



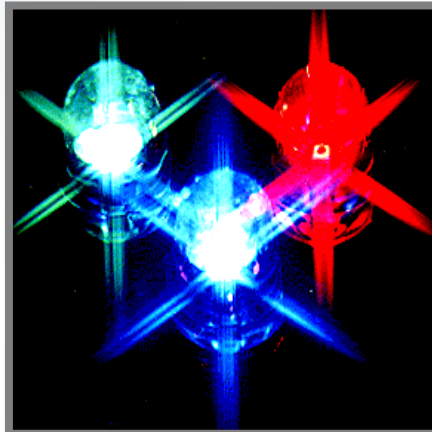
# Ab Initio Computation of Free Energies

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Department of Computational Materials Design*



Structural  
Materials



Optoelectronics

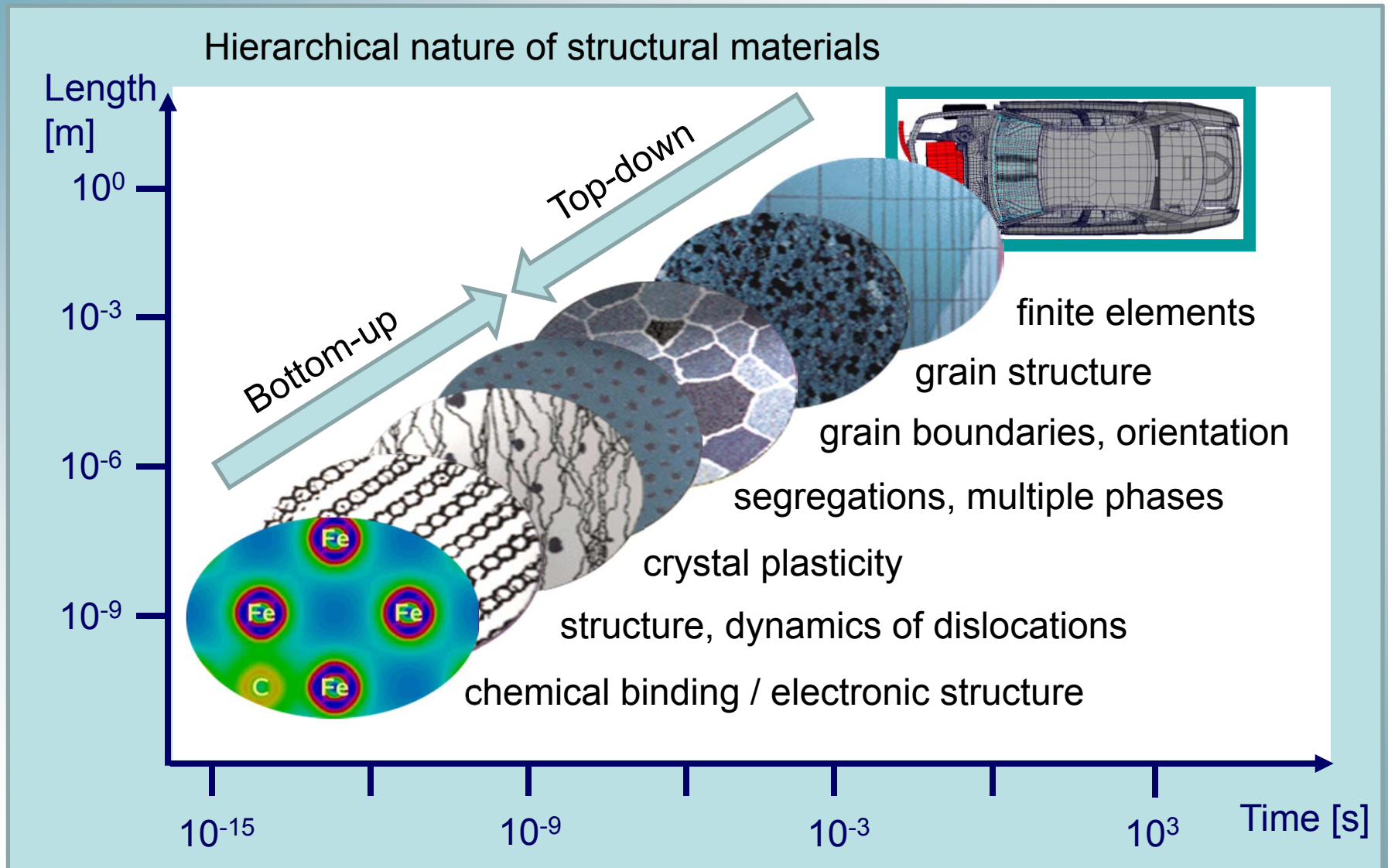


Corrosion



Bio-inspired  
materials

# Challenges in Materials Modeling

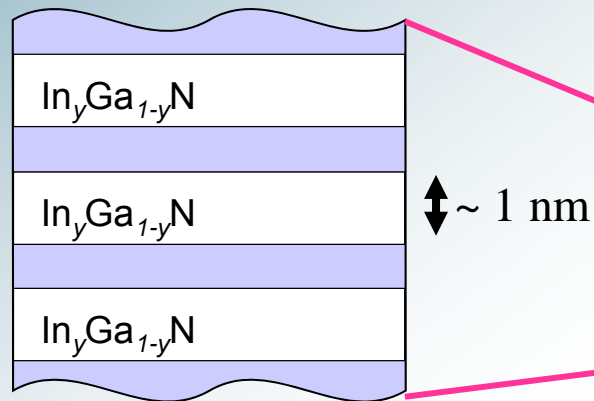


# Challenges in Materials Modeling



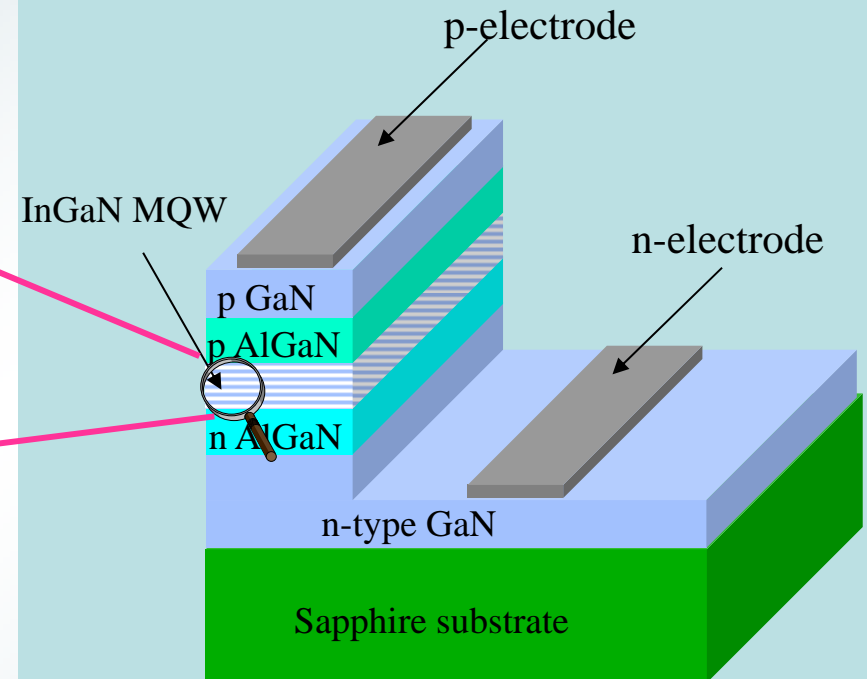
## Optoelectronic Devices

Structural design of the active region

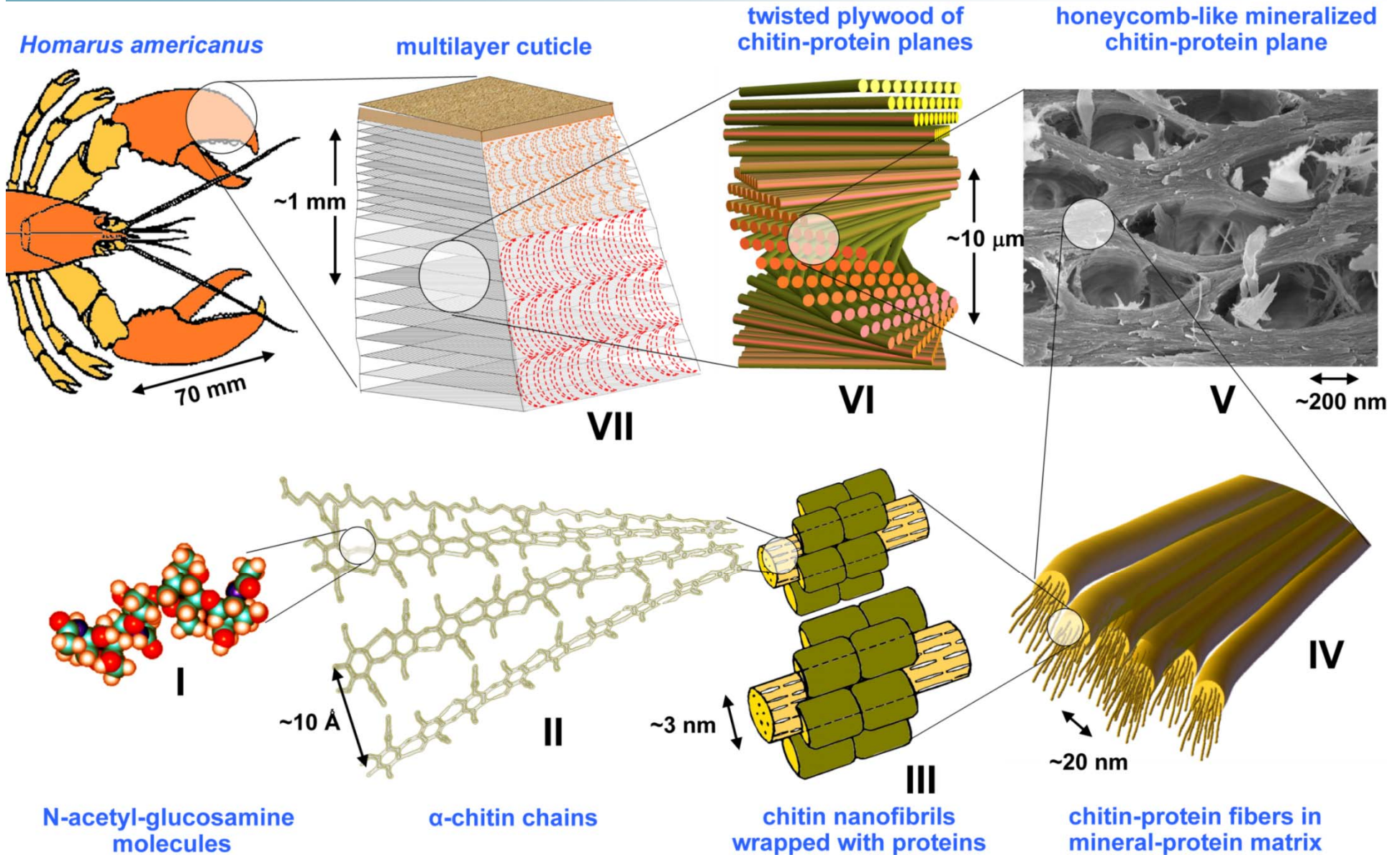


⇒ quantum well structure

### Device Structure: Blue Laser



# Challenges in Materials Modeling





## Development of accurate and reliable simulation tools for materials design

Are QM-based bottom-up methods able to address complexity and high temperature properties of modern materials?

Questions:

Is the accuracy at the most fundamental level sufficiently accurate?  
(How “robust” are today's available ab initio techniques?)

Can such an approach be used on present day computers?  
 (“smart” statistical approaches to sample huge configuration space)

Can we estimate error bars through all scales?

Key quantity:

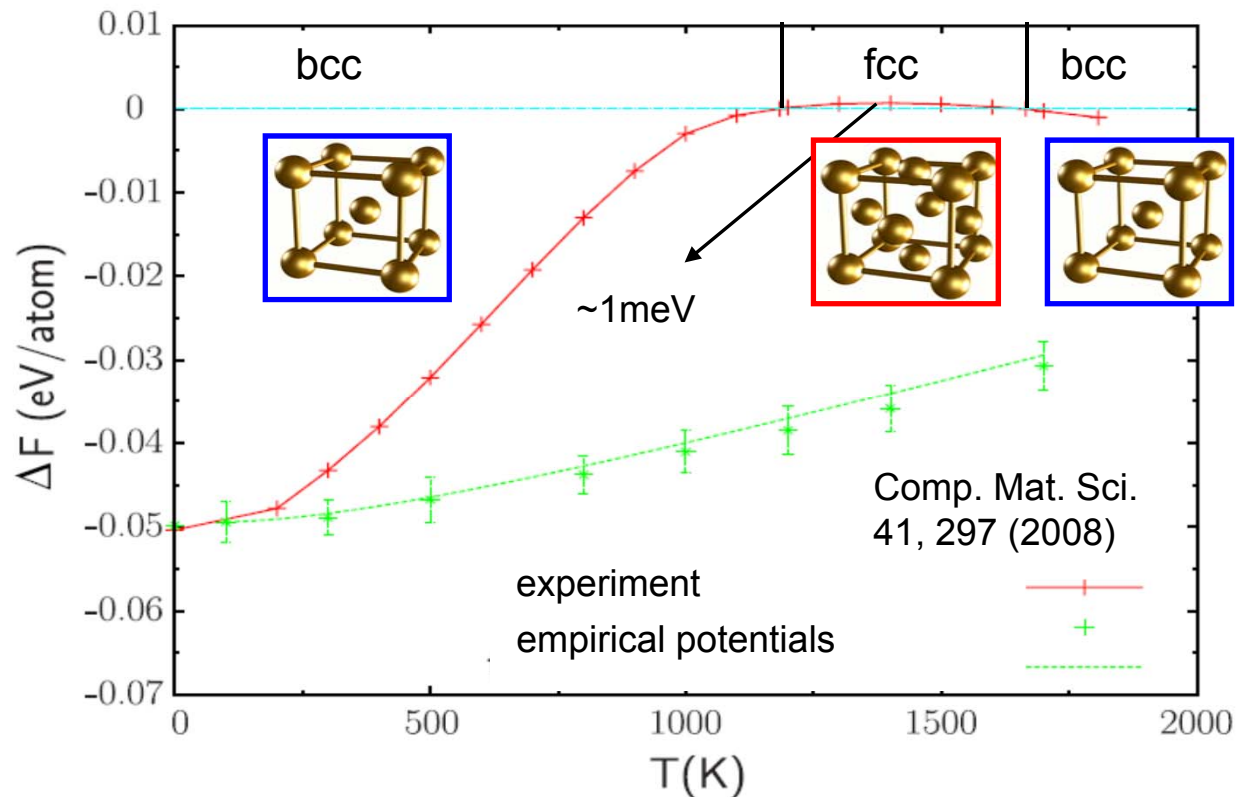
Free energy: Materials properties at finite temperatures

