Tools to Improve Interruption Management

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Outline

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- 4. Preventing Interruptions
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Introduction

Why do interruptions matter?

- Interruptions cause lost productivity due to resumption lag (about 23 minutes each time)
- Can create negative emotions in the workplace
- People are not very good at guessing when to interrupt
- By measuring how interruptible someone is, we can prevent problematic interruptions

Introduction

Different kinds of interruptions

- In-person interruptions
- Notifications
- Self-Interruptions







Introduction

Measuring Interruption

- Humans are bad at guessing how interruptible a co-worker is
- Using technology can allow us to better manage interruptions

Biometric sensors

- Electroencephalogram (EEG)
- Heart Rate, Interbeat
 Interval, Blood Volume
 Pulse
- Electrodermal Activity (EDA)



Machine Learning Algorithm

- A type of artificial intelligence that can learn over time
- Doesn't "learn" like humans do, can only be made to do a specific task well
- Uses training data to calibrate the algorithm

Machine Learning Example - Identifying apples with a Naive Bayes classifier

- Looks at each feature of the fruit independently to determine the probability it's an apple
 - Uses the training data to generate this probability
- Multiply the probabilities together, and compare it to the probability it is not an apple



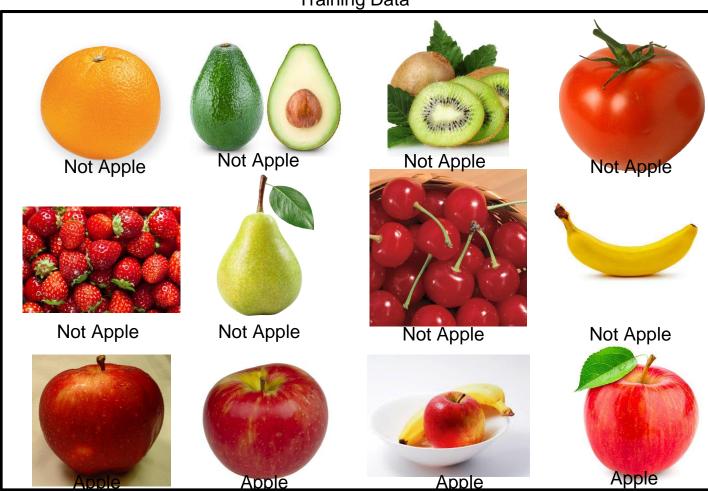








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Training Data

Measuring Interruptibility

Measuring Interruptibility

Bio-metric measurement of interruptibility

- Headband to measure brain activity and eye blinks
- Wristband to measure electrodermal activity
- Chest strap to measure heartbeat and interbeat interval
- Data is recorded from subjects and used as training data in a machine learning algorithm



Measuring Interruptibility

Bio-metric measurement of interruptibility

- Lab test and field test to gather data
- Participants worked on programming tasks while being interrupted
- Participants rated how interruptible they were, which labeled the training data
- Algorithm was successful 91.5% of the time in a lab test



Measuring Interruptibility

FlowTracker algorithm

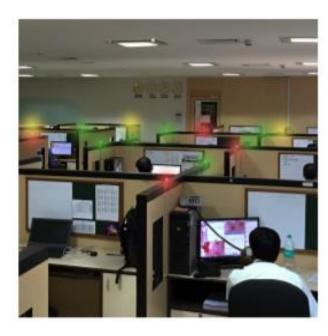
- Records clicks, key presses, and login state
- When activity reaches a peak, the user is classified as busy
- Smoothing algorithm is also applied so that the user's state doesn't switch too frequently

Preventing Interruptions

Preventing In-Person Interruptions

FlowLight

- Light displays interruptibility based on computer usage
- Synced with Skype availability status
- Tested on 449 employees of an international company
- Month long study



Preventing In-Person Interruptions

How does measuring interruptibility impact the workplace?

- The majority of workers said they were more productive and that the number of interruptions decreased
- Over 80% of participants continued using the FlowLight after the study
- Potential to be improved with biometric information



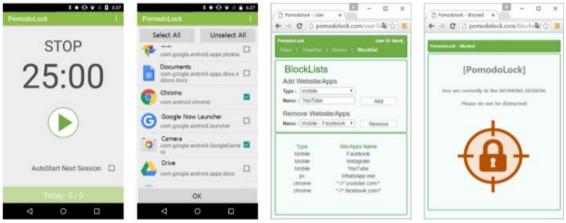


Preventing Email Interruptions

- Uses computer interaction data, tries to determine breakpoints between tasks
- Displays emails when you are at a breakpoint
- Improves levels of stress and feelings of hindrance
- Can be improved by incorporating priority of the message

Preventing Self Interruptions

- PomodoLock A phone/PC app developed based off of the Pomodoro Technique
 - 25 minutes of work, 5 minute break
 - During work time, you are blocked from distracting apps and sites



Conclusions

- Interruptions are costly to software developers
- Current technology allows us to prevent and manage the timing of interruptions
- Use of these technologies leads to increased productivity and other positive effects

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