

# TaskyApp: Inferring Task Engagement via Smartphone Sensing

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# Interruptibility and Task Engagement

Location

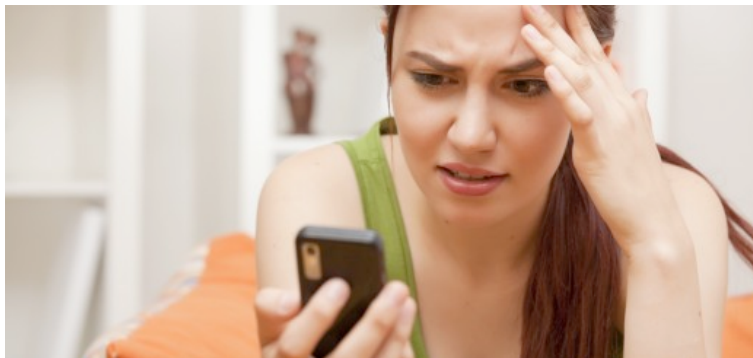
Sender

Movement

Content

Time of day

Task  
engagement



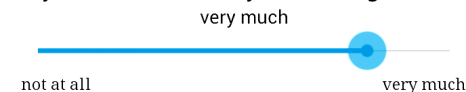
# Interruptibility and Task Engagement

- Link between task engagement and opportunity to interrupt (self-reported)
  - More skilled a person is, less she will be irritated by an interruption
  - **More challenging** a task is to a person, **more irritated** she will be with an interruption
  - More concentrated a person is on a task, more she will be irritated by an interruption

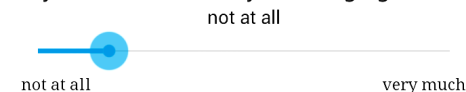
## Activity engagement

Describe your current activity.

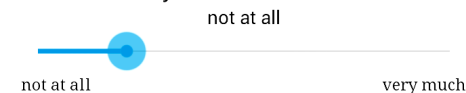
Is your current activity interesting?



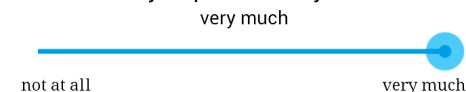
Is your current activity challenging?



How well are you concentrated on it?



Is the activity important for you?

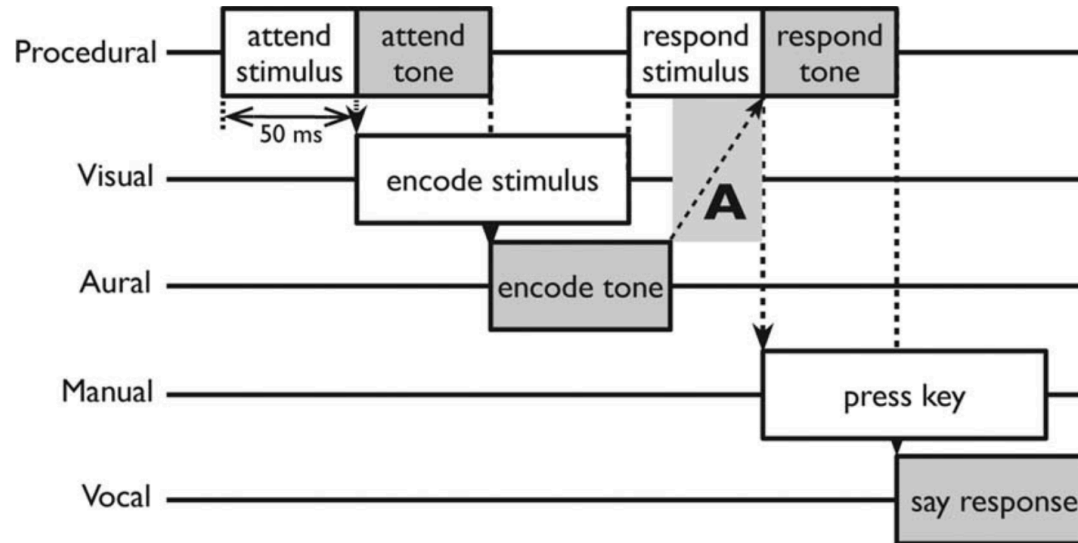


How skilled are you at it?



