

Materials Design through COMPUTICS: Large-Scale Density-Functional Calculations for Nanomaterials

➤ Computational Nanoscience

- What is Nano? What is Interesting?

- ✓ *Quantum Theory Required. Science and Technology are close to each other*

- ✓ *Carbon nanotube, Graphene, Si nanowire,...
Nanoshapes are decisive*

➤ Computics: Collaboration among Materials Science, Computer Science and Mathematics

- Density-Functional Theory (DFT)

- ✓ *Material is an Interacting Many-Electron System*

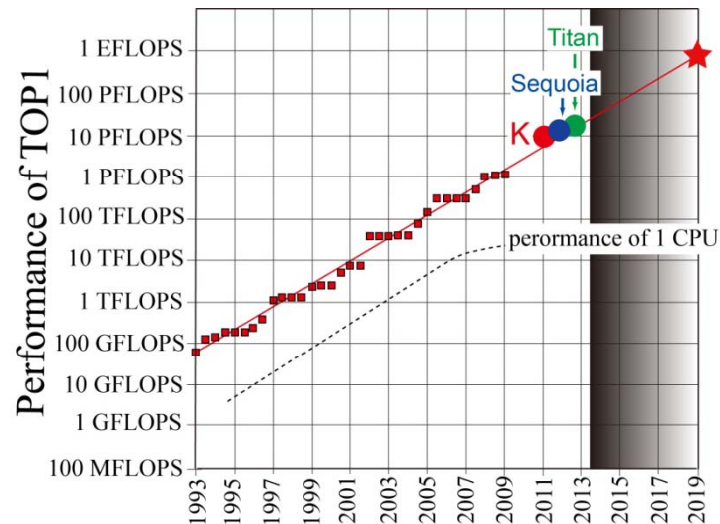
- HPC in DFT Calculations: RSDFT code

- ✓ *Mathematical Methodology, Algorithm*

- ✓ *How to Collaborate*

Computics: Fusion of Materials Science and Computer Science

Computers that are changing rapidly



✓ Saturation of performance of a single computation processor

Breakdown of Moore's Scaling Law

✓ Multi-core massively parallel architecture now

Kobe-K: 80,000 CPU x 8 cores

✓ Accelerator introduced in the current and next generation

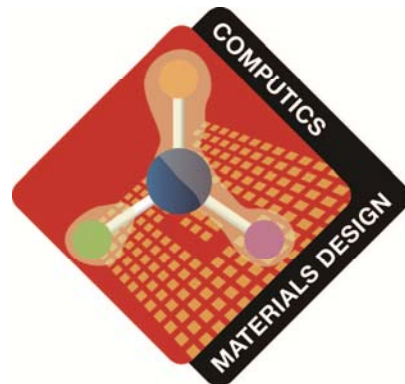
Supercomputer that is a monster

Collaboration between Computational Materials Science and Computer Science is Imperative

- Developments of Mathematical Methodology and Algorithms Suitable to Current and Future Hard wares
- Tuning of High Performance Computation codes based on deep knowledge of hard wares

Computics: Fusion of Materials Science and Computer Science

Computics,
A New Approach to
Materials Science in 21st Century,
in addition to *Mathematics*
that plays an important role from
Newton's Principia



<http://computics-material.jp/>



In collaboration with,

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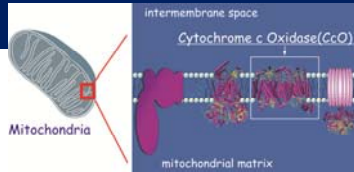


*Miwako Tsuji (Univ Tsukuba)
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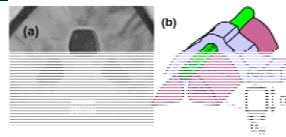
*'''
who are in the field of either
materials science
or computer science.*



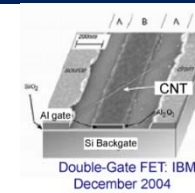
Challenges in Nanoscience



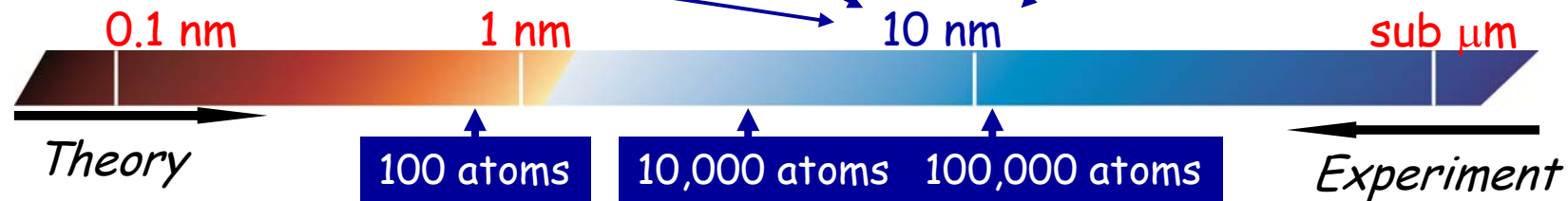
Cytochrome c Oxidase
X-ray determined structure
more than 10,000 atoms involved



Si-Nanowire FET
TokyoIT, IBM,...



FET with CNT
IBM 2004



*Challenges along
3 Directions in Science*

Space

Large scale:
Nano world =
100,000-atom
world

Strongly Correlated System,
Excitation Spectrum

Accuracy

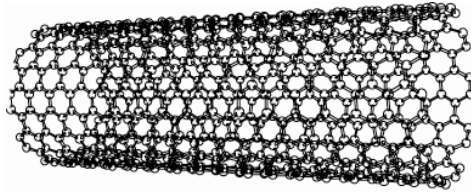
From sub-Femto sec
with electrons to
Nano sec with ions

Time

What is interesting in Nano?

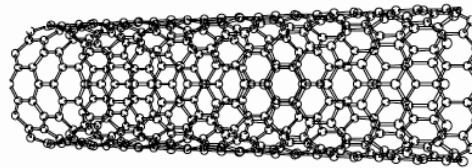
Carbon Nanotube

Metal



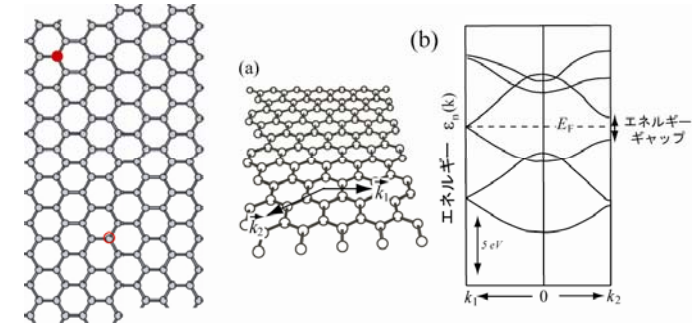
(n,n) tube

Semiconductor

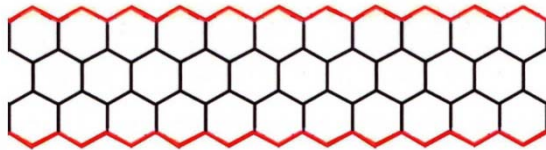


(n,0) tube

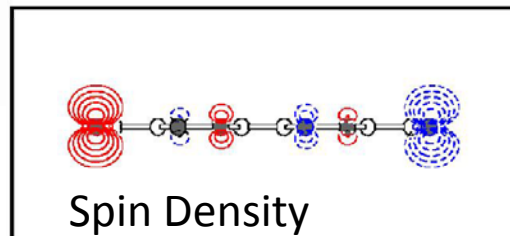
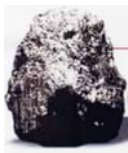
Hamada, Sawada & Oshiyama: Physical Review Letters 68, 1579 (1992)
Woldoer eta l., Nature 391, 59 (1998); Odom et al., Nature 391, 62 (1998)



Edge states in zigzag graphene nanoribbon → Nanomagnet?



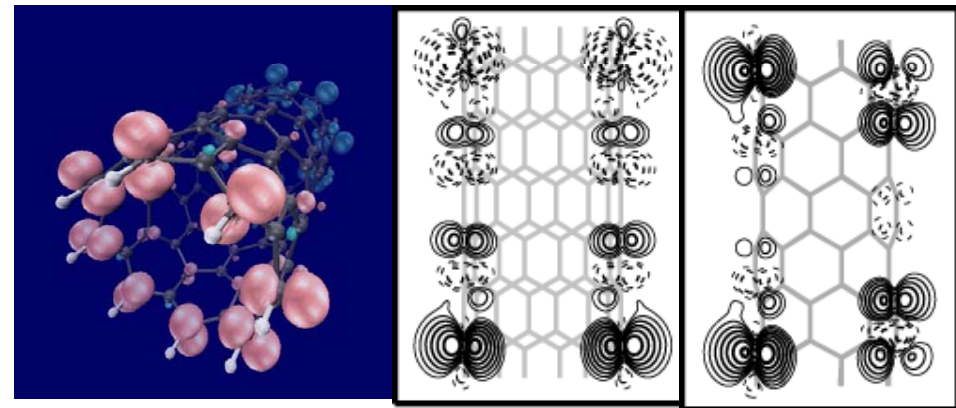
All carbon Magnet?



