



"Engineerable Al" Project

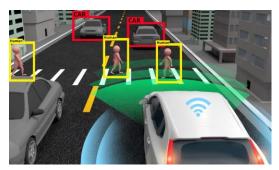
Engineerable Al Techniques for Practical Applications of High-Quality Machine Learning-based Systems

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Background

- Active investigation for industrial applications of machine learning (ML)-based Al systems
- → Increasing demands for quality assurance
 - ■Not only "higher accuracy for the given dataset"
 - Quality-sensitive application domains and customers



Autonomous Driving



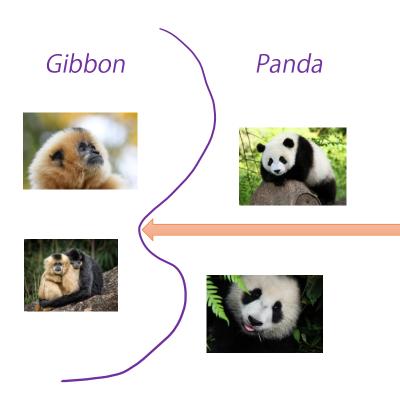
Medical Diagnosis



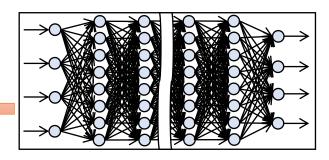
Process Automation

Background

Unique difficulties in ML: data-driven behavior and uncertainty



Software 2.0 as a collection of enormous parameters in deep neural networks



- Unclear boundary of functionality
- Unpredictability of quality to be achieved
- Weakness against noisy inputs

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Behavior (classification this time) is generated from training data

[http://free-photos.gatag.net/]

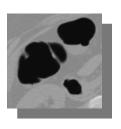
Problems



From discussions/surveys with 200+ practitioners

Medical Diagnosis

Want AI to detect cases overlooked by human doctors!



Failure with polyps at specific locations

Autonomous Driving

Want to ensure safety of Al under various situations!



Failure with pedestrians under specific situations

Difficult with very few data!

Rare cases more difficult for Al

Years for collecting data necessary in using deep learning

Difficulty with unstable assurance activities!

Enormous target situations

Trial-and-error for fixing something, then others broken





Our Vision

"Hard to Engineer" problems, why?

By building the whole functionality only with data

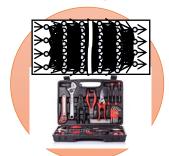
- Dependency on large data
- Uncontrollability of detailed behavior
- **→**Towards "Engineerable AI"

Effectively tailored for quality requirements

By incorporating "knowledge" not only data

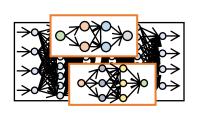
- Knowledge-incorporated construction
- Knowledge-driven improvement

Al Techniques + Software Engineering



Technical Approach

- (1) Knowledge-incorporated construction
 - Design models to reflect domain requirements

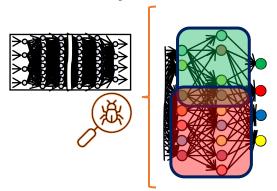


Example:

design modules to deal with individual types of cases

Achieve high performance with fewer data with more controllability and accountability

- (2) Knowledge-driven debugging
 - ■Improve models by identifying causes of success/failure



Example:

identify parts responsible for successful behavior and those for undesirable behavior

Controlled fix for targeted issues without unexpected degradation

Proof-of-Concept

- (1) Knowledge-incorporated construction
 - Design models to reflect domain requirements

Medical Diagnosis



Want AI to detect cases overlooked by human doctors!



Done: higher performance diagnosis with only 100 data



Difficult cases with less data

- (2) Knowledge-driven debugging
 - ■Improve models by identifying causes of success/failure

Autonomous Driving



Want to ensure safety of Al under various situations!

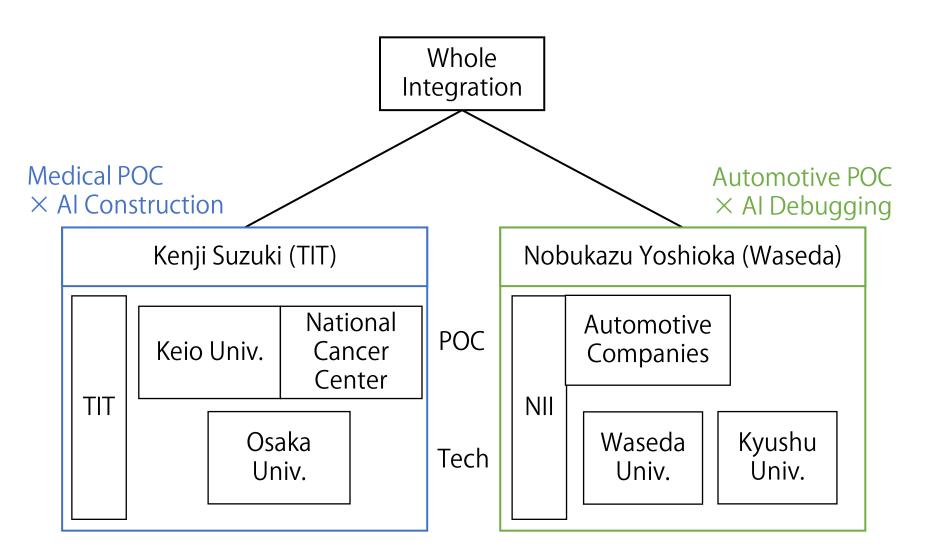


Done: controlled fix with little degradation in significant situations



Safety with enormous situations

Project Teams



Summary

- Engineerable Al project
 - For high-quality ML-based systems
 - Tackling difficulties in dependency on large data and uncontrollability of deep learning

Medical Diagnosis



Want AI to detect cases overlooked by human doctors!

Autonomous Driving



Want to ensure safety of Al under various situations!

Engineerable Al techniques

Knowledge-incorporated construction Knowledge-driven debugging