

Ubiquitous and Mobile Computing

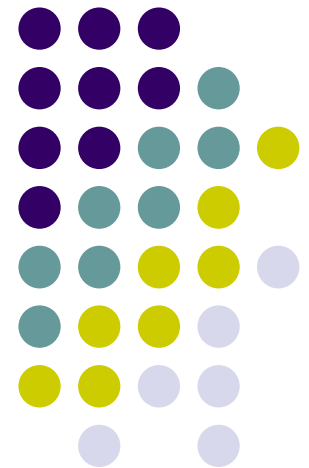
CS 528: Sandra Helps You Learn: the More You Walk, the More Battery Your Phone Drains

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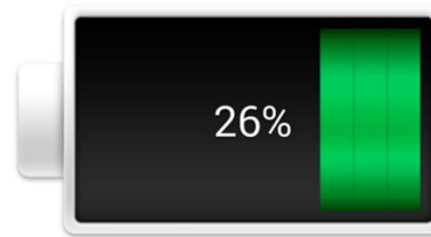




Outline

- Motivation and Introduction
- Related Work
- Methodology
- Evaluation
- Limitations and Discussion
- Conclusion
- References

Motivation

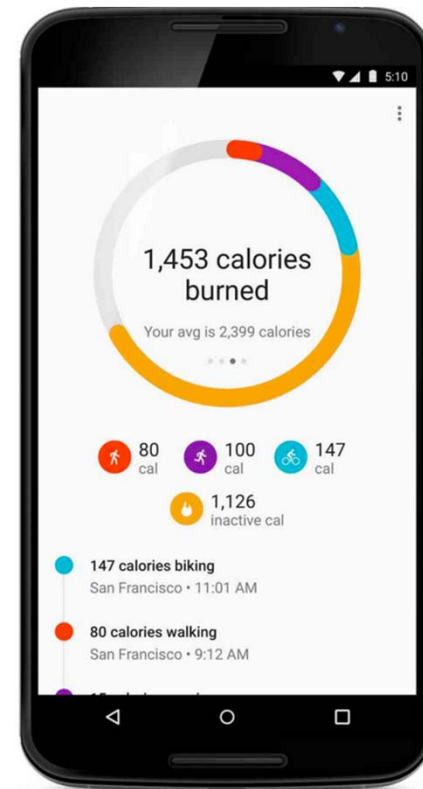


Power saver
Optimize battery life OFF

Extreme power saving mode
Put phone in extreme power saving mode to extend battery life OFF

Usage
View battery usage details

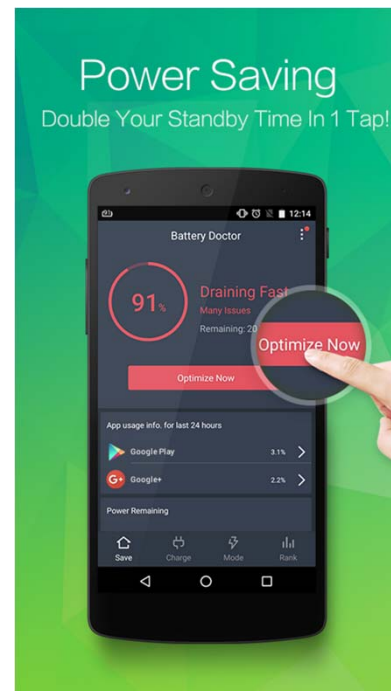
History
23h 23m 42s on battery





Why?

- Empirical practices help users develop battery model
- Research prototypes and commercial apps supplement the model

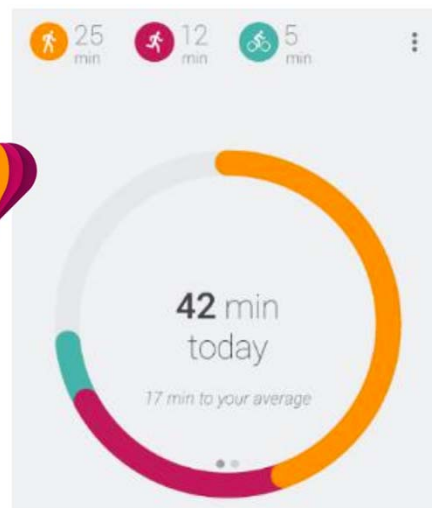




Why?

