



National Institute for Material Science



Research Trend of Data Driven Materials Research and its impact to Materials Science

Toyohiro Chikyow, Deputy Director

Center for Material Research by Information Integration (CMi2)

*Materials Data Science Group
Materials Data & Integrated System (MaDIS)
National Institute for Materials Science (NIMS)*

2019.9.6 CMD Workshop at Osaka Univ.

contents



NIMS

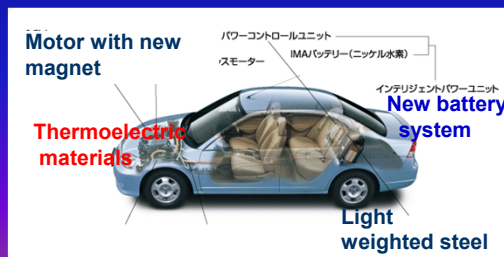


- **Materials Research : Past and present**
- **Present status of Materials Informatics**
- **What is “ machine learning or AI” ?**
- **History of
High Throughput Experimentation.**
- **Materials Research in future**
- **Conclusion**

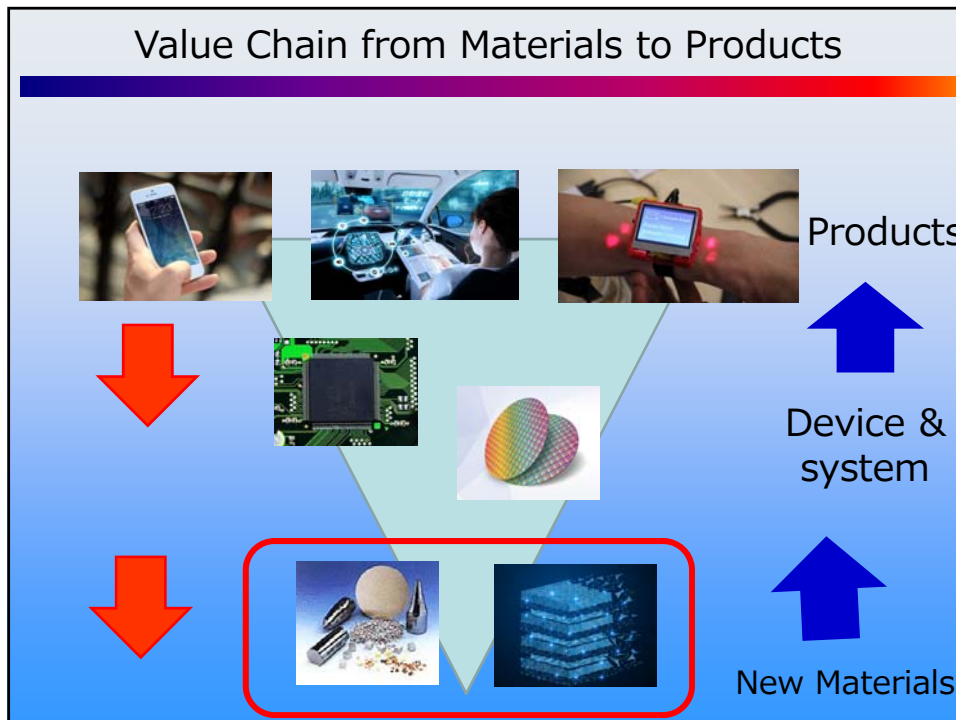
Materials innovation is the key to realize sustainable society



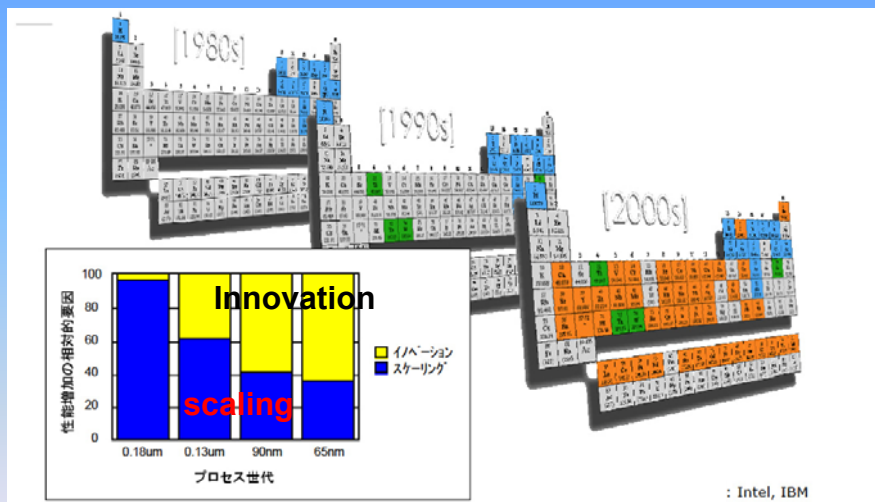
Super low power device and long life battery



Value Chain from Materials to Products

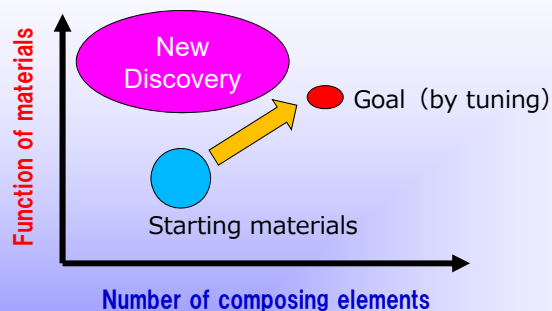


New Materials Innovation in nano device



“Innovation” is driven by “new materials” !

Why new discovery is difficult in materials



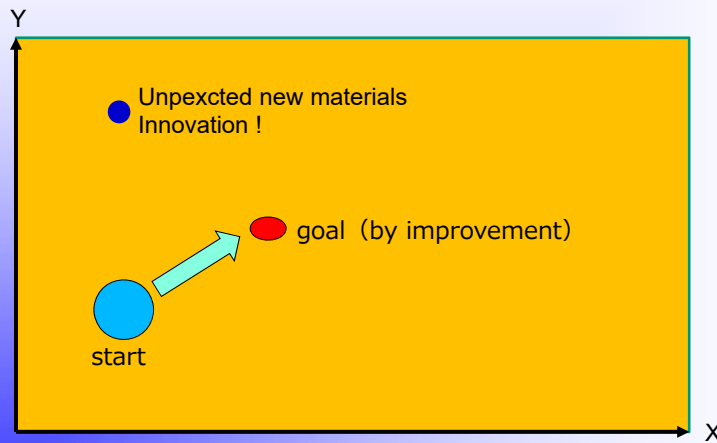
Starting Basic elements \rightarrow Basic+A \rightarrow Basic+A+B \rightarrow Basic+A+B+C

A:improvement Disadvantage A:Improvement B:Improvement disadvantage A:Improvement B:Improvement disadvantage C:Improvement

