

Machine Learning

LG전자 고급 데이터 사이언티스트 양성 과정

이영기

서울대학교 컴퓨터공학부



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SEOUL NATIONAL UNIVERSITY

*Success is not final. Failure is not fatal
It is the courage to continue that counts. Winston Churchill*

Quick Introduction

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- Phone: 02-880-1726
- E-mail: youngkilee@snu.ac.kr
- Research Homepage:
<http://youngkilee.blogspot.com>



Quick Introduction

- **Education**

- ✓ **Ph.D.**, Computer Science, KAIST, Korea, 2012
- ✓ **B.S.**, Computer Science, KAIST, Korea, 2003

- **Experience**

- ✓ Assistant Professor, Computer Science and Engineering,
Seoul National University, Korea, September 2018 - Present
- ✓ Assistant Professor, School of Information Systems,
Singapore Management University, Singapore, 2013 - 2018

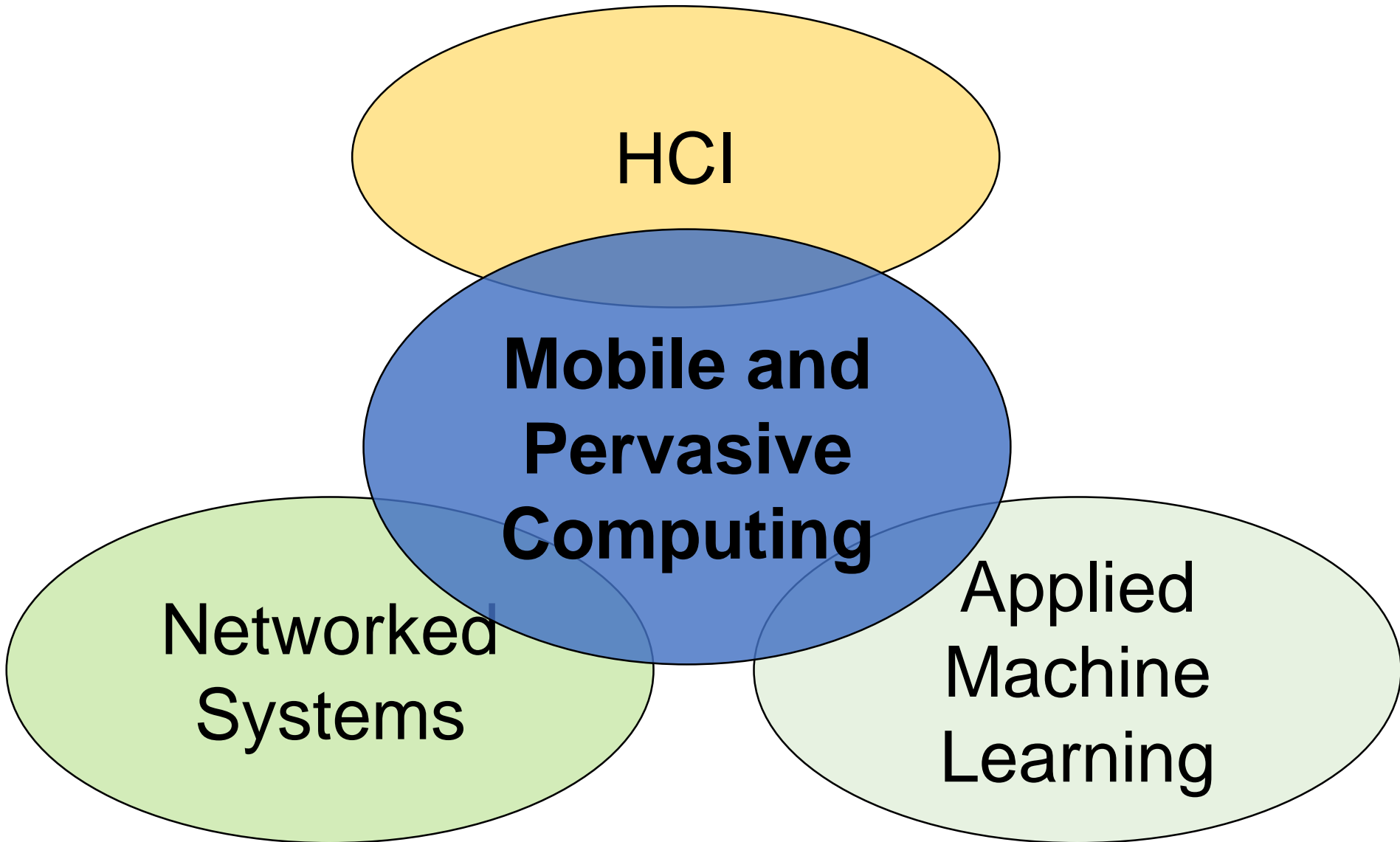
- **Research Interest**

- ✓ Human Behavior and Context Sensing
- ✓ Mobile and Embedded Machine Learning
- ✓ Systems and Tools for Emerging Devices
- ✓ Large-Scale Data Analytics Platform