- CSAs (Continuous Sensing Apps) introduce new major factors governing phones' battery consumption
- Nontrivial persistent battery drain
- Mobility-dependent battery drain

Google Fit:
activity tracking



Moves:
activity/place
tracking



Accupedo:
pedometer



Dieter:
pedometer



Related Work

- **Battery Interface**

Found and provided useful battery information

Not focusing on background-running CSAs

- **Battery Management**

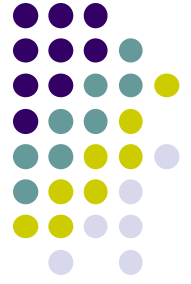
Proposed battery management strategy

Not aiming to help CSA users manage battery by themselves

- **Battery Diagnosis**

Detected abnormal battery drain and its causes

Not letting users aware of the impact of mobility conditions



Key Research Questions

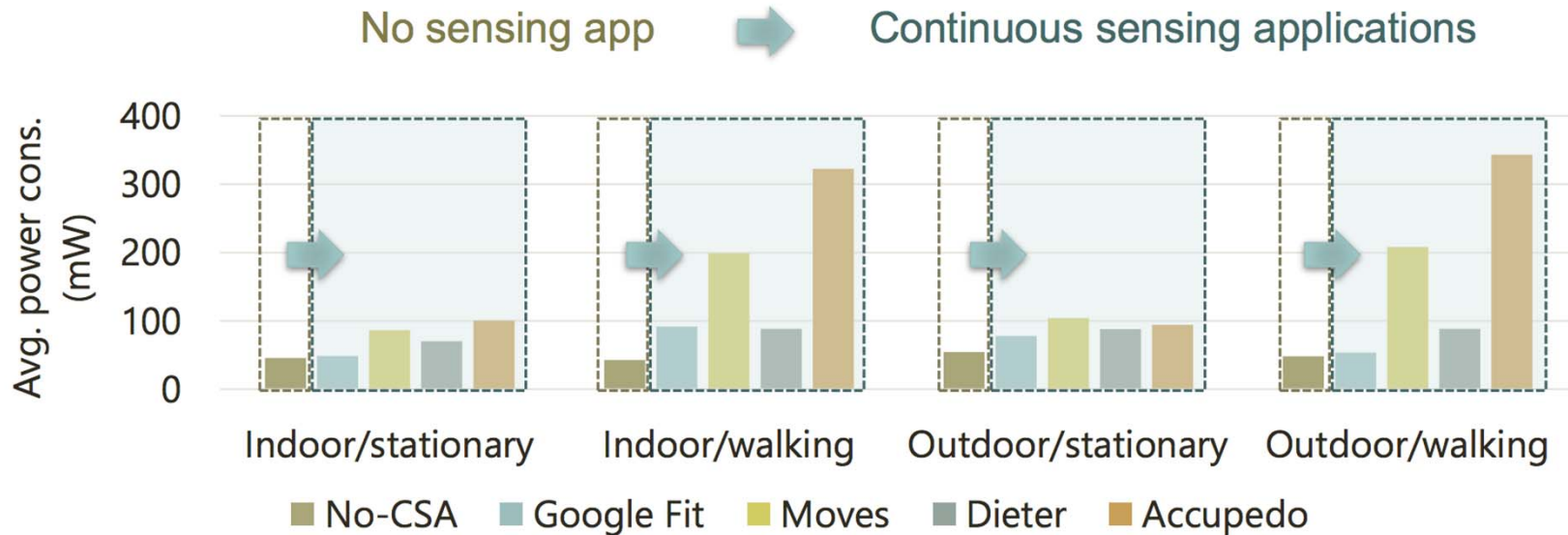
- Why do CSAs make current practices no longer effective?
 1. Quantitative impact: Nonlinear battery drains of CSAs
 2. User perceptions with CSAs' battery behaviors
- How can we help users' battery management?

