Human Behavior and Context Sensing

It takes many good deeds to build a good reputation, and only one bad one to lose it.

Benjamin Franklin





Human-Centered
Computer Systems Lab

Overview

- Objective
 - ✓ To understand the computational aspect of life-immersive computing

Plan C

Plan B

Plan A

- Content
 - √ Human behavior and context sensing
 - ✓ Computational challenges of context sensing
- After this module, you should be able to
 - ✓ Have a high-level understanding of human behavior and context sensing and their computational challenges

Notes on Project

We have 9 amazing project teams!

 Think about how to work smoothly through various communication channels.

- Create a GitHub repo and share it with Kyungjin.
 - ✓ GitHub ID: kyungjin-lee

Life-Immersive Mobile Computing

Sense real-world situations and human behavior



Extract and infer useful insights and Knowledge



Pothole Monitoring



Sleep Quality Monitoring



Location-aware Alarms



Physical Activity Diary



Provide
what people need
right on time & place



Bus Stop Queue Estimation



Proactive Advertisement

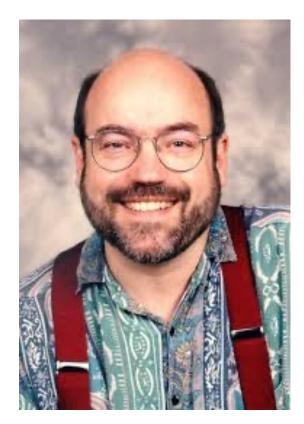
Life-Immersive Mobile Computing

 The first step toward realizing Mark Weser's vision for ubiquitous computing

The Computer for the 21st Century

Mark Weiser

The most profound technologies are those that disappear. They weave themselves into the fabric of everyday life until they are indistinguishable from it.



Key Building Block: Behavior and Context Sensing



Research Trends

Comprehensive/ detailed behavior

✓ Centimeter-level indoor localization

- ✓ Eating
- ✓ Smoking
- ✓ Shopping
- ✓ Dancing
- ✓ Drumming

- ✓ Turn-takings
- ✓ Linguistic contents
- ✓ Emotional expressions

External Behavior



Location



Physical Activity



Conversation

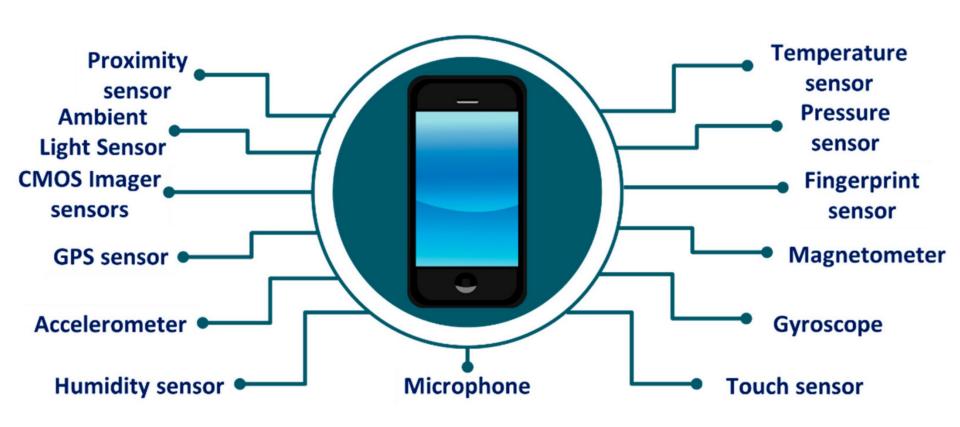
Internal States

- √ Heartrate
- ✓ Stress
- ✓ Mood
- ✓ Sleep quality
- ✓ Distractibility
- ✓ Intention
- ✓ Engagement
- ✓ Attention
- ✓ Mindfulness
- ✓ Emotion
- ✓ Anxiety
- Depression
- ✓ Boredom
- ✓ Fatigue