Agenda for Today

- Introduction to the class
- Introduction to the machine learning
- Introduction to the statistical learning
 - ✓ Chapter 2 of our textbook
- Lab for the chapter 2

My Areas of Research



Research Intro: Mobile Computing



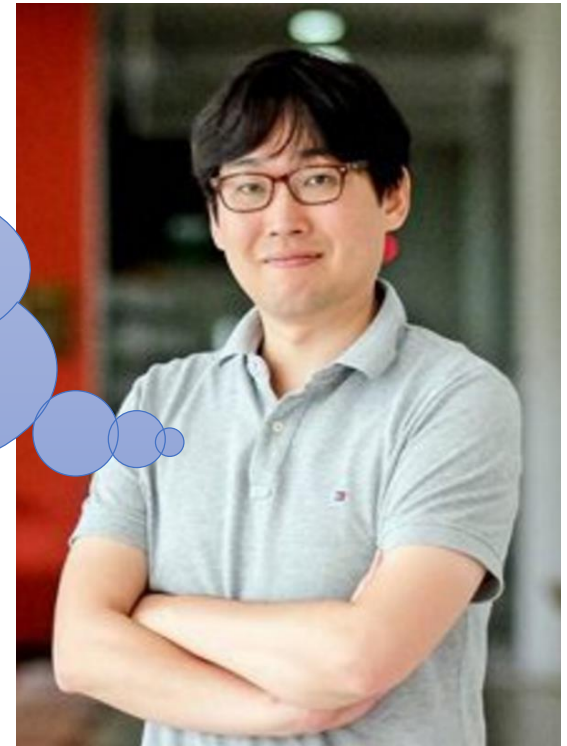
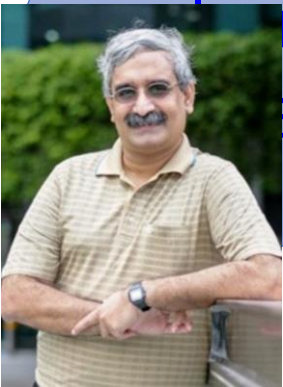
“Mobile” “Computer” or Beyond?



Group-Aware Mobile Ads

- Location-based mobile ads are often of no use.
- Group-aware promotion to satisfy the entire group.
(used in Resort World Sentosa in Singapore).

Korean BBQ: 50% off for
lunch today!!! Wow, this is a
great deal. But...



- Based on our group detection system, GruMon [SenSys 14].


Independent Living Assistant

*Not only to enhance physical wellness,
but also for socially and mentally healthy life*



- On-going project at SMU (initial testbed being deployed to 100 elderly who live alone)

Life-Immersive Mobile Computing



Sense human
behavior, emotion, and
surrounding contexts



Extract useful
insights and knowledge



Provide
what people need
right on time & place



Sleep Quality Monitoring



Pothole Monitoring



Location-aware Alarms



Physical Activity Diary



Bus Stop Queue Estimation



Proactive Advertisement

Key Building Block: Context Sensing

Comprehensive/ detailed behavior

- ✓ Centimeter-level indoor localization
- ✓ Eating
- ✓ Smoking
- ✓ Shopping
- ✓ Dancing
- ✓ Drumming
- ✓ Turn-takings
- ✓ Linguistic contents
- ✓ Emotional expressions