Sandra

Standby powers under different mobility conditions



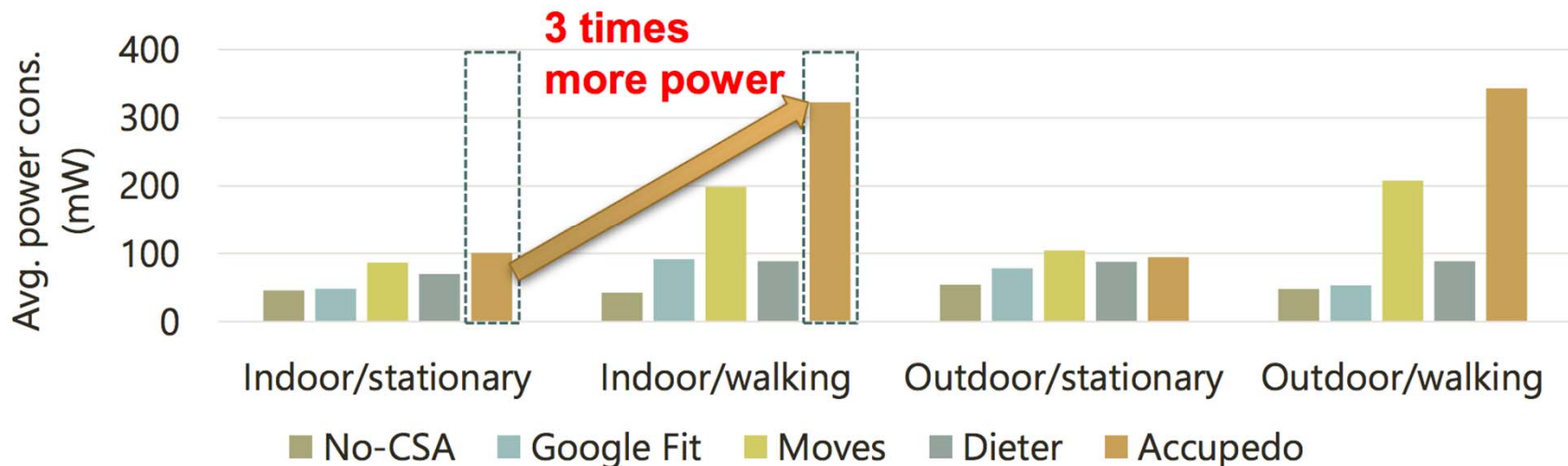
- CSAs drain extra stand-by power
- Average increment: **171%** compared to No-CSA



Standby powers under different mobility conditions



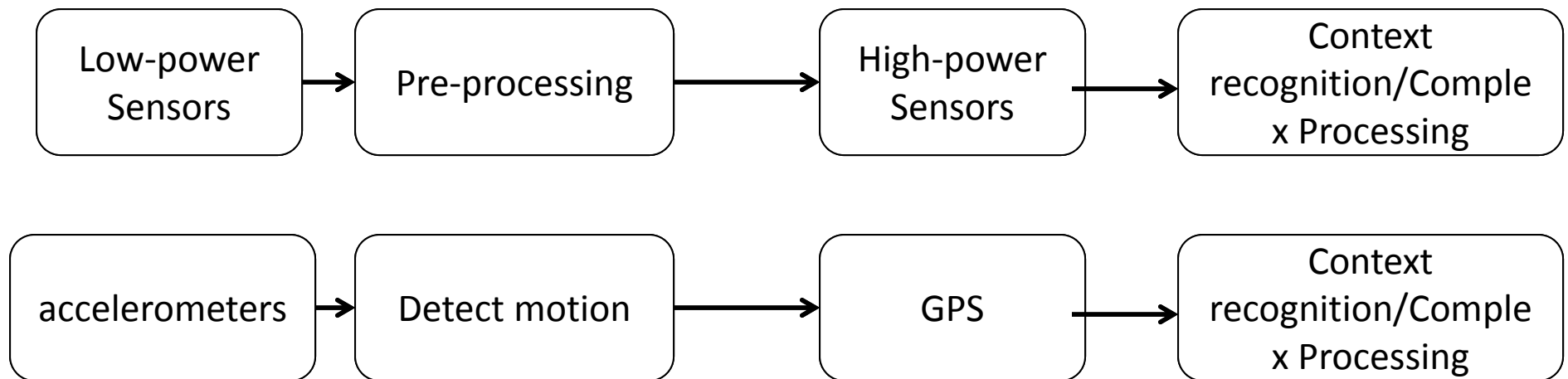
- CSAs drain extra stand-by power
- Average increment: 171% compared to No-CSA
- Extras vary depending on user's mobility condition