$$A > B > C$$

Extension of the Data Field & Space

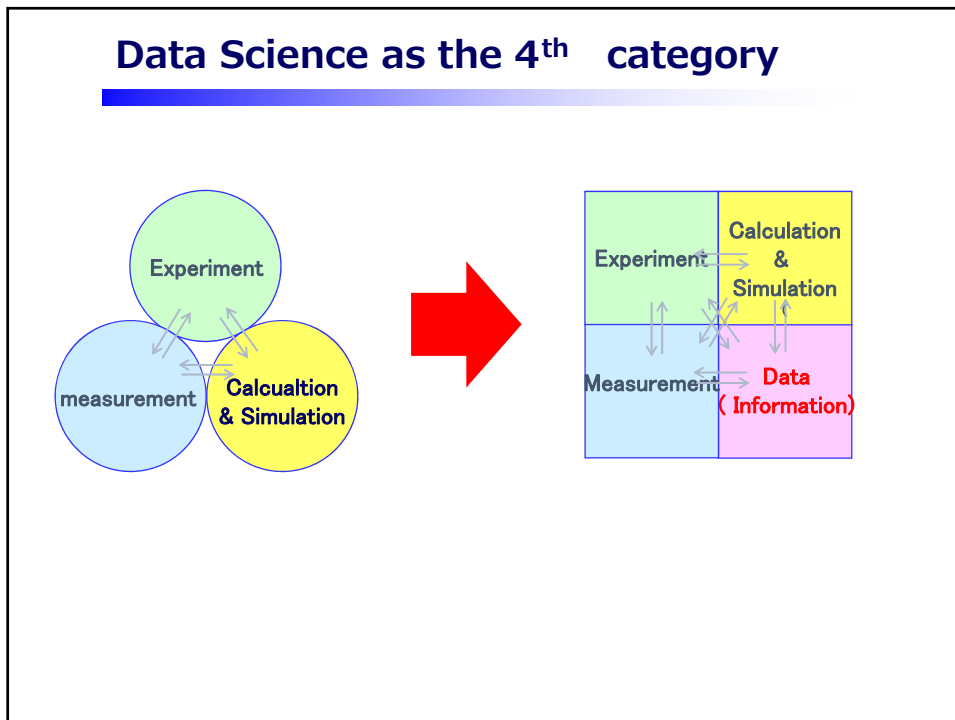
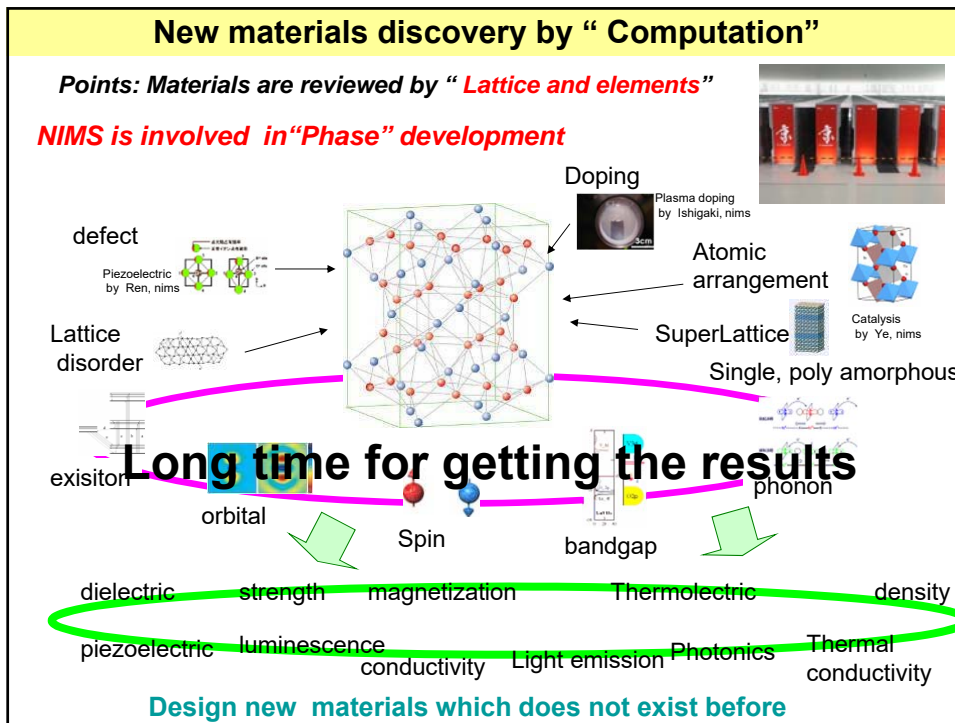
National Institute for Material Science



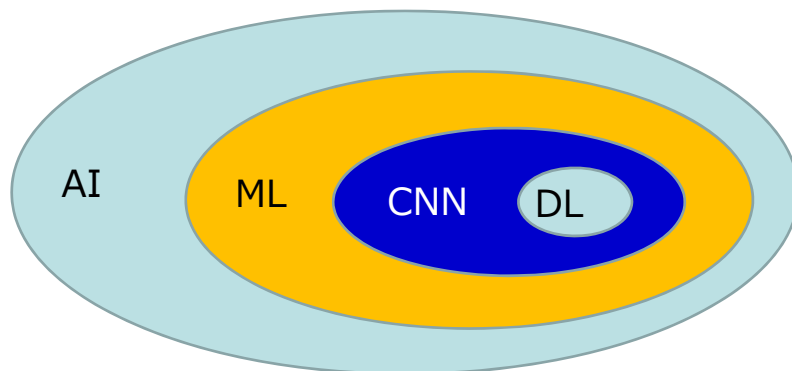
A large variety of the composition

PERIODIC TABLE OF THE ELEMENTS

1	1A																	18	18A						
1	H																	2	He						
2	Li	2A	Be											10	Ne										
3	Na	3A	Mg											16	Ar										
4	K	4A	Ca	Sc	Ti	V	Cr	Mn	Fe	Co	Ni	Cu	Zn	31	Ga	32	Ge	33	As	34	Se	35	Br	36	Kr
5	Rb	5A	Sr	Y	Zr	Nb	Mo	Tc	Ru	Rh	Pd	Ag	Cd	47	In	48	Sn	49	Sb	50	Te	51	I	52	Xe
6	Cs	6A	Ba	La-Lu	Hf	Ta	W	Re	Os	Ir	Pt	Au	Hg	81	Tl	82	Pb	83	Bi	84	Po	85	At	86	Rn
7	Fr	7A	Ra	Ac-Lr	Rf	Db	Sg	Bh	Hs	Mt	Uun	Uuu	Uub	111	Uut	112	Uuq	113	Uup	114	Uuh	115	Uus	116	Uuo
8			La	Ce	Pr	Nd	Pm	Sm	Eu	Gd	Tb	Dy	Ho	Er	Tm	Yb	Lu								
9			Ac	Th	Pa	U	Np	Pu	Am	Cm	Bk	Cf	Es	Fm	Md	No	Lr								

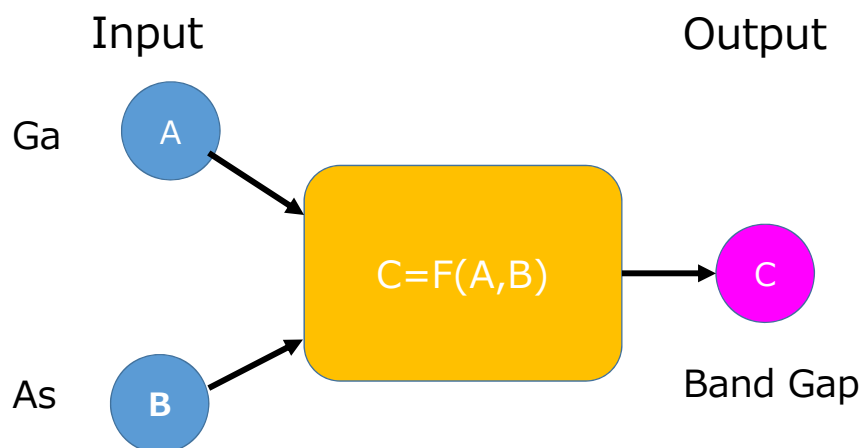


AI, ML CNN and Deep Learning



But data must be there for the AI

How we can use Machine Learning



$$C = d_1A_1 + d_2B_1 + d_3A_2 + d_4B_2 + \dots$$

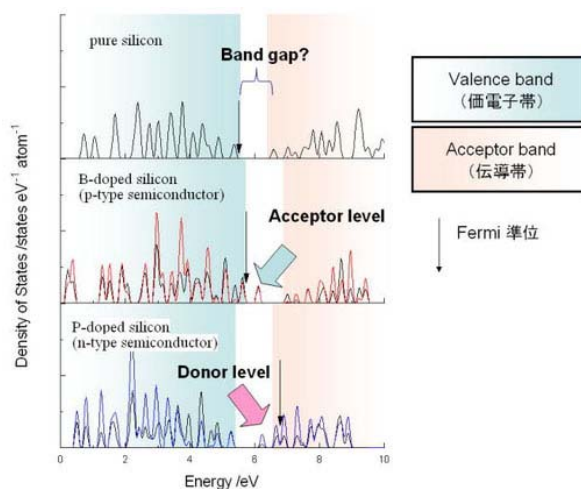
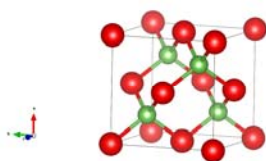
A_1, \dots, B_1, \dots : Descriptors

What is the descriptor ?

