

Advances in Artificial Intelligence: Diverse and Collaborative Learning

Session co-chairs:

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Ion Matei, Palo Alto Research Center

Huge progress and success in recent **AI** based on **deep learning**



AlphaGo (Silver et al., 2016)

- ✓ Representation learning
- ✓ Monte Carlo tree search



Google Home

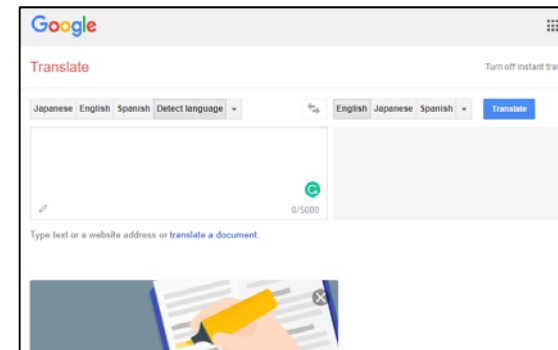
- ✓ Speech recognition & synthesis

<https://www.engadget.com/2018/06/12/google-home-handles-three-requests-at-once/>



Self-driving car (Google and so on)

- ✓ Pedestrian detection
- ✓ Semantic segmentation



Google Translate

- ✓ Neural machine translation

<https://translate.google.com/>

Shogi/ Japanese chess



IBM Watson



Deep learning era

Wining in ILSVRC (2012)
Google's cat (2012)
M&A of DeepMind (2013)
FB/Baidu's AI lab. (2013)
Alpha Go (2016)
Google Home (2017)

Application to robots and cars

Self-driving Vehicle
Pepper

Statistical
Natural language processing

Application to
Search engine

Task ontology

Watson(2011)

LOD (Linked Open Data)

Siri(2012)

Calo project
bot

Shogi(2012-) Go

Machine learning

Web, Big data

Medical diagnosis

MYCIN
DENDRAL

Expert system

Ontology

Eliza

Dialog system

Winter of AI...

Chess (1997)
Deep Blue

Search
Game, puzzle

Planning
STRIPS

1956

1970

1980

1995

2010

2015

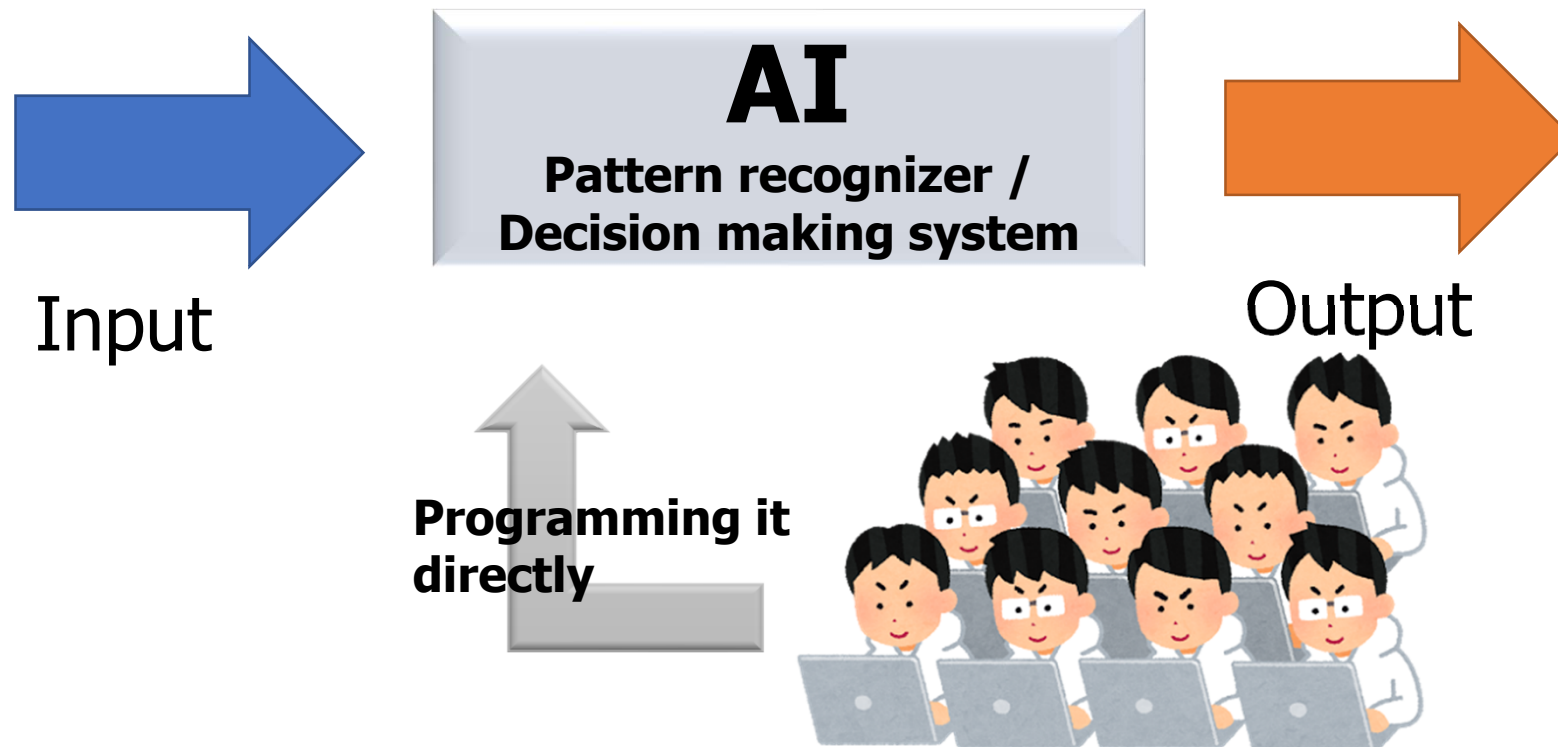
The 1st AI boom
(Search, Induction)