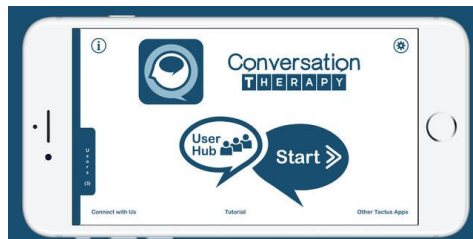
Generic external context



Location



Physical Activity



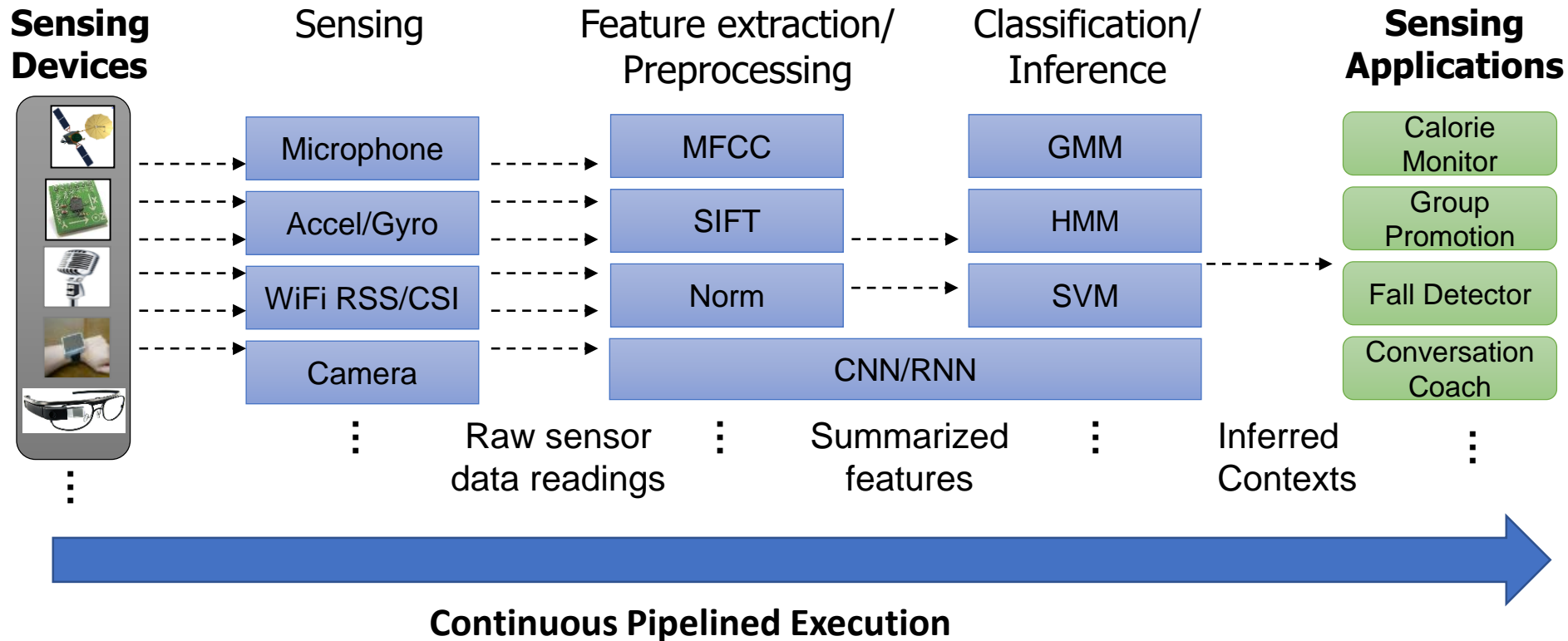
Conversation

Internal States

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

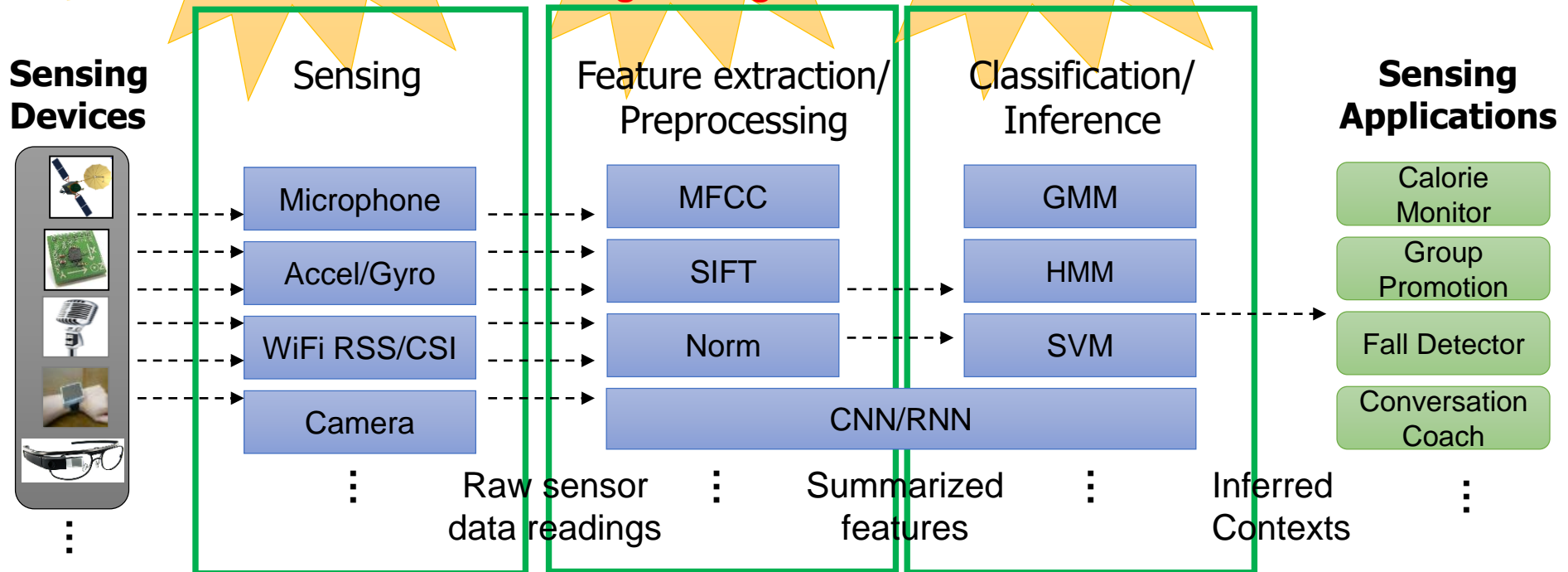
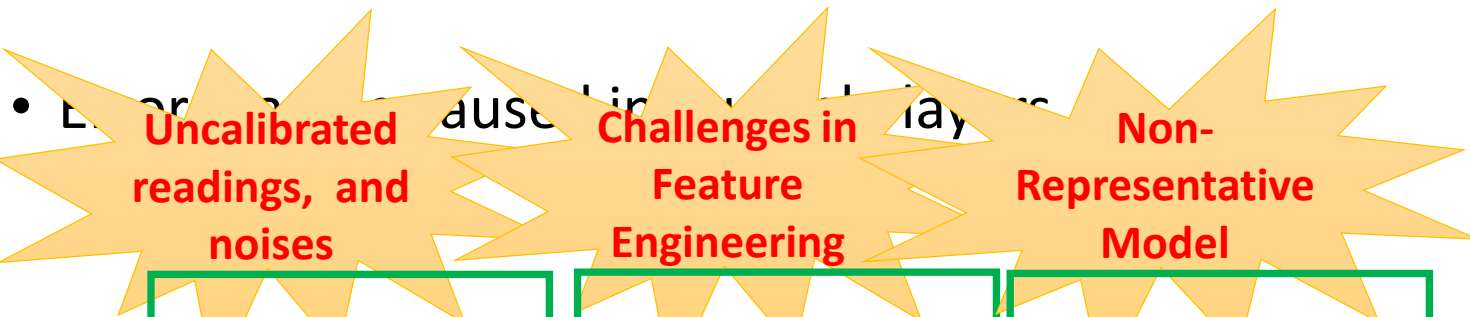
Common Computational Flow