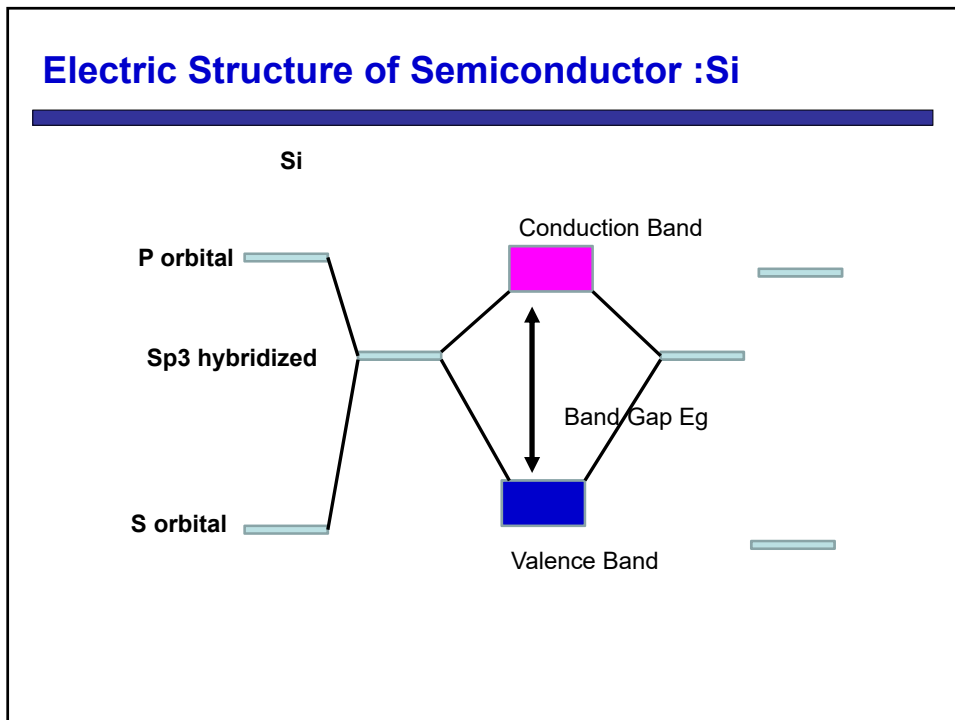
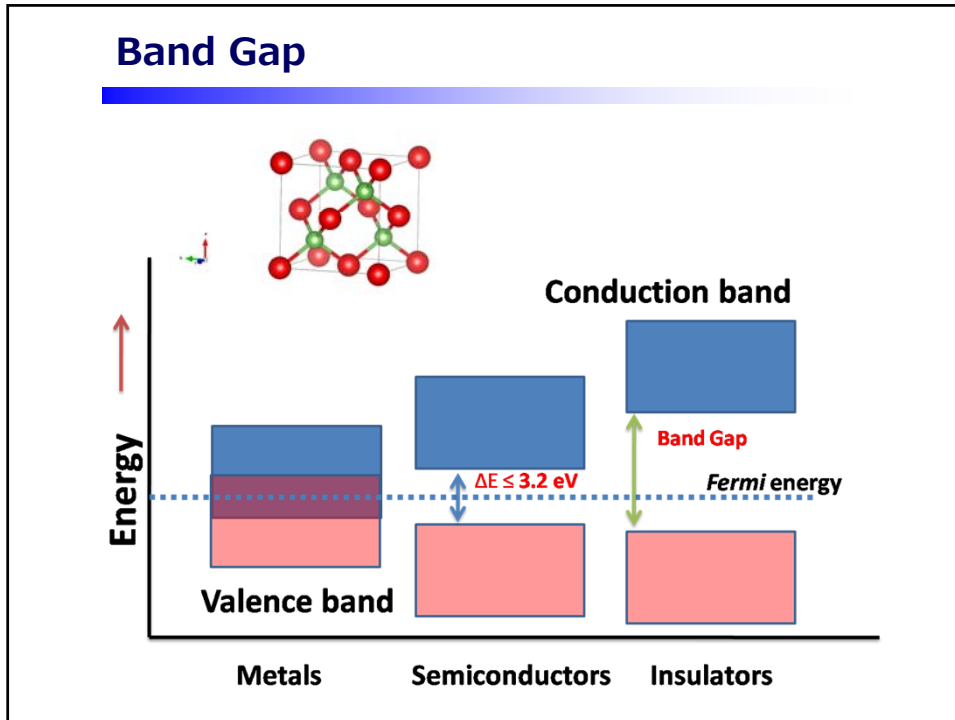
Descriptor

- 1) AI people:
"Just the parameter to fix the regression. No meaning"
- 2) Materials Informatics people:
"Those are important parameters to determine the property of the materials "

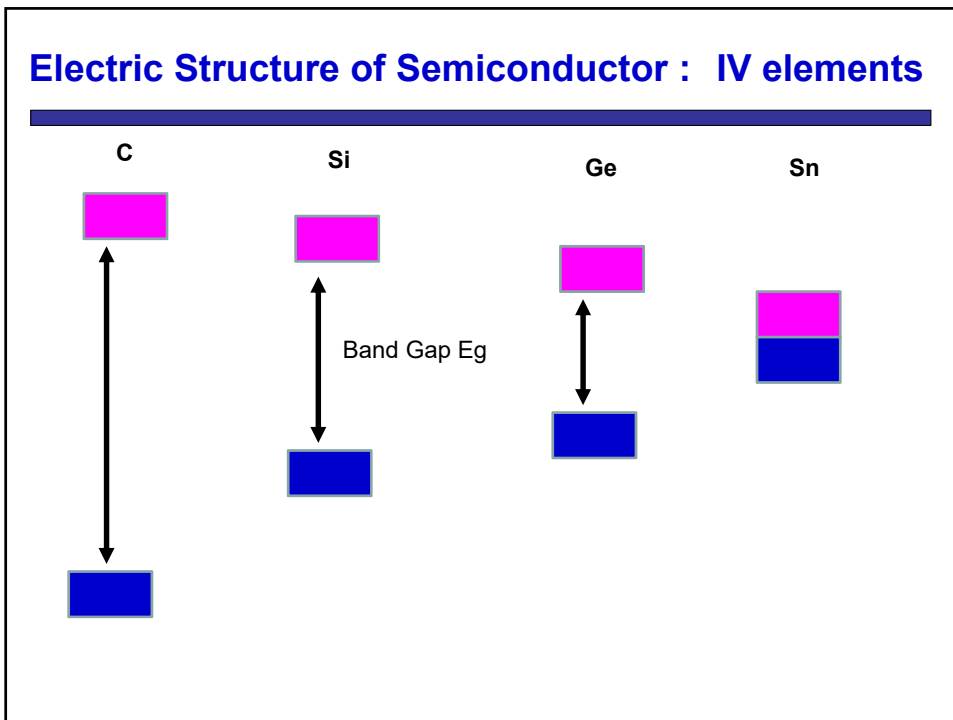
Band Gap



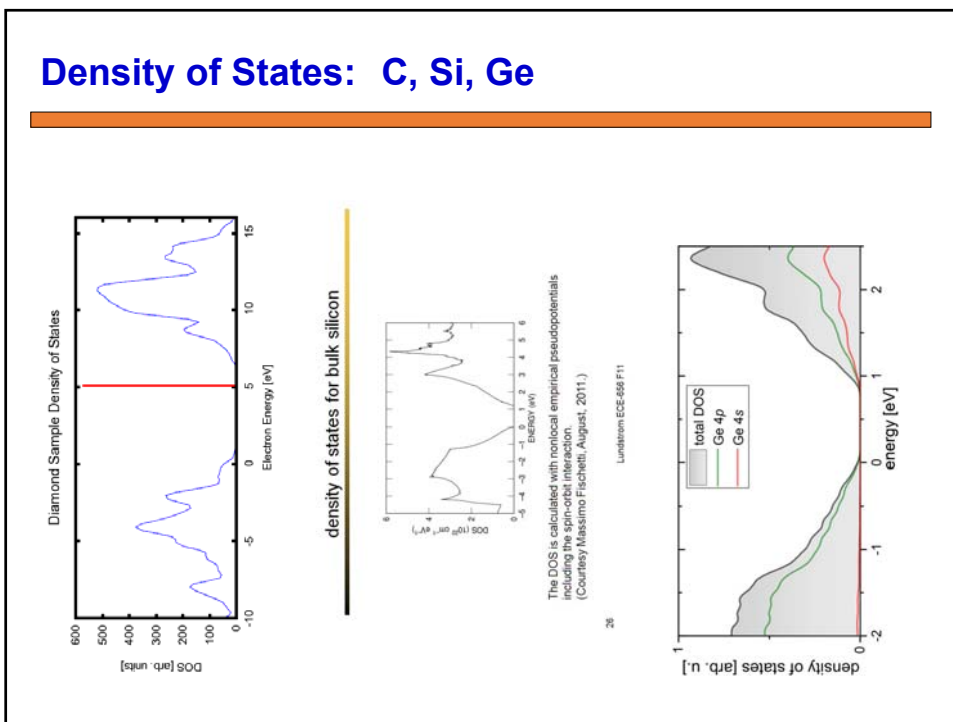
First principle calculation by Prof.Masanobu Nakayama, Nagoya Inst.of Tech