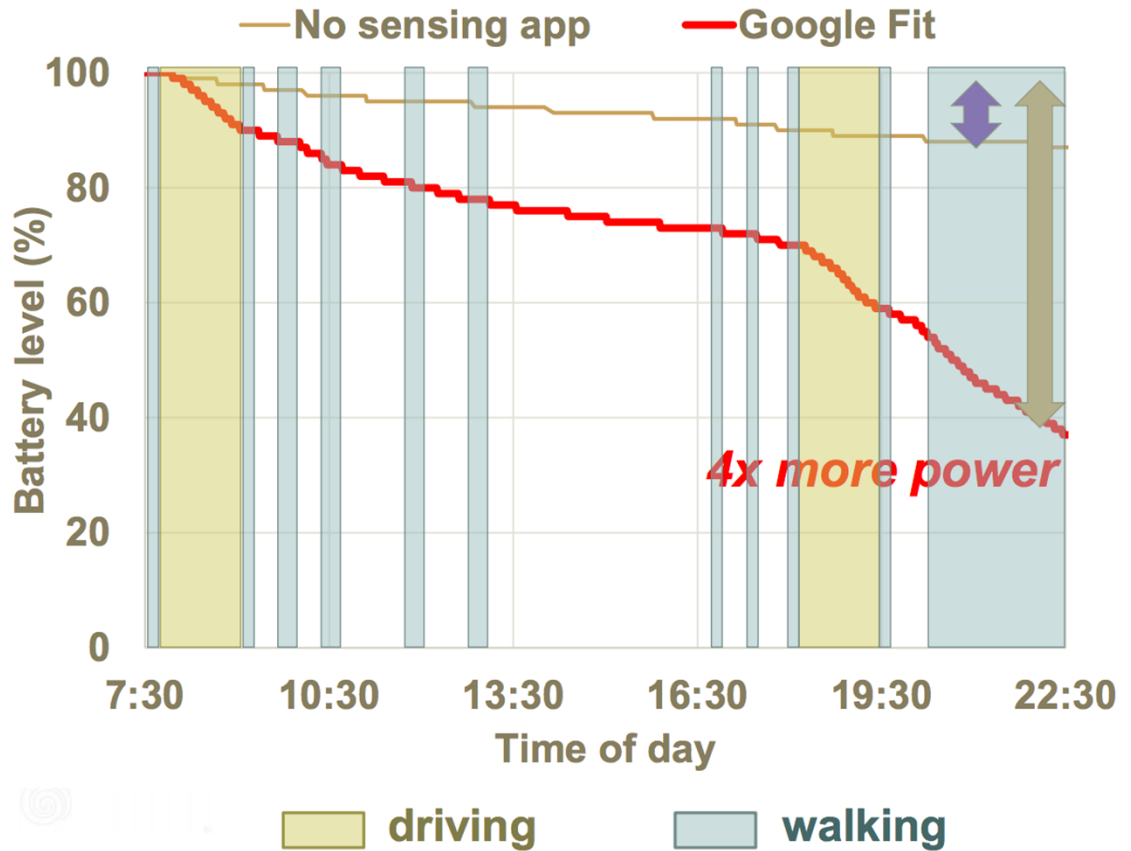
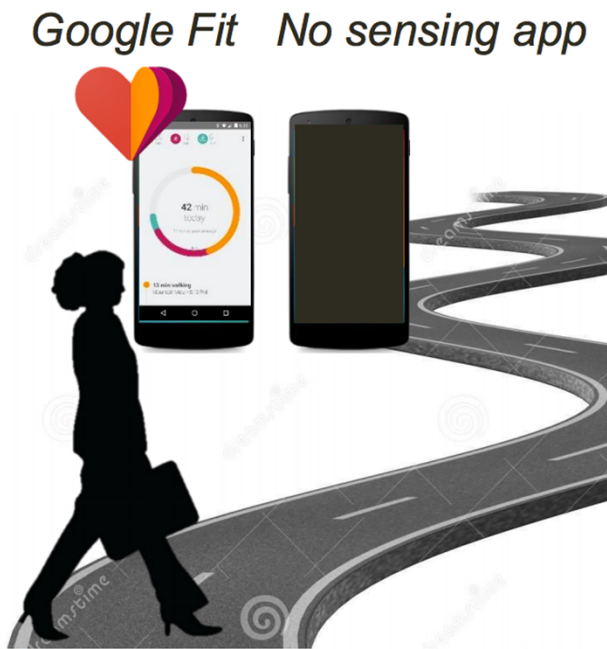


Why mobility-dependent?