Up-spin

Down-spin

Electron Spin is polarized

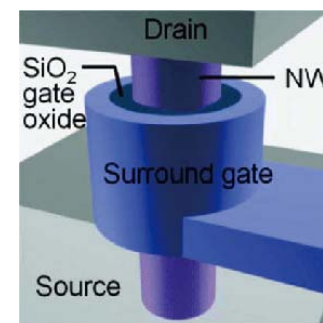
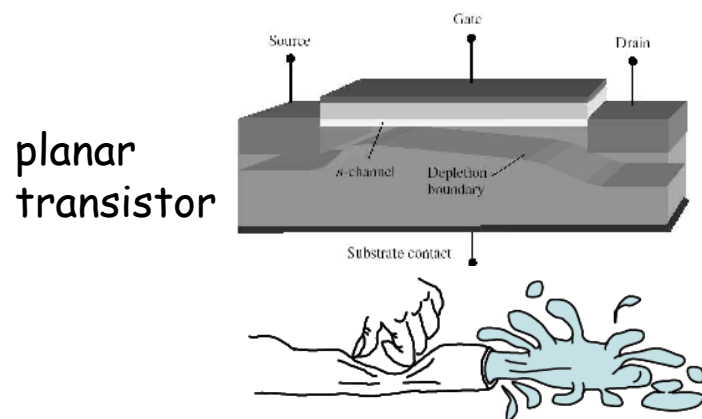


Okada & Oshiyama: Physical Review Letters 87, 146803 (2001);
J. Phys. Soc. Jpn 72, 1510 (2003)

Prediction of Electron States of Si Nanowires

*Si Nanowire, a Booster
in the next-generation semiconductor technology*
More Moore → More than Moore

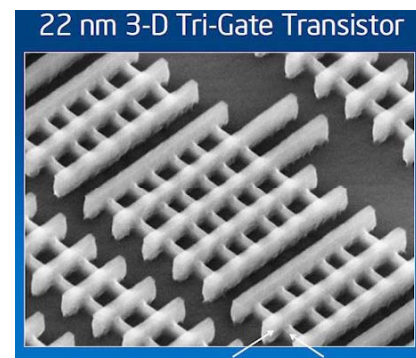
International tech roadmap
for semiconductors,
<http://www.itrs.net/>



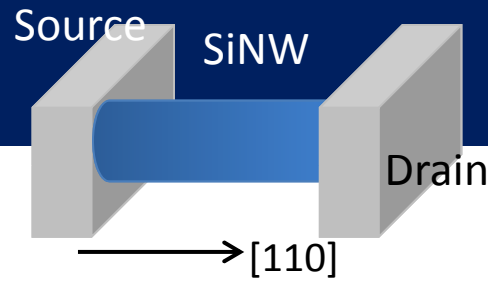
Surrounding gate transistor



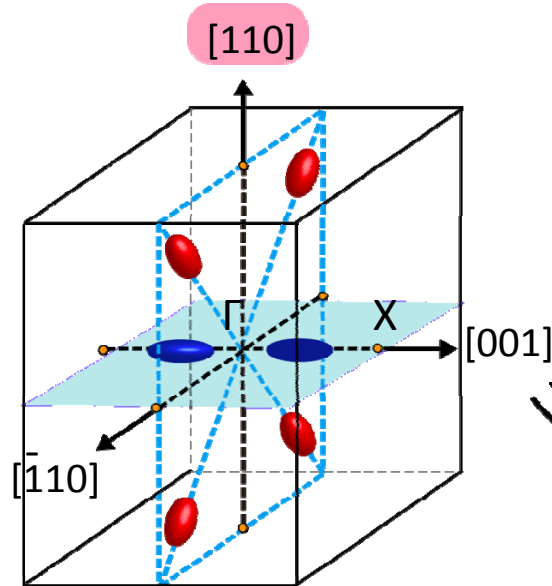
- Gate Controllability
 - Suppress short-channel effects
 - Suppress leaks at off state
 - save energy
- Ballistic Conduction



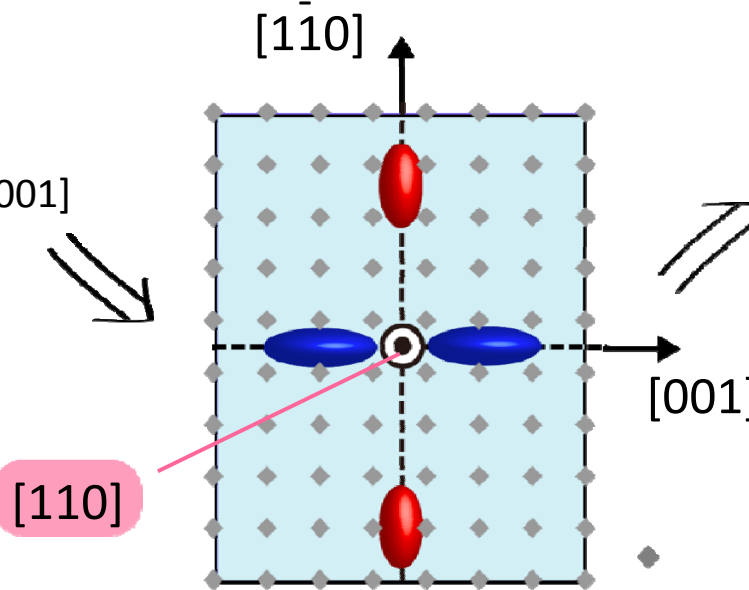
*Actually tri-gate by
Intel in 2011*



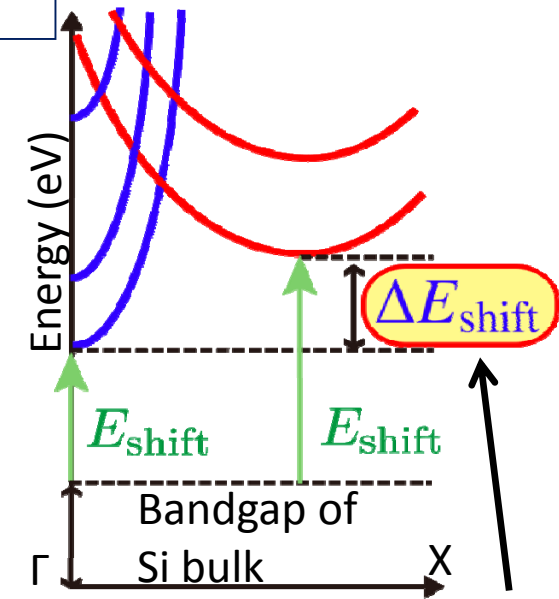
Quantum Confinement in Si (110) Nanowires