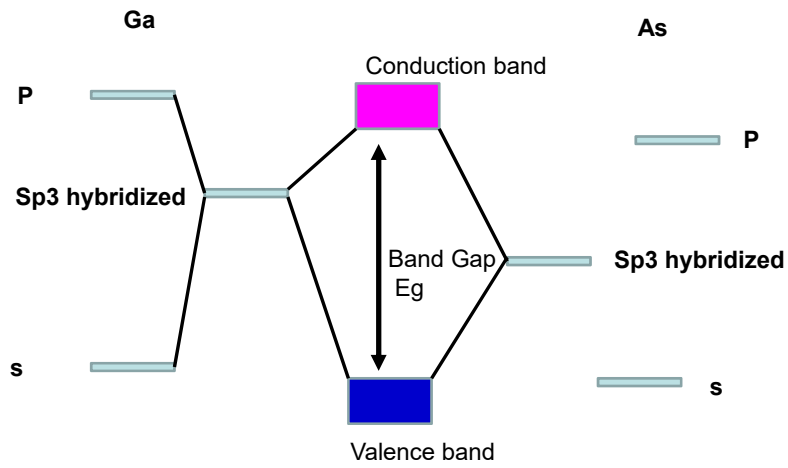
Electric Structure of Semiconductor : IV elements



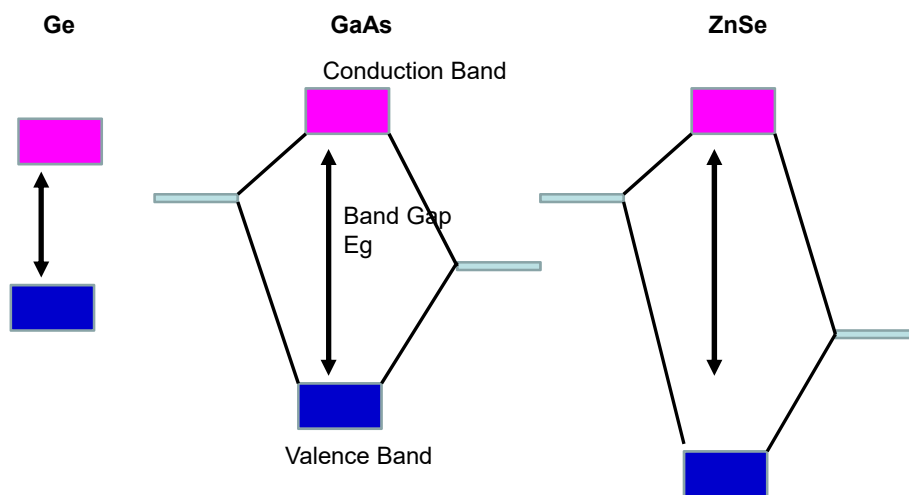
Density of States: C, Si, Ge

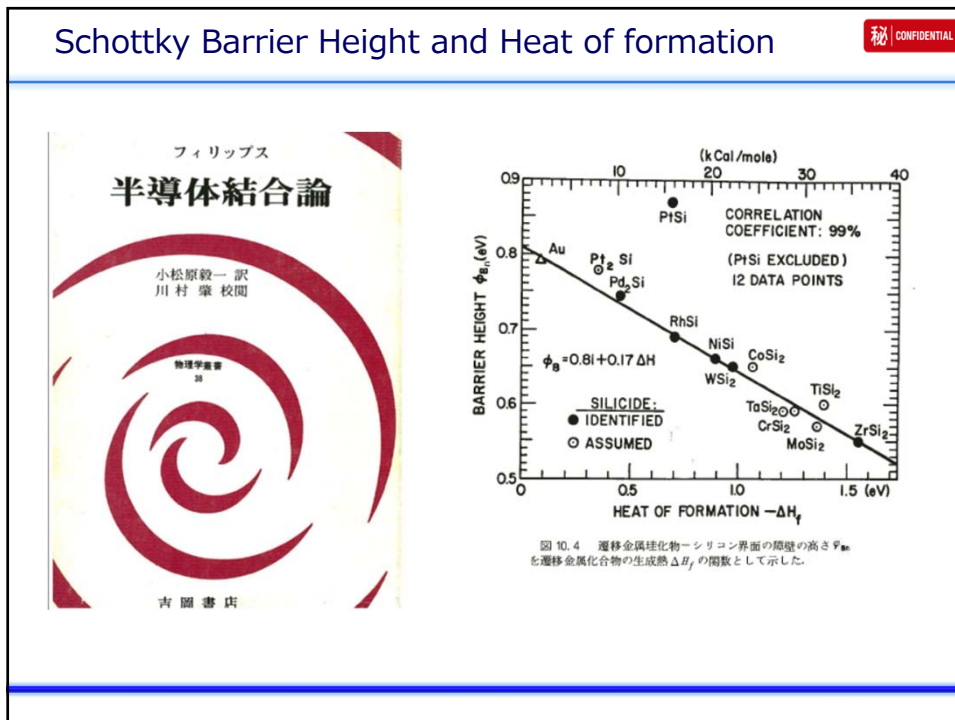
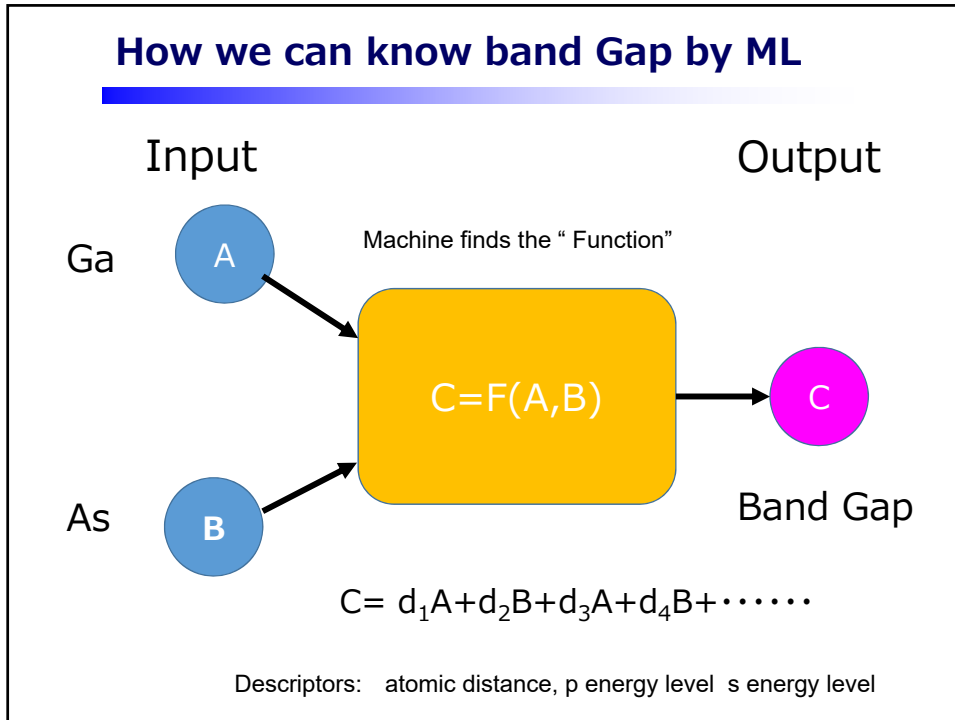


Electric Structure : GaAs



Electric Structure of Semiconductor : IV, II-V, II-VI



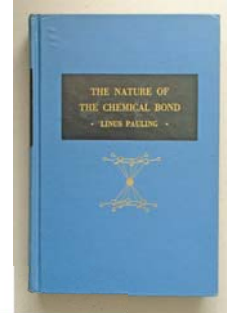


We want to design new materials ! (Electronegativity)

Heat of formation $\Delta H_f \propto X(\alpha) - Y(\beta)$

$X(\alpha)$: **Electronegativity** of X : α
 $Y(\beta)$: **Electronegativity** of Y : β

From "Nature of Chemical Bond" by Linus Poling



H 2.1																				
Li 1.0	Be 1.5															B 2.0	C 2.5	N 3.0	O 3.5	F 4.0
Na 0.9	Mg 1.2															Al 1.5	Si 1.8	P 2.1	S 2.5	Cl 3.0
K 0.8	Ca 1.0	Sc 1.3	Ti 1.5	V 1.6	Cr 1.6	Mn 1.5	Fe 1.8	Co 1.8	Ni 1.8	Cu 1.9	Zn 1.6	Ga 1.6	Ge 1.8	As 2.0	Se 2.4	Br 2.8				
Rb 0.8	Sr 1.0	Y 1.2	Zr 1.4	Nb 1.6	Mo 1.8	Tc 1.9	Ru 2.2	Rh 2.2	Pd 2.2	Ag 1.9	Cd 1.7	In 1.8	Sb 1.9	Te 2.1	I 2.5					
Cs 0.7	Ba 0.9	La 1.2	Hf 1.3	Ta 1.5	W 1.7	Re 1.9	Os 2.2	Ir 2.2	Pt 2.2	Au 2.4	Hg 1.9	Tl 1.8	Pb 1.8	Bi 1.9	Po 2.0	At 2.2				
Fr 0.7	Ra 0.9	Ac 1.1	Th 1.3	Pa 1.5	U 1.7	Np 1.3														

We want to design new materials !