- Adopting conditional sensing pipelines for energy saving



Day-long battery drains under real life mobility variations



User perceptions with CSAs' battery behaviors



24 participants



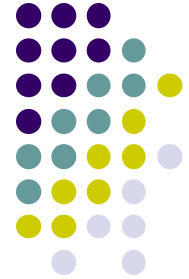
Sensing apps



1:1 Interview



- Limited understanding about CSA's operation and battery drains
- Suspicion of erratic battery drains, but don't know why



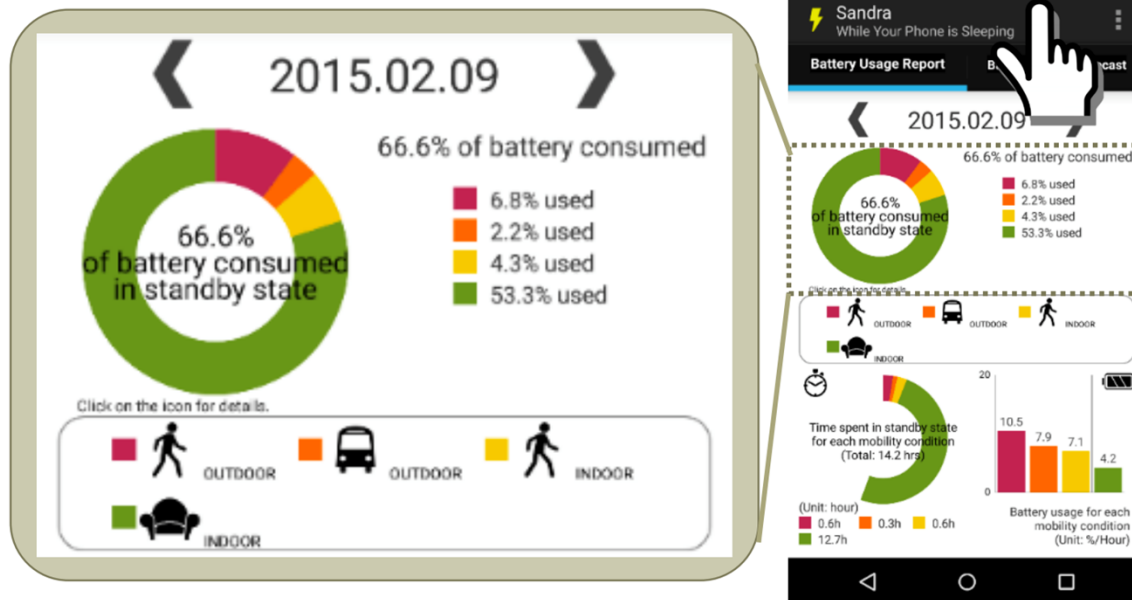
Design

- Purposes

- Make users understand mobility-dependent power behavior
- Help users take necessary actions in advance