Six conduction-band valleys in Brillouin Zone



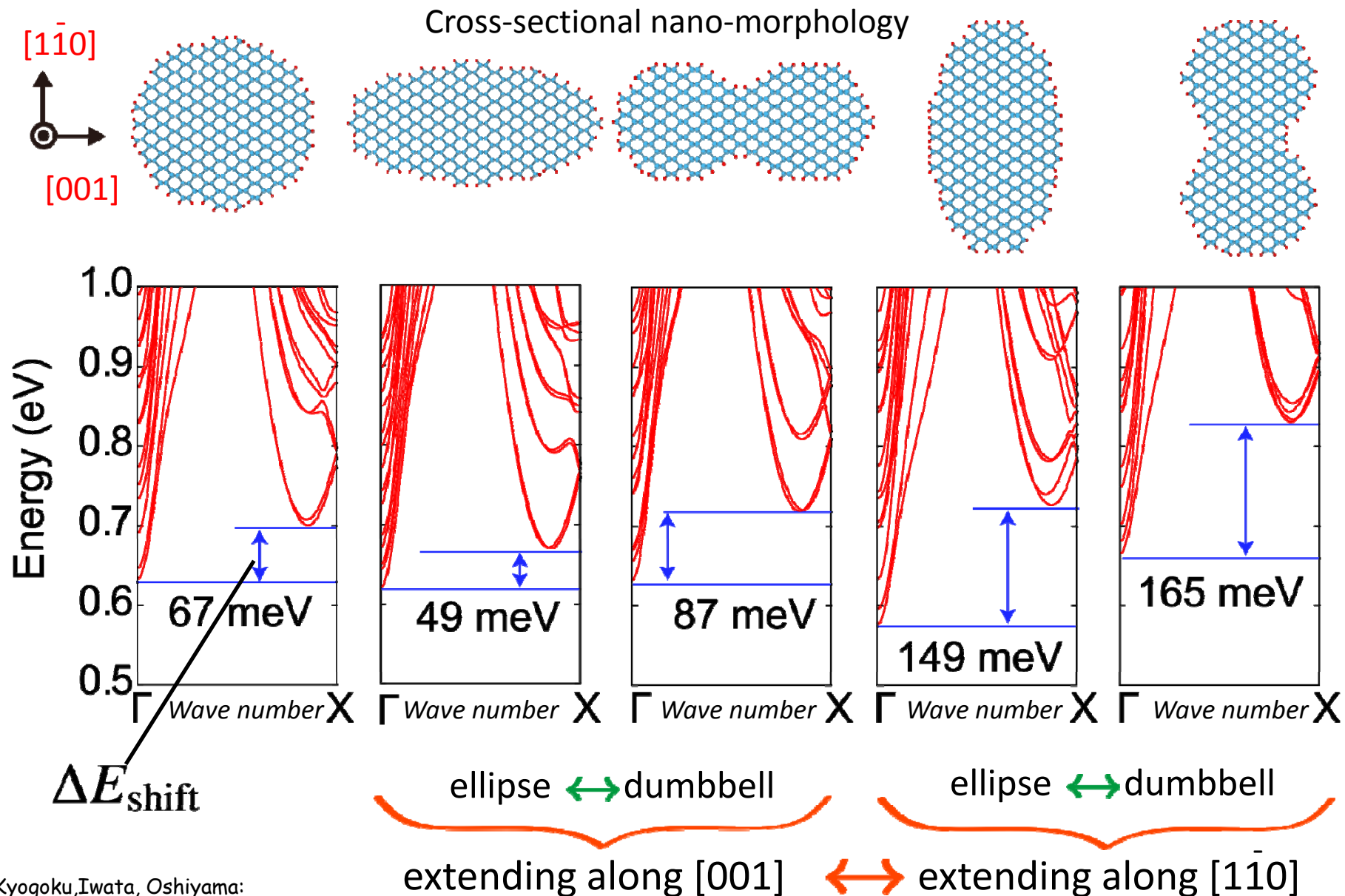
Conduction band structure of [110]-SiNW



This quantity is decisive in the number of the conduction channels.

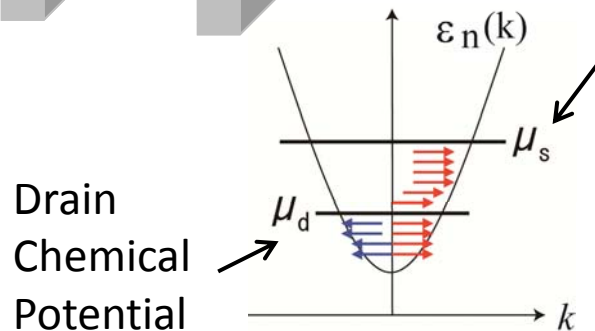
Properties of SiNWs are determined by **the cross-sectional nano-morphology**

Conduction Band Structures of Various Si(110) NWs





Ballistic Transport in Si Nanowires

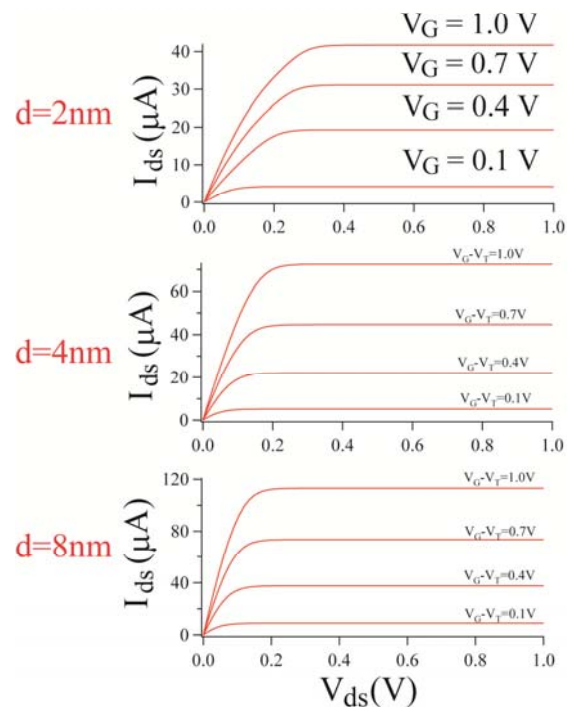


Source Chemical Potential

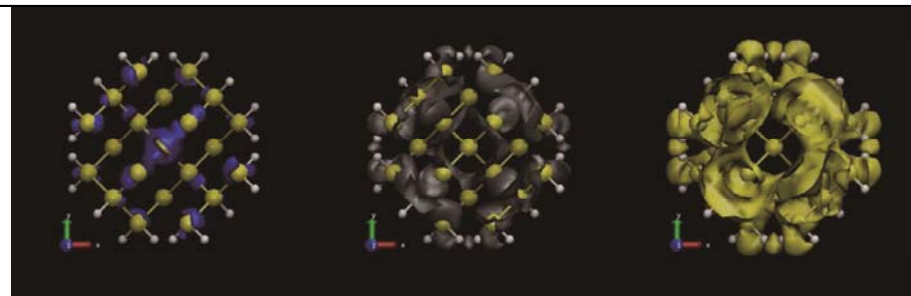
$$I = \frac{q}{\pi \hbar} \sum_n \int d\varepsilon_n [f(\varepsilon_n, \mu_s) - f(\varepsilon_n, \mu_d)]$$

$$j(r) = \frac{q}{im} \sum_n \left\{ \int_{v_k > 0} dk f(\varepsilon_n(k), \mu_s) [\varphi_{nk}^*(r) \nabla \varphi_{nk}(r) - \varphi_{nk}(r) \nabla \varphi_{nk}^*(r)] \right. \\ \left. + \int_{v_k < 0} dk f(\varepsilon_n(k), \mu_D) [\varphi_{nk}^*(r) \nabla \varphi_{nk}(r) - \varphi_{nk}(r) \nabla \varphi_{nk}^*(r)] \right\}$$

Current-Voltage Characteristics Si(110) nanowires



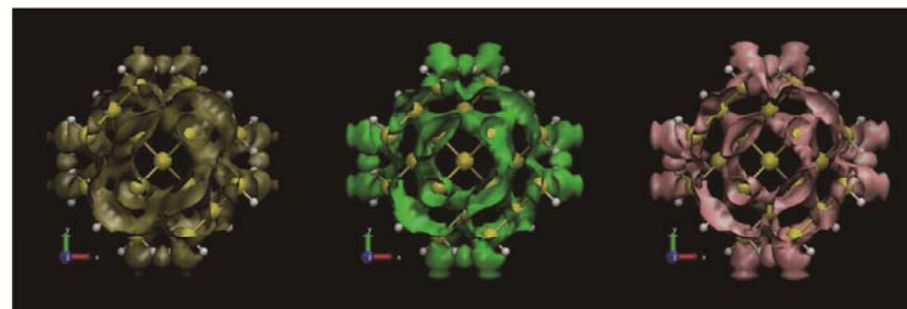
Current Density depending on the gate voltage



$V_G = 0.1V$

$V_G = 0.3V$

$V_G = 0.5V$



$V_G = 0.6V$

$V_G = 0.8V$

$V_G = 1.0V$



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- ✓ *How to Collaborate*

Density Functional Theory (DFT)

Total energy of a material is a functional of the electron density:

$$E[n(\mathbf{r})] = \langle \Psi | \hat{H} | \Psi \rangle = \langle \Psi | \hat{T} + V_{\text{nucI}} + V_{\text{ee}} | \Psi \rangle$$

$$= T_S[n(\mathbf{r})] + \int v_{\text{nucI}}(\mathbf{r})n(\mathbf{r})d\mathbf{r} + \frac{1}{2} \iint \frac{n(\mathbf{r})n(\mathbf{r}')}{|\mathbf{r} - \mathbf{r}'|} d\mathbf{r}d\mathbf{r}' + E_{\text{XC}}[n(\mathbf{r})]$$

*Quantum effects:
exchange and correlation*

When the electron density is written as $n(\mathbf{r}) = \sum_{i: \text{occupied}} |\varphi_i(\mathbf{r})|^2$

By minimizing $E[n]$ with respect to $n(\mathbf{r})$, we obtain Euler equation,

$$\left\{ \begin{array}{l} \left[-\frac{1}{2} \nabla^2 + v_{\text{eff}}(\mathbf{r}; n(\mathbf{r})) \right] \varphi_i(\mathbf{r}) = \varepsilon_i \varphi_i(\mathbf{r}) \\ v_{\text{eff}}(\mathbf{r}) = v_{\text{nucI}}(\mathbf{r}) + \int \frac{n(\mathbf{r}')}{|\mathbf{r} - \mathbf{r}'|} d\mathbf{r}' + \frac{\delta E_{\text{XC}}[n]}{\delta n(\mathbf{r})} \end{array} \right. \quad \text{Kohn-Sham equation}$$

KS equation should be solved self-consistently.
(Most time consuming process)

→ total energy, geometry optimization, physical properties,...

