

# Group D: Big Data & AI

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# UQ and Numerical Methods in Kinetic Theory

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work with Martin Frank<sup>†</sup>, Jonas Kusch<sup>†</sup>, Kerstin Küpper<sup>★</sup>

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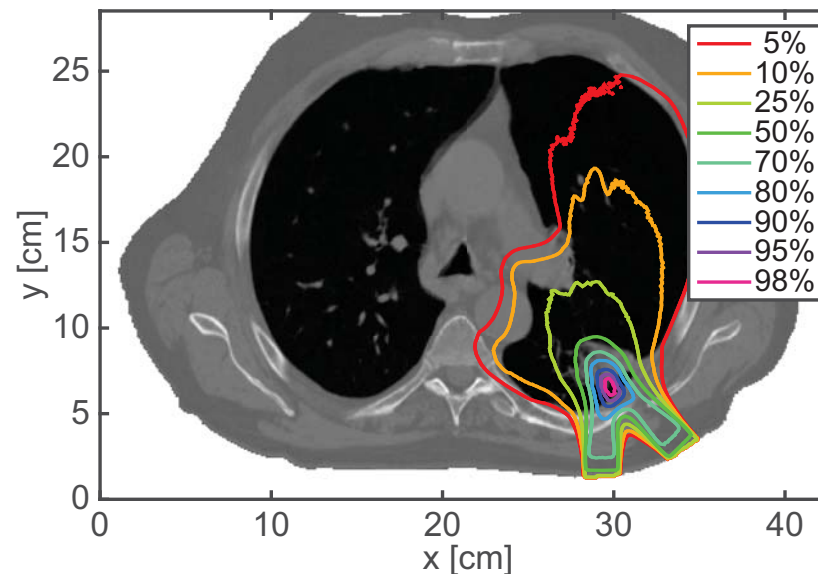
<sup>★</sup> Next Kraftwerke GmbH

RIKEN AICS HPC Youth Workshop 2018, Kobe



# What we are doing...

- ① Design numerical schemes for solving PDEs
- ② Application in radiotherapy
- ③ Quantification of uncertainties



Dose distribution by K. Küpper

# Challenges we are facing...

Designing numerical schemes:

- ① "optimal" parameters are often hard to find by theory
- ② which methods works best in which cases?

UQ:

- ① single simulation is already costly
- ② therefore hard to measure sensitivities of outputs w.r.t inputs
- ③ too much data to deduce (or even quantify) dependencies

# How Big Data & ML might help...

Designing numerical schemes:

- ① run experiments with different parameters and label outputs
- ② use this data to train ML algorithms to make predictions
- ③ only feasible with high performance machines

UQ:

- ① cluster medical data and reduce dimensionality
- ② train with given experimental / simulation data
- ③ triggering dependencies without running full simulations



# Simulation of Implanted Muons in high-Tc Superconductor System, $\text{YBa}_2\text{Cu}_3\text{O}_6$

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**Muon Spin Resonance Lab, Department of Condensed Matter Physics, Hokkaido  
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**Advanced Meson Science Lab, RIKEN Nishina Center**

