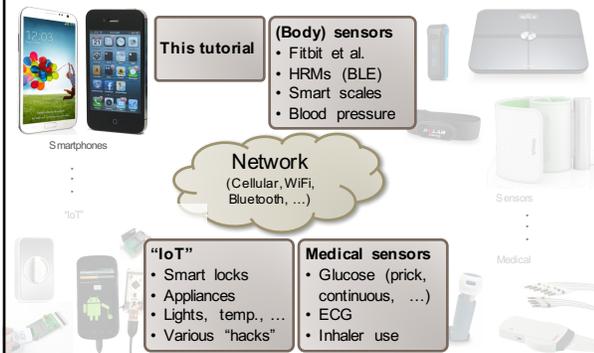
This tutorial focuses on:

- Work with a substantial analytics component
- Data collected via smartphones (although we'll touch on others sensors briefly, but we won't go into sensing or ubiquitous computing territories—much)

## Looking forward...

- Mobile phone penetration rapidly increasing
  - For many people, a smartphone will be their first computer
- All of these technologies are becoming mainstream
- Sensors are becoming cheaper and easier to hook up
- So, what's beyond (just) the mobile (smart)phone?

## Mobile devices



## Coming everywhere:

E.g., wearable mobile devices



## Cheap...



Ten years ago:

Mica Mote (Crossbow)

- Atmel ATmega 103L @4MHz
- 128KB flash / 4KB SRAM
- 916MHz radio transceiver (38.4Kbps)

~\$300 per mote w/sensors

Today:

e.g., RFDuino (\$21)

- Nordic ARM Cortex-M0 (32bit)
- Bluetooth 4.0 (BLE)
- or, ESP8266 (**\$2-3 !!**)
- 32bit processor (Xtensa IP core)
- 2-4GB flash !
- Full WiFi & IP stack !!

10-100x cheaper  
More capable  
Popular\*

\*RFDuino raised \$373K out of \$5K goal on Kickstarter

## Cheap... and ubiquitous

- It's easier than you think!
- Proliferation of open-source, open-hardware tools:
  - Arduino ecosystem (AVR and ARM), mbed, BeagleBone, RasPi, ...
  - ESP8266, NodeMCU, ...
  - Sensors in forms for easy prototyping (breakout boards, etc)
  - Wireless modules (BLE, Xbee, ...)
- Very active hacker (maker) communities
- Cloud platforms (e.g., Imp, Xively, IFTT, Spark Core, ...)
- Some mainstream interest (e.g., Android Accessory APIs)
- Co-design of sensing and analytics
  - Already a trend in mHealth

## Larger picture: venues

In addition to data mining / web + social media venues:

- Medical health informatics
    - Many...
    - Good collection: <http://mhealth.imir.org/collection/view/51>
  - Ubiquitous computing
    - Mobile sensing workshop
    - Urban computing workshops
    - New urban computing conferences
  - Networking
    - PhoneSense
    - MobiCASE
- Many of these areas are starting "analytics" workshops

## The rest of this tutorial

- 9:25 Spiros
  - Mobile technology overview
  - Mobile sensing: localization
- 9:25 – 10:05 Tina
- Local-based social networks
  - Mobile advertising
- 10:05 – 10:30 Katharina
- Resource-constrained Graphical Models for App Usage Mining

Very interdisciplinary area, we *had* to leave many things out

## Tutorial resources

<http://mobilemining.clusterhack.net/>

- Link also on conference website
- These slides (handouts)
- Links
  - References,
  - Datasets,
  - Other useful material

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