Continuous sensing and analytics of user activities, location, emotions, and surroundings with mobile/IoT/wearable devices



Challenge 1: Inference Accuracy

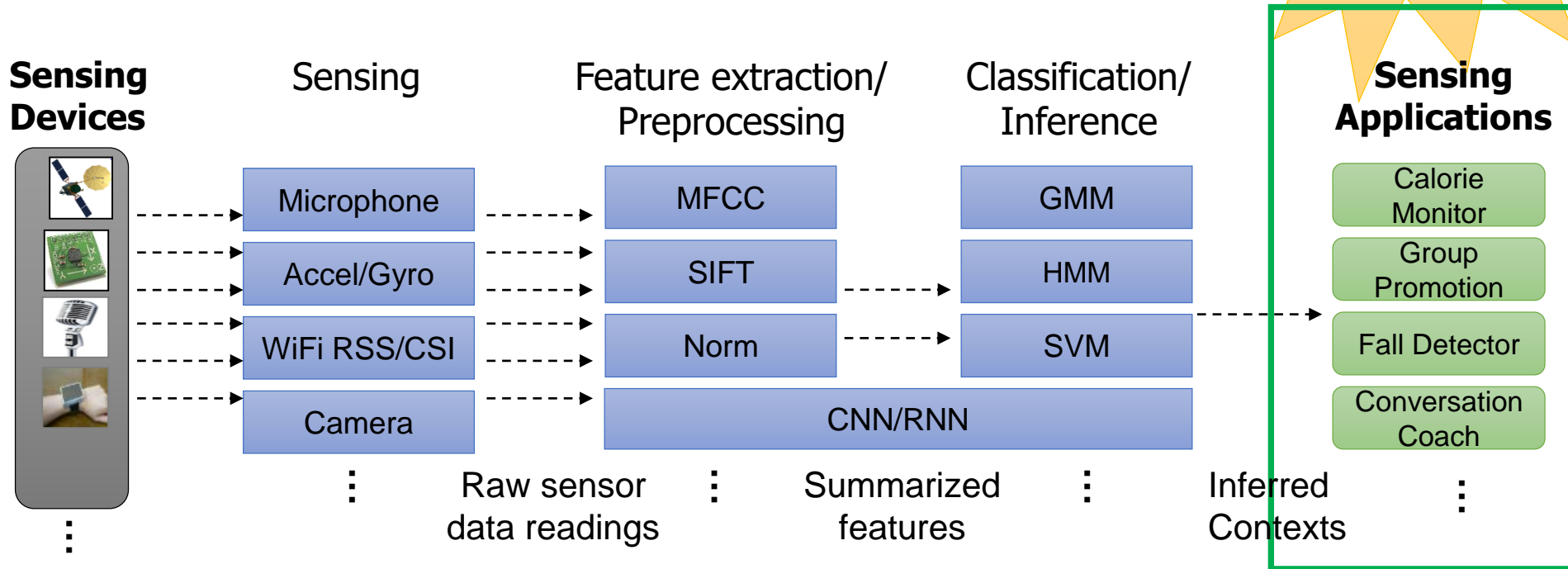
- > 90% accuracy is extremely challenging.