# Theory of Multitasking

- Interference when two or more threads ask for the same resource at a time

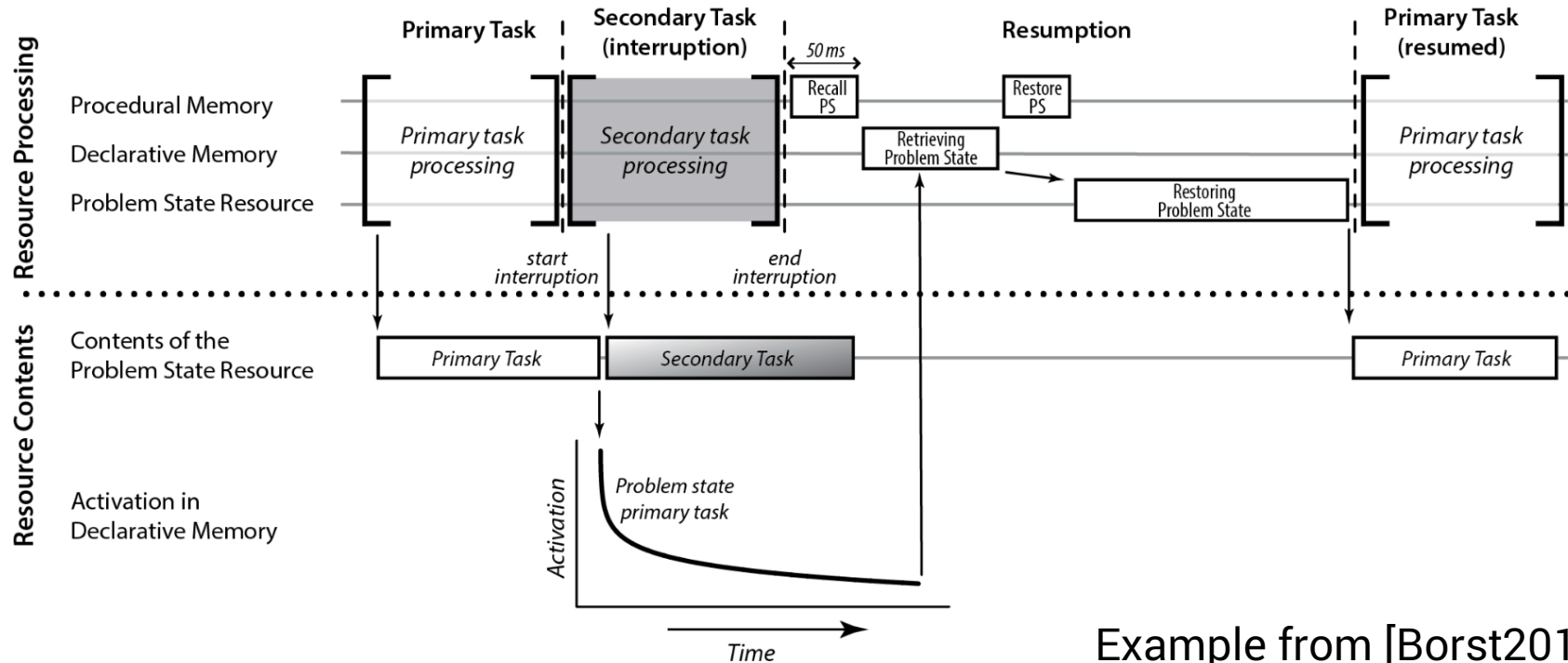


Example from [Borst2010]



# Theory of Multitasking

- Complex tasks require problem state saving/retrieving



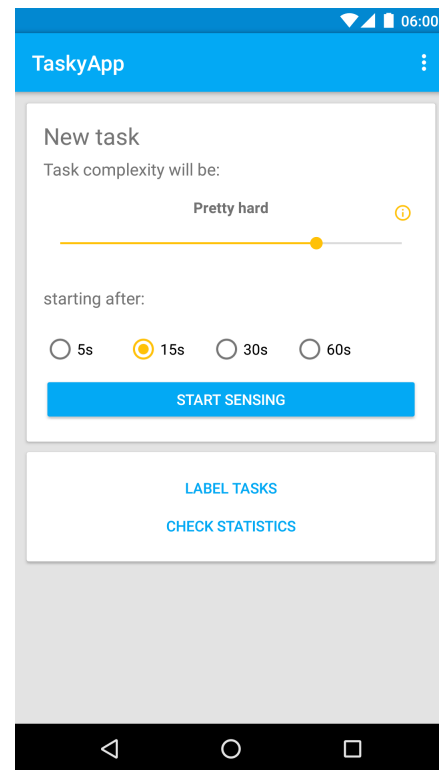
Example from [Borst2015]