Solving Kohn-Sham Equation

Introducing complete basis set $\chi_n(\mathbf{r})$ to expand Kohn-Sham wave function (orbitals):

$$\varphi_i(\vec{r}) = \sum_{\mu} c_{i\mu} \chi_{\mu}(\vec{r})$$

KS equation is converted to matrix equation:

$$\sum_{\mu'} H_{\mu\mu'}^{KS}(\{c_{i\mu}\}) c_{i\mu'} = \varepsilon_i \sum_{\mu'} S_{\mu\mu'} c_{i\mu'}$$

that is solved usually by iterative techniques such as **conjugate gradient** or **residual minimization**.

The most standard basis set has been the **plane wave basis set**:

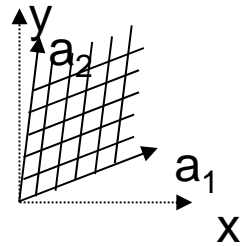
$$\chi_{\mu}(\vec{r}) = e^{i(\vec{k} + \vec{G}_{\mu}) \cdot \vec{r}}$$

However, in the plane-wave scheme, several quantities such as $n(\mathbf{r})$, $v(\mathbf{r})$ etc is Fourier-transformed and Inverse-Fourier-transformed frequently, which becomes a heavy burden on massively parallel-architecture computers.

⇒ Do it all in real space!

Solving Kohn-Sham Equation Using Real-Space Finite-Difference Scheme

Introducing mesh points with spacing H in real space,

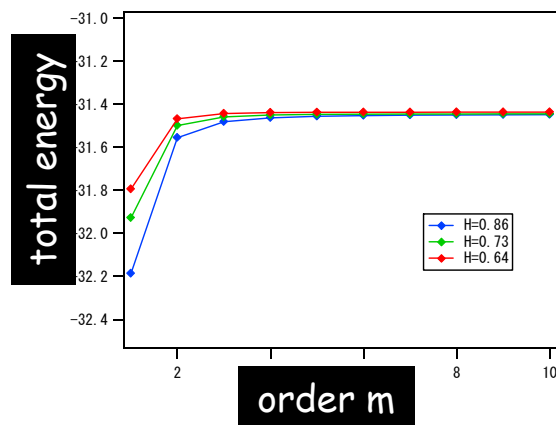


Kohn-Sham differential equation is converted to a M -th order finite-difference equation:

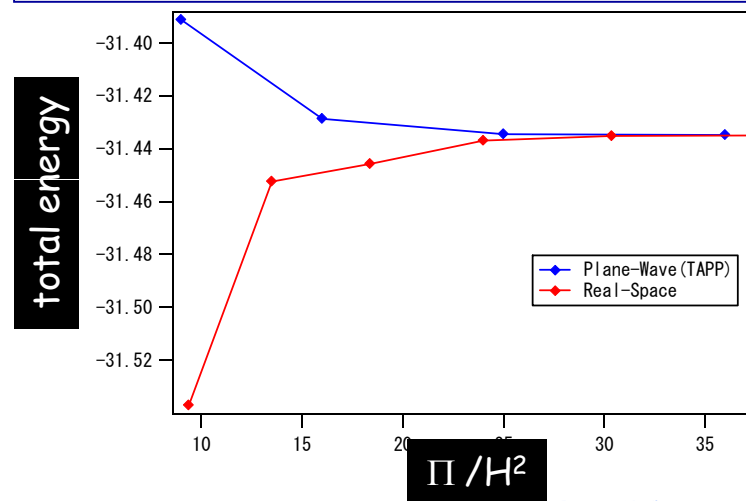
$$\frac{\partial^2}{\partial x^2} \varphi_j(x, y, z) = \sum_{m=-M}^M c_m \varphi_j(x + mH, y, z)$$

All the quantities computed at each mesh point, KS Hamiltonian expressed as a matrix.

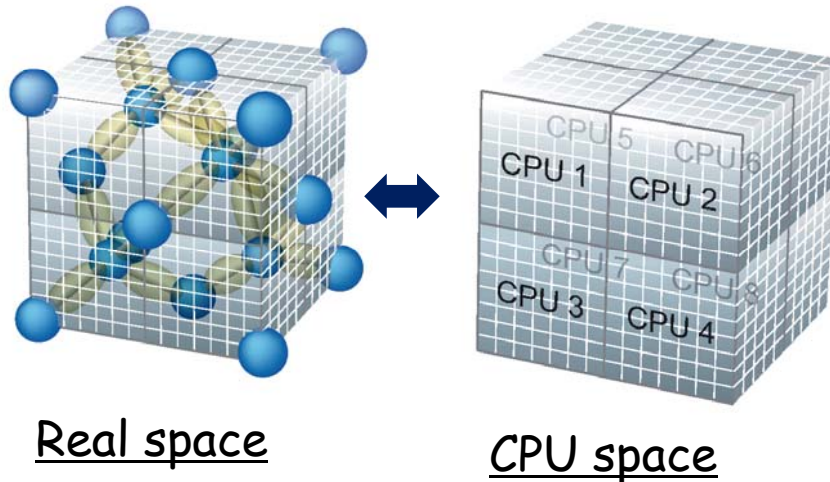
Check of finite difference



Check of mesh spacing
 \Rightarrow systematic accuracy improvement



Advantages of Real-Space DFT (RSDFT) in Parallel Computing



On multi-core parallel machines:

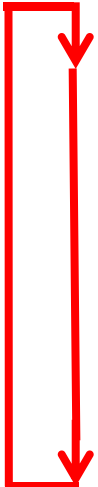
- Huge number of 3D mesh points divided into moderate-size cells
- Each cell treated by a single node or core
- Also, electron states (orbitals) are grouped and each group is treated by a single node or a core : **Hybrid parallelization**
- **MPI for CPU and OpenMP for core**

- Almost free from FFT, reducing communication burden
⇒ high efficiency
- Flexible boundary condition to wave-functions
⇒ targets expanded including charged objects
- Utilize inherent locality of the system
⇒ prospect to efficient $O(N)$ scheme

Solving Kohn-Sham Equation

$$\left[-\frac{1}{2}\nabla^2 + v_{\text{eff}}(\mathbf{r}_k; n(\mathbf{r}_k)) \right] \varphi_i(\mathbf{r}_k) = \varepsilon_i \varphi_i(\mathbf{r}_k)$$

Getting Self-Consistent Field (SCF) by Iterative Computations

- 
- 1 (CG) Conjugate gradient method $O(N^2)$
 - 2 (GS) Ortho-normalization by Gram-Schmidt method $O(N^3)$
 - 3 Density and potential update $O(N)$
 - 4 (SD) Subspace diagonalization $O(N^3)$

GS and SD should be accelerated!

Memory is not the problem so far: $O(N^2)$

107,292-atom SiNW:

$$\varphi_i(r_k) = 576 \times 576 \times 576 \text{ (mesh } k) \times 229,824 \text{ (state } i) \times 8 = 351 \text{ TBite}$$

Gram-Schmidt Ortho-normalization: Active use of Level 3 BLAS in $O(N^3)$ Computation

$$\varphi_1 = \psi_1$$

$$\varphi_2 = \psi_2 - \varphi_1 \langle \varphi_1 | \psi_2 \rangle$$

$$\varphi_3 = \psi_3 - \varphi_1 \langle \varphi_1 | \psi_3 \rangle - \varphi_2 \langle \varphi_2 | \psi_3 \rangle$$

$$\varphi_4 = \psi_4 - \varphi_1 \langle \varphi_1 | \psi_4 \rangle - \varphi_2 \langle \varphi_2 | \psi_4 \rangle - \varphi_3 \langle \varphi_3 | \psi_4 \rangle$$

$$\varphi_5 = \psi_5 - \varphi_1 \langle \varphi_1 | \psi_5 \rangle - \varphi_2 \langle \varphi_2 | \psi_5 \rangle - \varphi_3 \langle \varphi_3 | \psi_5 \rangle - \varphi_4 \langle \varphi_4 | \psi_5 \rangle$$

$$\varphi_6 = \psi_6 - \varphi_1 \langle \varphi_1 | \psi_6 \rangle - \varphi_2 \langle \varphi_2 | \psi_6 \rangle - \varphi_3 \langle \varphi_3 | \psi_6 \rangle - \varphi_4 \langle \varphi_4 | \psi_6 \rangle - \varphi_5 \langle \varphi_5 | \psi_6 \rangle$$

.....

DGEMM on K:

96.6 % efficiency to the peak performance
by making the necessary data kept on L1D
and L2 cache

Most calculations can be performed as
Matrix \times Matrix operations!