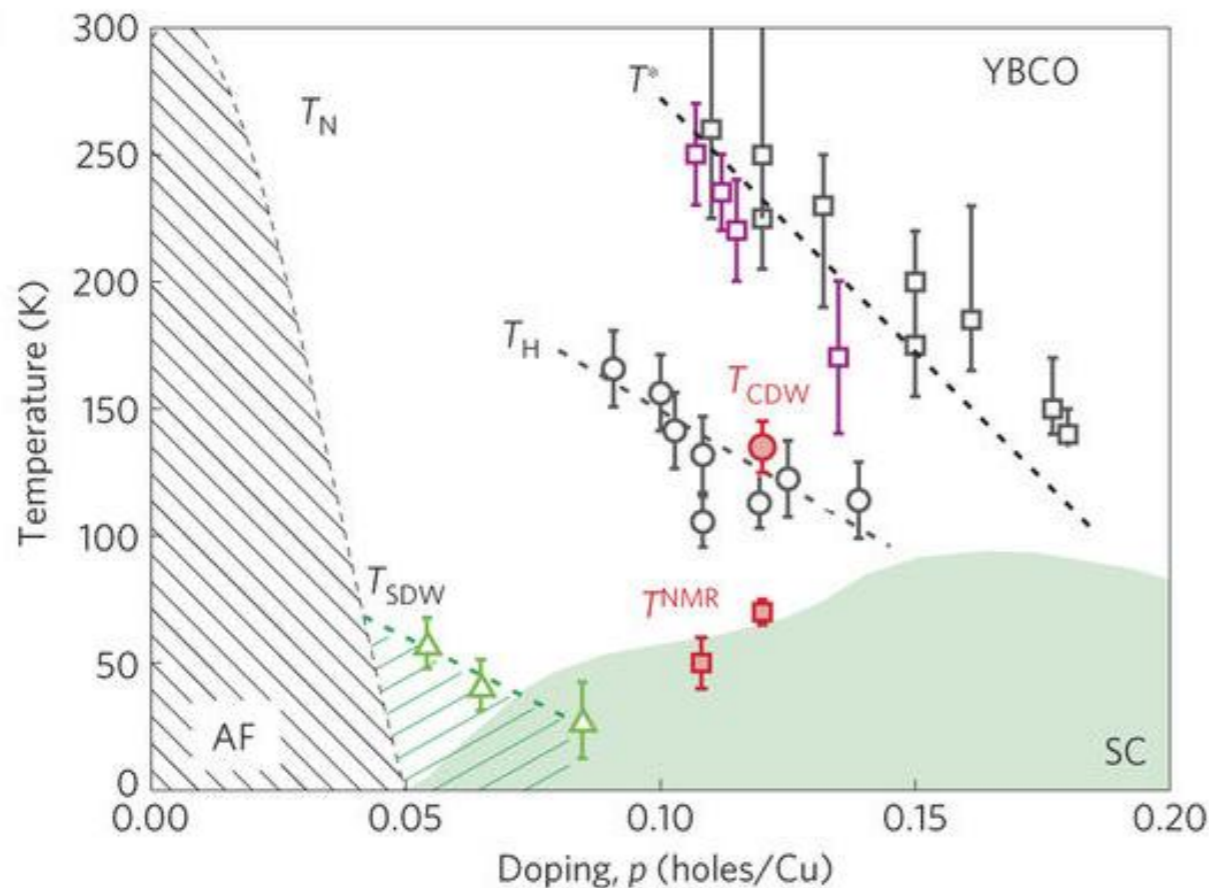