Smartphone Sensors

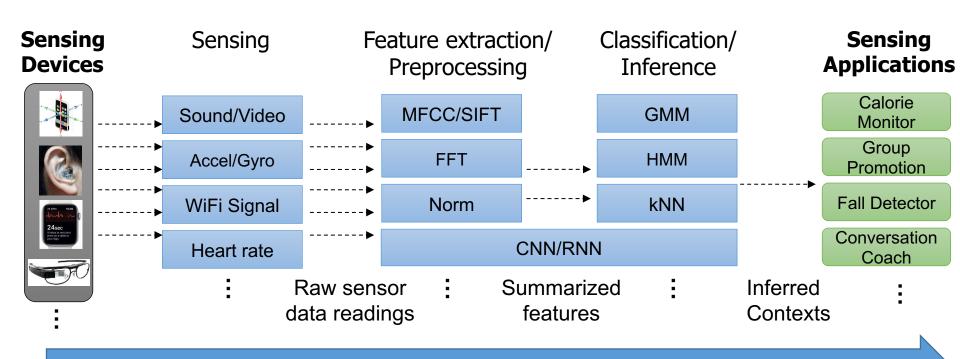


Sensor-rich Environments



Common Computational Flow

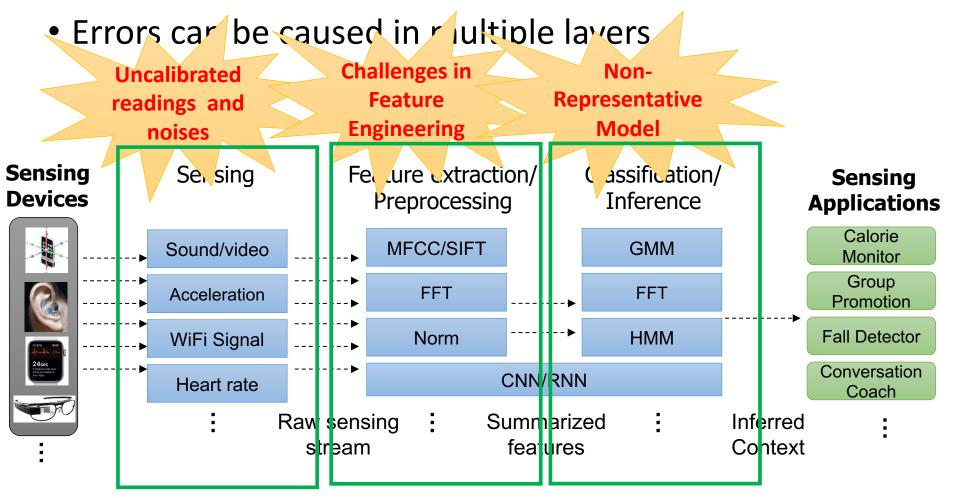
- 1. Continuous sensing of various low-level signals.
- 2. Extraction of user behavior and contexts using machine learning.



Continuous Pipelined Execution

Challenge 1: Inference Accuracy

> 90% accuracy is extremely challenging.



ML View vs. Mobile Computing View

Vision and text data

Vs.

Multi-modal sensory data from mobile/wearable/IoT devices

Clean and organized data from controlled environment

Vs.

Noisy and un-organized data from real worlds

Pre-collected, static set of data (no control on data and users)

Vs.

On-the-fly, dynamic set of data (control on data and users)

Challenge 2: Application Usability

- The inference results are not 100% correct.
- App design should overcome the inaccuracy.

Results Sensing Sensing Feature extraction/ Classification/ Sens.ng **Devices** Inference Preprocessing **Applications** Calorie MFCC/SIFT **GMM** Sound/video **Monitor** Group **FFT** Acceleration FFT **Promotion Fall Detector HMM** Norm WiFi Signal Conversation CNN/RNN Heart rate Coach Raw sensing Summarized Infe red stream features Context

App Design with

Inaccurate

Human-Al Interaction (1/2)

Microsoft

Project Name: Guidelines for Human-Al Interaction

- **Goal**: Recommend best practices for how AI systems should behave upon initial interaction, during regular interaction, when they are inevitably wrong, and over time total 4 steps.
- <u>Guidelines for Human-Al Interaction</u> (CHI 2019)
 - The paper collected 150 AI-related design recommendations and synthesized them into 18 guideline sets, helping practitioners to design better human-centric AI-infused systems through this guidelines.
 - A user study was conducted with 49 HCI practitioners to evaluate the
 effectiveness of applying these guidelines to various existing products and
 whether the guideline is clear.