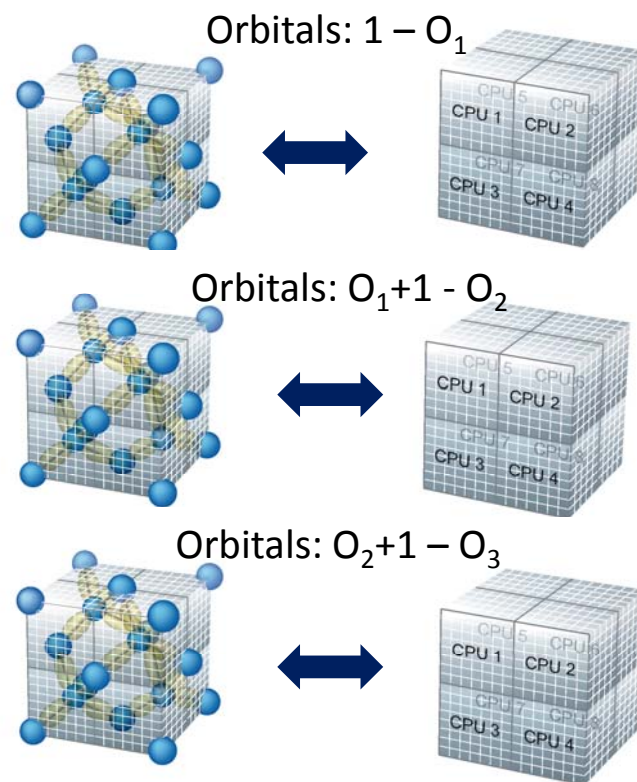
$$[\varphi_3, \varphi_4, \varphi_5, \varphi_6, \dots] = [\psi_3, \psi_4, \psi_5, \psi_6, \dots] - [\varphi_1, \varphi_2] \begin{bmatrix} \dagger \varphi_1 \\ \dagger \varphi_2 \end{bmatrix} [\psi_3, \psi_4, \psi_5, \psi_6, \dots]$$

Real Space Parallelization + Orbital Parallelization

Kohn-Sham equation

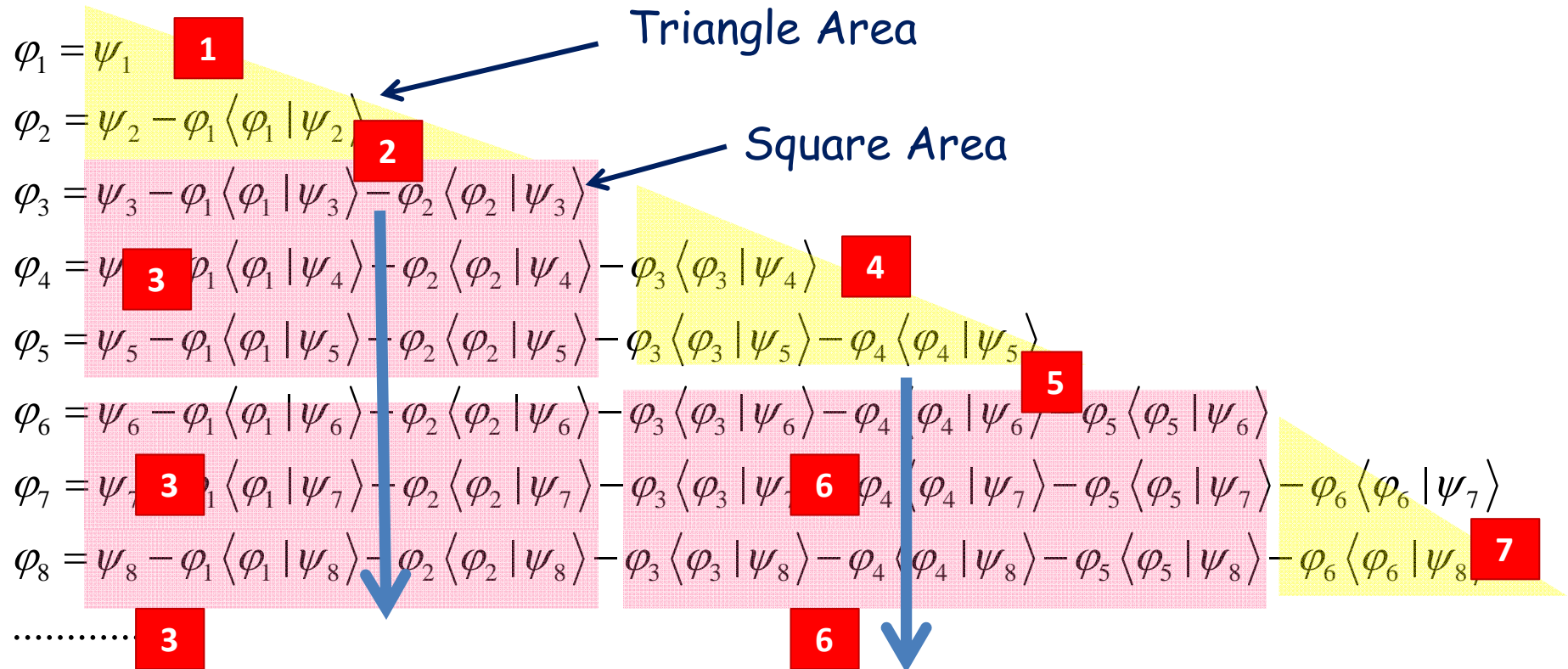
$$\left[-\frac{1}{2} \nabla^2 + v_{\text{eff}}(\mathbf{r}_k; n(\mathbf{r}_k)) \right] \varphi_i(\mathbf{r}_k) = \varepsilon_i \varphi_i(\mathbf{r}_k)$$

Real Space parallel $\{k\}$ + *Orbital Parallel* $\{i\}$



- Reduce communication cost
- Guarantee high performance in massively parallel computations

Gram-Schmidt Ortho-normalization: Orbital Parallel Coding of Matrix Operations

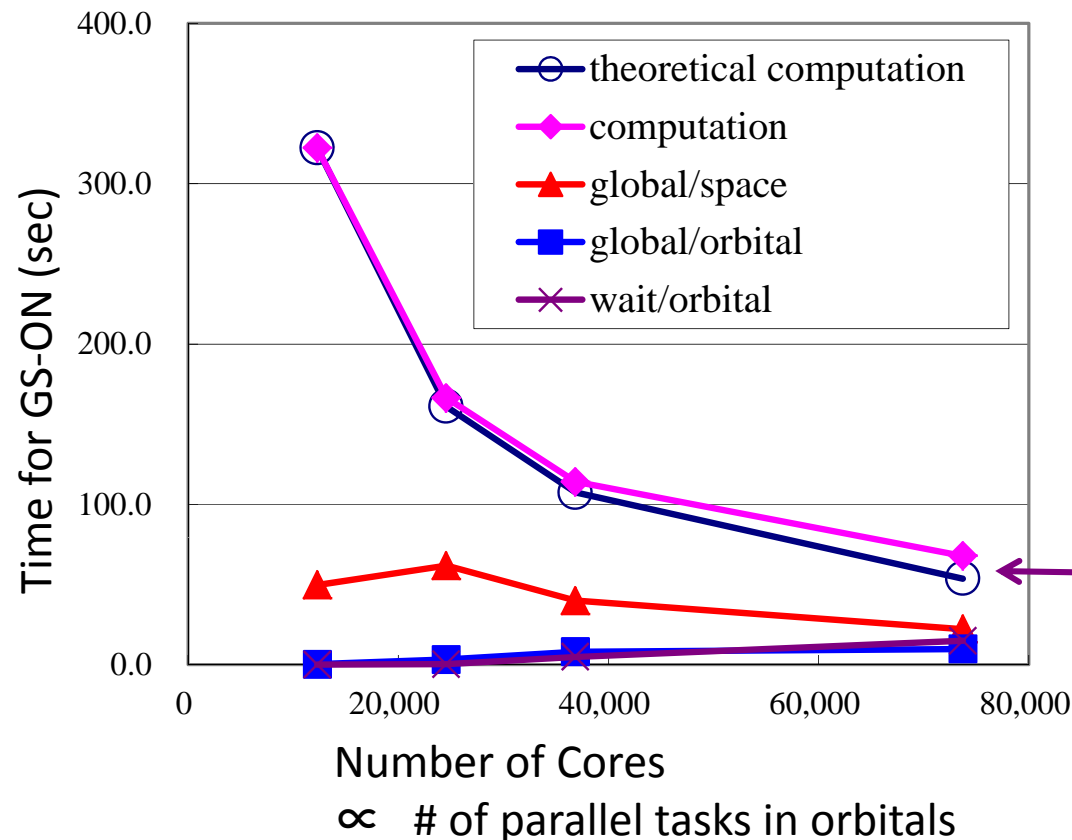


1. Calculate the triangle area
2. Transfer the calculated values to the square area
3. Calculate the square area in parallel

Efficiency of Gram-Schmidt Ortho-normalization

62 % (55,296 nodes of K) - 80% (1000 CPU of PACS-CS)
efficiency to the peak performance

Strong Scaling in Orbital Parallelization for GS ON



Si Nanowire 19,848 atoms
[Grids:320x320x120, Orbitals:40,992]