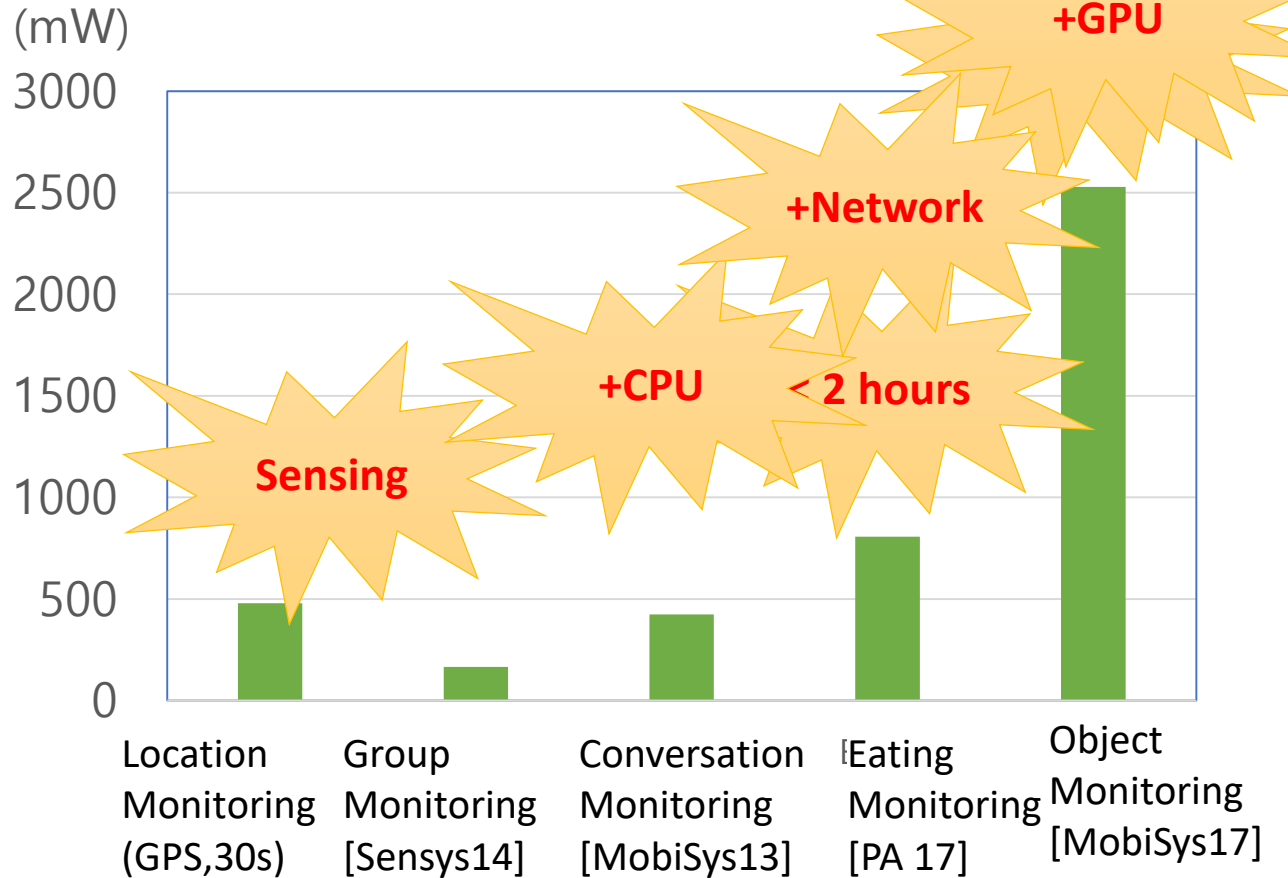
Challenge 2: Application Usability

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

**App Design with
Inaccurate
Results**

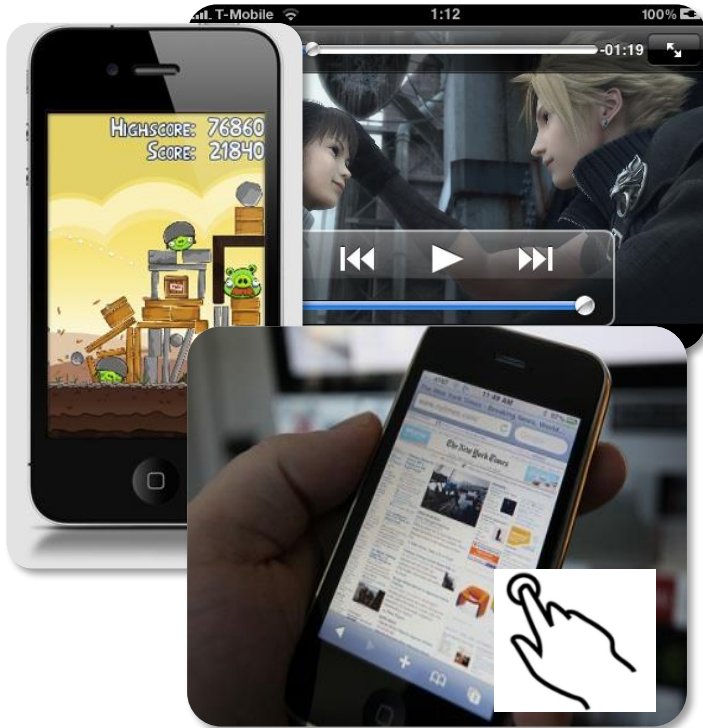


Challenge 3: Power Scarcity



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

Challenge 4: New Operational Mode



Small display, user mobility
→ A single user-interactive application

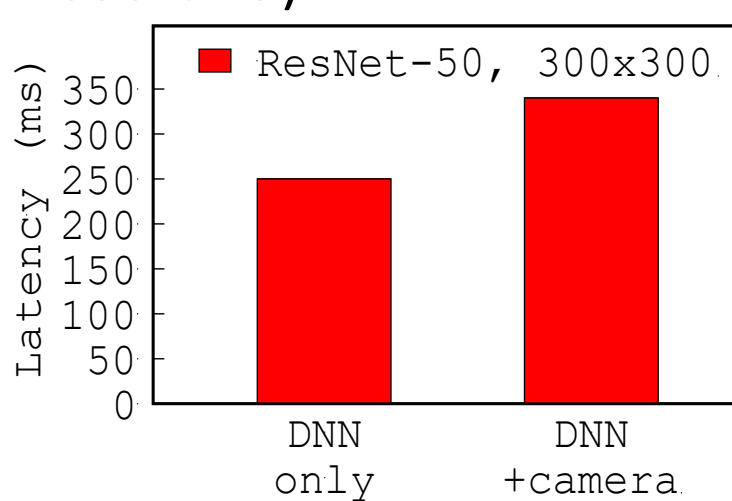
Vs.



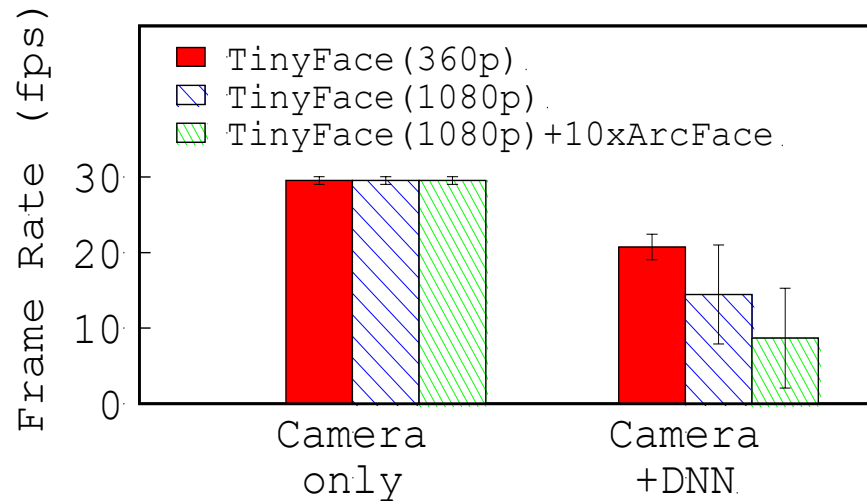
Mobile sensing : *autonomous, situation-aware services*
→ Concurrent background sensing applications

Challenge 5: Resource Contention

- Ran face detector continuously (with a TinyFace CNN-based Model) and a foreground camera rendering concurrently.
- Measured frame rates on Google Pixel 3 XL (Qualcomm Adreno 630 GPU).



DNN Execution Time



Frame Rate of Camera Rendering

DNN Execution and Rendering Performance

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???



Individual Apps Solve All These?