Physica 100B (1980) 1-28
 © North-Holland Publishing Company

COHESION IN ALLOYS - FUNDAMENTALS OF A SEMI-EMPIRICAL MODEL

A. R. MIEDEMA
Philips Research Laboratories, 5600 MD Eindhoven, The Netherlands
 and
 P. F. de CHÂTEL and F. R. de BOER
Natuurkundig Laboratorium, University of Amsterdam, 1018 XX Amsterdam, The Netherlands
 Received 7 December 1979

Table Ia

Metal	ρ^* (V) [10, 13-15]	$\epsilon^{1/2}$ ($\text{eV}^{-1/2}$) [10, 13-15]	$\nu^{3/2}$ ($\text{eV}^{-3/2}$) [10, 13-15]	ν^0 (eV^{-3}) [6]	βF_{at} (kJ/mole) [26]
Sc	3.25	1.27	6.1	1200	6.6
Ti	3.65	1.47	4.8	2050	11
V	4.25	1.64	4.1	2600	14
Cr	4.65	1.73	3.7	2400	14
Mn	4.65	1.61	3.8	1600	4.4
Fe	4.93	1.77	3.7	2550	12
Co	5.10	1.75	3.5	2550	13
Ni	5.20	1.75	3.5	2450	12
Y	3.20	1.21	7.3	1100	7.2
Zr	3.80	1.39	2.8	1950	12
Nb	4.00	1.62	4.9	2700	18
Mo	4.65	1.77	4.4	2950	26
Tc	3.30	1.81	4.2	3050	26
Ru	3.40	1.83	4.1	3050	26
Rh	3.40	1.76	4.1	2750	23
Pt	3.45	1.87	4.3	3100	16
La	3.05	1.09	6.0	900	5.5
Hf	3.55	1.43	5.6	2300	15
Ta	4.05	1.63	4.9	3050	22
W	4.80	1.81	4.5	3300	31
Re	3.40	1.86	4.3	3650	33
Cu	3.40	1.85	4.2	3500	35
In	3.55	1.83	4.2	3100	25
Pb	3.65	1.78	4.4	2550	18
Th	3.30	1.28	7.3 ^a		
U	4.05	1.36	5.6 ^a		
Pa	3.80	1.44	5.2 ^a		
Ce	4.55	1.47	3.7	1850	9.3
Al	4.45	1.39	4.7	1250	10
Au	5.15	1.57	4.7	1550	18

^aRoom temperature allotropic.

$$\Delta H = f(c) [-Pe(\Delta\phi^*)^2 + Q(\Delta n_{ws})^2]$$

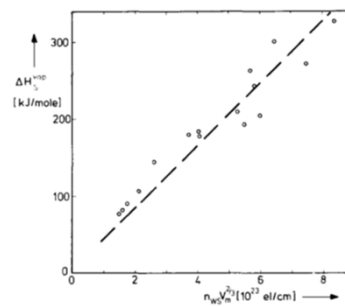
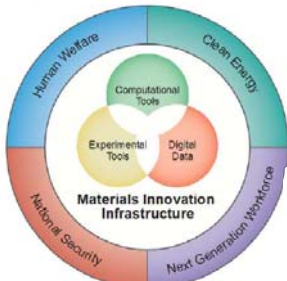



Fig. 4. The linear relation, implied by figs. 2 and 3, between the heat of vaporization per unit atomic surface, $\Delta H_{\text{vap}}^0/V_m^{2/3}$, and the electron density, $n_{ws} V_m^{2/3}$, for non-transition metals. In this figure the divalent metals with their unusually low heat of vaporization have been omitted.

$$\Delta H_f \propto \text{electron density}$$

-Materials Genome Initiative-

*To help businesses discover, develop, and deploy new materials **twice as fast**, we're launching what we call the **Materials Genome Initiative**. The invention of silicon circuits and lithium ion batteries made computers and iPods and iPads possible, but it took years to get those technologies from the drawing board to the market place. We can do it faster.*

-President Obama, Carnegie Mellon University, June 2011

Materials Informatics (Data-Driven Materials Research)

① Creating of DB

- Individual Experimental Data
- Individual Computational Data
- Bibliographic DB
- MatNavi

② Collection and Integration of data

Integrated DB

annotation

③ Data-mining, Machine Learning, Modeling

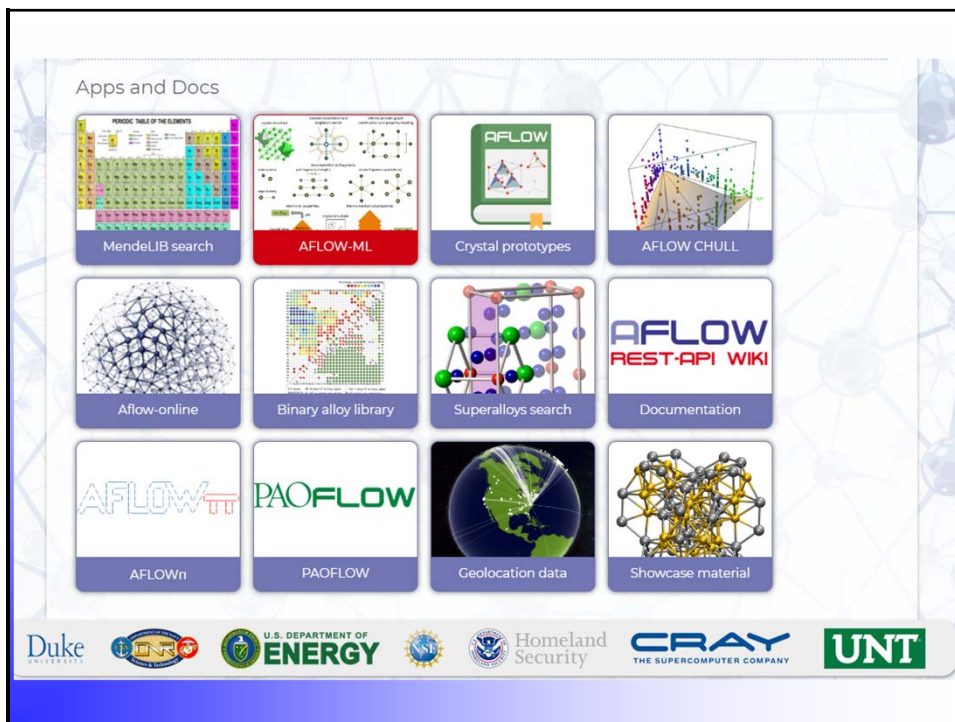
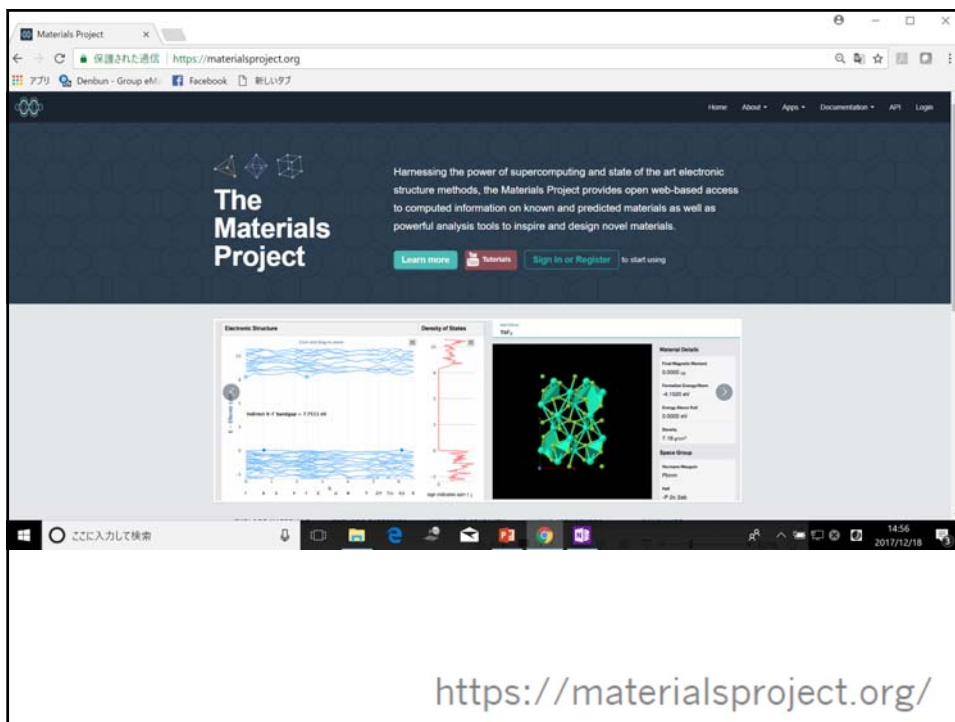
- Rule-based system
- Discripiter

production system

- Selection of candidate
- Visualization of correlations

- Verification by experiment
- High-throughput **new materials' discovery**
- Visualization, Validation
- Understanding by theory
- Understanding rules between structures and functions

26
From the JST-CRDS



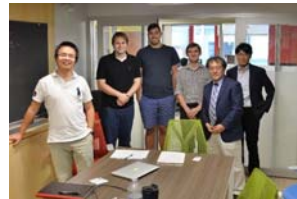
Harvard University :USA

秘 CONFIDENTIAL

Clean Energy Project: Prof. Alán Aspuru-Guzik

The Clean Energy Project (CEP) is a virtual high-throughput discovery and design effort for the next generation of plastic solar cell materials.