# Solid-Solid Phase Transitions



Key quantity: Free energy

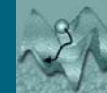
Free energy difference between bcc and fcc iron



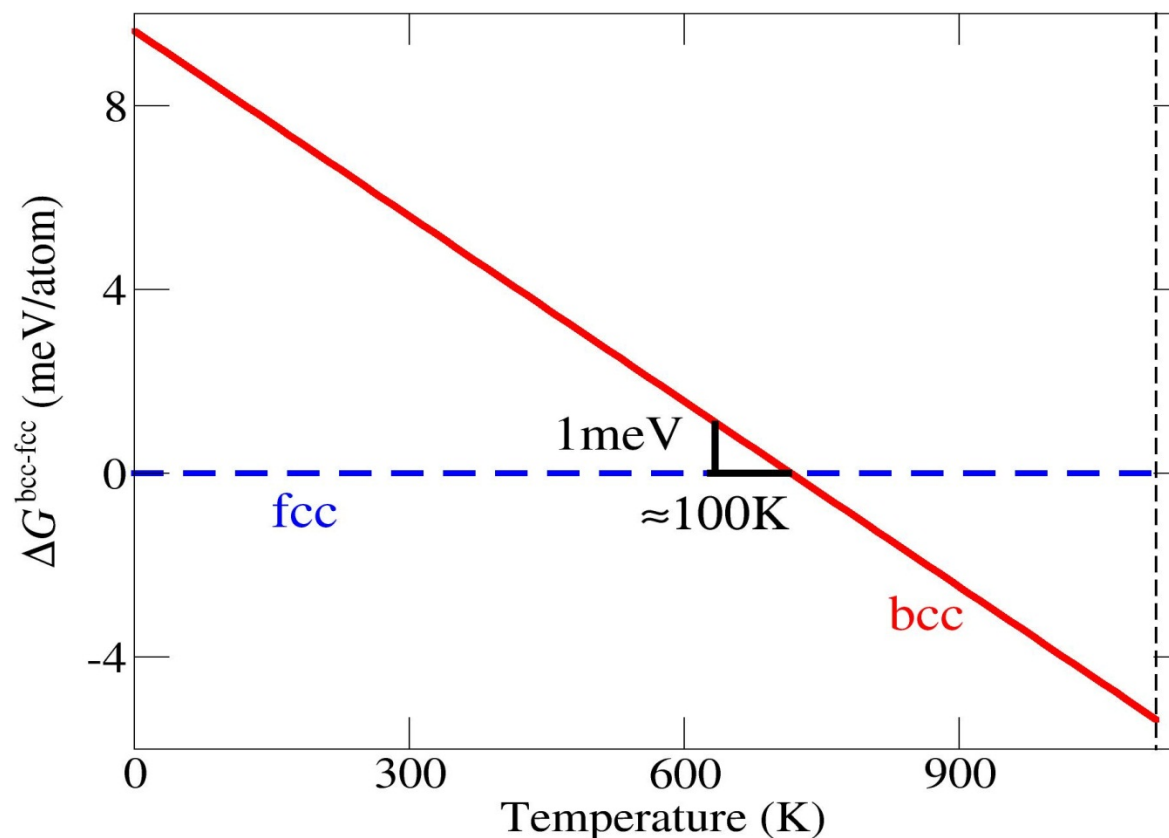
Energy resolution better 1 meV!

- Can we achieve such accuracy with present day DFT functionals?
- Can we “afford” statistical averages with such an accuracy?

# Free energies: What accuracy is needed?



## Free energy difference between bcc and fcc calcium



Target:

Numerical precision  $< 1 \text{ meV}$

→ remaining error purely due to xc-functional



**DFT has been originally developed as a ground state ( $T=0K$ ) theory**

**→ How to extend it to finite temperatures?**

# Ab initio computed Free energies

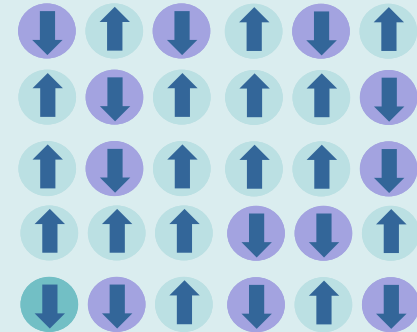


## Approach (schematic):

Energy for any configuration/phase can be computed:

$$E_{tot} \left( \left\{ \vec{R}_I, Z_I, \vec{\sigma}_I, f_i, \dots \right\} \right)$$

→ applicable to any system: bulk, surface, nano, ...



All possible excitation mechanisms can be described:

→ vibrational, magnetic, electronic, chemical, ...

Statistical averages provide thermodynamic quantities:

$$Z(V, T, x) = \left\langle e^{-E^{BOS}(\{\vec{R}_I, Z_I, \sigma_I, f_i, \dots\}_V) / k_B T} \right\rangle_{V, T, x}$$

Knowledge of partition function allows to derive any thermodynamic quantity!

# Ab initio Thermodynamics



➡ Key: Calculate partition function

$$Z(\hat{A}, T) = \sum_{\{\vec{R}_I\}_{\hat{A}}} e^{-E^{BOS}(\{\vec{R}_I, Z_I\}_{\hat{A}})/k_B T}$$

10<sup>7</sup> configurations  
a few hours

# Ab initio thermodynamics: Fundamentals



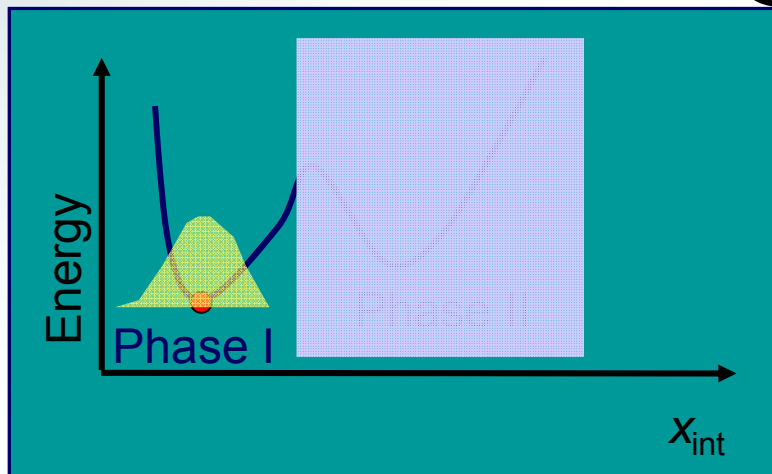
➔ Key: Calculate partition function

$$Z^{vib}(V, T) = \sum_{\{\vec{R}_I\}_V} e^{-E^{BO}(\{\vec{R}_I\}_V)/k_B T}$$

DFT
Experiment

Reduce ↓ dimensionality

$$Z_{Phase}^{vib}(V, T) = N_{deg} e^{-E^{tot}(\{\vec{R}_I\}_V^{Phase})/k_B T} \sum_{\{\vec{R}_I\}_V} e^{-E^{vib}(\{\vec{R}_I\}_V^{Phase})/k_B T}$$



equil. config.

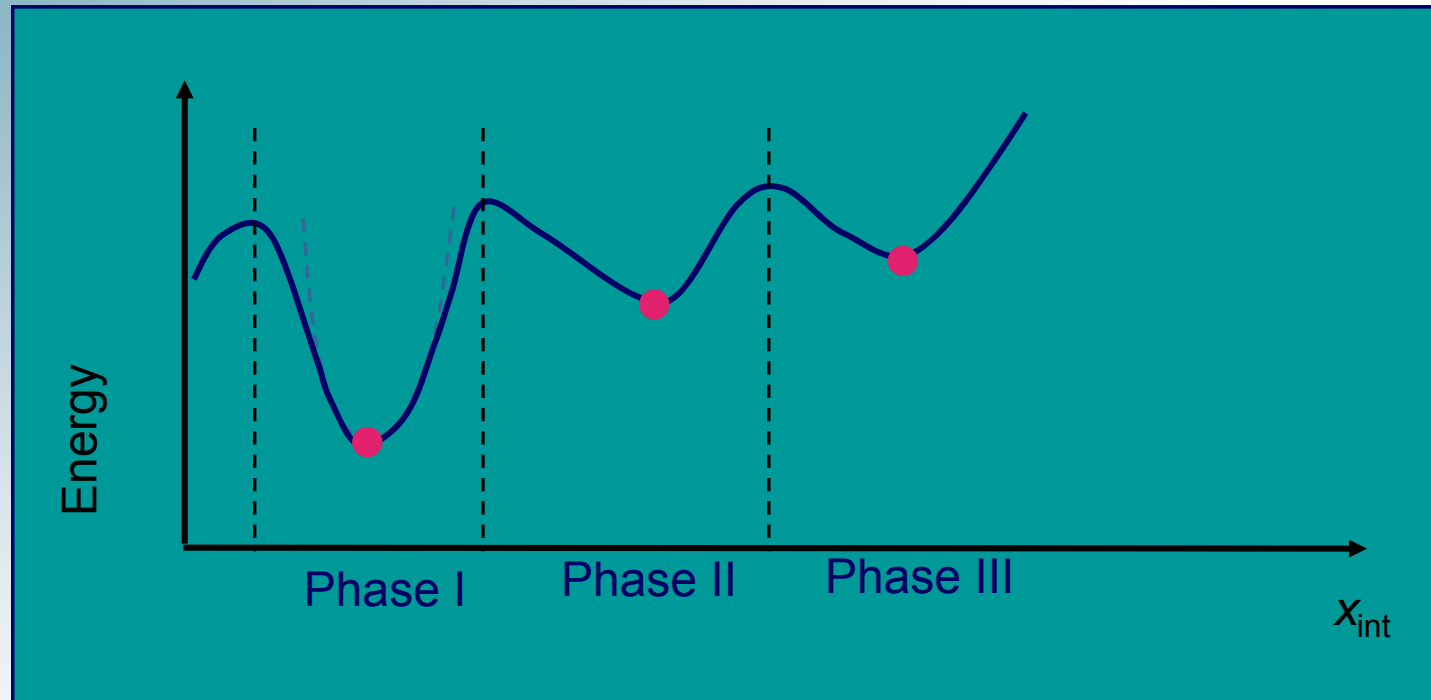
$$D_{I,J} = \frac{\partial E^{tot}(\{\vec{R}_I\}_V^{Phase})}{\partial u_I \partial u_J}$$

vibronic config.

# Ab initio thermodynamics: Fundamentals



Consider Born-Oppenheimer Surface



Divide high dimensional phase space into well defined regions  
→ split up energy in corresponding contributions

$$E^{\text{BOS}}\left(\left\{\vec{R}_I, Z_I\right\}\right) = E_{\sigma}^{\text{min}} + E_{\sigma}^{\text{harm}} + E_{\sigma}^{\text{anharm}}$$

# Ab initio Thermodynamics



$$Z(V, T, x) = \left\langle e^{-E^{BOS}(\{\vec{R}_I, Z_I, \sigma_I, f_i, \dots\})/k_B T} \right\rangle_{V, T, x}$$

$$= \sum_{\{\vec{R}_I, Z_I, \sigma_I, f_i, \dots\}} e^{-E^{BOS}(\{\vec{R}_I, Z_I, \sigma_I, f_i, \dots\})/k_B T} \Big|_{V, T, x}$$

Statistical averages over coordinates, magnetic moments, occupations, chemical compositions

Adiabatic approximation

$$= \sum_{\{\vec{R}_I\}} e^{-E^{BOS}(\{\vec{R}_I\}; \{Z_I, \sigma_I, f_i, \dots\}_{fixed})/k_B T} \Big|_{V, T, x}$$

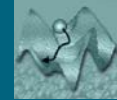
Vibrational excitations

$$\times \sum_{\{f_i\}} e^{-E^{BOS}(\{f_i\}; \{\vec{R}_I, Z_I, \sigma_I, \dots\}_{fixed})/k_B T} \Big|_{V, T, x}$$

Electronic excitations

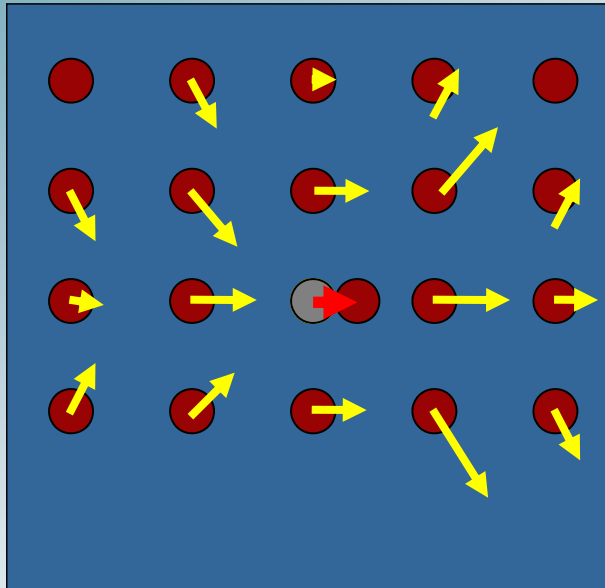
+ cross terms

e.g. electron-phonon interactions



# Quasiharmonic approximation + electronic excitations

# Vibronic excitations



$$\left[ T_{ion} + E^{BOS} \left( \underline{\underline{A}}, \tau_1, \dots, \tau_b \right) \right] \Lambda = E \Lambda$$

$$F_j = - \frac{\partial E^{BOS}(\tau_1, \tau_2, \dots)}{\partial \tau_j}$$

Dynamical matrix:

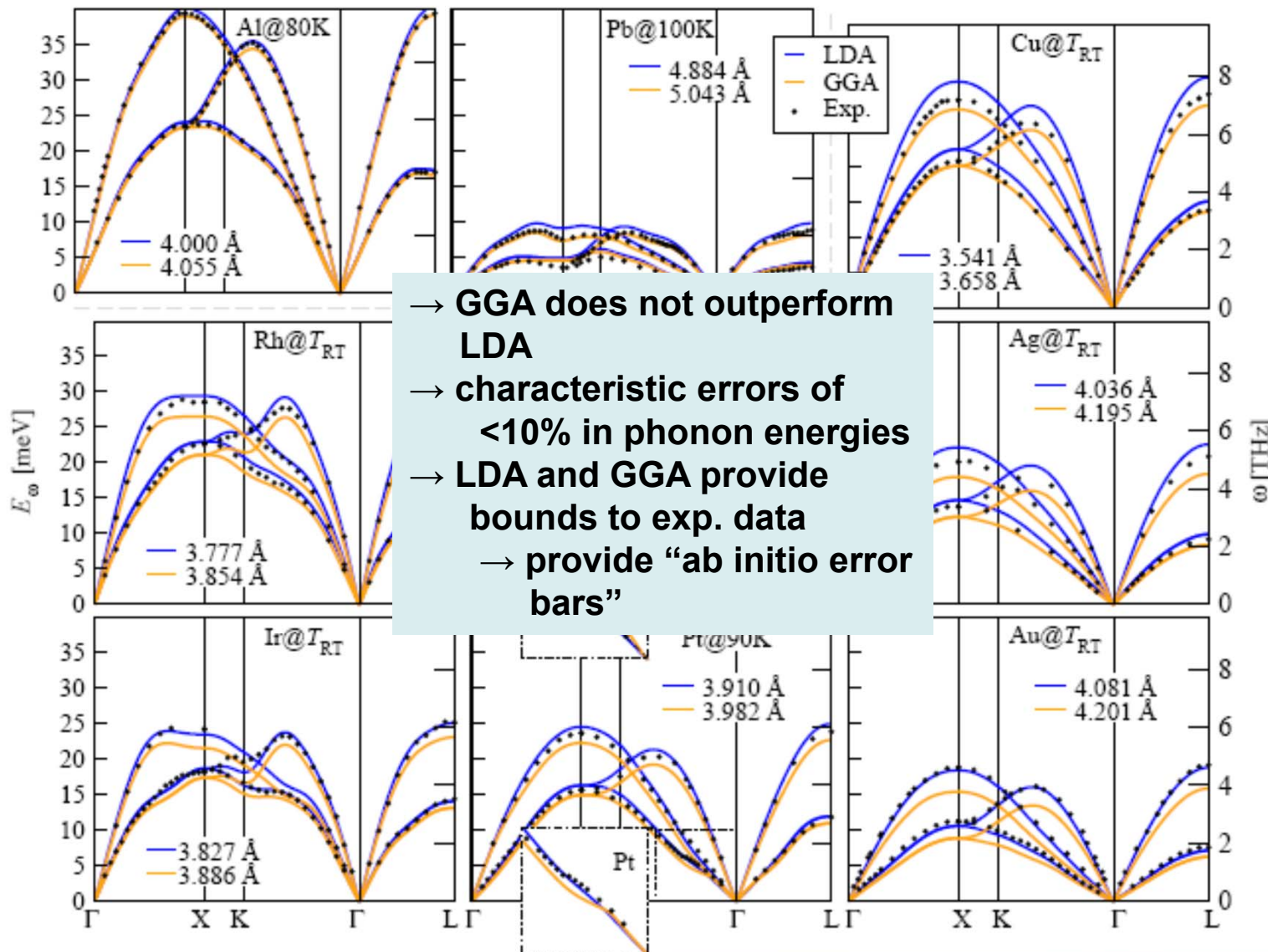
$$D_{ij} = - \frac{\partial F_i(\tau_1, \tau_2, \dots)}{\partial \tau_j}$$

Harmonic ansatz for extensions  $u_i$ :

$$M_i \ddot{u}_i + \sum_j D_{ij} u_j = 0 \iff \omega^2 u_i = \sum_j D_{ij} u_j$$

$$F(T, \underline{\underline{A}}) = E^{BOS} + k_B T \sum_q^{BZ} \sum_i^{3r} \left[ \frac{1}{2} \frac{\hbar \omega_i}{k_B T} + \ln \left[ 1 - \exp \left( - \frac{\hbar \omega_i}{k_B T} \right) \right] \right]$$

# Vibrational Properties: Phonons

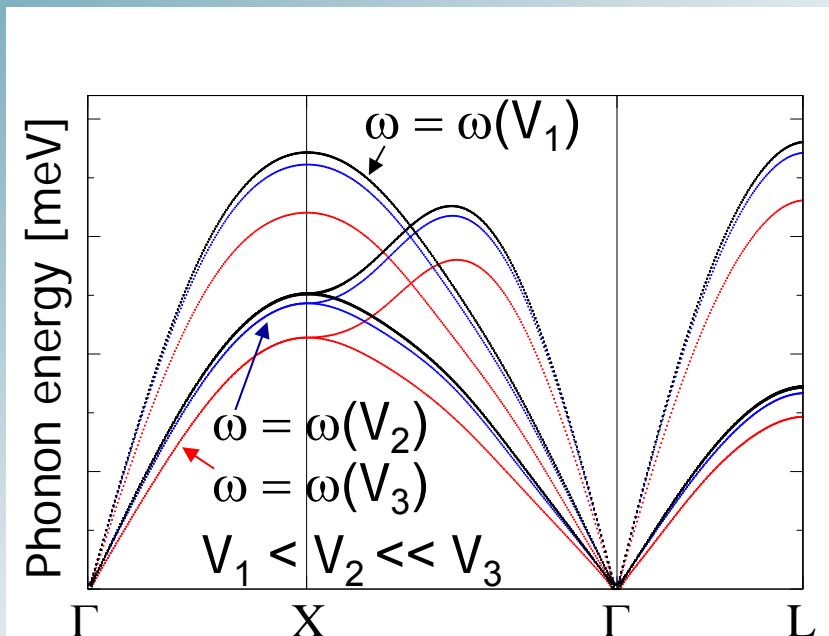


Grabowski, Hickel, JN, PRB 76, 024309 (2007)

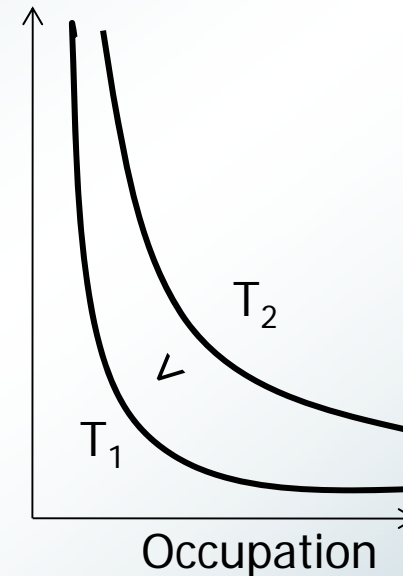
# Extension to Finite Temperatures



## Quasiharmonic approximation



+



Phonons (V)

+

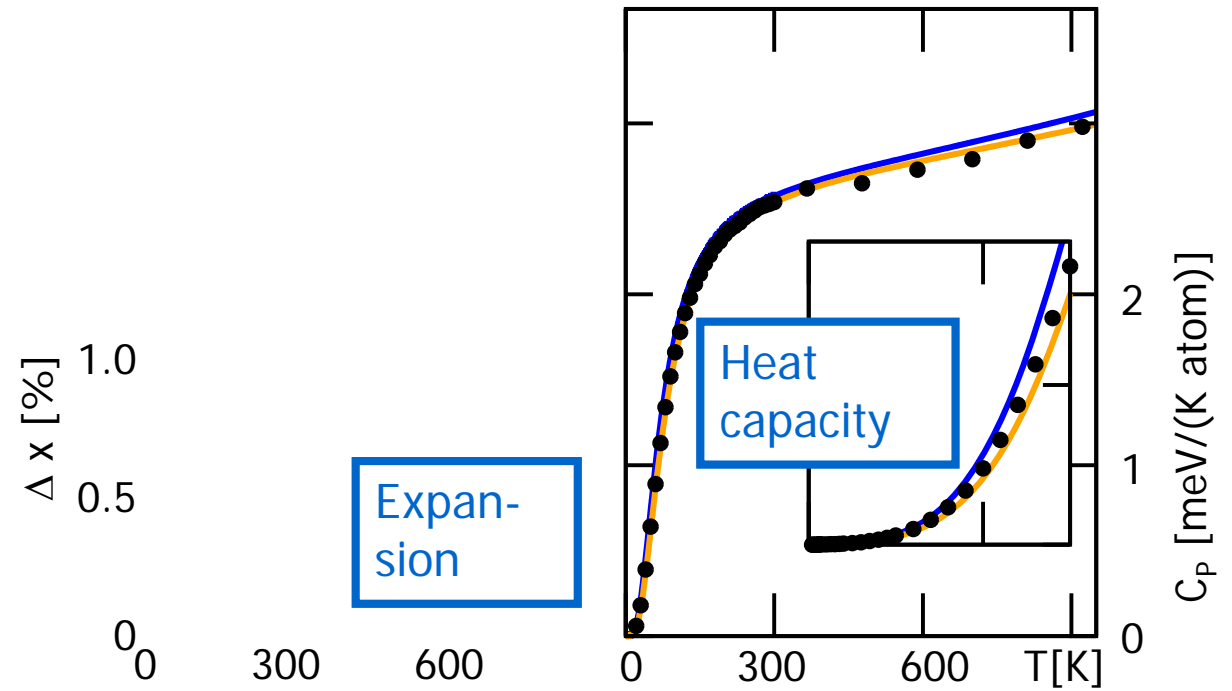
Bose-Einstein (T)

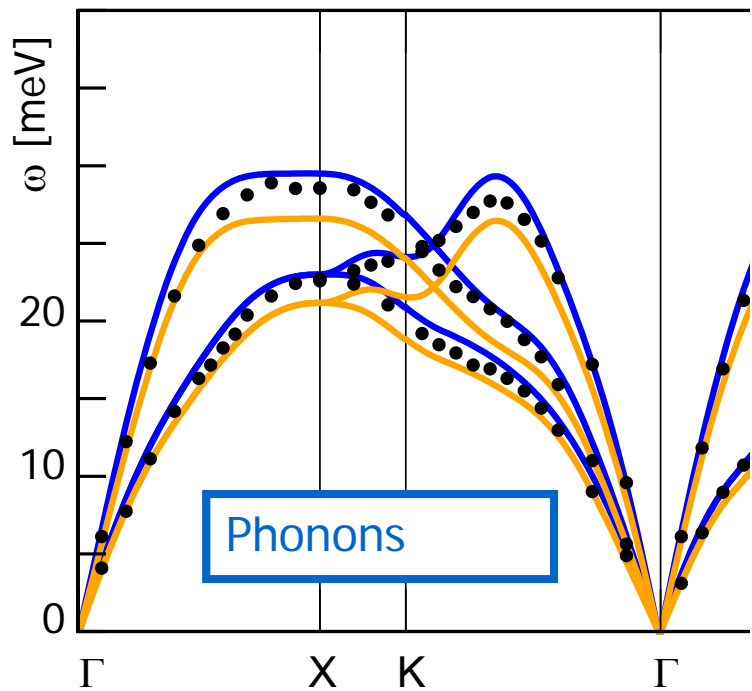
=  $F^{\text{vib}}(T, V)$

$$F(T, \underline{\underline{A}}) = E^{\text{BOS}} + \sum_q^{\text{BZ}} \sum_i^{3r} \left[ \frac{1}{2} \hbar \omega_i + k_B T \ln \left[ 1 - \exp \left( -\frac{\hbar \omega_i}{k_B T} \right) \right] \right]$$



200 400 600 800 T[K]





Rh

$\Delta x$  [%]

