



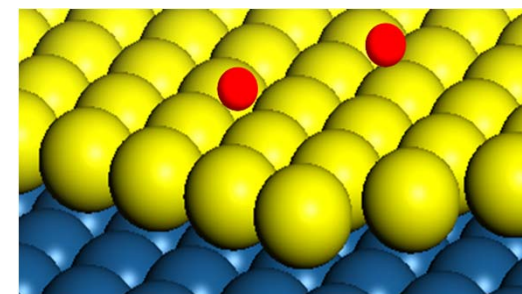
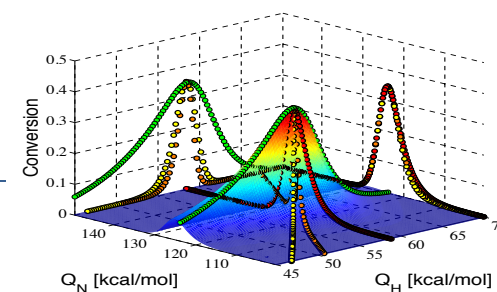
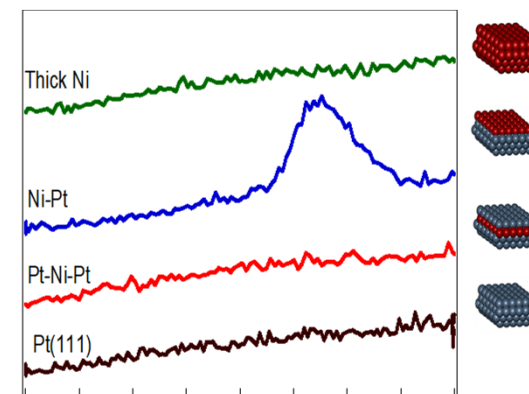
In Silico Prediction of Materials for Energy Applications

Dion Vlachos

June 27, 2018

University of Delaware

Department of Chemical and Biomolecular Engineering
Catalysis Center for Energy Innovation (CCEI),
an Energy Frontier Research Center



*CCEI is an Energy Frontier Research Center funded by the
U.S. Department of Energy, Office of Basic Sciences*





Outline

- Energy, sustainability, and technological needs
- Grand computational challenges
- Topics
 - Catalyst prediction
 - Uncertainty
 - Design of active site
 - Materials gap
- Outlook

CATALYSIS

Synthesis,

multiscale modeling and

***in situ* characterization**

form the pillars of modern

catalytic science



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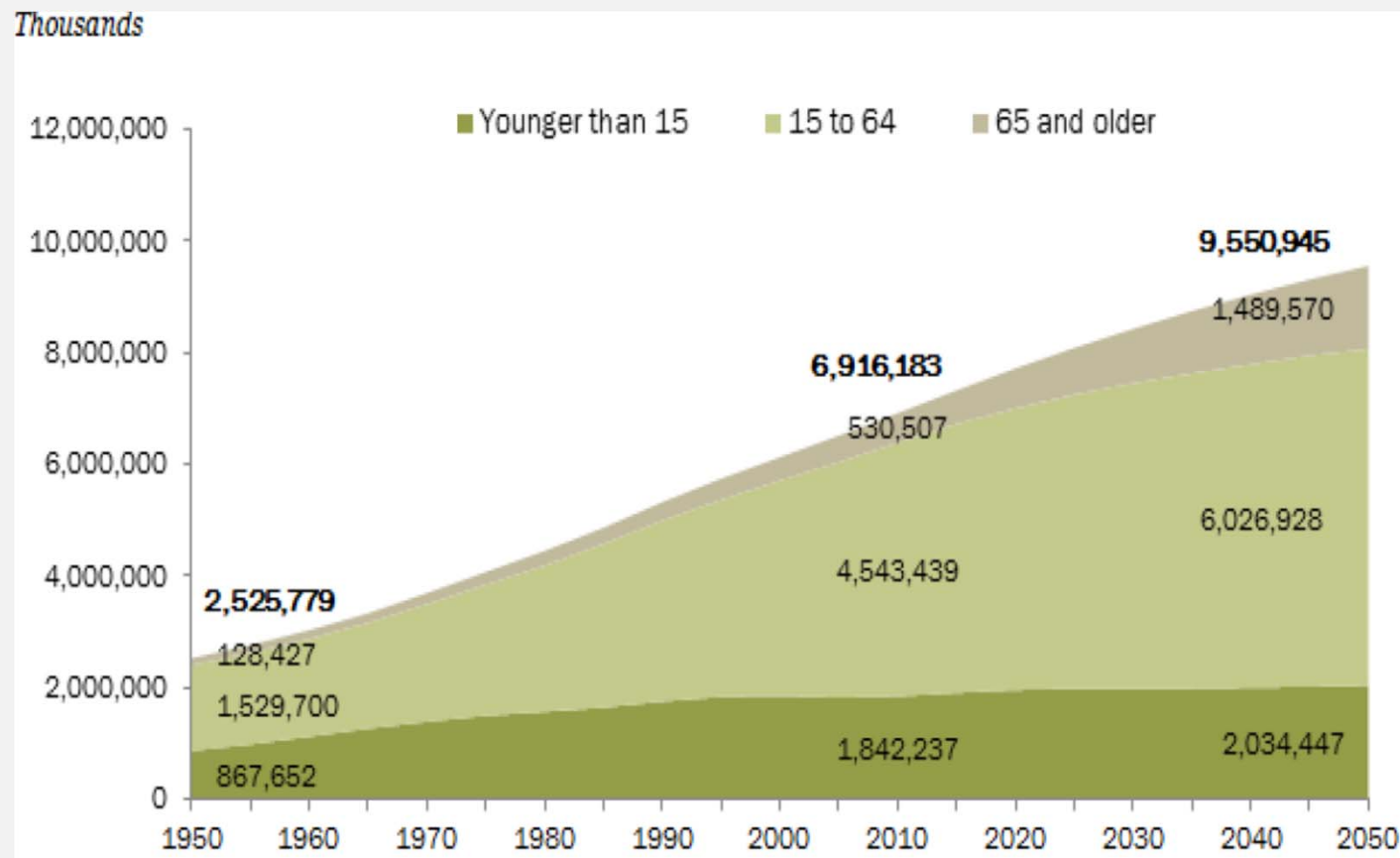
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Societal Challenges