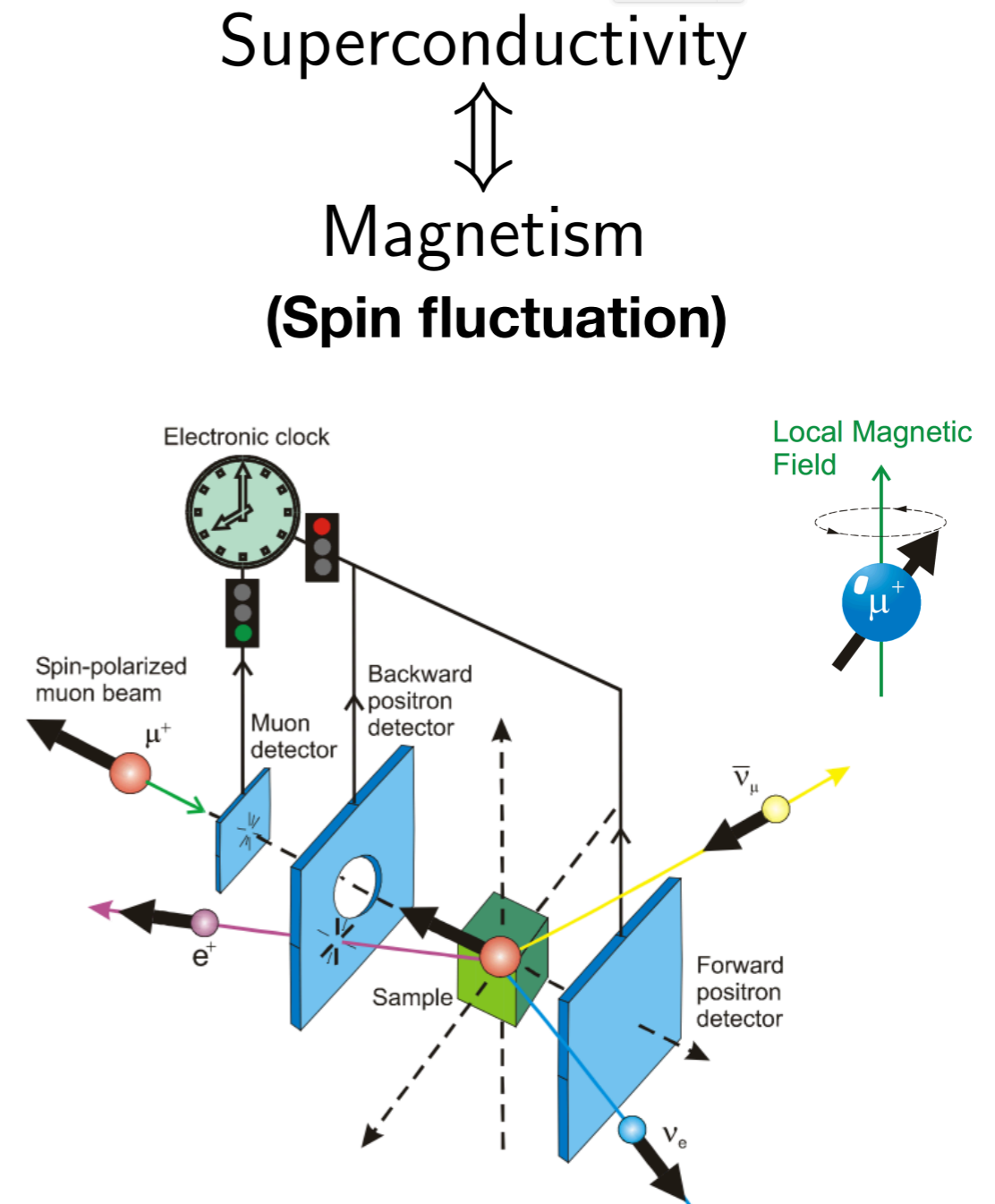