1.0

0.5

0

L

300

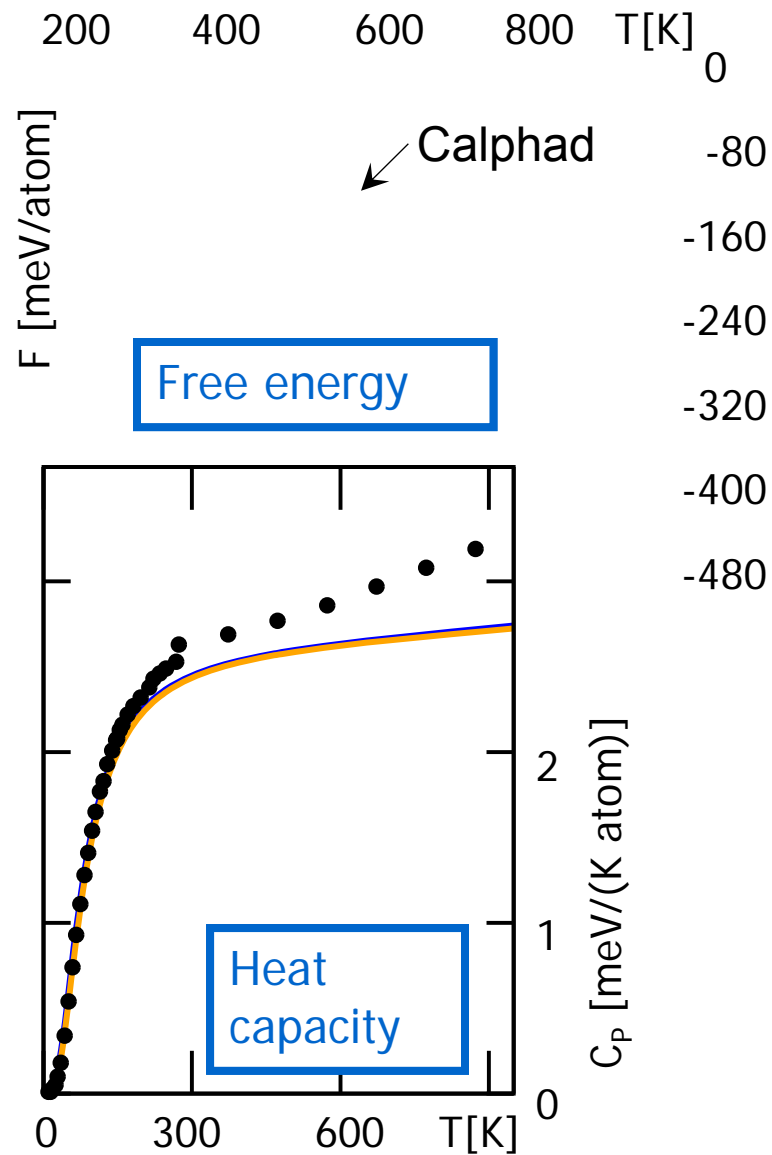
600

GGA

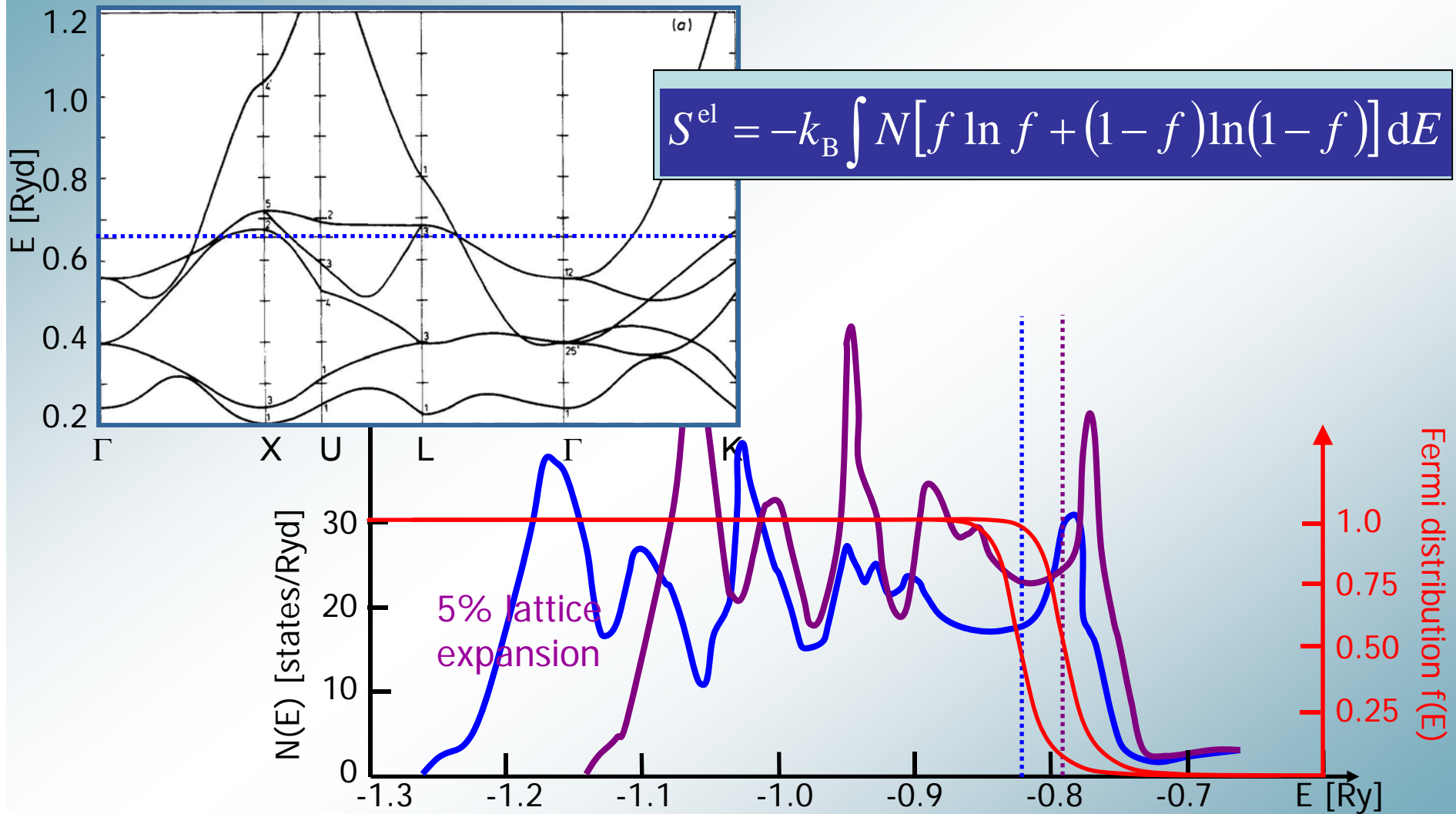
LDA

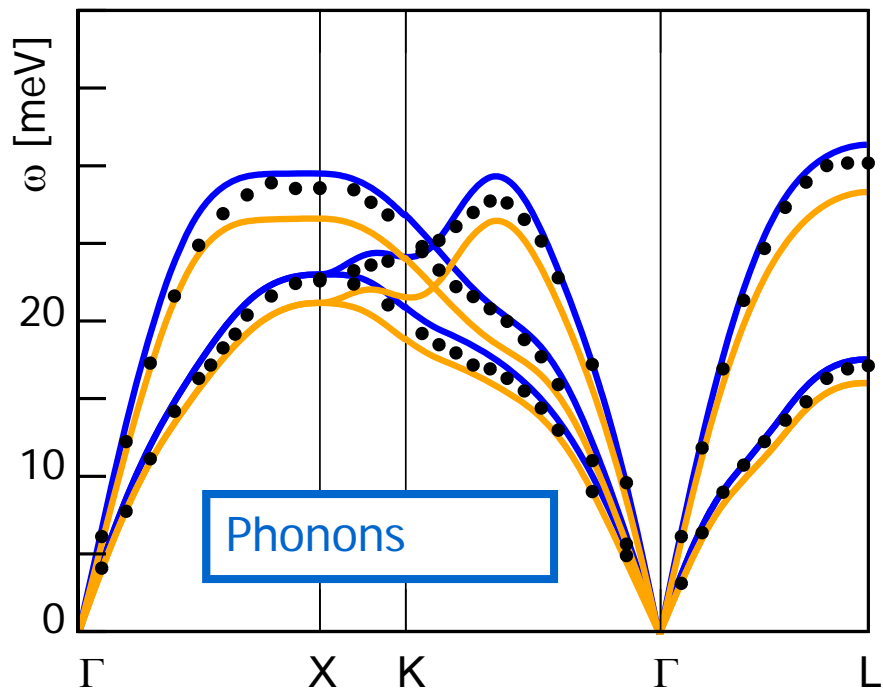
Exp.

Expansion



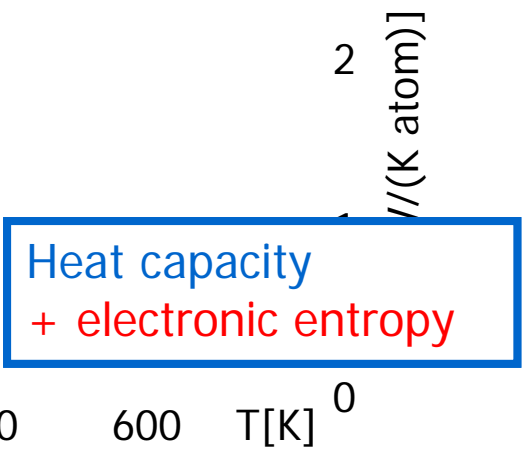
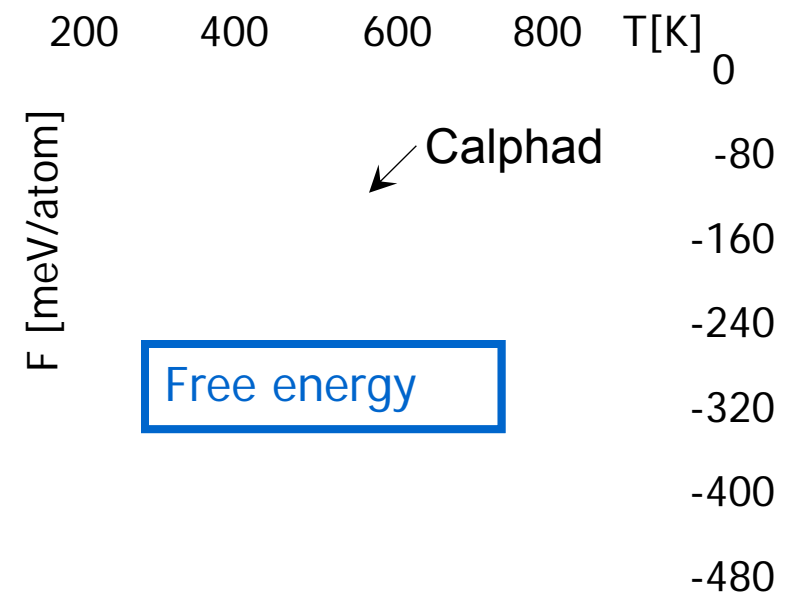
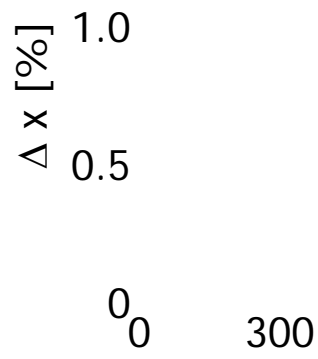
# Contribution of electronic excitations to free energy

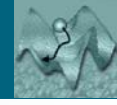




— GGA  
— LDA  
● Exp.

Rh

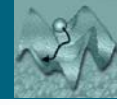




**Are contributions beyond quasiharmonic approximation relevant?**



# Specific Heat at High Temperatures



## Über die spezifische Wärme fester Körper bei hohen Temperaturen.

Von **M. Born** und **E. Brody** in Göttingen.

(Eingegangen am 9. Juli 1921.)

Einleitung. Nach dem Gesetz von Dulong und Petit bzw. seiner Verallgemeinerung durch Neumann, Regnault und Kopp soll die Atomwärme bei konstantem Volumen  $C_v$  im festen Aggregatzustande sich mit wachsender Temperatur dem Grenzwerte

$$C_v = 3R = 5,96 \text{ cal grad}^{-1}$$

nähern ( $R$  ist die absolute Gaskonstante). Bei vielen Substanzen aber wird dieser Wert merklich überschritten, und zwar um so mehr, je näher die Temperatur dem Schmelzpunkt kommt. Es gibt hauptsächlich zwei Ursachen, auf die man diese Erscheinung zurückführen kann: Einmal das Anwachsen der Schwingungsamplituden der Atome über den Bereich hinaus, in dem Proportionalität zwischen Ausschlag und zurückziehender Kraft besteht; sodann das Auftreten neuer Freiheitsgrade. Besonders bei den Metallen liegt es nahe, an die Freiheitsgrade der Leitungselektronen zu denken, die bei hohen Temperaturen einen Anteil am Wärmehalt übernehmen könnten. Aber das Ansteigen der spezifischen Wärme bei der Annäherung an den Schmelzpunkt spricht mehr für die erste Hypothese; denn das Schmelzen muß doch in einem Anwachsen der Atomschwingungen zu solcher Stärke bestehen, daß das Gefüge des Kristallgitters zerstört wird, und es ist wahrscheinlich, daß sich diese heftige Agitation der Atome schon vorher thermisch bemerklich macht.

### Question:

- origin of additional degrees of freedom at high T

### Assumption:

- anharmonicity
- electrons in conduction band

### Present proposals:

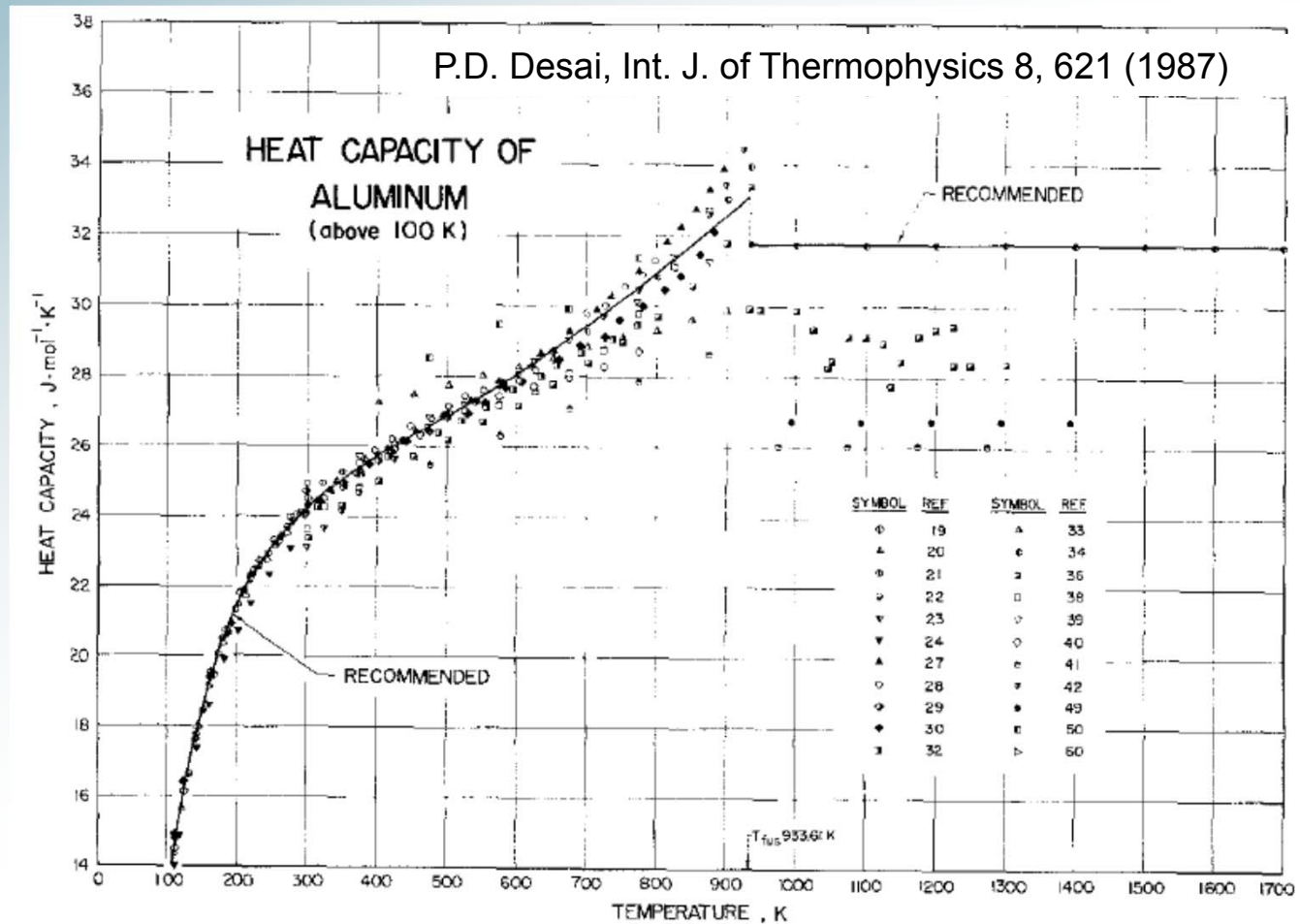
- vacancies
- di-vacancies and interstitials
- self-critical vacancy conc.
- anharmonicity
- electrons in conduction band

...