The 2nd AI boom
(Knowledge engineering)

The 3rd AI boom
(Machine learning, DL)

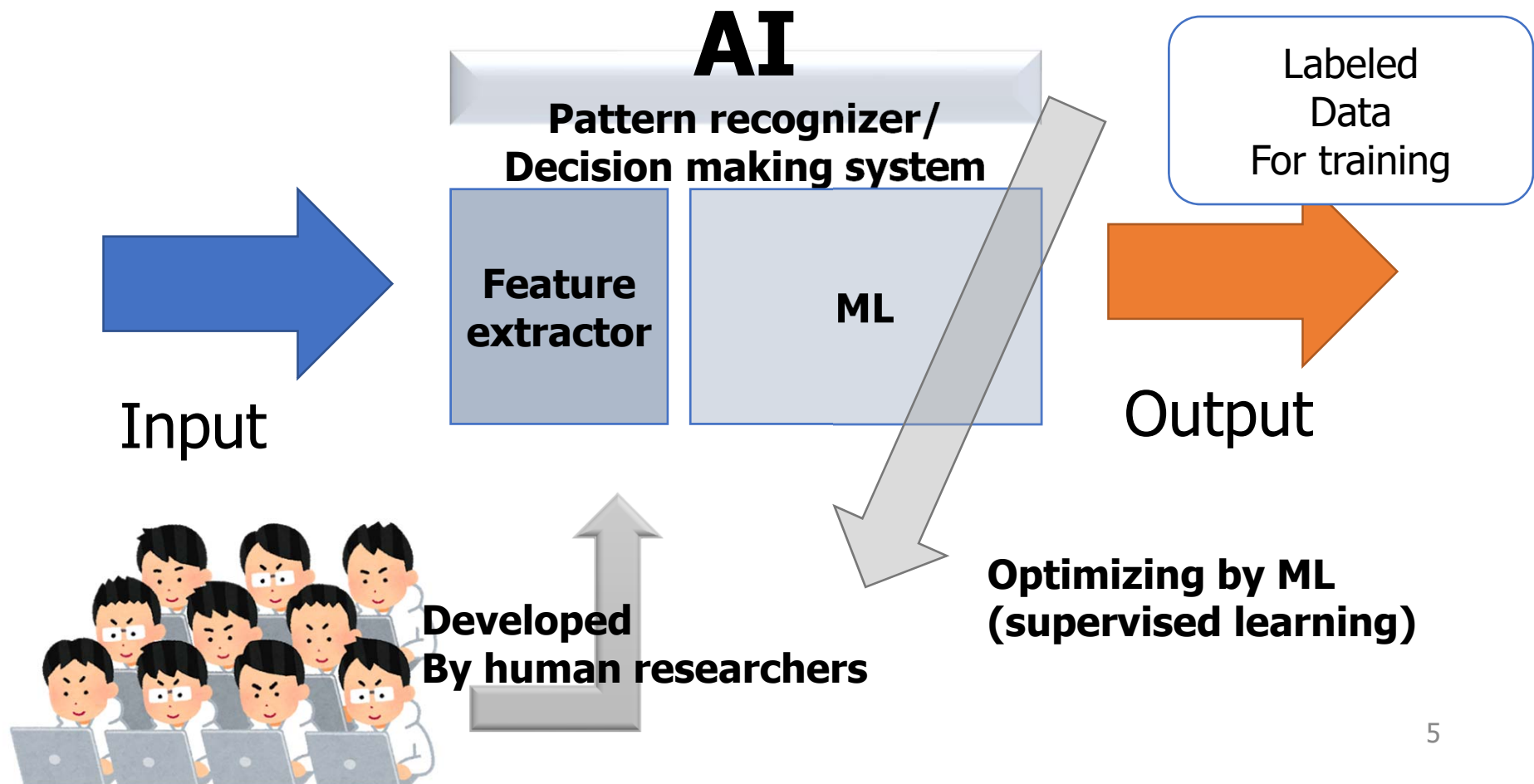
Rule-based AI

- Representative of the 2nd AI boom (1970s-)