- Interfaces

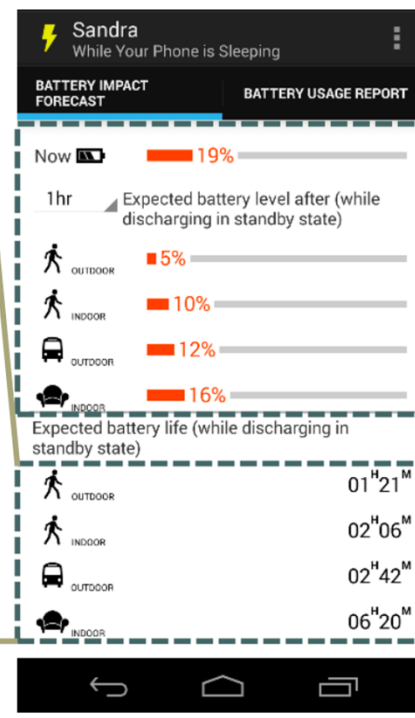
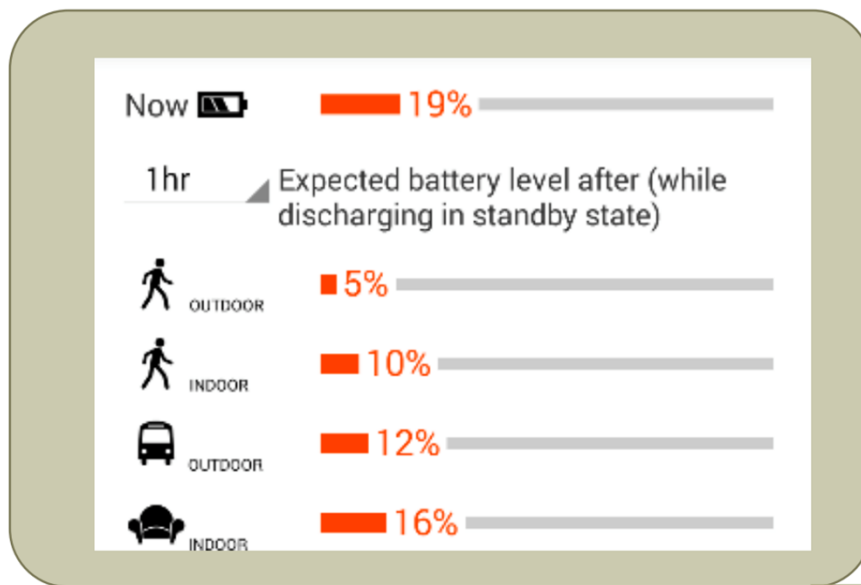
Provides a retrospective battery use summary



Design

- Interfaces(ctd.)

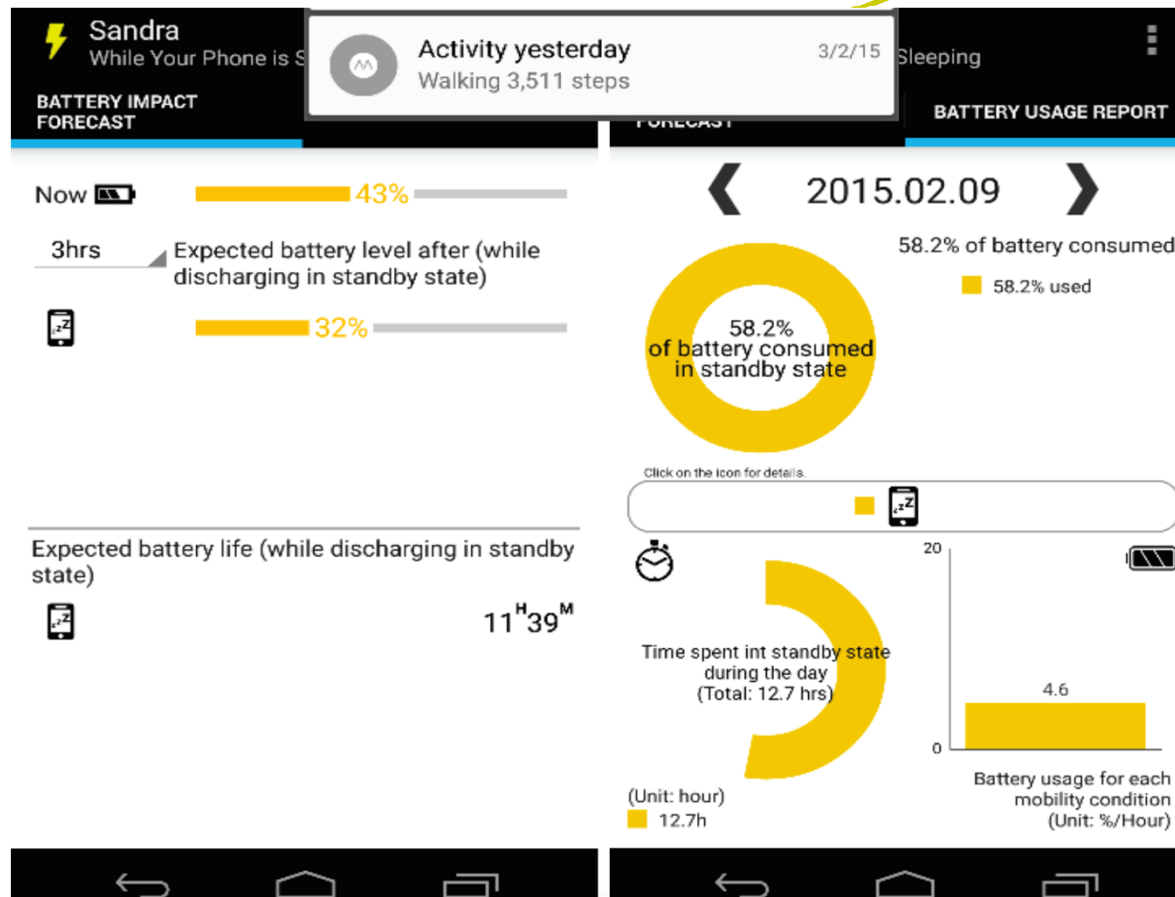
Provides expected standby times for a set of commonly occurring mobility conditions