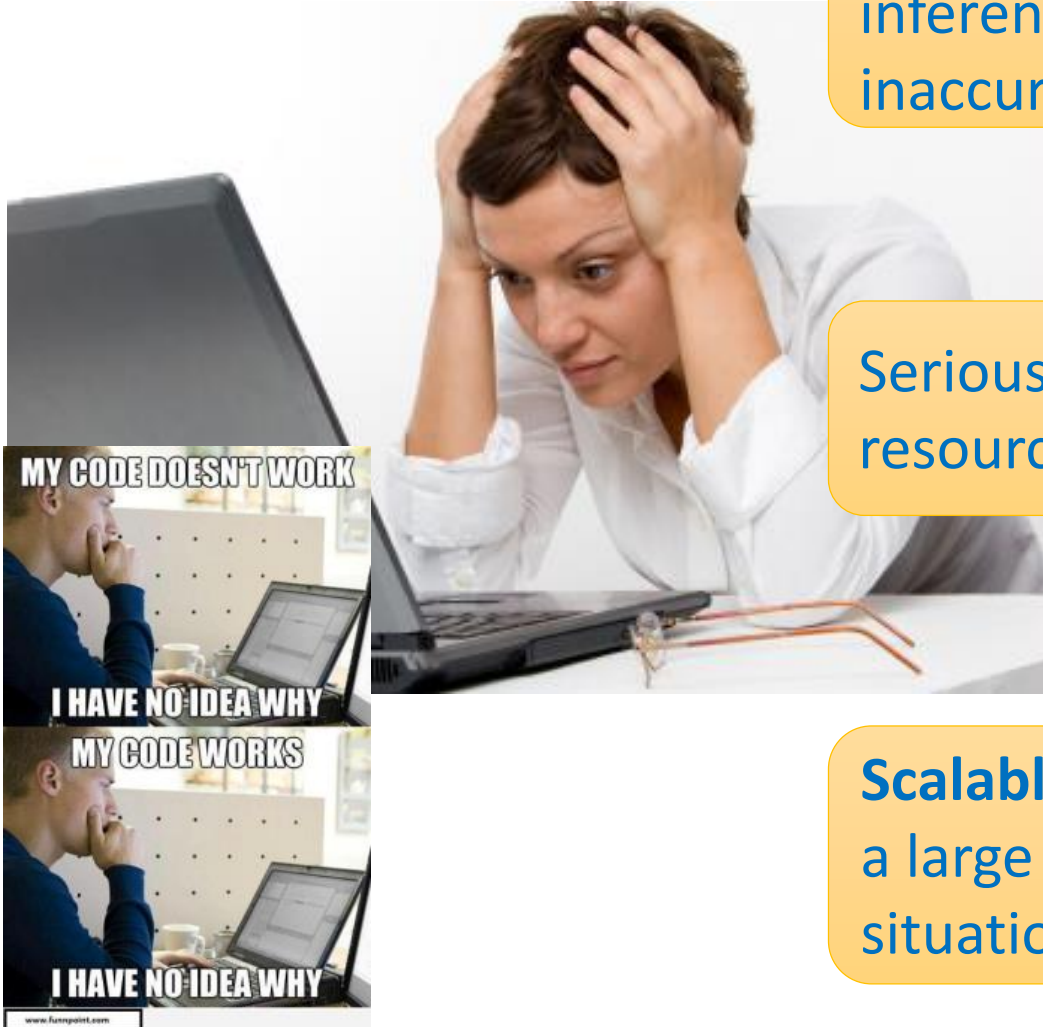
Complexity to implement **accurate** inference logics and handling inaccuracy in app design



Serious **optimization** in battery and resource usage



Scalable deployment and testing with a large pool of real users in real-life situations



Full-Fledged Mobile Sensing Platform

Life-Immersive Sensing Applications

"Notify me when the user is in a group of 3"

Group-aware App

Fall monitoring

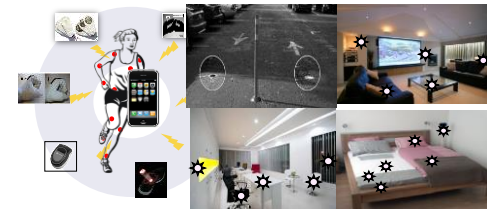
Simple and Intuitive Context Specification



Context Sensing and Analytics Platform

(on mobile/IoT/wearable devices and clouds)

Abstraction of Inference Logic and Runtime Resources



A rich set of mobile/IoT/wearable devices

Course Objectives

- Upon completion of the course, you should be able to:
 - ✓ Understand key concepts and technical underpinnings of various machine learning techniques.
 - ✓ Apply machine learning models to various real-world problems.



Class Timings

4주차	2/3	2/4	2/5	2/6	2/7
오전	데이터마이닝 (심규석)	데이터마이닝 (심규석)	데이터마이닝 (심규석)	데이터마이닝 (심규석)	데이터마이닝 (심규석)
오후	기계학습 (이영기)	기계학습 (이영기)	기계학습 (이영기)	기계학습 (이영기)	기계학습 (이영기)
5주차	2/10	2/11	2/12	2/13	2/14
오전	기계학습 (이영기)	기계학습 (이영기)	기계학습 (이영기)	심층학습 (윤성로)	기계학습 (이영기)
오후	심층학습 (윤성로)	심층학습 (윤성로)	심층학습 (윤성로)	심층학습 (윤성로)	기계학습 (이영기)

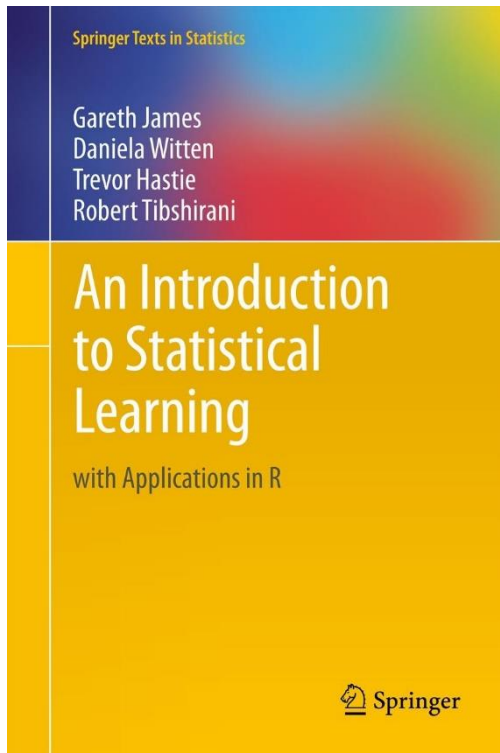
Pre-Requisites

- Took an undergraduate “introduction to statistics” and “introduction to linear algebra” courses.
- Familiar with python.