# Challenges

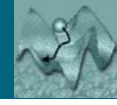


## Experiment



### Error sources

- Radiation losses ( $\sim T^3$ )
- Defects, imperfections



**How to sample over  $10^7$  configurations with  
ab initio accuracy?**



**Can we use empirical potentials to describe  
anharmonic contributions**

# Coarse graining configuration space



## Main challenge:

Reduce number of (ab initio) configurations by several orders of magnitude

$$\langle U(T) \rangle = \frac{1}{N} \sum_i^N E^{BOS}(\{\vec{R}_I(t_i)\})$$

from MD or MC

## Two major concepts

Thermodynamic integration

Free energy perturbation

# Thermodynamic Integration

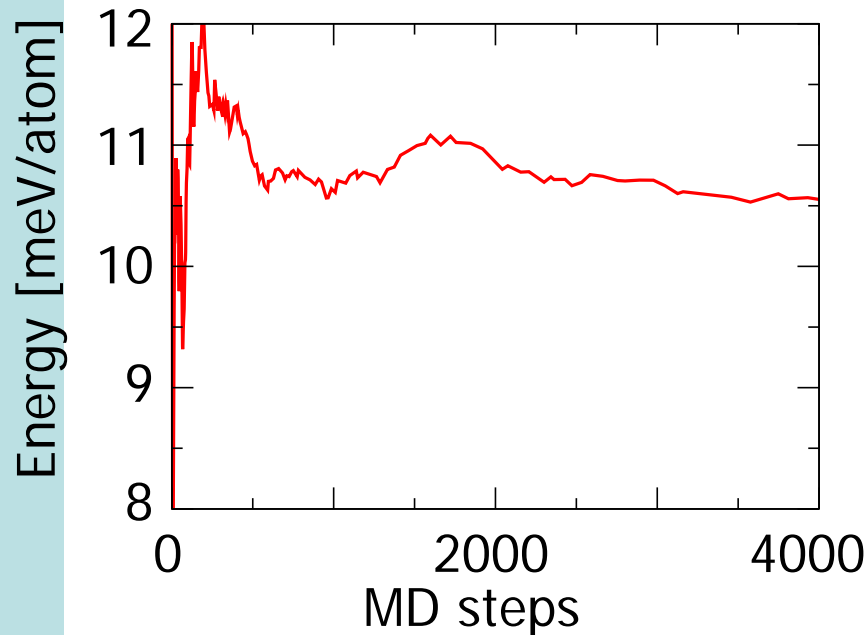


**Key idea: Compute free energy change between reference system A and real system B**

$$\Delta F(A \rightarrow B) = \int_0^1 \left\langle \frac{\partial U(\lambda)}{\partial \lambda} \right\rangle_{\lambda} d\lambda$$

with  $U(\lambda) = U_A + \lambda(U_B - U_A)$

MD simulation in TI-step



Application straightforward if good reference is available

Typically the number of configurations can be reduced by 1-2 orders of magnitude  
→ several  $10^4$  configurations

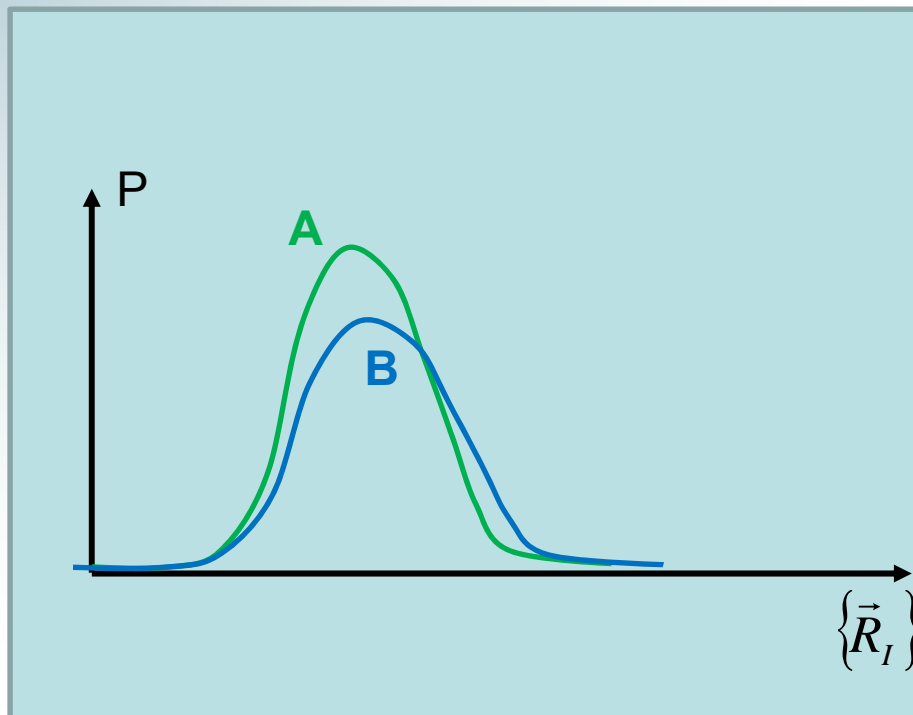
Not affordable on highest ab initio level

# Free energy perturbation



**Key idea: Compute free energy change between reference system A and real system B**

$$\Delta F(A \rightarrow B) = -k_B T \ln \left\langle \exp \left( -\frac{E_B - E_A}{k_B T} \right) \right\rangle_A$$



Performance increases with quality of reference

For large differences to reference the method becomes inefficient/fails

For the targeted accuracy less efficient than thermodynamic integration

# How to boost coarse graining?



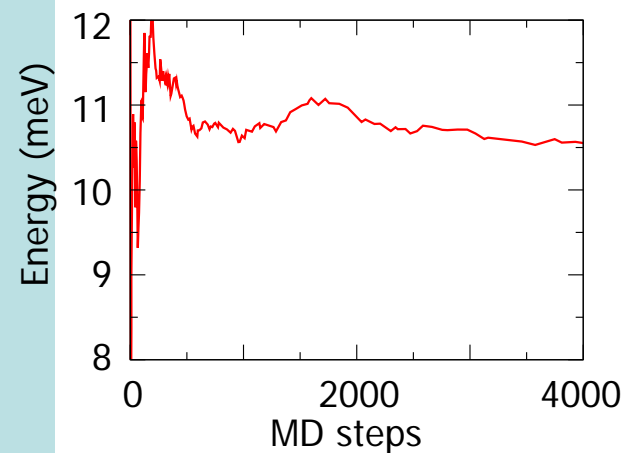
Cycles in thermodynamic integration

$$\Delta F(A \rightarrow B) = \int_0^1 \left\langle \frac{\partial U(\lambda)}{\partial \lambda} \right\rangle_{\lambda} d\lambda$$

$\lambda$  integration (2-10 steps)

→ Improved reference reduces step number

Thermodynamic average  
( $10^3 \dots 10^4$  MD steps)



Performance bottleneck

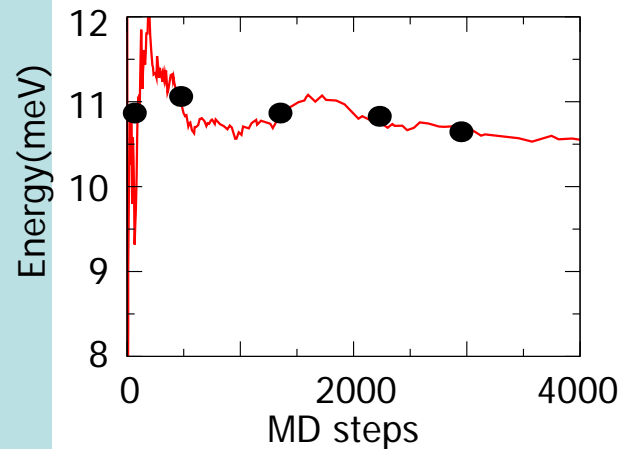


## $\lambda$ integration (2-10 steps)

- Improved reference reduces step number
- Here: Quasiharmonic reference

## Thermodynamic average ( $10^3 \dots 10^4$ MD steps)

- Use free energy perturbation approach
- Use low/medium converged DFT as reference

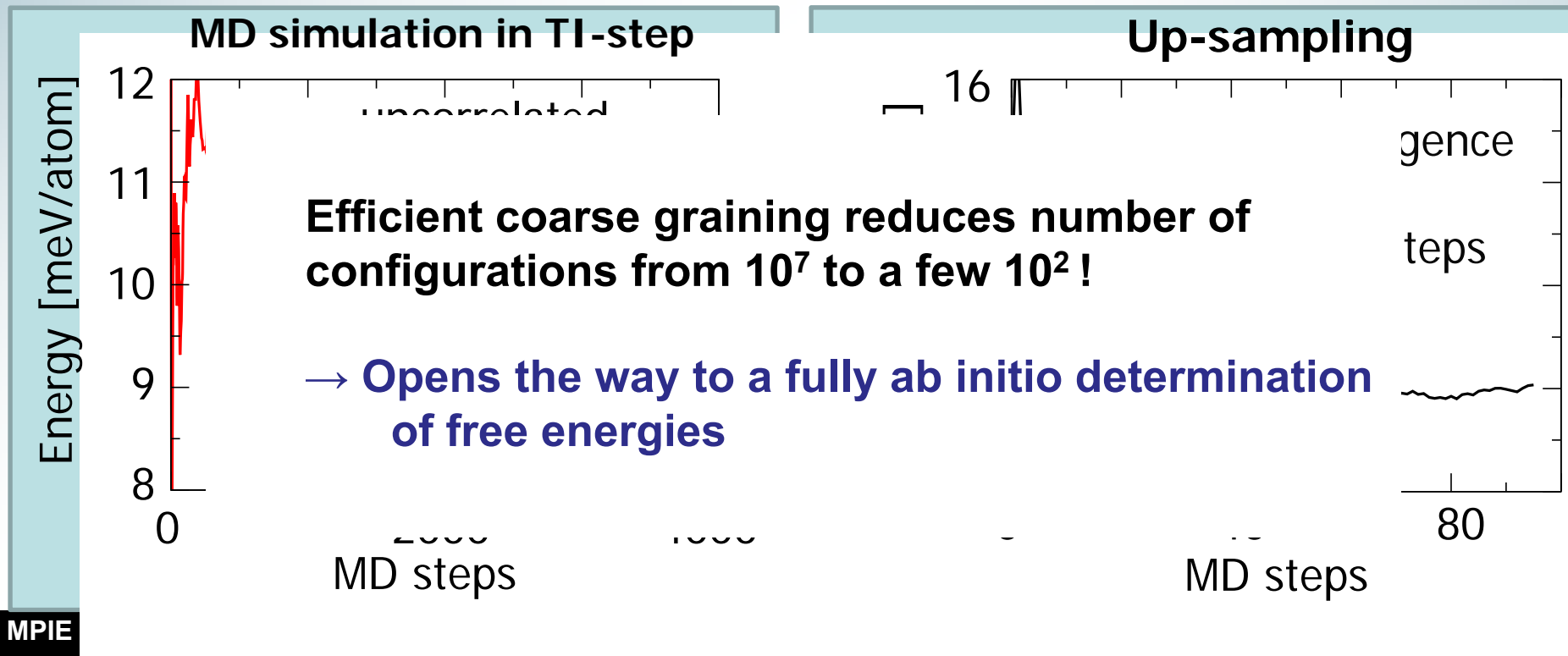
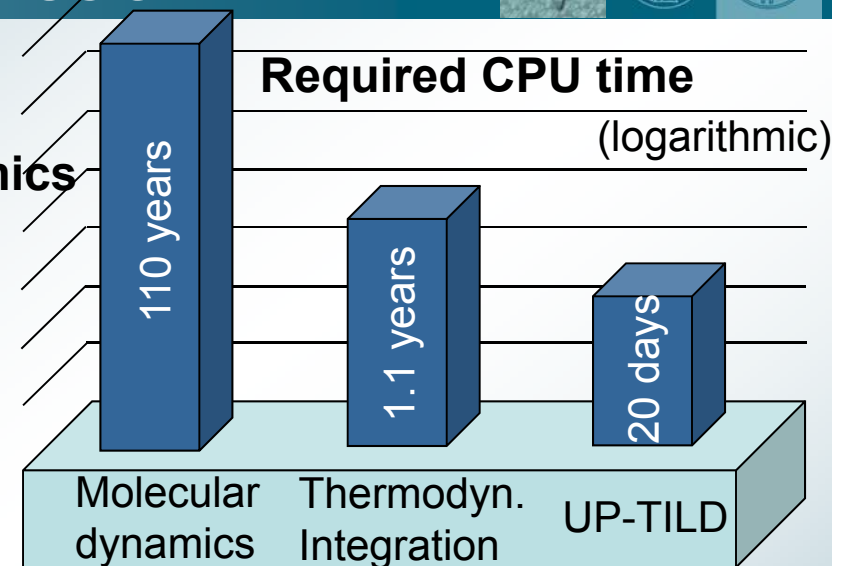


→ Reduces number of configurations by  $10^2$

# Performance of the new approach

- Wavefunction extrapolation
- Thermodyn. integr. (TI) with Langevin dynamics
- Ensemble average instead of time average (parallelization)
- **Up-sampling** in  $\lambda$ -TI step
- etc.

Grabowski et al., Phys. Rev. B79, 134106 (2009).