Number of parallel tasks
- 1,536 (fixed) in space
- 1, 2, 3, 6, in orbitals

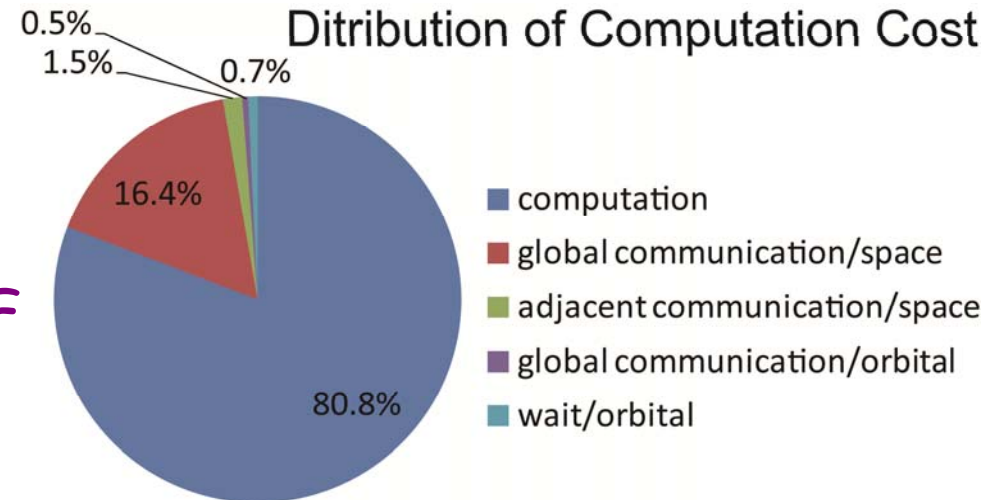
Ideal efficiency

RSDFT on K at Kobe



- ✓ 107,292-atom DFT calculations for Si Nanowires
using 55, 296 nodes

- 3.08PFLOPS = 43.6% efficiency to peak performance
- Single iteration in SCF = 5,500 sec



- ✓ 10,000-atom DFT calculations

- Getting SCFields costs one day using several hundreds of nodes (less than 1 % of the total resource) with 34 % - 70 % efficiency
- Good strong scaling (78 % - 88 %) for 4,608 ~ 15,552-atom systems using 768 ~ 12,288 nodes



Computation cost and Memory: Achieved values

➤ Memory: order N^2

- 107,292-atom SiNW

$$\varphi_i(r_k) = 576 \times 576 \times 576 \text{ (mesh } k) \times 229,824 \text{ (state } i) \times 8 = 351 \text{ TBite}$$

- SiNW with 10,000 atoms

$$\varphi_i(r_k) \approx 3.5 \text{ TBite}$$

➤ Computation Cost: order N^3

- 107,292-atom SiNW with 55,296 nodes (442,368 cores)

- ✓ Solve KS equation with given $n(r) = 5,500 \text{ sec}$
- ✓ Get SCF potential = 5 days, geometry optimization = 50 days
- ✓ Several hundreds PFLOPS (of course exa) machine makes it

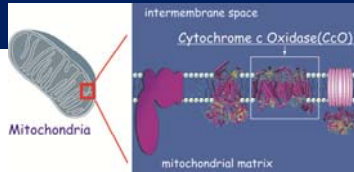
- SiNW with 10,000 atoms with 2000 - 3000 nodes

- ✓ Get SCF potential < 1 day, geometry optimization = 1 week

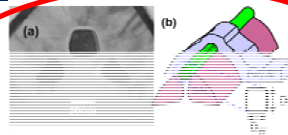


- HPC is the issue: perform $O(N^3)$ computations efficiently with some accelerators (GPU?).

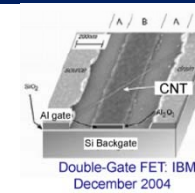
Challenges in Nanoscience



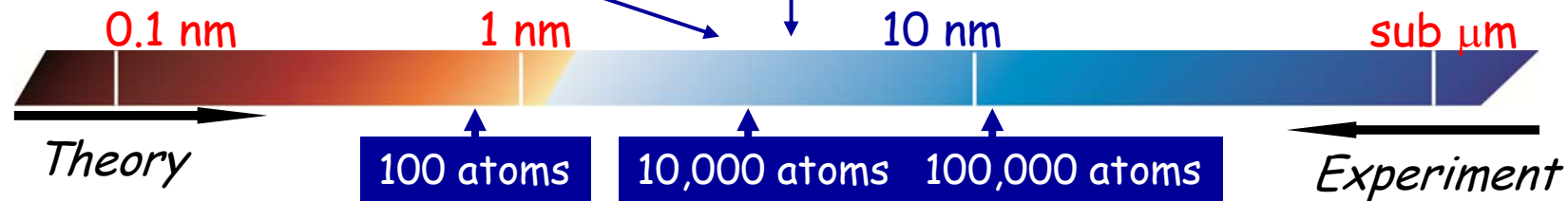
Cytochrome c Oxidase
X-ray determined structure
more than 10,000 atoms involved



Si-Nanowire FET
TokyoIT, IBM,...



FET with CNT
IBM 2004



*Challenges along
3 Directions in Science*

Space

Larger scale:
Nano world =
100,000-atom
world

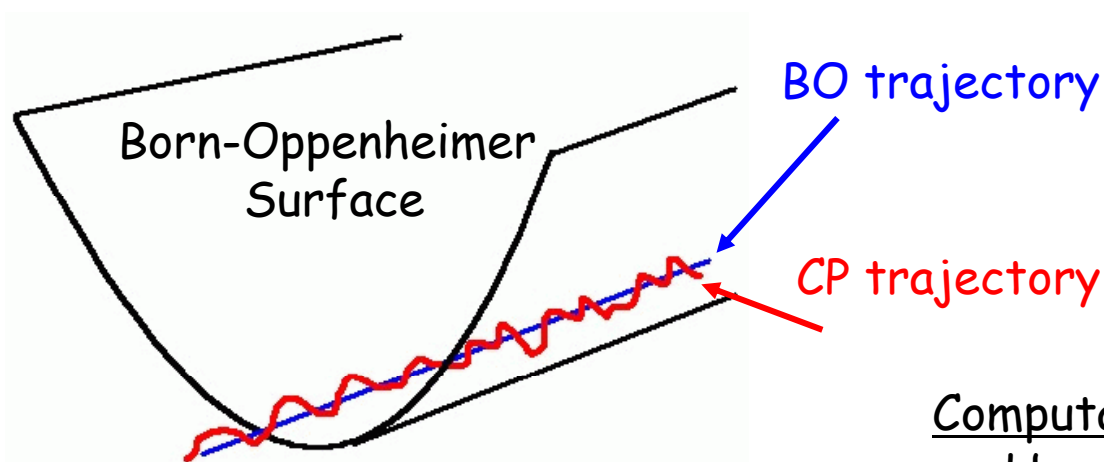
Strongly Correlated System,
Excitation Spectrum

Accuracy

From sub-Femto
sec with
electrons to Nano
sec with ions

Time

Car-Parrinello Molecular Dynamics



Computation Cost

Hamiltonian operation $\propto N^2$

Force Calculation $\propto N^2$

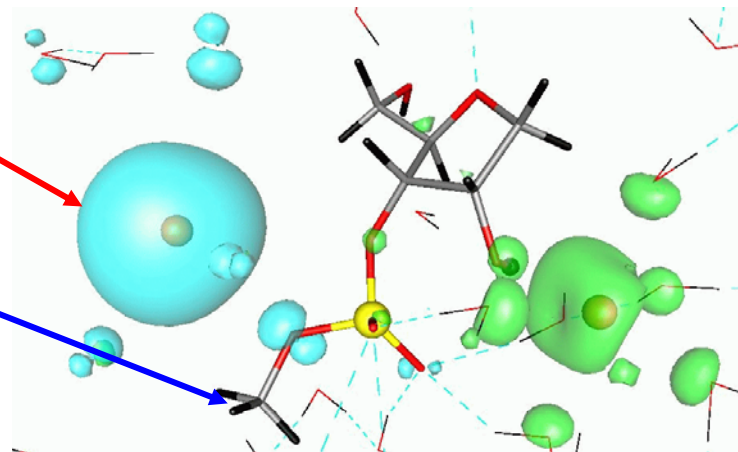
Lagrange Multiplier $\propto N^3$

$$\mu \ddot{\psi}_i = - \frac{\delta E^{DFT}}{\delta \psi_i^*} + \sum_j \Lambda_{ij} \psi_j$$

$$M_I \ddot{\mathbf{R}}_I = -\nabla_{\mathbf{R}_I} E^{DFT}$$

$$\mu_q \ddot{\alpha}_q = - \frac{\partial E^{DFT}}{\partial \alpha_q}$$

External Parameters: T, P, \dots



Car-Parrinello MD + MetaDynamics

Collective (reaction) coordinates: s_α

$$L = L_{CP} + \sum_{\alpha} \frac{1}{2} M_{\alpha} \dot{s}_{\alpha}^2 - \sum_{\alpha} \frac{1}{2} k_{\alpha} (s_{\alpha}(R) - s_{\alpha})^2 - V(s, t)$$