- Organic device materials
- Calculation data



The NOMAD Laboratory A European Centre of Excellence

CSC is part of two recently funded European Union's Horizon 2020 projects. The NoMad (Novel Materials Discovery) project will develop "big-data analytics" for materials science. CSC's role in the project is to support the researchers in HPC code optimization and utilization of PRACE systems.

National project on materials informatics
 “Materials Research by Information Integration” Initiative (MI²I)



Two approaches to material and device development



Conventional approach

Based on the fundamental physics, e.g. DFT, we investigate material properties

$$\left[-\frac{1}{2} \nabla^2 + w(\vec{r}) \right] \Phi(\vec{r}) = E\Phi(\vec{r})$$

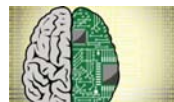


Direct problem scheme

Social and urban infrastructures are based on materials science




Inverse problem scheme



By getting help of the data science, we explore materials having a desirable properties.

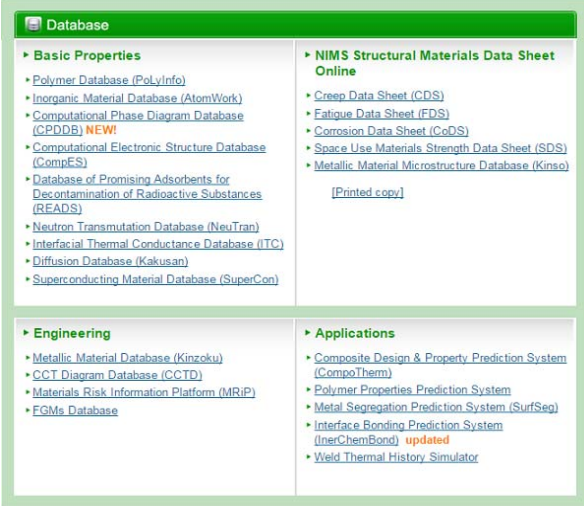
Materials Informatics

National project on materials informatics
 “Materials Research by Information Integration” Initiative (MI²I)



“MatNavi” is one of the world's largest materials databases provided by NIMS


See the web site,
http://mits.nims.go.jp/index_en.html



“MatNavi” consists of ~20 database (polymer, inorganic materials, superconductivity, etc.) with high reliability.

“MatNavi” provides data visualization tools and simple prediction simulator of material properties.


National project on materials informatics
 “Materials Research by Information Integration” Initiative (MI²I)

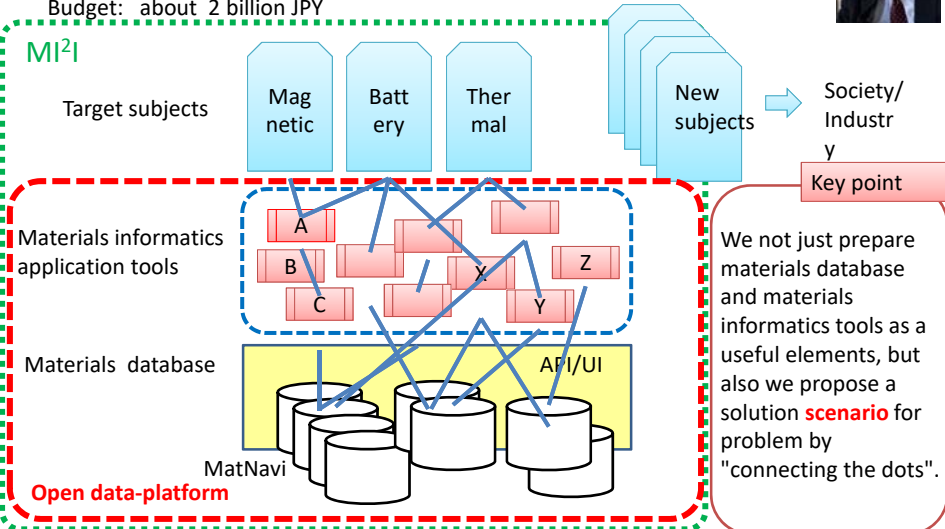


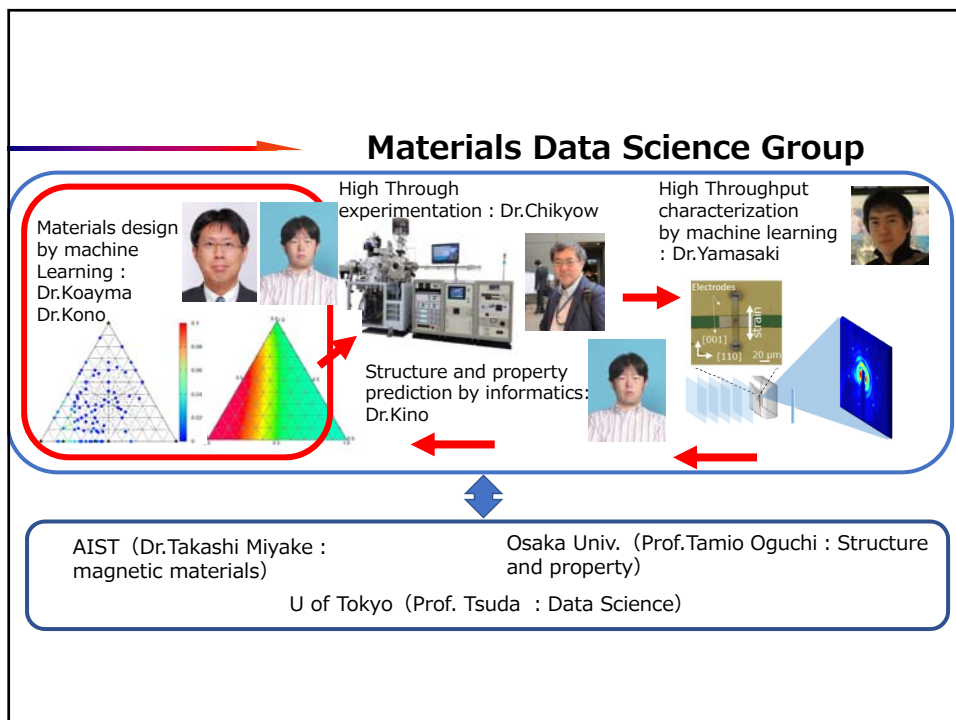
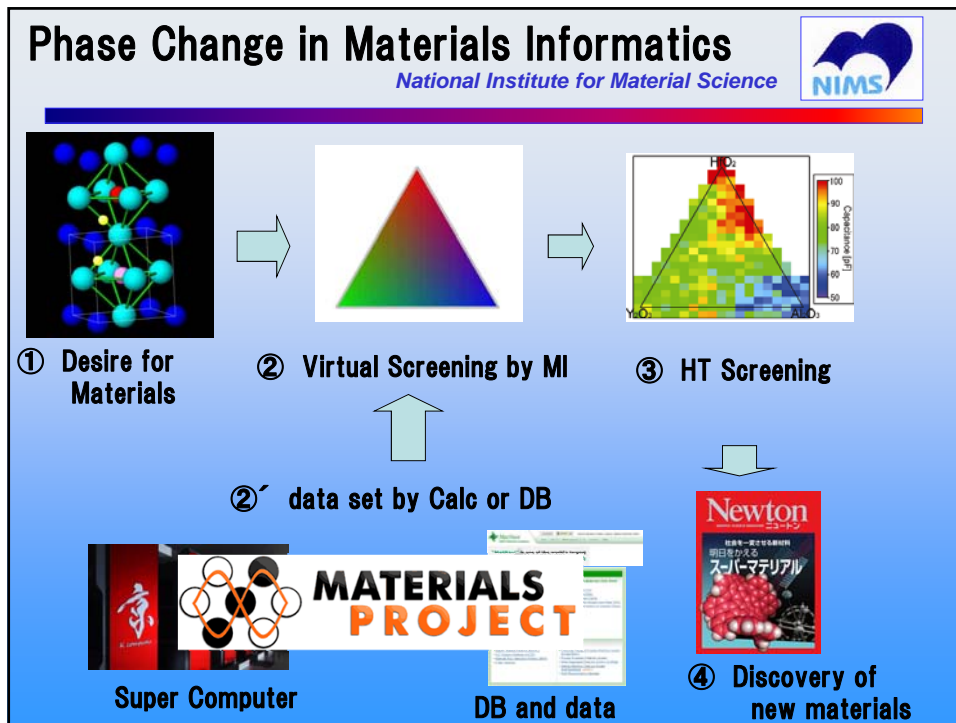
Start : July 1, 2015 Project leader: Satoshi ITOH (NIMS)