Textbook

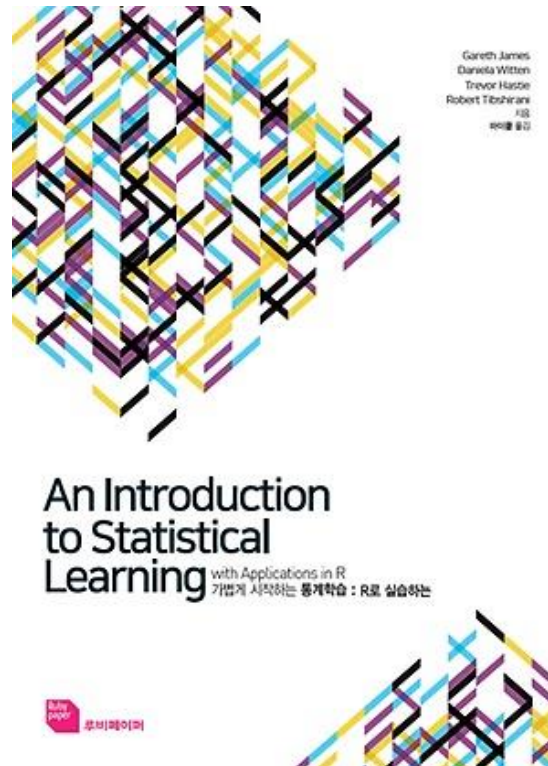
- Week 1

✓가볍게 시작하는 통계학습



- Week 2

✓Lecture Slides



Philosophy of the Textbook

- It is important to understand the ideas behind the various techniques, in order to know how and when to use them.
- One has to understand the simpler methods first, in order to grasp the more sophisticated ones.
- It is important to accurately assess the performance of a method, to know how well or how badly it is working [simpler methods often perform as well as fancier ones!]
- This is an exciting research area, having important applications in science, industry and finance.
- Statistical learning is a fundamental ingredient in the training of a modern data scientist.

Labs

- <https://gitlab.com/chphch/ISLR-python>
- Version
 - ✓ Python $\geq 3.5.2$
 - ✓ requirements.txt

Tentative Lesson Plan

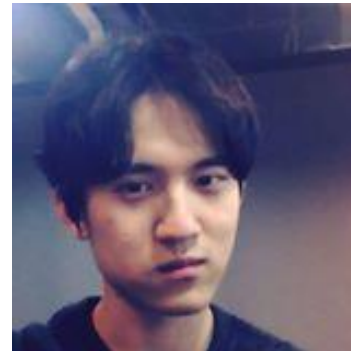
Day	Lecture Topic	Evaluation Milestones
1	Class Intro & Intro to Machine Learning	
2	Linear Regression and Classification	
3	Cross Validation and Variable Selection	
4	Non Linearity and Tree-based Models	
5	Support Vector Machines and Unsupervised Learning	Midterm Quiz (30%)
6	Hidden Markov Models	
7	Introduction to Convolutional Neural Networks	Mini Project Due (30%)
8	Application 1: Face Detection and Recognition	
9	Application 2: Human Activity Recognition	
10	Application 3: Healthcare Applications	Final Exam (40%)

Teaching Assistant

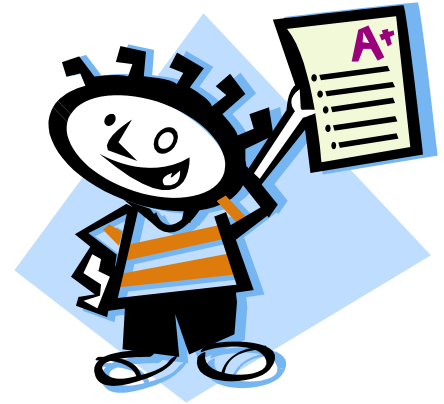
- Jingyu Lee (이진규)
- Changmin Jeon (전창민)
- Hyunwoo Jeong (정현우)
- How to contact TAs?

✓ Email to hcs.lab.2019@gmail.com.

✓ TAs will check the kakaotalk chatroom.



Assessment



- Project (30%)
- Midterm Exam (30%)
 - ✓ Week 1 Friday, Closed Book, Week 1 Materials
- Final Exam (40%)
 - ✓ Week 2 Friday, Closed Book, All Materials
- Useful link
 - ✓ Answers to textbook exercises:

<http://blog.princehonest.com/stat-learning/>

Quick Intro of Everybody

- Brief Intro.
- What ML problems you are interested / solving.

Most Importantly ...

- Let's work hard but have fun!