usual CP Lagrangean

fictitious kinetic energy

Restrain potential:

coupling with other coordinates

Gaussian potentials added

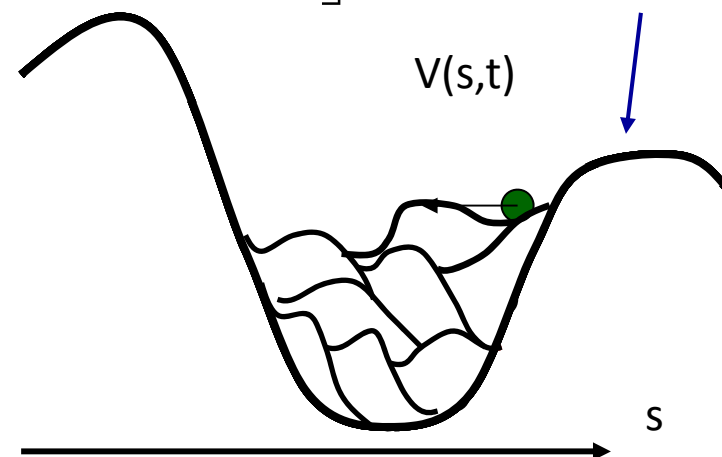
where the history dependent function V is

$$V(\mathbf{s}, t) = \sum_{t_i < t} W_i \exp\left(-\frac{1}{2} \frac{(\mathbf{s} - \mathbf{s}_i)^2}{(\Delta s^{\perp})^2}\right) \exp\left[-\frac{1}{2} \frac{((\mathbf{s}_{i+1} - \mathbf{s}_i)(\mathbf{s} - \mathbf{s}_i))^2}{(\Delta s_i^{\parallel})^4}\right]$$

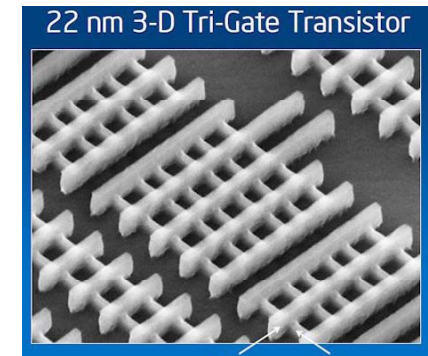
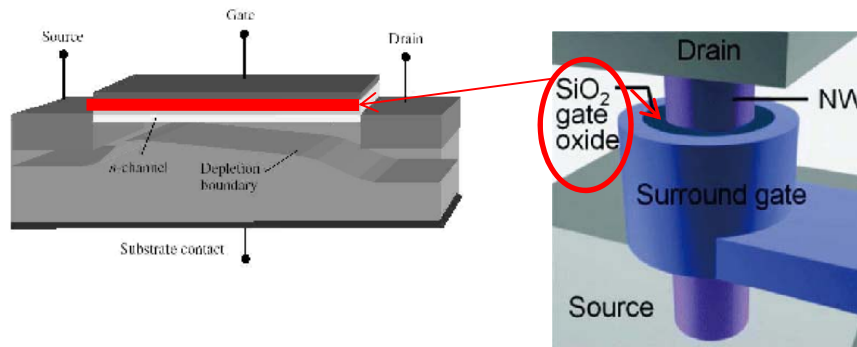
Free-energy landscape that we want to get

and sum of gaussians gives us Free-energy landscape

$$\lim_{t \rightarrow \infty} V(\mathbf{s}, t) = -F(\mathbf{s}) + \text{const.}$$



Silicon Oxidation: Essential Process in Device Fabrication

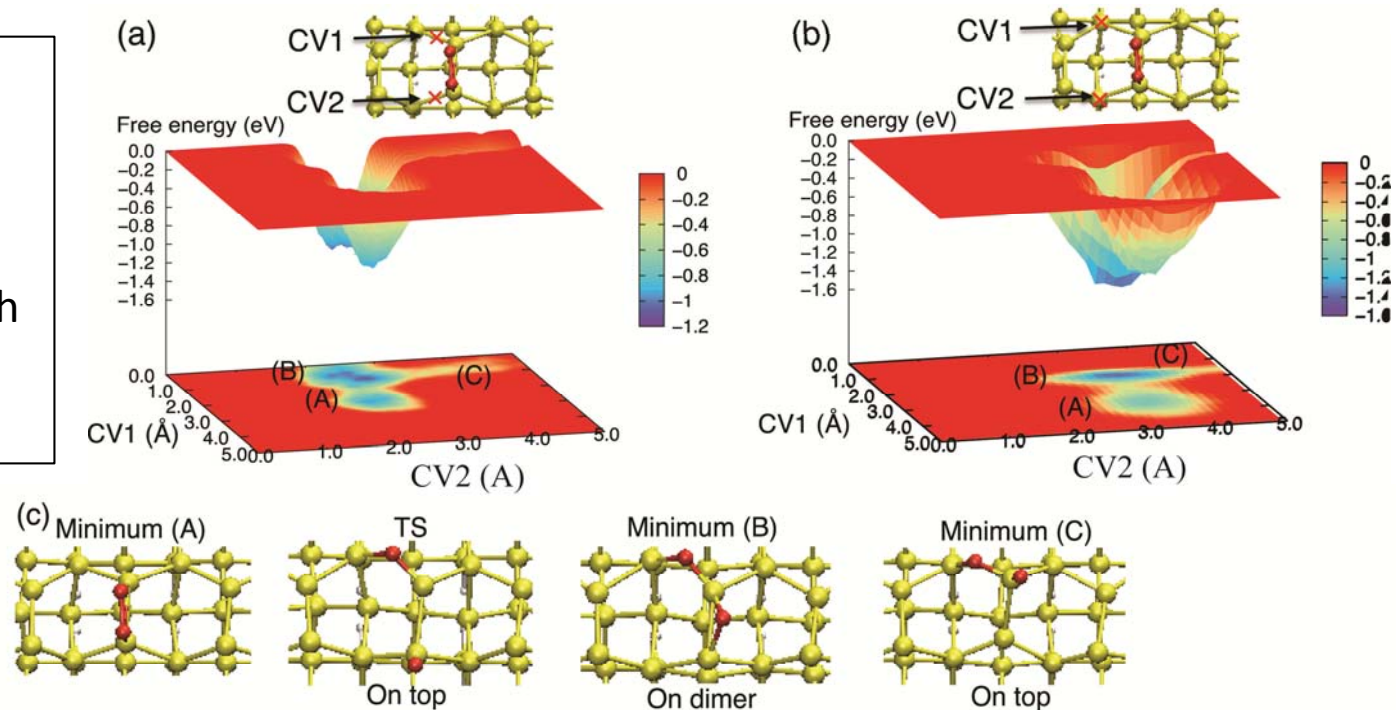


- ✓ Good Insulators for Gate Functions are Provided by Silicon Oxidation
 - ✓ Film Thickness is about 1 nm or less
 - ✓ New materials which have high dielectric constants are incorporated
- Atom-scale Understanding of Oxidation Processes

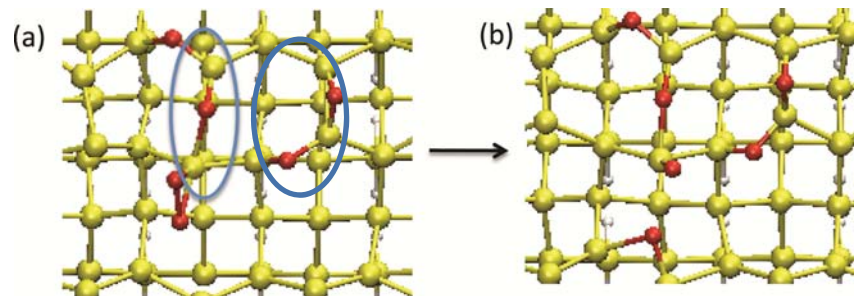
Si Oxidation: Free-energy Landscape

$$L = L_{CP} + \sum_{\alpha} \frac{1}{2} M_{\alpha} \dot{s}_{\alpha}^2 - \sum_{\alpha} \frac{1}{2} k_{\alpha} (s_{\alpha}(R) - s_{\alpha})^2 - V(s, t)$$

Stable
Geometry
Minimum(B)
is reached from
Geometry (A) with
free-energy
barrier of 0.8 eV



Minimum(B) is ubiquitous and further oxidation starts from there



Koizumi, Boero,
Shigeta, AO:
PRB 85, 205314 (2012)

Car-Parrinello Molecular Dynamics: Present and Future

➤ Present

- 100 - 300 atoms are extracted from nano-structures and used as a model → **Too small**
- Simulation continues only in 10 picoseconds at most → Meta-dynamics accelerate reactions,

But...