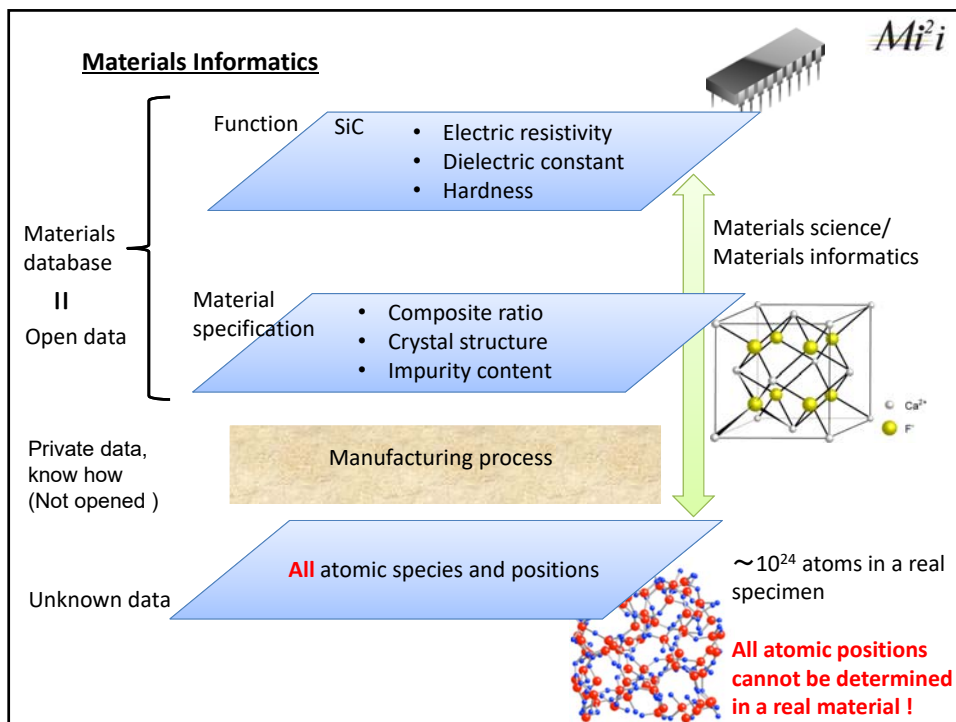
Period: 5 years

Budget: about 2 billion JPY



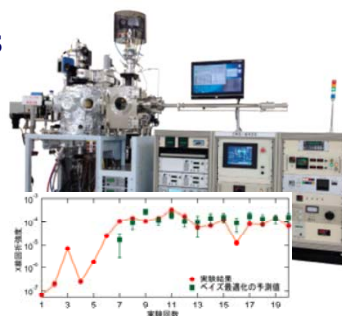
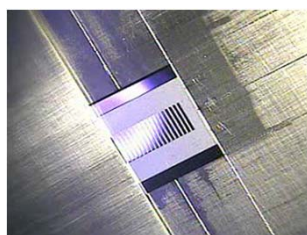
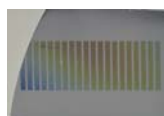




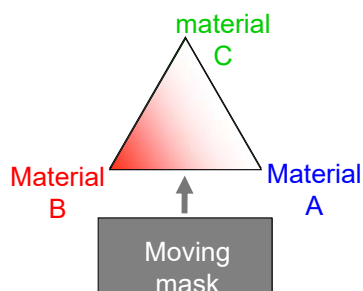
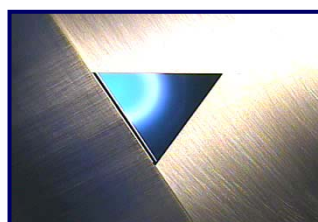
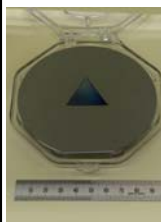


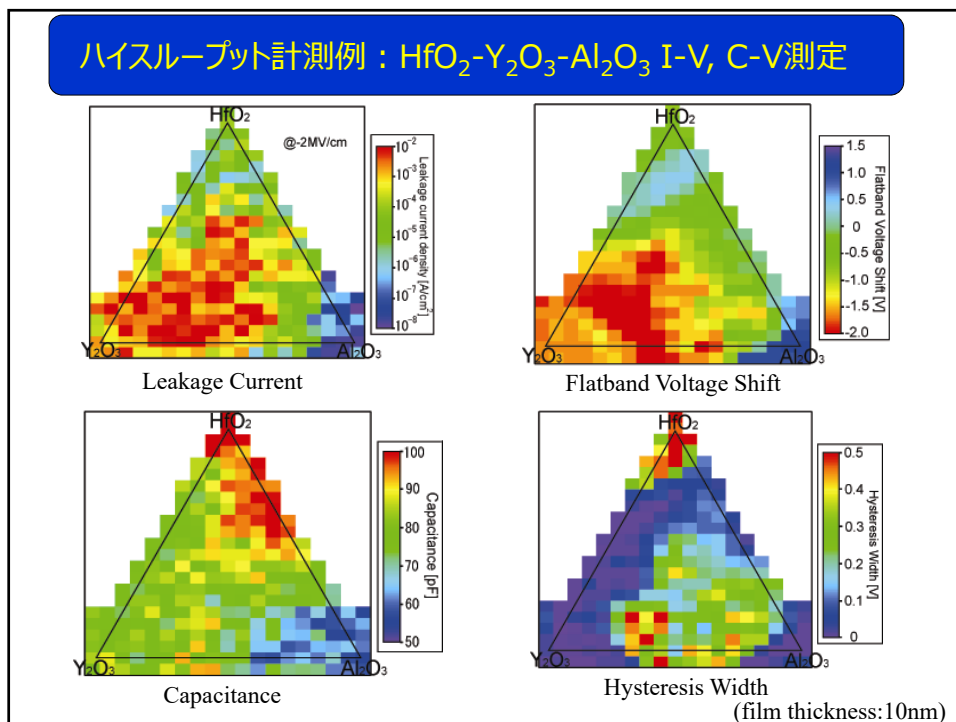
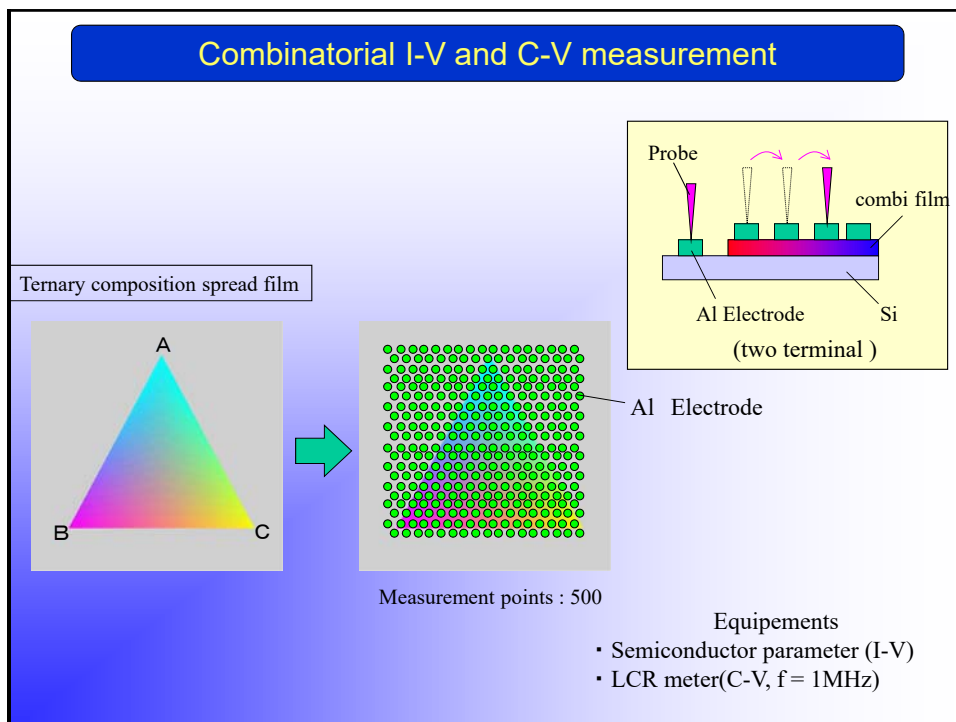
③ High Throughput Experimentation