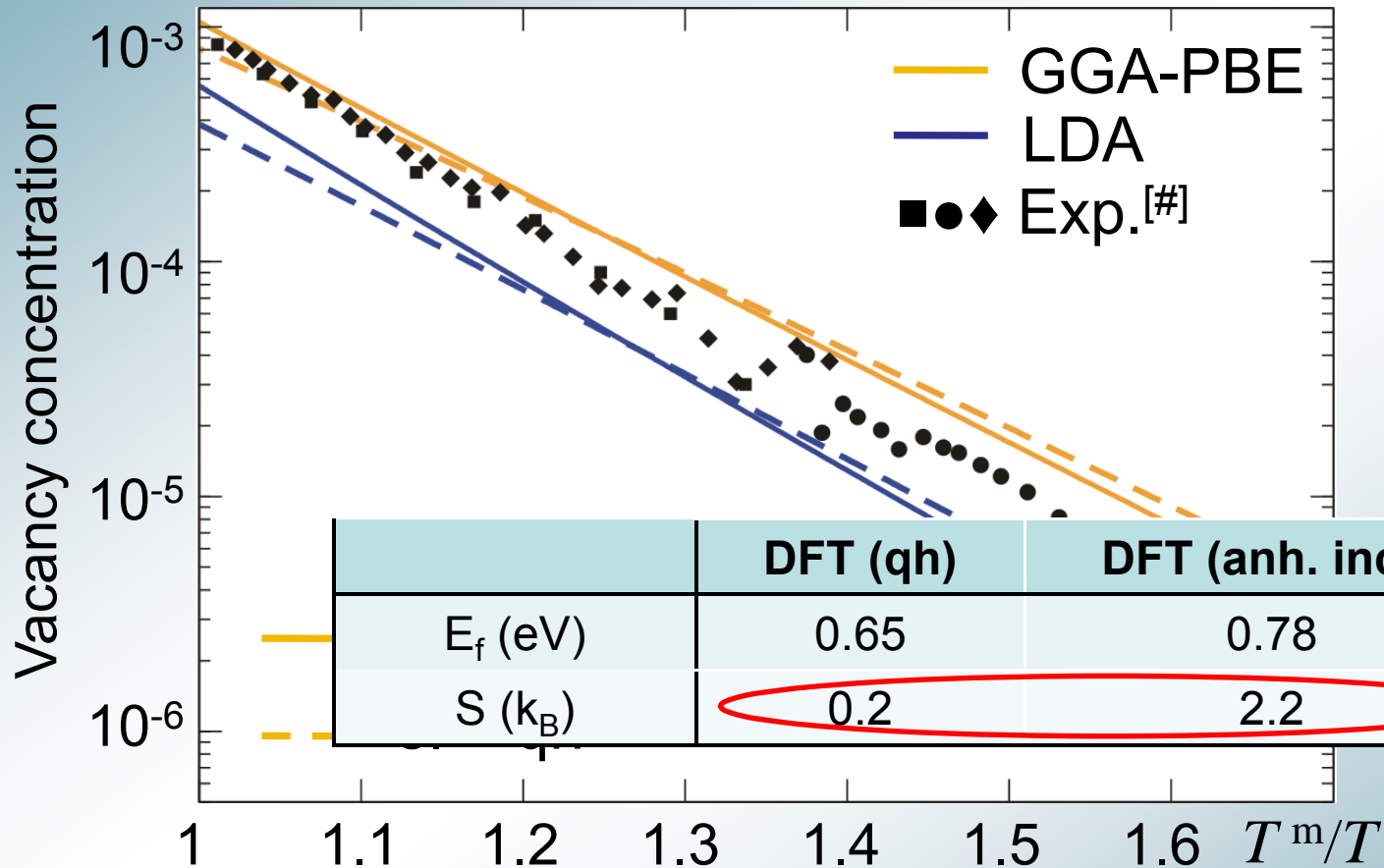
# Benchmark against experiment

# Vacancy concentration



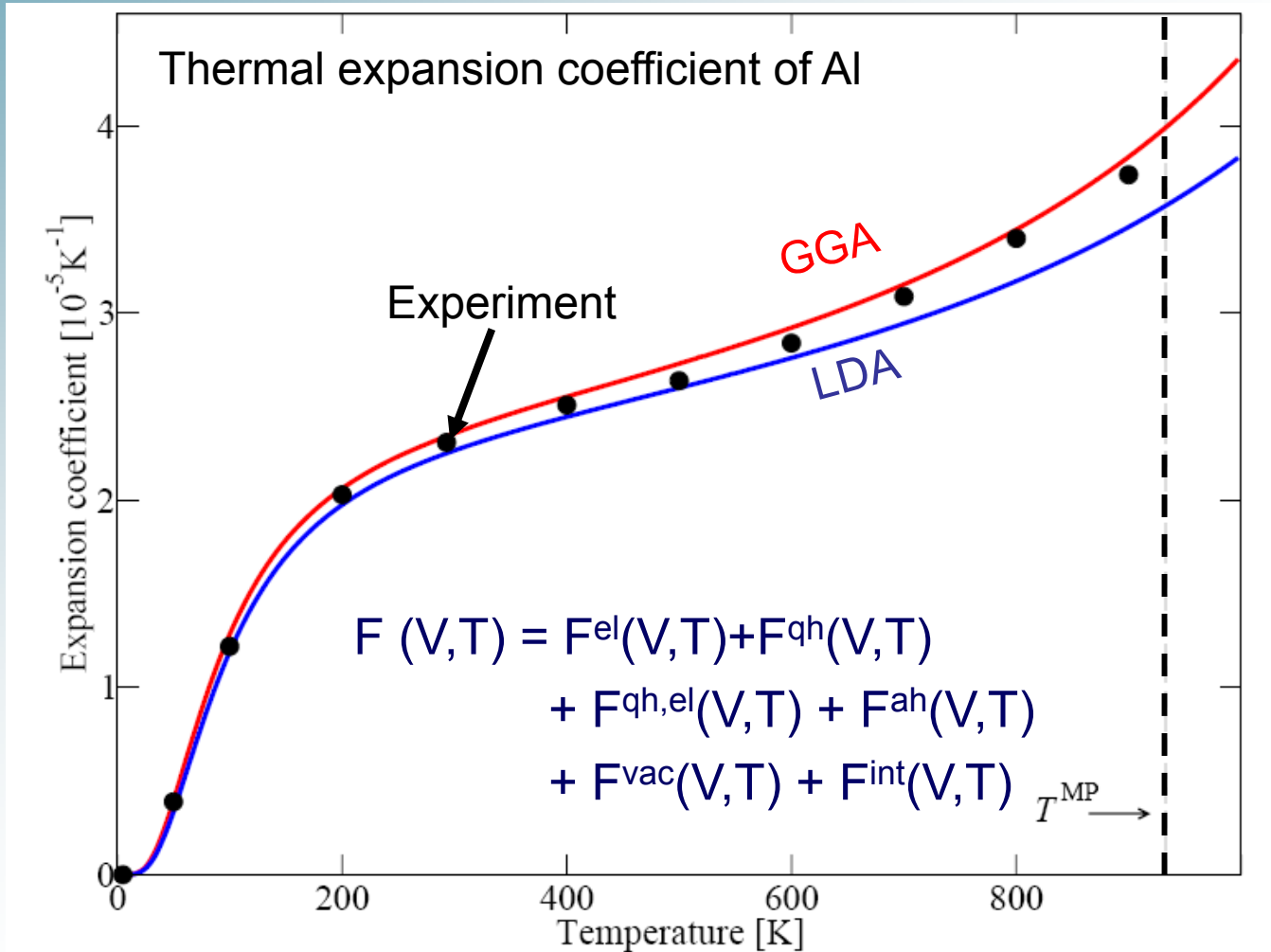
[#] R. O. Simmons and R. W. Balluffi, Phys. Rev. 117, 52 1960.  
Th. Hehenkamp, J. Phys. Chem. Solids 55, 907 1994.

fcc aluminum



el: electronic, qh: quasiharmonic,  
ah: anharmonic

# Thermal expansion coefficient



→ Excellent agreement with experiment

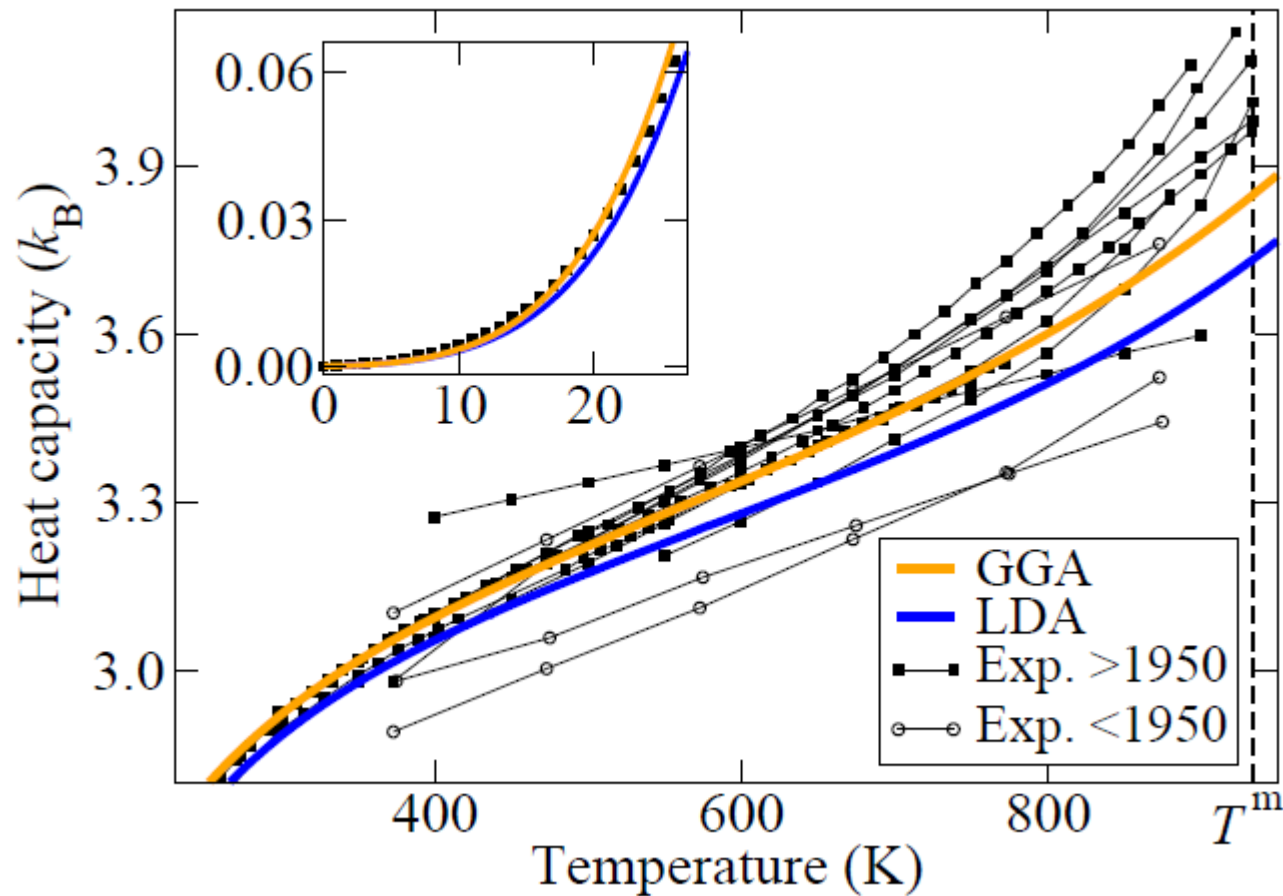
→ Systematic trend: LDA and GGA provide approximate measure of error bars

Grabowski, Ismer, Hickel, Neugebauer, Phys. Rev. B 79, 134106 (2009)