# Can we automatically infer task engagement?



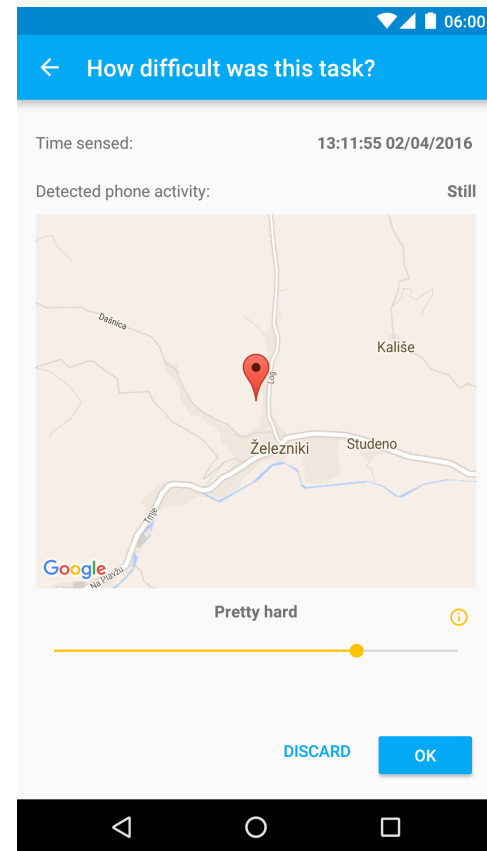
# TaskyApp

- Can smartphones sense that their users are busy (in an office setting)?
- TaskyApp data collection app
  - Background sensing of:
    - Device movement (raw and Google Activity Recognition reported), ambient sound, location
    - BT/WiFi sensing
    - Screen status, sound settings
    - Google calendar events
  - Data labelling via experience sampling and retroactive assisted labelling



# TaskyApp

- Data collection trial
  - Volunteering (with a chance of winning 50€)
  - Eight office workers for five weeks
    - 232 labelled instances (3035 unlabelled)
    - Most data between 8am and 6pm





# TaskyApp – Data Analysis

- Linear regression fit with task difficulty (1-5 on a Likert scale) as a dependent variable
  - **Movement data** gives the most informative features
  - The regression explains only a small part of the data

Variable	B(Std. Err.)	t (Sig)
Acc. Y mean	-.038 (.02)	-1.82 (.068)
Acc. Z mean	.026 (.02)	1.43 (.153)
Acc mean intensity	-.711 (.23)	<b>-3.04 (.003)</b>
Gyro. MCR	-.003 (.00)	<b>-4.06 (.000)</b>
Gyro. variance	.200 (.16)	1.24 (.217)
Hour of day	.067 (.02)	<b>3.49 (.001)</b>
<i>Reg.Constant</i>	<i>8.385 (2.31)</i>	<i>3.63 (.000)</i>

N=232;  $R^2=0.19$ ,  $F=8.64$  ( $p=.000$ )



# TaskyApp – Data Analysis

- Classify a task engagement moment as either **easy** or **difficult** depending on the sensed features
  - We experimented with different classifiers but Naïve Bayes seems to work best (probably due to the low amount of data)
    - 62.5% accuracy compared to 52.8% baseline
    - Also, leads to **favourable errors** – few difficult tasks predicted as easy

easy'	difficult'	
45 (19,4%)	62 (26,7%)	easy
25 (10,8%)	100 (43,1%)	difficult



# Task Engagement Inference

- Even in a restricted office setting smartphone-based task inference is challenging
- Movement features seem to be the most informative
- Next step – wearables
  - Sense heart rate and skin temperature



# Thank you!

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