

# Toward a high resolution temperature distribution map using crowdsourcing smartphone battery temperature

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# Introduction

- Temperature is a physical quantity that is very important to environment and human health
- Air temperature data, including urban one, is important for meteorology, energy demand planning, urban heat islands (UHI)<sup>1</sup>, effects of temperature on human health
- Air temperature data of cities is often not available, especially high temporal and spatial resolution data

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<sup>1</sup>UHIs are areas in urbans those have temperature higher than their surroundings

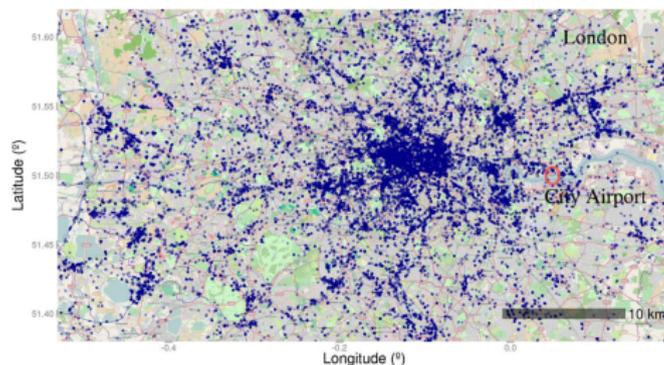
# Temperature data observation approaches

- Observation approaches:

Method	Temporal resolution	Spatial resolution	Accuracy
Weather station	High	Low	High
Satellite	Low	High	Good
Internet of Things	High	Average	High
Smartphone	High	High (in cities)	Good

- Each smartphone has a battery temperature sensor to avoid overheating
- The sensor can be used to measure air temperature indirectly

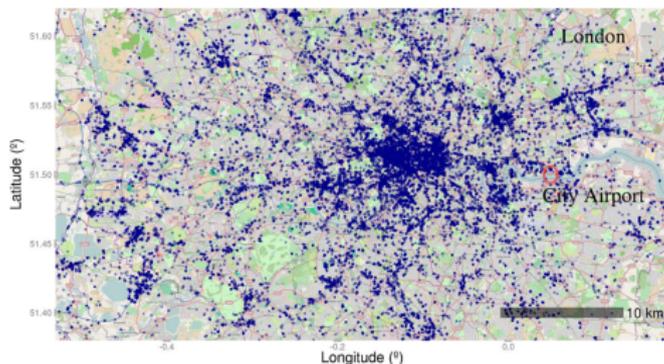
# Overeem's crowdsourcing approach



- Overeem et al.: Collect data from a large number of smartphones, then use a **heat transfer model (HM)**<sup>2</sup> to estimate air temperature:
  - 1 Filter: [Just plugged in + start charging] and [being turned on/off + discharging] and [battery temperature from 10-47°C]
  - 2 Average temperature readings by space and time
  - 3 Build a simple linear regression model to estimate air temperature from the averaged values

<sup>2</sup> A. Overeem et al., *Crowdsourcing Urban Air Temperatures from Smartphone Battery Temperatures*, *Geophysical Research Letters* **40**, 2013,

# Overeem's crowdsourcing approach



- Estimated temperature is representative for **areas**, such as cities
- Accuracy depends on number of readings
- Assessment: Compare to a weather station (distance 5-27 km)
- Android app: WeatherSignal (<https://goo.gl/Fefgoi>)
- Promising results in some cities: Rome, Sao Paulo<sup>3</sup>, Buenos Aires

<sup>3</sup> A. Droste, A. Overeem et al., *Crowdsourcing Urban Air Temperatures through Smartphone Battery Temperatures in Sao Paulo, Brazil*, Journal of

# Our approach: crowdsourcing + statistical model

- We use a **statistical model** (SM):
  - When a smartphone is in idle state, its battery temperature is correlated with its surrounding air temperature
  - We build a simple linear regression model to estimate air temperature from *idle* smartphone battery temperature readings
- Difference from HM approach: Estimated temperature is representative for **each smartphone** environment:
  - Each smartphone estimates air temperature independently
  - *Ability to have maps of temperature distribution*
- Assessment: Compare to a nearby thermometer

# Experiments design and model building

- Equipments:

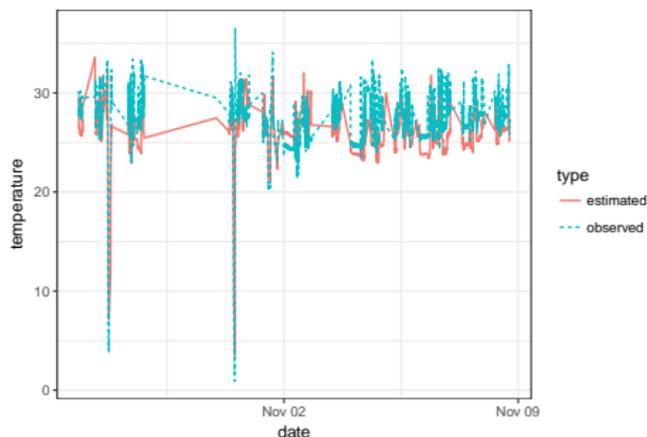
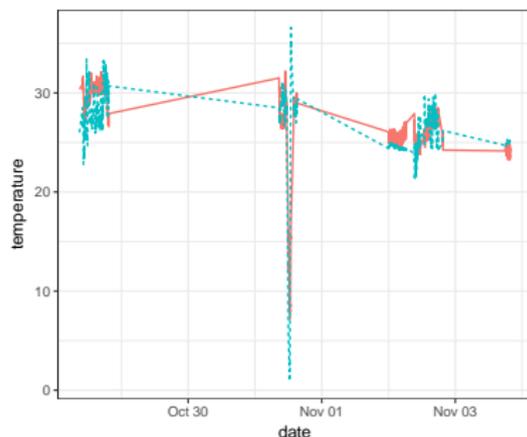
Smartphone model	Manufacturer	OS version	Air temperature sensor
Nexus 4	LG Electronics	Android 5.1	not available
Galaxy Note 2	Samsung	Android 4.4.2	not available
Galaxy Note 3	Samsung	Android 4.4.2	available

- Environments to collect data: 0 – 4°C, 20 – 25°C and 30 – 35°C
- A part of temperature data collected from the Note 3 is for training data, the rest and data from Nexus 4, Note 2 are testing data
- Our model for air temperature estimation:

$$T_{air} = 0.8765 \times T_{battery} + 0.4776, \quad (1)$$

coefficient of determination of the model is  $R^2 = 0.79$ .

# Experiment results



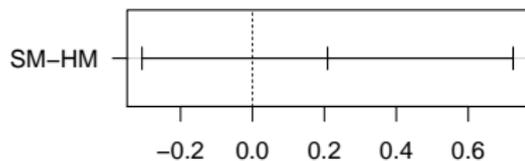
Left: Test result for Samsung Galaxy Note 2, right: test result for LG Nexus 4.

Smartphone model	ME	MAE	Correlation	$R^2$
Nexus 4	1.11	1.86	0.79	0.62
Galaxy Note 2	-0.82	2.03	0.74	0.55
Galaxy Note 3	0.00	0.95	0.89	0.79

ME: Mean error, MAE: Mean absolute error,  $R^2$ : Coefficient of determination

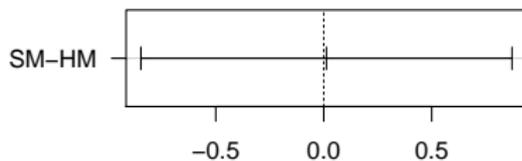
# Comparison of SM and HM approaches

95% family-wise confidence level



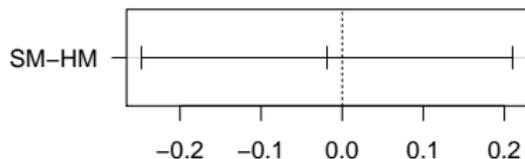
Differences in mean levels of approach

95% family-wise confidence level



Differences in mean levels of approach

95% family-wise confidence level



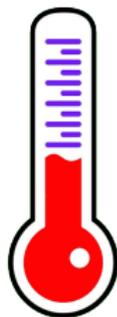
Differences in mean levels of approach

Left, right, bottom: Comparison using ME, MAE and  $R^2$  metrics

SM = Statistical model, HM = Heat transfer model

# Smart Thermometer: A Proof of Concept app

- Smart Thermometer (<https://goo.gl/G2BEPQ>) is an Android app for estimation of air temperature using Equation (1)



- Beta launch: end of 2015, public launch: Mar 2016. Current status:
  - More than 60K downloads, average 1,200 use per day, user rating = 3.2/5.0
  - Data from the app is not yet collected

# Summary

- SM is comparable to HM in ME, MAE and  $R^2$  metrics
- Advantages of SM:
  - Each smartphone estimates air temperature independently
  - Maps of temperature distribution
- Future works:
  - Additional predictor: *battery current* (Android 5.x or newer)
  - Use an independent temperature sensor for model building
  - Recognition of smartphone context: Indoor/outdoor, in pocket (clothing), in transportation means
  - Anonymous data collection, data fusion with other source of data (Landsat, IoT)

# Thank you for your attention!