Human-Al Interaction (2/2)



Project Name : People + Al guidebook



- The guidebook is designed to introduce the entire development process of Al-driven services, beyond developing machine learning or deep learning models.
- It was introduced in 2019 Google I/O, and more details are available at this presentation.



WWDC 2019 Human Interface Guidelines for Machine Learning

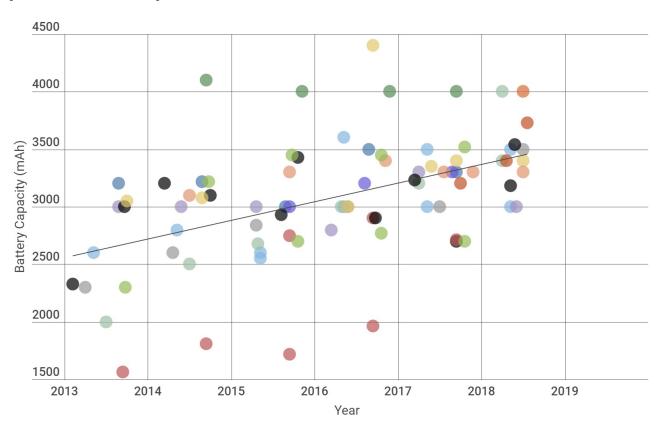
 Apple also suggests their own <u>Apple's Guidelines</u>, which is a much simpler set of advices on the practical app implementation.

Another good article to read:

<u>Re-examining Whether, Why, and How Human-AI Interaction Is Uniquely Difficult</u>
 <u>to Design</u> (CHI 2020, focuses more on UX design than on system design)

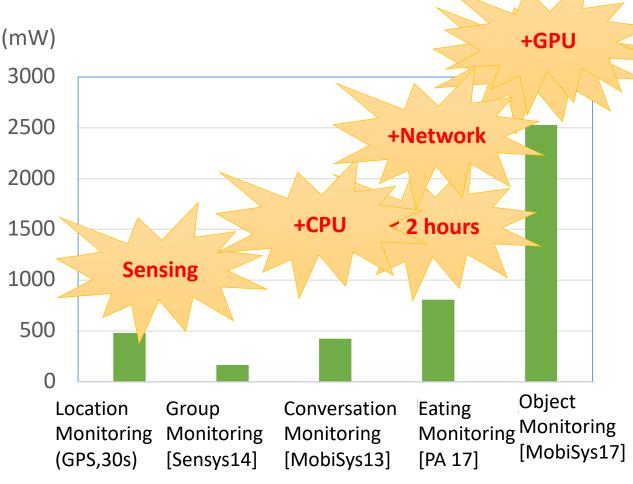
Challenge 3: Power Scarcity (1/2)

- Battery capacity hardly increases over time.
- It is much harder to improve 'performance per watt' compared to 'performance' for mobile devices.



Challenge 3: Power Scarcity (2/2)





- Measured with Samsung Note 4 (3220mAh battery)
- Used Samsung Gear (315 mAh battery) for Anapruna (eating detection)

Challenge 4: New Operational Mode



Vs.



A single user-interactive application

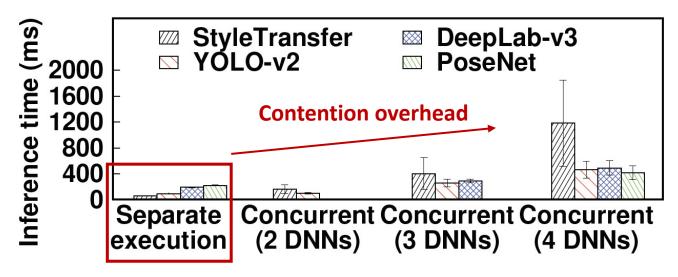
- Requiring continuous user attention
- Assumes limited interaction capabilities (e.g., small screen, touch interface)