Figure 1. Phase diagram of YBCO<sub>6+x</sub>  
(Chang et al., Nat. Physics, 8, 2012)

- Antiferromagnetic Mott insulator at  $x=0$
- Magnetic ordering is disappear with emergence of superconductivity.
- SDW coexistence with SC



Probe the microscopic magnetism by  $\mu$ SR technique



# Our Work

- ★ **Magnetic order parameter ==> Muon stop position**
- ★ **Try solve that problem by using computing approach**
- ★ **Supercell calculation to simulate muon behavior in the system (~ 384 atom, 3200 electron and more), high performance computing plays important role.**
- ★ **We need tons of calculation and analysis amount big data in different doping concentration (x) and parameters to propose an explanation of high-Tc superconductivity mechanism that driven by spin fluctuation**



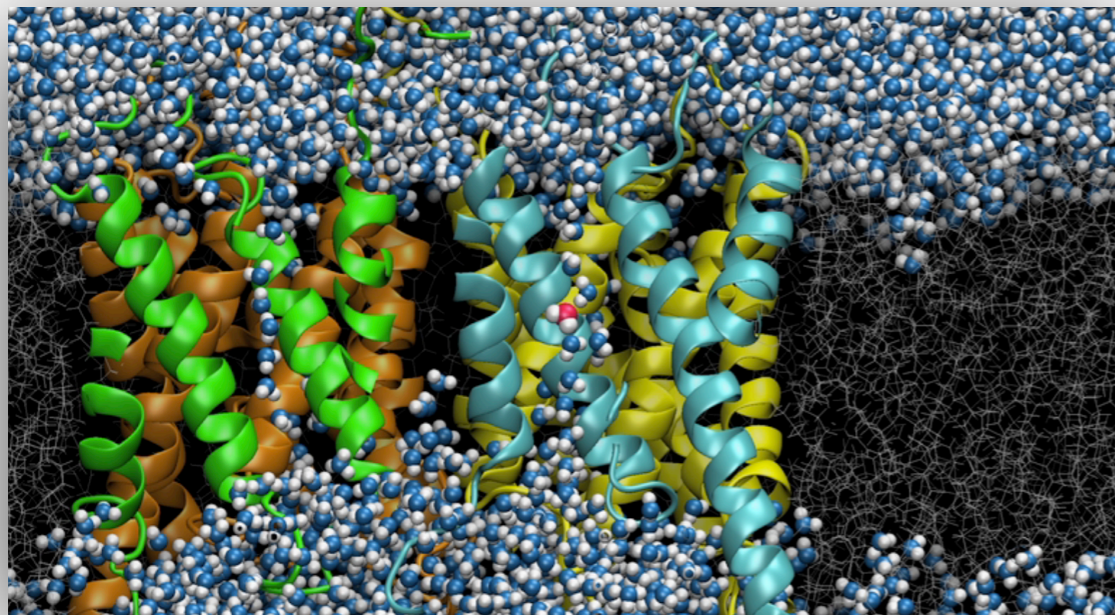