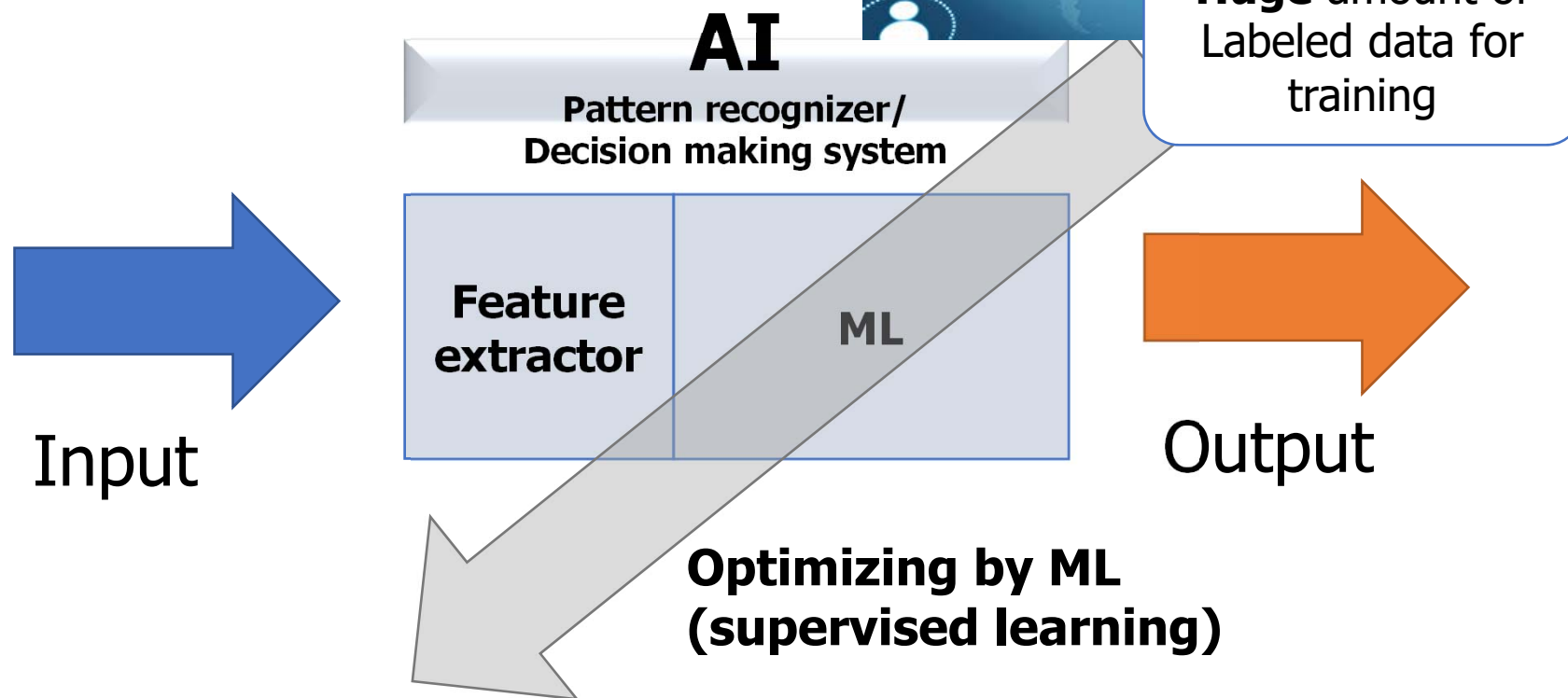
Progress in Machine Learning

- Progress in 1990s and 2000s



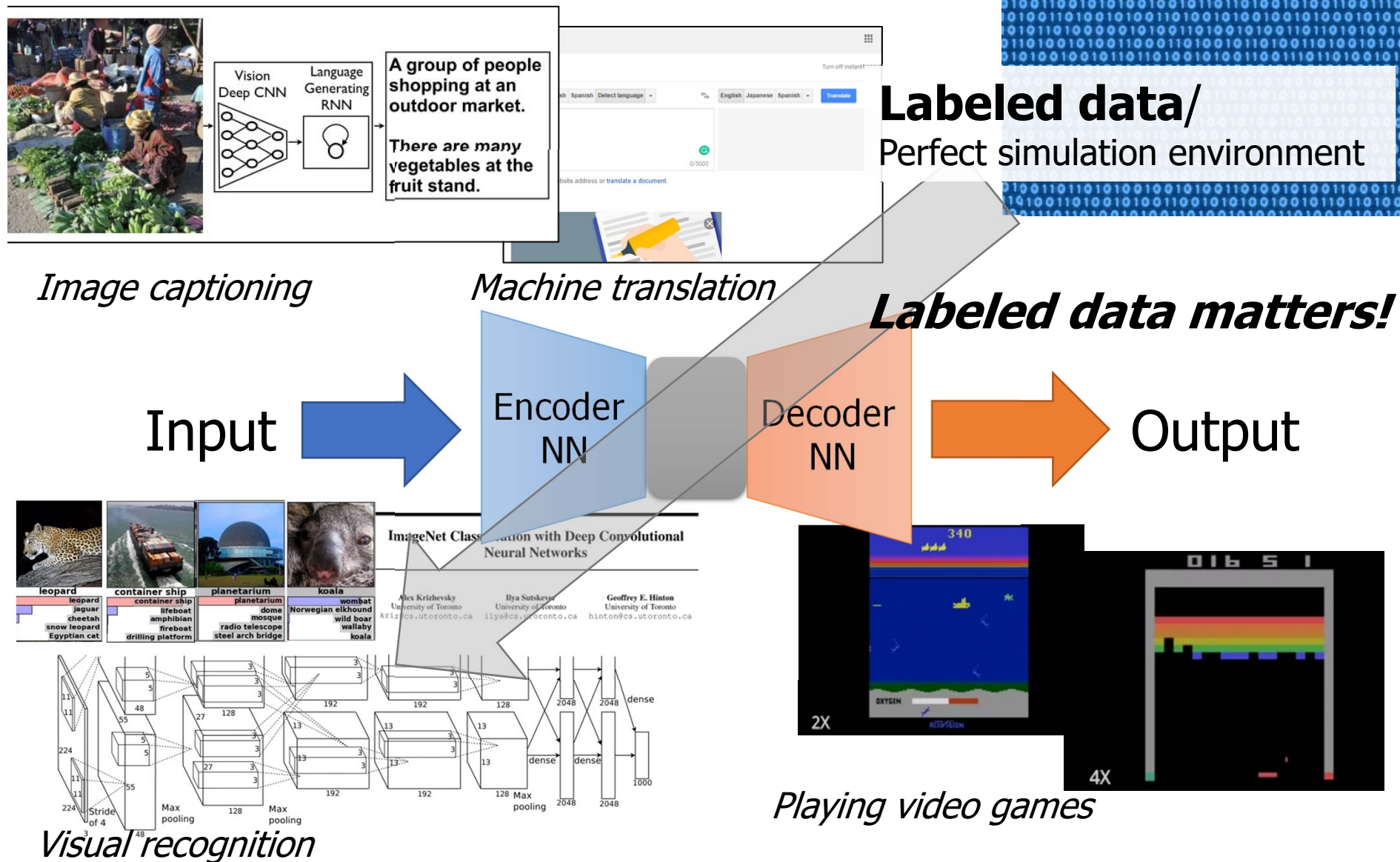
Deep learning

- The 3rd AI boom since around 2013



End-to-End learning

Encoder-decoder architecture



Labeled data make a machine intelligent!?

Deep learning solved everything?

Dialogue with AI researcher!

(Cynical but somehow typical)



Non-AI Guy



N: Hey! I want to use AI for my business to solve our problem. Can you give me some advice?

A: Okay. Do you have (labeled) data?

AI Guy



N: We have some. But, it's not enough, and other data has not been organized well.

A: Is it easy for you to prepare a large amount of labeled data?

N: No. I may be able to try. But, it will take a long time. We have several problems to do that.

A: Hmm. If you don't have enough data, I cannot help you. If you solve the problems and prepare a large amount of (labeled) data, please visit me again.

Then, the non-AI person gave up adapting AI/ML-based methods.