Binary combinatorial synthesis



Ternary combinatorial synthesis







National project on materials informatics
“Materials Research by Information Integration” Initiative (MI²I)

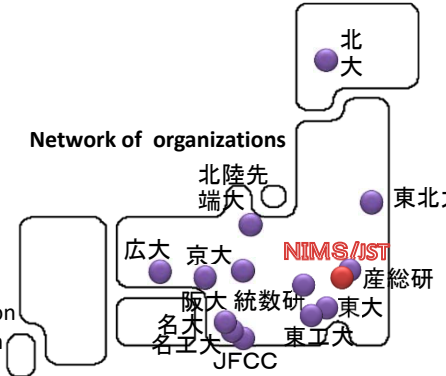
Construction of “all-Japan” structure , “connecting the dots”

“all-Japan” by inter-regional cooperation

- NIMS/JST as hub organization
- Academia network・・・Hokkaido Univ., Tohoku Univ., Univ.Tokyo, TUT, JAIST, Nagoya Inst.Tech., Kyoto Univ., Osaka Univ., Hiroshima Univ., AIST, IMS, JFCC
- Translational research satellites

“all-Japan” by inter-sector cooperation

- Industry-government-academia cooperation
- A new scheme of multi-client collaboration




Network of organizations

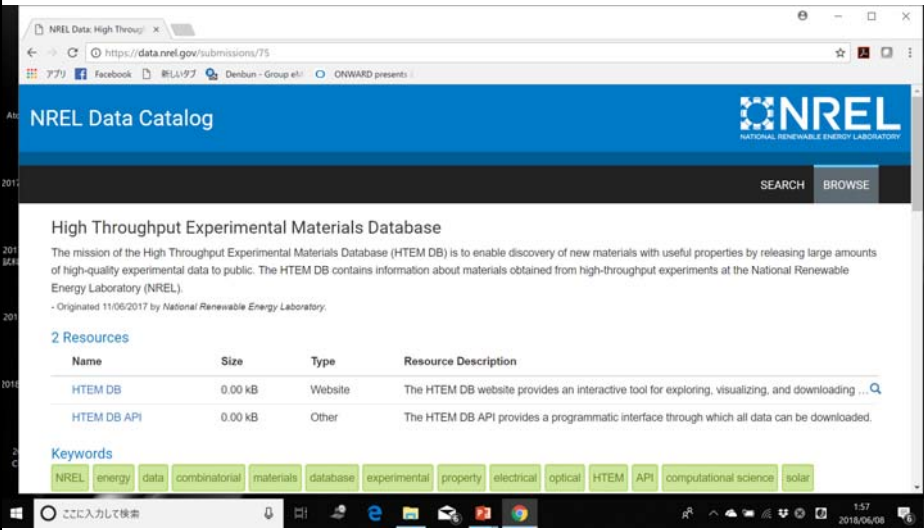
Government/ participation of Municipalities

MI²I consortium (4/1/2016 ~)

- Collaboration in equal partnership of participants
- Sharing of information and achievements as possible



NREL : Open experimental data



NREL Data Catalog

High Throughput Experimental Materials Database

The mission of the High Throughput Experimental Materials Database (HTEM DB) is to enable discovery of new materials with useful properties by releasing large amounts of high-quality experimental data to public. The HTEM DB contains information about materials obtained from high-throughput experiments at the National Renewable Energy Laboratory (NREL).

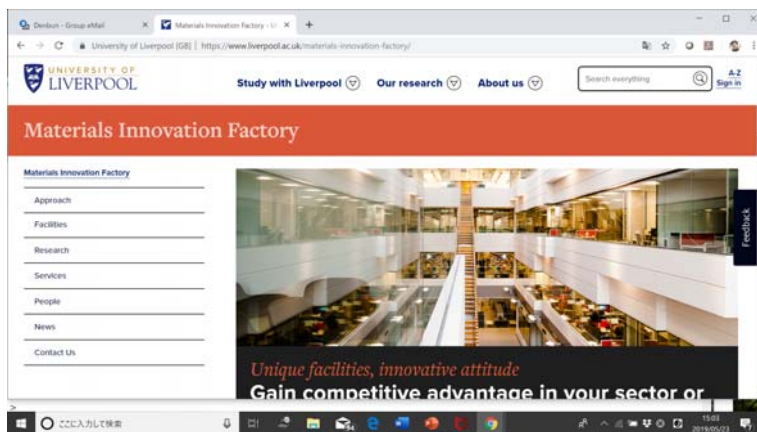
2 Resources

Name	Size	Type	Resource Description
HTEM DB	0.00 kB	Website	The HTEM DB website provides an interactive tool for exploring, visualizing, and downloading ...
HTEM DB API	0.00 kB	Other	The HTEM DB API provides a programmatic interface through which all data can be downloaded.

Keywords: NREL, energy, data, combinatorial, materials, database, experimental, property, electrical, optical, HTEM, API, computational science, solar

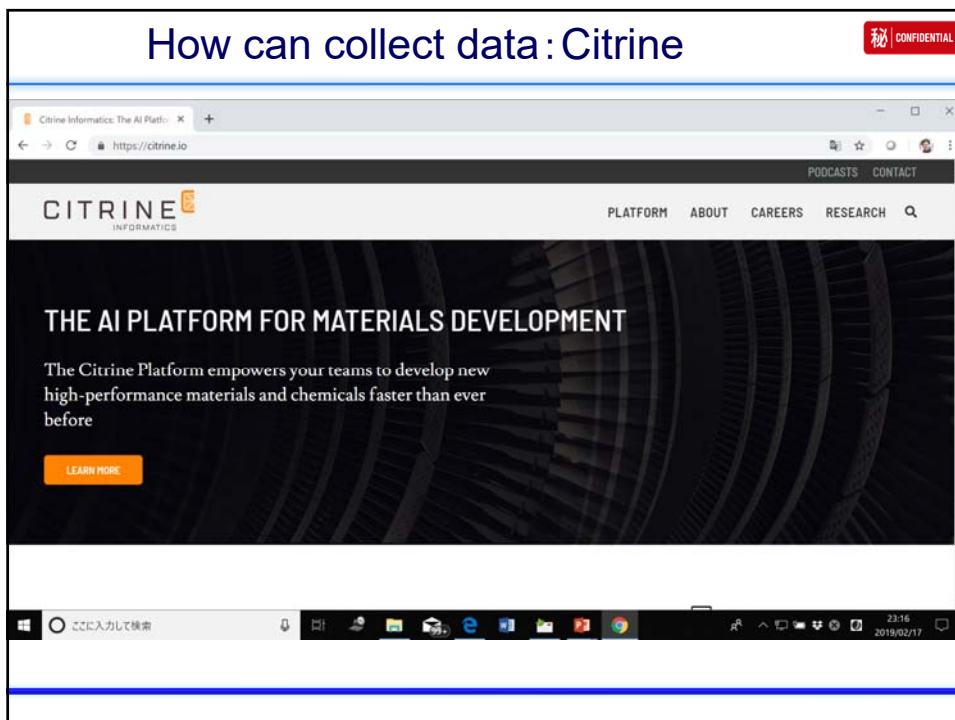
最近のトレンド: Materials Innovation Factory at University of Liverpool

秘 CONFIDENTIAL




How can collect data : Citrine

秘 CONFIDENTIAL



summary

- Data driven materials science will be the major trend in materials science
- Fusion of vertical screening by MI and high throughput experimentation will accelerate new materials discovery
- Descriptors give us hints to think different direction
=> from " Black box " to " White box "



MEGA EVOLUTION of MATERIALS

NIMS WEEK 2019

10/28 Mon

11/1 Fri

10/28 Mon
@Tsukuba

NIMS Open House \ 80 Advanced Labs /
Open House: Lab & Facility
<Enterprise / University students>
 Guided Lab Tours / Lectures on Materials Science etc.

30 Wed
@Tokyo

NIMS Award / Academic Symposium
Materials Innovations Driven by AI + DATA!
Winning Lecture Pioneer of data driven materials science Prof. Gerbrand Ceder (UC Berkeley)
 Global material database editor Dr. Pierre Villars (MPDS)

31 Thu
@Tokyo

NIMS Technology Transfer \ 80 Research topics /
Latest Research Exhibition 2019
Invited Lecture Launched Research center with NIMS "HondaJet" Engine development Mr. Sanford Browne (L'Oréal)
 Mr. Yoshihiko Wajima (Honda R&D)

11/1 Fri
@Tokyo

STAM 20th Anniversary Symposium
Open Science and Materials Research
 Evolve into Data journal / Material development using Open Data

Admission Free WEB Registration Required

National Institute for Materials Science (NIMS)