# All-atom molecular dynamics

Reproduce the dynamics of a system by using atomic model

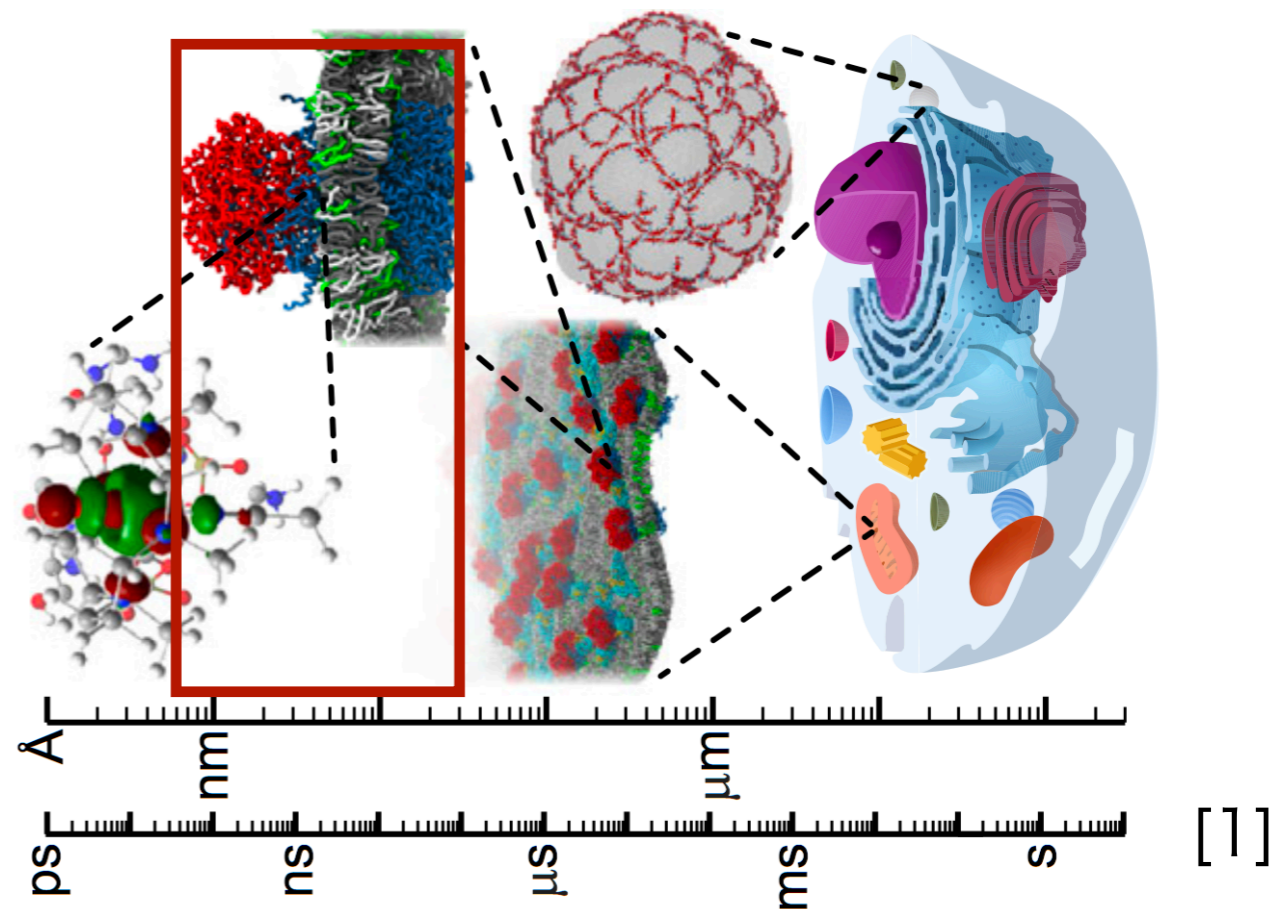
Newton's equation of motion

$$m \frac{d^2 \mathbf{r}}{dt^2} = \mathbf{F}$$

Water in aquaporin



This make it possible to analyse a phenomenon in a atomic scale



Calculation cost is high

[1] H. I. Ingólfsson, C. Arnarez, X. Periole and S. J. Marrink, *J. Cell. Sci.* **129**, 257 (2016).

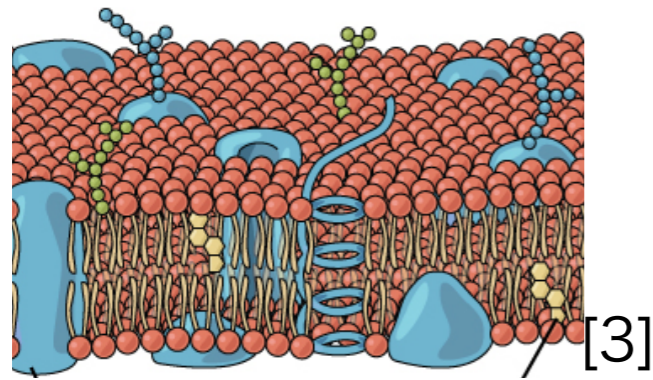
[2] E. Yamamoto, T. Akimoto, Y. Hirano, M. Yasui, and K. Yasuoka, *Phys. Rev. E.* **89**, 022718 (2014).

# Challenge: Analysis by machine learning

Analyse the atomic data which is obtained by molecular simulation

## Analysis example

Short time simulation



ns ~  $\mu$ s



Learn

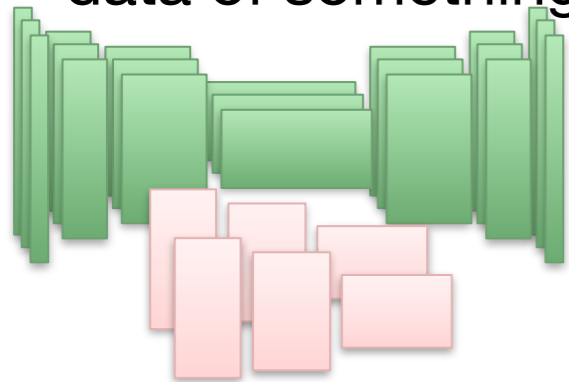
Predict the long-time-scale property

Motion  
Fluidity  
etc.