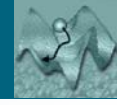
# Heat Capacity



## Heat capacity of Al

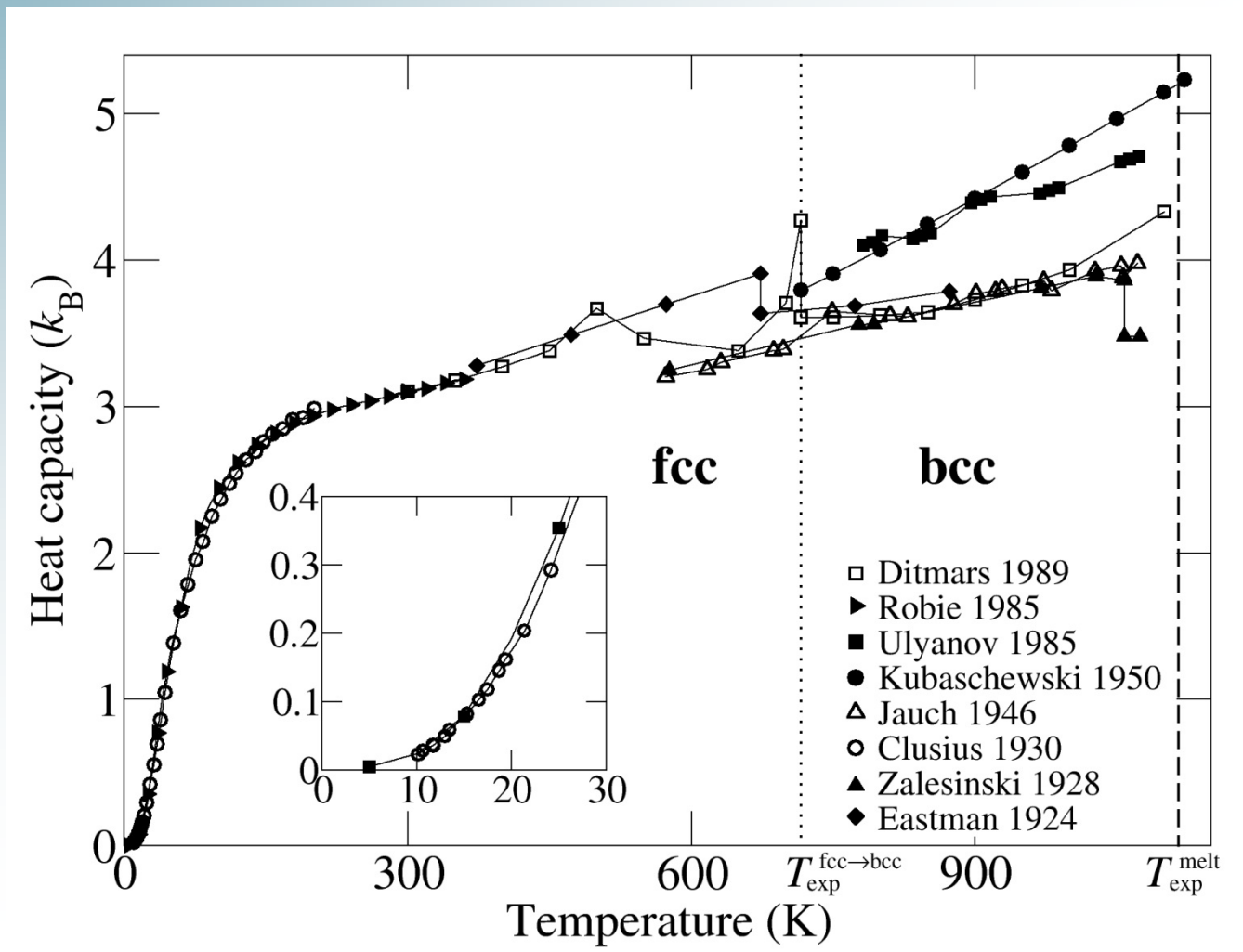


→ DFT gives lower bound to all recent experiments

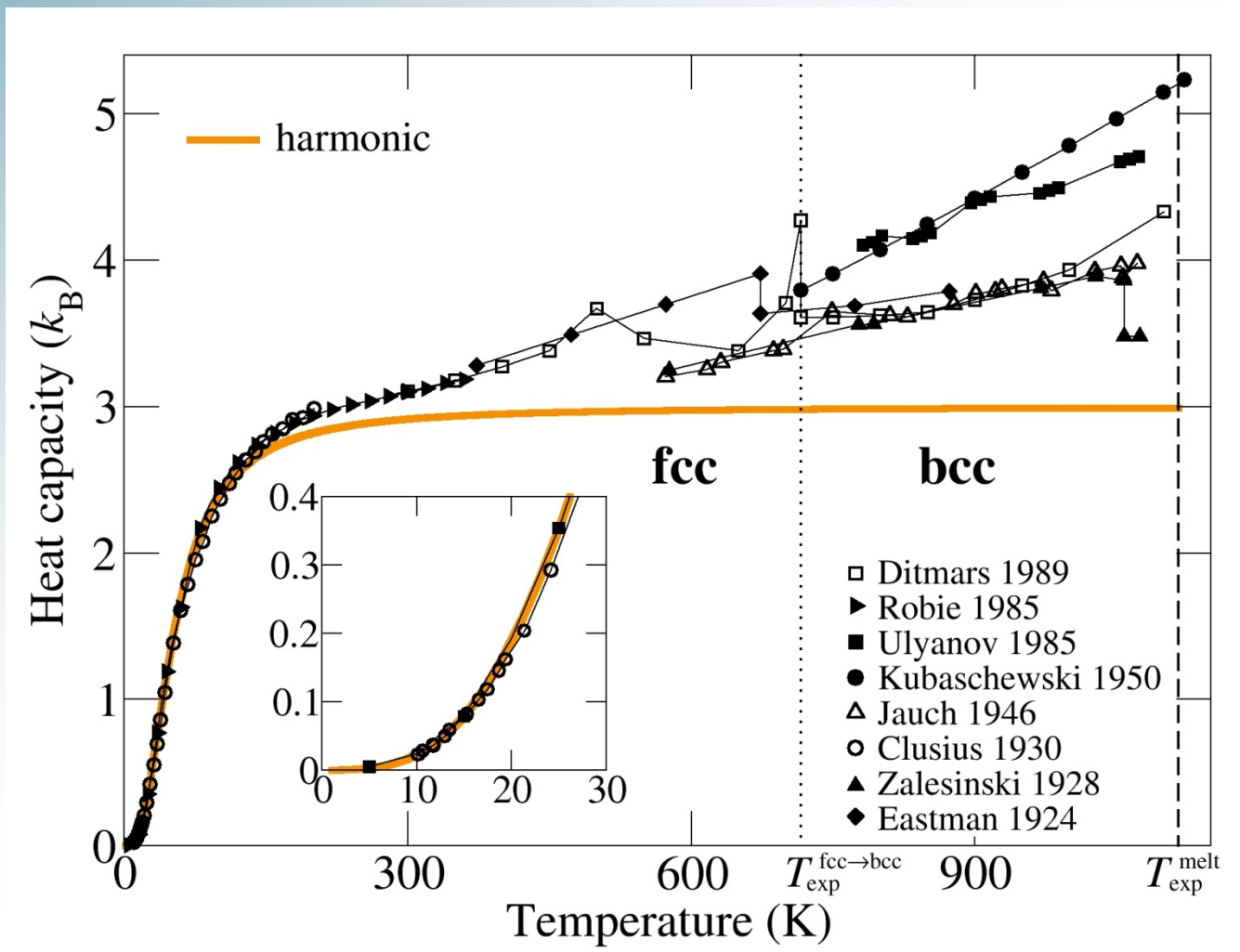


# Applications

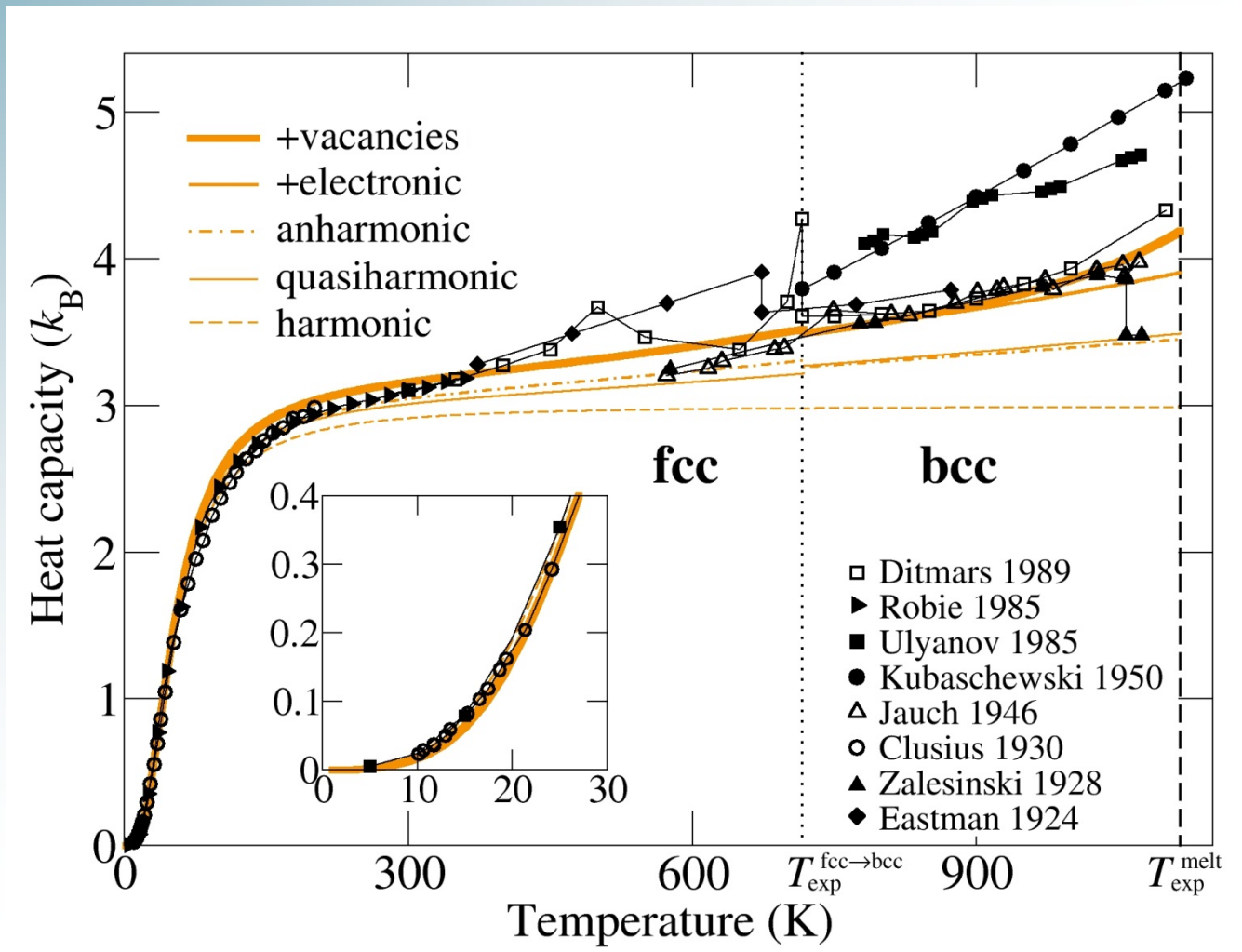
# Calcium: Heat capacity



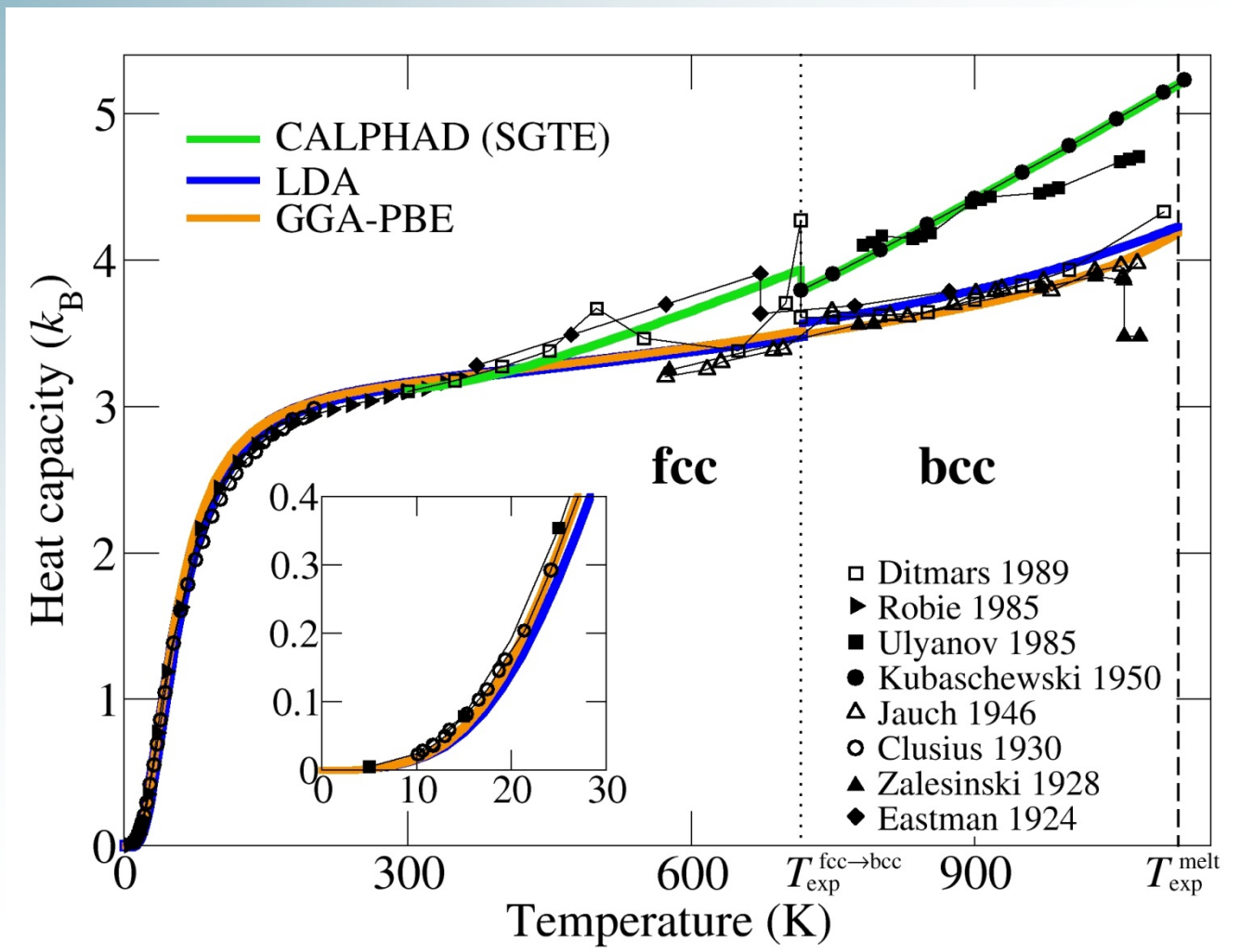
# Calcium: Heat capacity

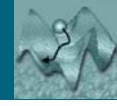


# Calcium: Heat capacity



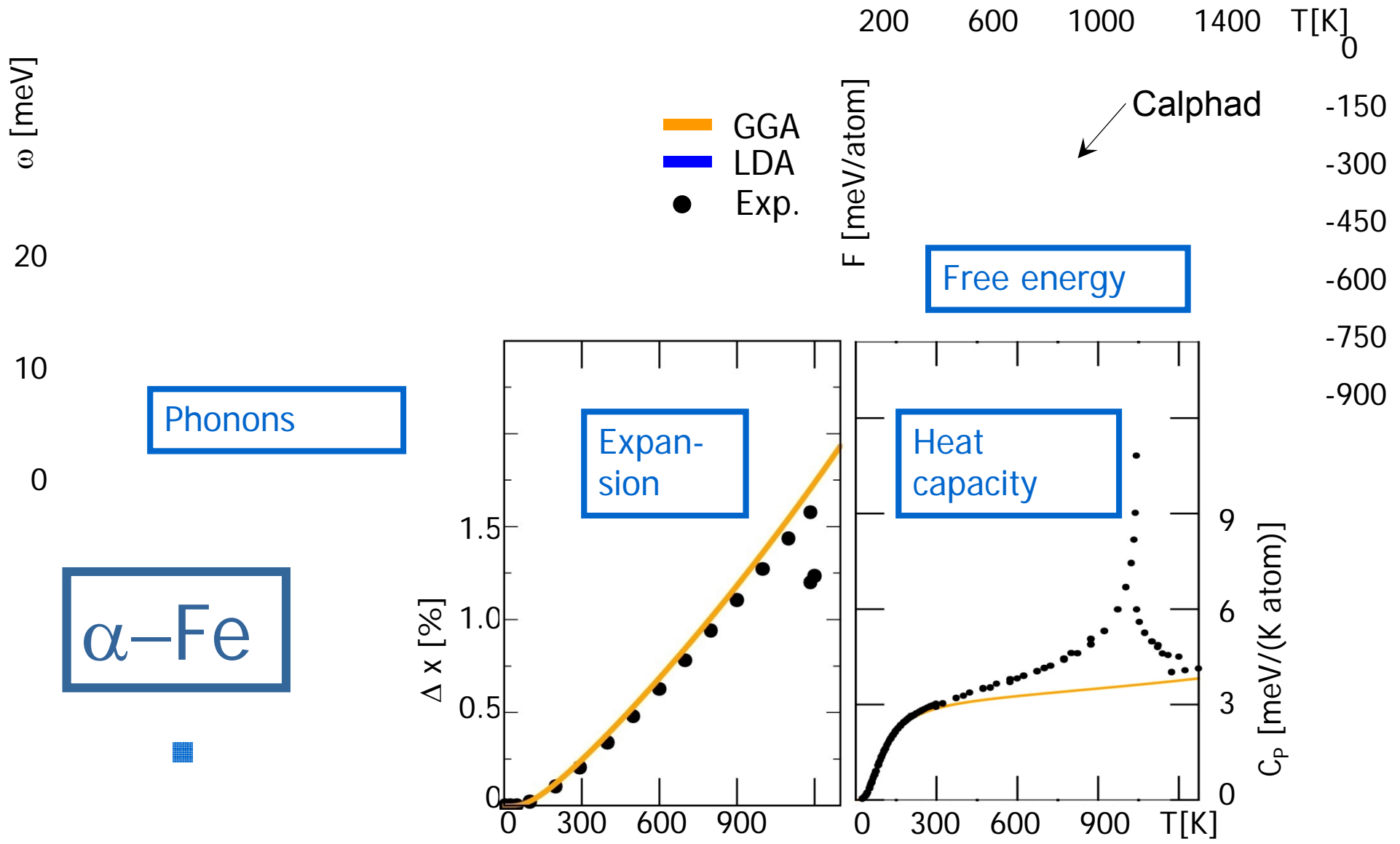
# Calcium: Heat capacity





# Magnetic Excitations

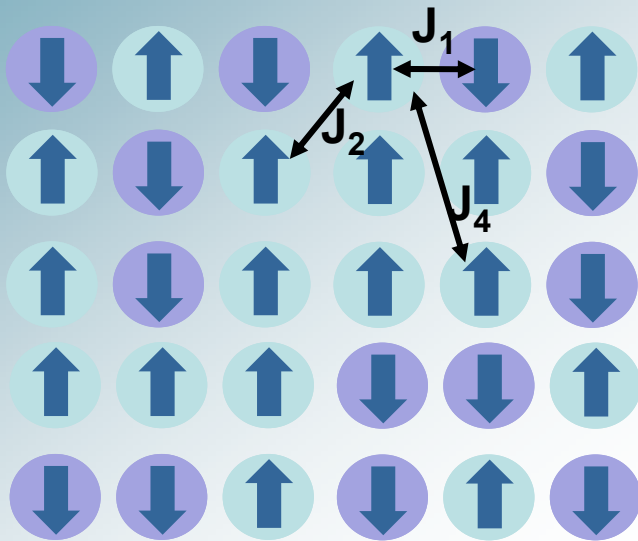




# Magnetic Excitations



$T > 0K$

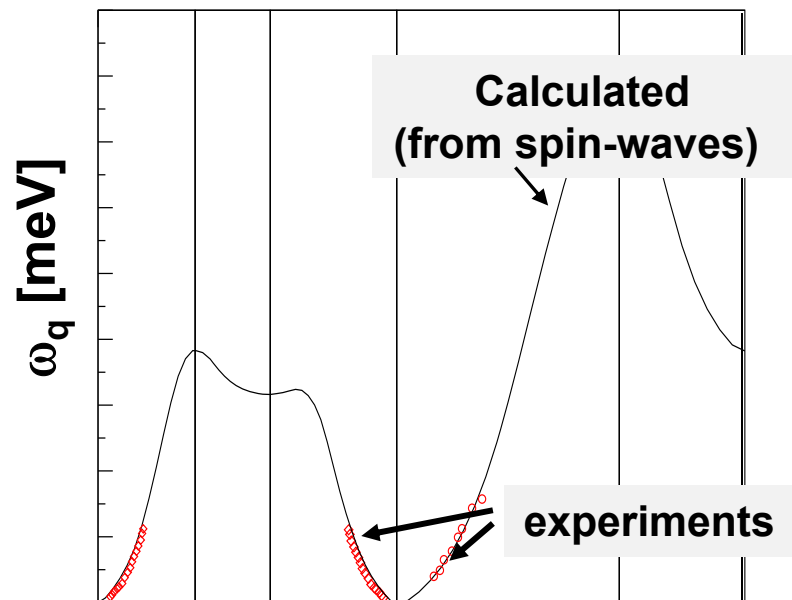


Construct spin Hamiltonian  
(e.g. Heisenberg model)

$$H = - \sum_{i,j} J_{i,j} \mathbf{S}_i \cdot \mathbf{S}_j$$

Typically 30 and more neighbors  
have to be included!