Design

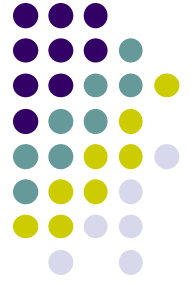


- Interfaces(ctd.)



Sandra-lite

- Downgraded Sandra
- Single standby life expectation
- Without per-mobility breakdown



Implementation

- Mobility Condition Monitoring
 - Leverage context information generated by continuous sensing applications by using open APIs
- Calculation of battery drains
 - Drain rate(%hour): average decrease in battery levels per hour= $\frac{\text{total battery decrease}}{\text{total duration}}$ (for a mobility condition)
- One concern: Sandra overhead
 - Incurs only a marginal cost(Power overhead: only 3-7 mW), so not a big deal



Evaluation

- Experimental Setup



- Conduct two questionnaires for using Sandra and Sandra-lite in a 5-point Likert-scales(Strongly Disagree, Disagree, Neutral, Agree, Strongly Agree)



Evaluation

- Q1: “Dis it bring changes to your existing understanding about your phone’s stand-by battery drain?”



Realizing that the phone consumes different power

Mo
understanding(p-value 0.023)



Positive impact on sensing applications



Evaluation

- Q3: “Did you find it helpful in managing your phone’s

In-situ arrangement:
Recharging the phone



Acquiring
new everyday practices:
Turning off GoogleFit on driving



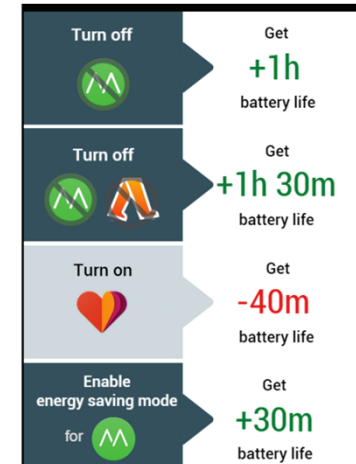
Feeling less nervous
under limited battery:
Before sleeping



Mobility-aware battery information was perceived as useful (p-value= 0.005)

Limitations and Discussions

- Limitations
 - Study Limitation of Target Participants
 - Other Factors that Affect Nonlinear Battery Drain
- Extensions
 - Notify expected increase of a phone's battery
 - Future mobility pattern-based battery advisor
 - Context-dependent recharge alert



Please connect charger

2hours of expected battery life, based on your past mobility pattern: less than 10% remaining

[DETAILS...](#) [GOT IT!](#)

Please connect charger

2hours of expected battery life, based on your past mobility pattern: less than 18% remaining

[DETAILS...](#) [GOT IT!](#)



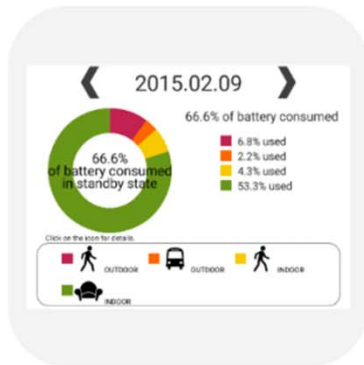
Conclusion



Mobility-dependent battery drain



Real devices, real users,
real-life situations

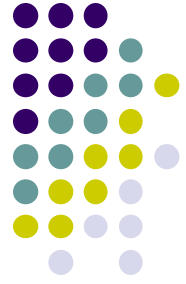


Sandra



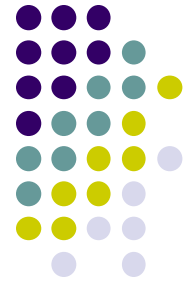
Usefulness of mobility-aware
battery information

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