Concurrent background-running applications

- Intermittent user interaction
- Autonomous, situation-aware services

Challenge 5: Resource Contention (1/2)

Multi-DNN GPU contention degrades inference speed

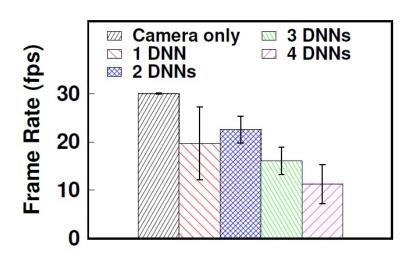


Can run at 2 fps when coordinated perfectly

Measurements on Xiaomi MACE over LG V50 (Qualcomm Adreno 640 GPU)

Challenge 5: Resource Contention (2/2)

Rendering-DNN GPU contention degrades frame rate.



--- Camera only --- 3 DNNs
2 DNNs

30
25
20
15
10
5
0
2 4 6 8 10
Time (s)

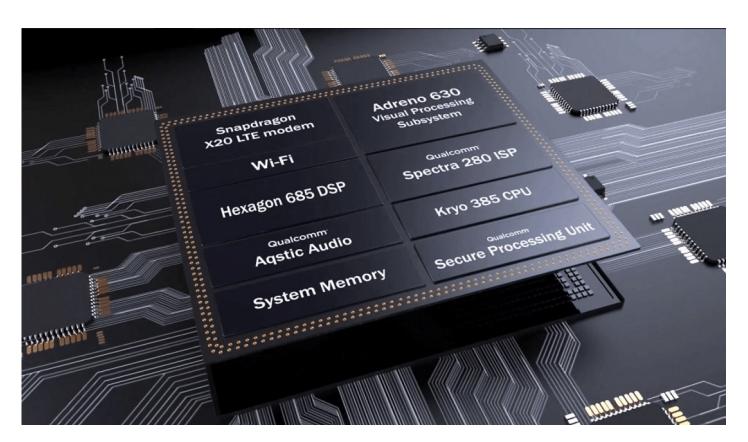
Average frame rate

Frame rate over time

Measurements on Xiaomi MACE over LG V50 (Qualcomm Adreno 640 GPU)

Mobile Processors and Architecture

- Many heterogeneous processors
- Single system-on-chip design with integrated architecture



Challenge 6: Poor Scalability

Amazing mobile s

How to test with

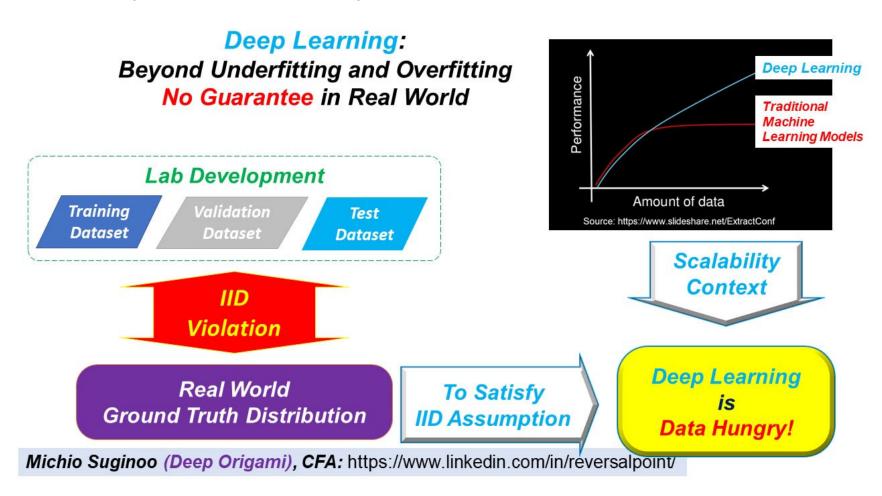
Lets test it with lab users and a small number of real users and consider it "real-world".

Wow! It does not work!
Need access to real venues
With real users on real devices
HOW???



IID Assumption in Machine Learning

• It is difficult to know if "Independent and Identical Distribution Assumption" (IID assumption) holds or not.



Individual Applications Solve All These?