Magnons



Koermann, et al., PRB 78, 033102 (2008)

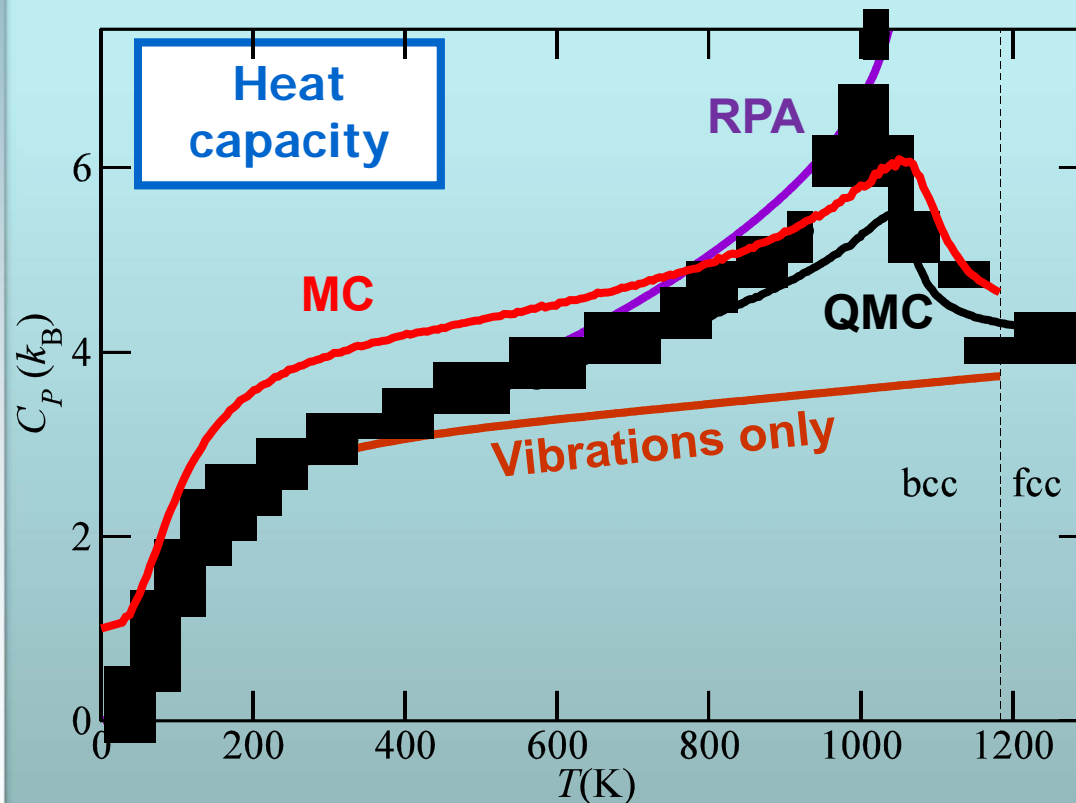
# Finite Temperature Magnetism



Construct and solve magnetic Hamiltonian:  $H^{mag} = -\sum_{i,j} J_{i,j} \mathbf{S}_i \cdot \mathbf{S}_j$

→ from DFT  
→ >30 neighbors

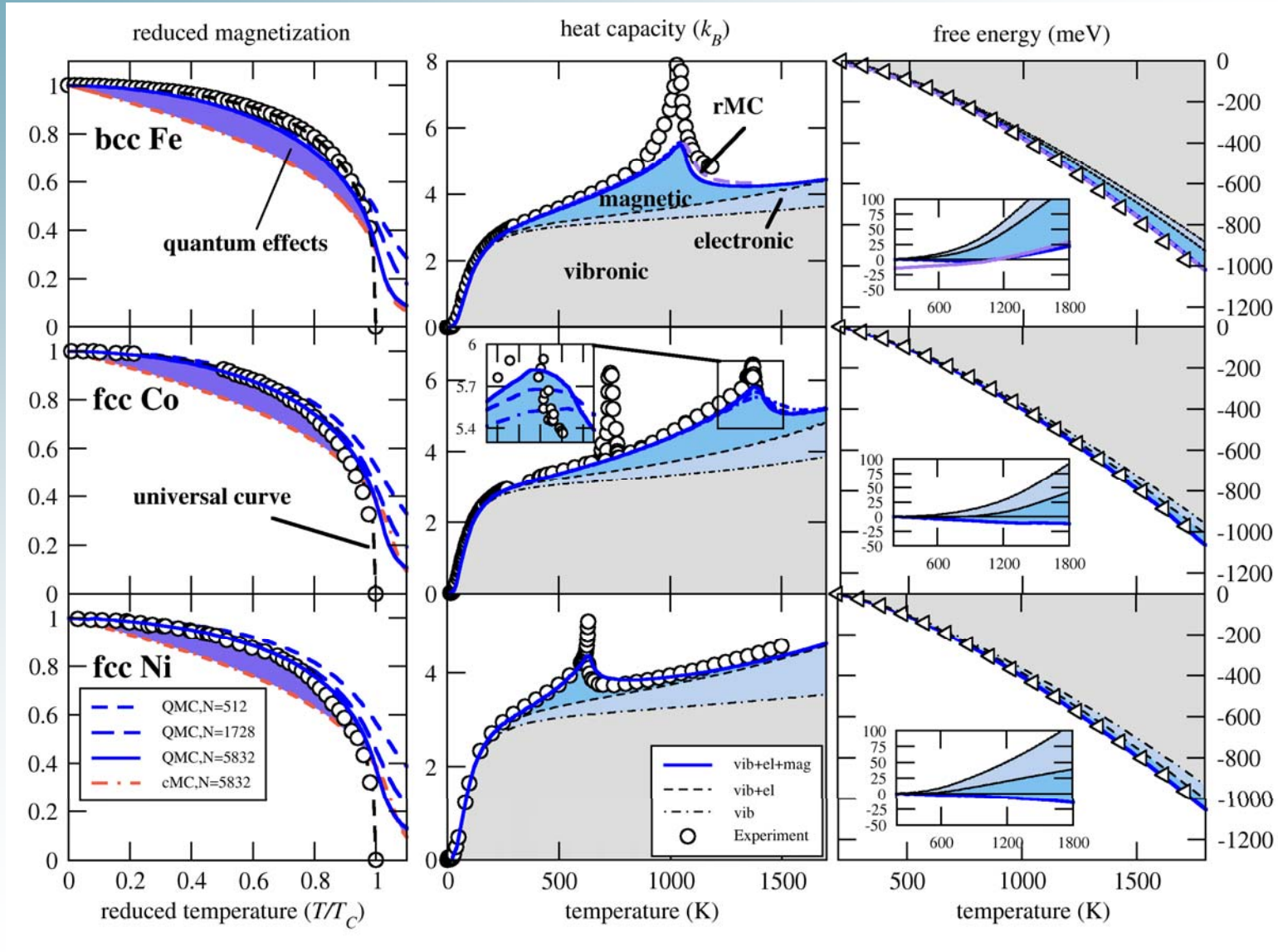
Example: Heat capacity of  $\alpha$  iron



Hierarchy of approaches:

- classical Monte-Carlo (**MC**):  
→ fails below  $T_C$
- analytical approach (**RPA**):  
→ only  $<T_C$
- quantum Monte-Carlo (**QMC**):  
→ map on eff.  $H^{mag}$   
→ works everywhere

# Magnetization, heat capacity, free energies

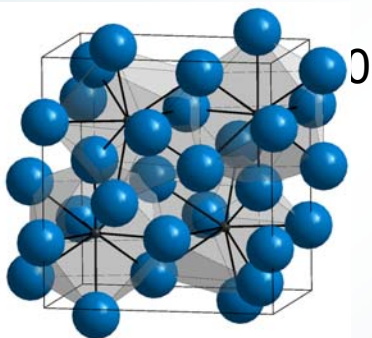
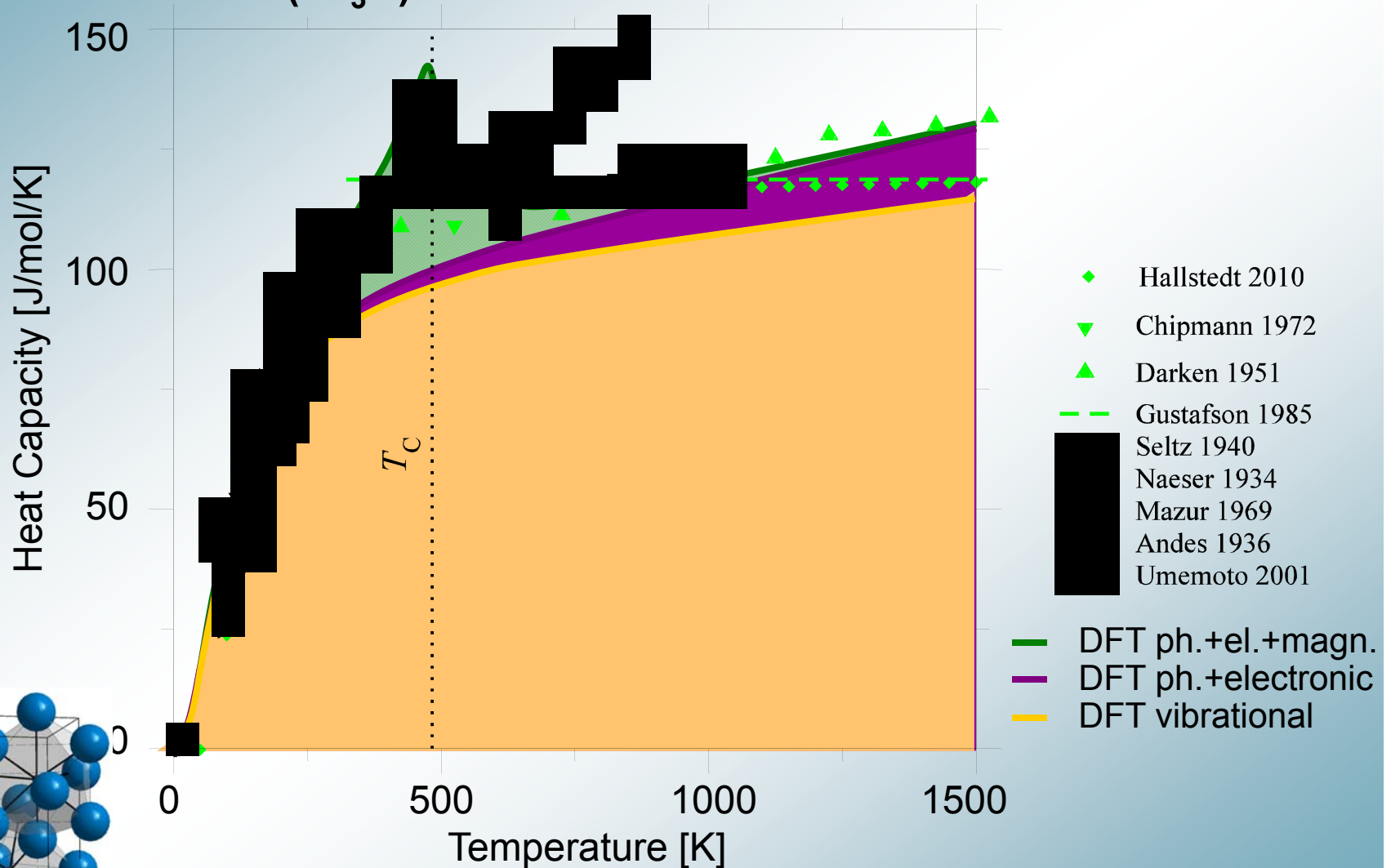


F. Körmann, A. Dick, T. Hickel, and J. Neugebauer, Phys. Rev. B 83, 165114 (2011)

# Heat capacity of alloys



## Example: Cementite ( $\text{Fe}_3\text{C}$ )



□ G. Naeser, Mitt. Kais.-Wilh.-Inst. Eisenforsch. 16 (1934) 207210.

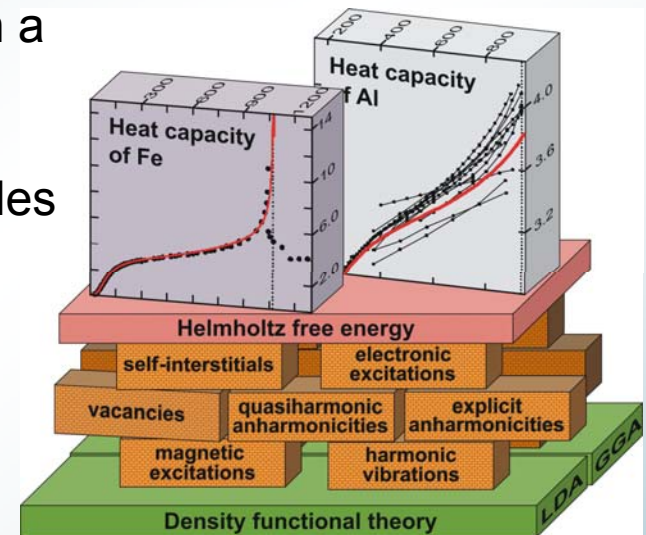


# Conclusions



## Description of T-dependent materials properties by DFT

- Toolset to compute all free energy contributions with a numerical accuracy of  $\sim 1$  meV
- Partitioning into various excitation mechanism provides new insight (e.g. dominating entropy contributions close to melting)
- Accuracy in finite temperature effects often close to or even beyond experiment



## Powerful basis to design multiscale approaches for real-world materials

- Steel design, understanding/overcoming failure mechanisms (e.g. H embrittlement)
- predicting experimentally not accessible phase transitions
- Other materials systems: Lightweight alloys, semiconductor surfaces, phase change and shape memory alloys, biological nanocomposites etc.