➤ Future

- Real-Space CPMD required → **now crossover at 1000-atom system**
- New scheme for Phase-space sampling???

Visualization

Visualization

Not for demonstration, publicity, advertisement etc,
But for the research tool.

Not for explaining what we already know,
But for exploring what we don't know.

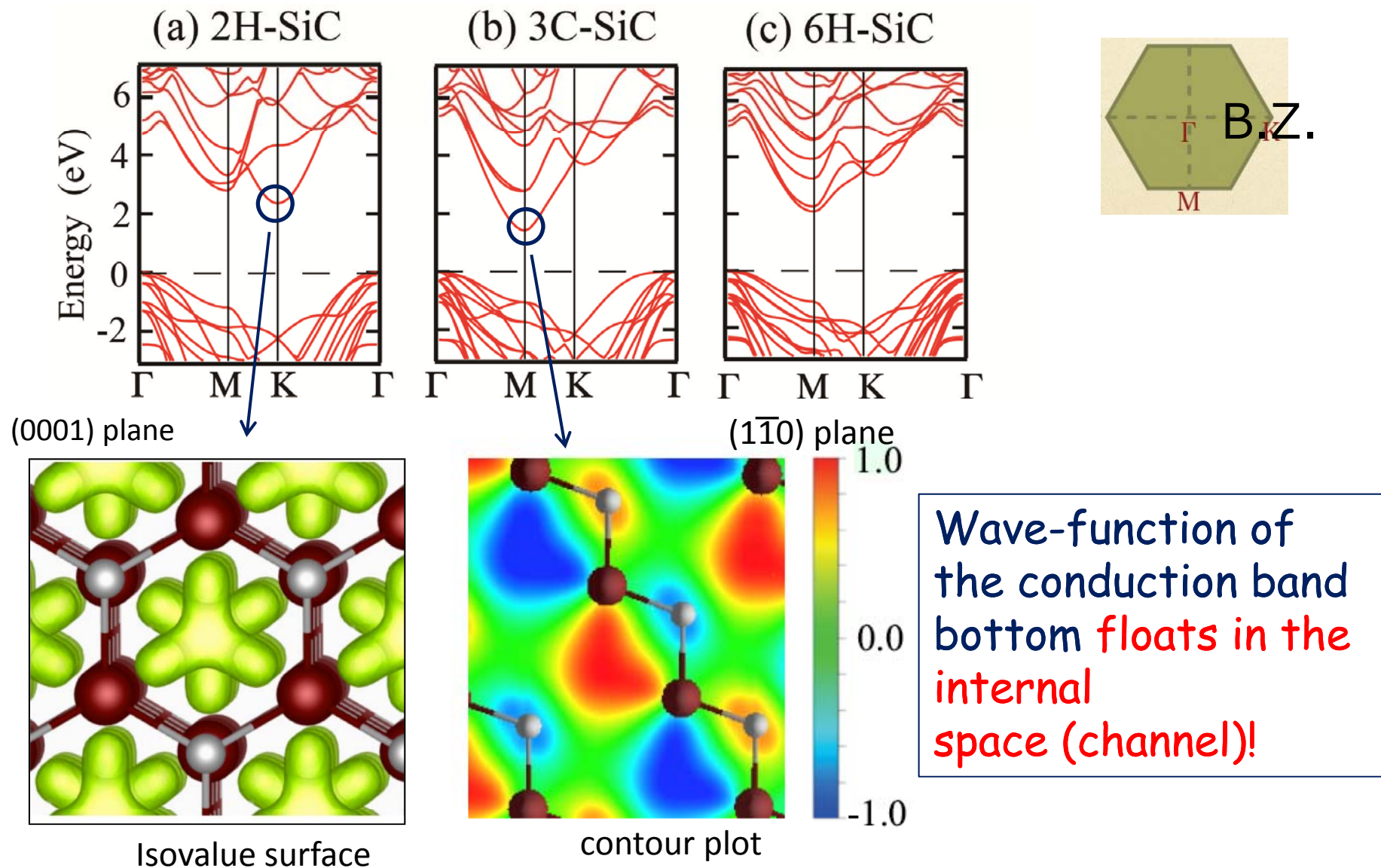
*get into materials, see the position of nuclei,
see the electron clouds, think about possible reactions,
imagine new properties,,,,,, then do simulations.*



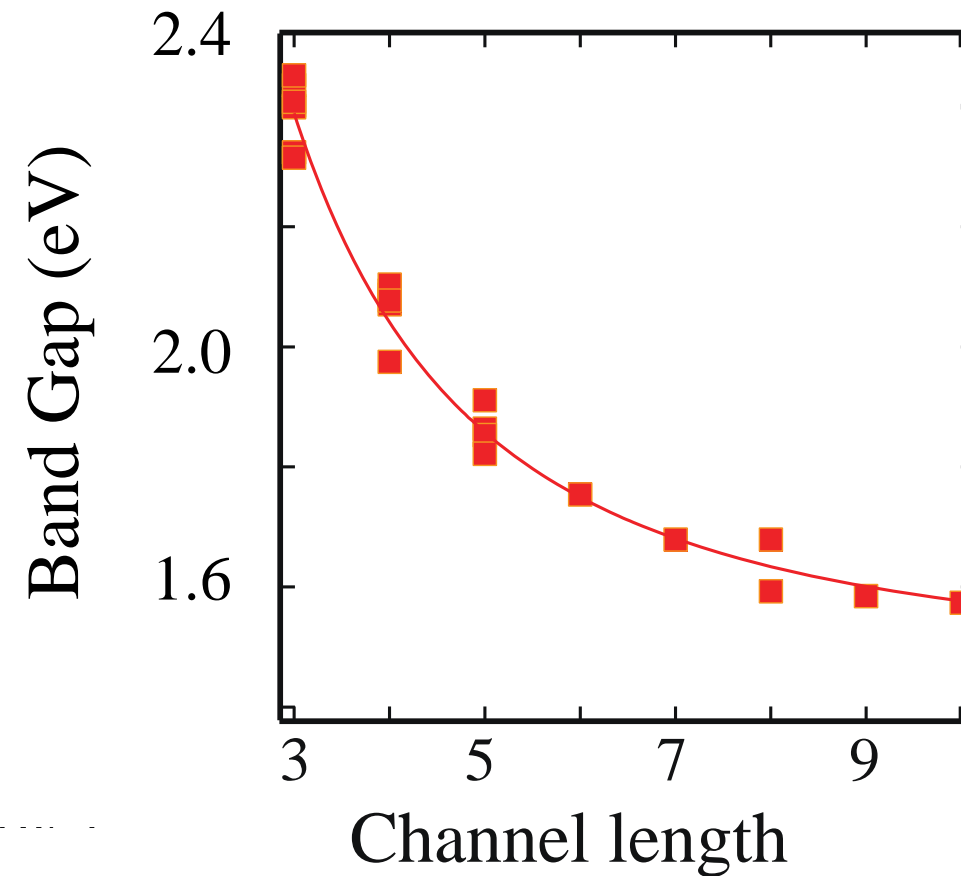
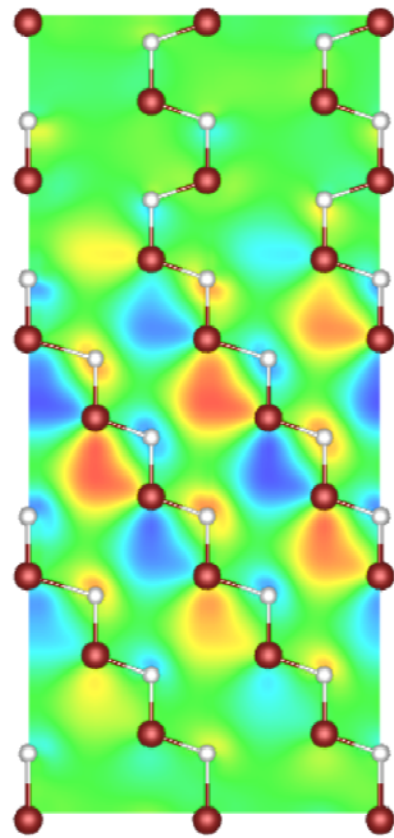
Fantastic Voyage 1966



Calculated Energy bands and Kohn-Sham orbitals of SiC



Volume of the internal space makes it



Materials Design through COMPUTICS: Large-Scale Density-Functional Calculations for Nanomaterials

➤ Computational Nanoscience

- What is Nano? What is Interesting?

- ✓ *Quantum Theory Required. Science and Technology are close to each other*

- ✓ *Carbon nanotube, Graphene, Si nanowire,...
Nanoshapes are decisive*

➤ Computics: Collaboration among Materials Science, Computer Science and Mathematics

- Density-Functional Theory (DFT)

- ✓ *Material is an Interacting Many-Electron System*

- HPC in DFT Calculations: RSDFT code

- ✓ *Mathematical Methodology, Algorithm*

- ✓ *How to Collaborate*