ms

**Cut the simulation cost**

data of something

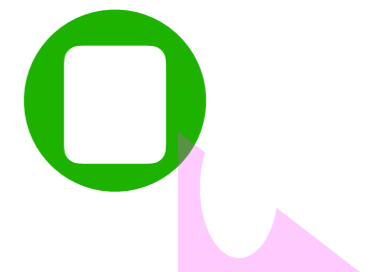


Learn

make a physical interpretation

Shape of something

**Make it easier**



# Uncertainty Quantification for Tsunami-drafted objects Based on Big Data

2018/02/06

RIKEN AICS HPC Youth Workshop FY2017

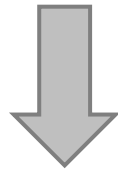
Yoshiki Mizuno

1<sup>st</sup> year of Master's Program

School of Engineering, The University of Tokyo

# Introduction: Uncertainty Quantification

- In my master's program, we have tried to **quantify uncertainties** for objects drafted by tsunami waves.
- Tsunami waves are generated by a lot of **uncertain** factors such as earthquakes, volcanic eruptions, and underwater landslides<sup>\*1</sup>.



Behaviors of tsunami-drafted objects damage certain structures in an **uncertain** way.



**A Ship Drafted by Tsunami Waves** <sup>\*2</sup>

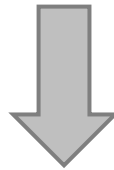
<sup>\*1</sup> A. Sarri, S. Guillas, and F. Dias,

"Statistical emulation of a tsunami model for sensitivity analysis and uncertainty quantification", 2012.

<sup>\*2</sup> <http://www.asahi.com/special/10005/TKY201104070204.html>

# Uncertainty Quantification and Big Data

- **Big data** related to these uncertainties enables us to analyze behaviors of tsunami-drafted objects.
- Bayesian framework is one of the possible approaches to infer the behaviors based on such **big data** <sup>\*1</sup>.



**Uncertainty quantification** has a strong potential to mitigate damage caused by tsunami-drafted objects.

<sup>\*1</sup> A. Sarri, S. Guillas, and F. Dias,

“Statistical emulation of a tsunami model for sensitivity analysis and uncertainty quantification”, 2012.



# How Big data and Deep learning may replace traditional compilers in HPC

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# My motivation

- Lot of approaches exists to write efficient code:
  - DSLs,
  - Runtimes,
  - Libraries, ...
- Although they are great, they have flaws:
  - Rely on compiler (may not take the best choices)
  - Imply overheads
  - Ask for lot of information
  - ...



# Current work

## 1. Improve collaboration between domain scientists & optimization specialists

- Through a DSL specifying the application's semantics

## 2. Assists optimization specialists in the simulation code optimization processing

- Using the semantics and source-to-source compiler
  - Asking for information not specified in the semantics
- Nothing automatic
  - You know what you want to do, I help you achieve this goal

➤ But what if I could guess this information ?





# Current work to infinity...

- I would know:
  - what to do (semantics)
  - how to do it (optimization choices)
  - parameters to do it efficiently
- It would ease optimization specialists job
  - Thus increase productivity
- How can I find these parameters ? Big data !
  - $F(\text{semantics, choices}) = \text{Parameters}$



# ...and beyond !

- But data may not be enough
  - We have to analyze it
  - Could make mistakes
- Let's use Deep learning for that
  - Find correlation impossible to make for us
- Next level
  - Compilers can't make next to optimal choices in every situation
  - Deep learning may achieve that:
    - Give it what you want to do (semantics)
    - And it will generate how it should be done for efficiency