Harnessing the Web for Population-Scale Physiological Sensing: A Case Study of Sleep and Performance

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Human Cognitive Performance

- Ability to perform mental actions and processes including attention, memory, reasoning, decision making, planning, etc.
- High cognitive performance important for:
 - Productivity [Colten & Altevogt, 2006]
 - Learning outcomes [Kelley et al., 2015]
 - Accident risk [Dinges, 1995]
- Laboratory setting: Performance varies throughout day [Van Dongen & Dinges, 2000] and is decreased after sleep loss [Dinges, 1995]

Existing Research

- Laboratory setting:
 - Induce sleep deprivation
 - Regular, intrusive, artificial performance tasks
 - Missing real-world influences incl. motivation, mood, illness, behavioral compensation (e.g., caffeine), and complex sleep patterns



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Lack of scalable methods to characterize real-world cognitive performance & sleep [Roenneberg, 2013]

Open Research Questions

1. How does cognitive performance vary in the real world?

2. How do real-world sleep patterns impact performance?

Challenges

- Real-world more complex than laboratory
 Need much larger dataset
- But existing methods don't scale!
- Need cognitive performance measurements
 - Annoying: Regular, intrusive, artificial performance tests
 - Should use performance on real tasks
- Need sleep measurements
 - Can't observe in lab or control as before
 - Can't trust subjective reports
- How can research progress outside the laboratory?

Our Key Insight

Use existing interactions with technology as a sensor into real-world cognitive performance.

Harnessing Search Engine Interactions

- Search engines are used repeatedly every day, awake or sleepy, by billions of people
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Measuring Sleep

- Use wearable device
 - Many search engine users already own device



 Objective measures of time in bed by clicking "Start" & "I'm awake" (plus accelerometer-based algorithm)

Real-World Sleep & Performance at Scale

- Our insights enable study of realworld performance & sleep at scale
- 400x larger study than ever before

Dataset

- Cohort: 32k users over 18 months
 - US representative age, BMI, sleep; mostly male (93%)
 - (Opt-in to link Bing searches & Band data)
- Performance: 75M interaction tasks
 - Keystroke time (and click time)
 - Bing search engine



- Sleep: 3M nights of sleep
 - Microsoft Band



How does cognitive performance vary in the real-world?

Diurnal Performance Variation



- Performance far from constant (31% variation)
- Slowest during typical sleep times (circadian rhythm)

Error bars (all figures): 95% confidence interval

Robustness of Results

Findings are robust. Not explained by...

- Effects of individual users / Population differences
 - Observe true within-person variation
- Type of query
 - Control for click entropy to capture query intent (navigational vs. informational)
 - Similar results for specific queries like "facebook"
- Learning effects
 - Few queries repeat; show no signs of learning effects
- Weekend vs. weekday effects
- Network latency dynamics

How can we model real-world performance variation?

Three biological processes drive performance variation

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- Three biological processes drive performance variation
 - 1. Circadian rhythm (C):

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- Three biological processes drive performance variation
 - Circadian rhythm (C): 1. time-dependent, behaviorindependent, near 24h oscillations
 - 2. Homeostatic sleep drive (H): the longer awake, the more tired you become
 - 3. Sleep inertia (I): performance impairment experienced immediately after waking up
- Hard to disentangle effects
 - Many factors, highly correlated
 - Lab: Forced desynchrony protocol
 - Our method: Variation across millions of real-world interactions (web search)



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Statistical Model

- Generalized Additive Model
 - Intercept
 - Keystroke (control for key pressed: "A", "a", "@", …)
 - Time of day (circadian rhythm)
 - Time since wakeup (homeostatic sleep drive & sleep inertia)

$$y_{i} = \alpha + f^{k}(x_{i}^{k}) + f^{t}(x_{i}^{t}) + f^{w}(x_{i}^{w}) + \epsilon_{i}$$
Keystroke time Residual

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- Parameter learning
 - Fine-grained discretization functions (non-parametric)
 - Least squares optimization



Model Estimates: Time of Day



- Model identifies underlying circadian rhythm
- Consistent with lab-based studies

Model: Time After Wakeup



- Model identifies underlying homeostatic sleep drive and sleep inertia consistent with lab-based studies (validation)
- New insights: It was impossible to measure real-world cognitive performance at scale. Now we can!

How do real-world sleep patterns impact performance?

Sleep Loss over Multiple Nights

- 1. Can we observe an additive effect of multiple nights with little sleep?
- 2. How long does it take to recover from sleep loss? (Given real-world sleep patterns)
- Measure performance over 7 days after zero (SS), one (SI), or two (II) insufficient nights of sleep (less than 6h)

Recovering from Sleep Loss



Recovering from Sleep Loss



Performance by Sleep Timing



Our Contributions

- New method: Use existing technology interactions to study sleep and cognitive performance
 - Large-scale, real-world (outside of laboratory)
 - Continuous, non-intrusive measurements of realistic tasks
- New insights: Real-world performance is not constant but exhibits variation based on time of day and complex sleep patterns. We are the first to quantify these effects.

Our Contributions

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- Population-scale Physiological Sensing
 - Physiology: Branch of biology dealing with the functions and activities of living organisms and their parts
 - Learn about biological functions through user activity logs

Acknowledgments

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Joining faculty job market end of 2017. Please let me know about opportunities at your institution.

Ask me anything!

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