Reality

in practical application of AI

- ✓ We **don't have** a large amount of **labeled data** for our target domain.
- ✓ We often have massive **unlabeled data** obtained from web sites, social networks, sensor networks, and POS data.
- ✓ Our target domain is often not a static dataset, but often a **dynamic physical system**, e.g., industrial plants, robots, and networked IoT devices observing human activities.

Practical challenge in AI

- ❑ Let's use a small labeled dataset and a large amount of unlabeled data. (Semi-supervised learning)

Anomaly mining

- ❑ Let's make use of constraints of physical systems

Physical system analytics

How can we tackle with the "data sparsity" problems to solve real-world problems?

Relational data analysis

- ❑ Let's carefully model the observed unlabeled data (Unsupervised learning)

Robot control

- ❑ Let's make a robot collect data

US-side

Speakers



1. Leman Akoglu
(Carnegie Mellon University)

"Anomaly Mining: Beyond Detection"



2. Ion Matei
(Palo Alto Research Center)

"Learning Algorithms for Physical Systems:
Challenges and Solutions"

Perception
Big data

Action
Physical systems

JP-side



4. Iku Ohama
(Panasonic)

"Statistical Models for Discovering
Knowledge from Relational Data"

3. Takamitsu Matsubara
(Nara Institute of Science and Technology)



"Sample-Efficient Reinforcement Learning
for Real-World Robot Control"

Advances in Artificial Intelligence: Diverse and Collaborative Learning



□ 1st session (chair: Tadahiro Taniguchi)



1. Anomaly Mining: Beyond Detection
Leman Akoglu, Carnegie Mellon University
2. Learning Algorithms for Physical Systems: Challenges and Solutions
Ion Matei, Palo Alto Research Center



□ 2nd session (chair: Ion Matei)



3. Sample-Efficient Reinforcement Learning for Real-World Robot Control
Takamitsu Matsubara, Nara Institute of Science and Technology
4. Statistical Models for Discovering Knowledge from Relational Data
Iku Ohama, Panasonic



Enjoy the talk and the discussion of Advanced AI session!