Growing and Aging Population



Human needs

- Medicine
- Food
- Energy
- Water

The nexus of food-energy-water requires **more and cleaner** energy and chemicals



Meeting Tomorrow's Renewable Energy and Sustainability Challenges

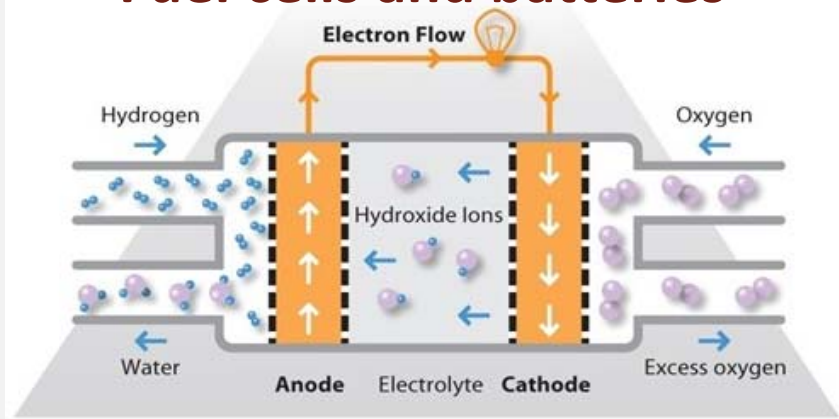
CO₂, CH₄ activation/RAPID



Efficiency/RAPID



Fuel cells and batteries



www.fuelcelltoday.com/technologies/afc

Renewables





Meeting Tomorrow's Renewable Energy and Sustainability Challenges **Requires**

- New materials (catalysts, adsorbents, membranes)
 - Emerging feedstocks: shale gas, biomass, biogas, food waste, CO₂, water,...
 - Modular manufacturing – portable and distributed processing
 - Clean energy, e.g., from solar or wind, beyond electricity
 - ✓ Efficient electrification of chemical plants (H₂ production, purposely tuned processes, localized heating, bond activation)
-
- **Diversified and distributed feedstocks and renewable energy drive a paradigm shift**



Outline

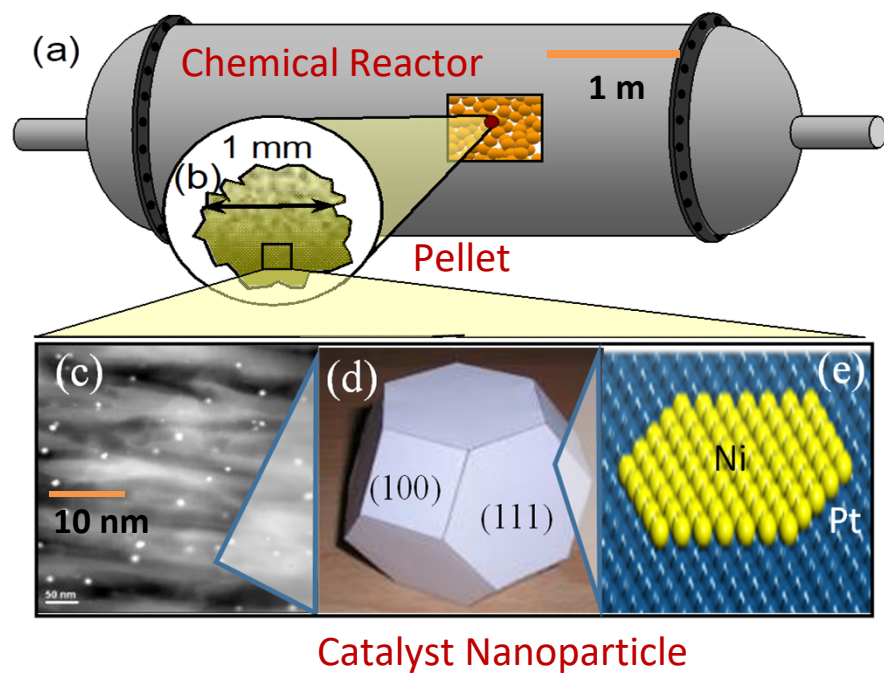
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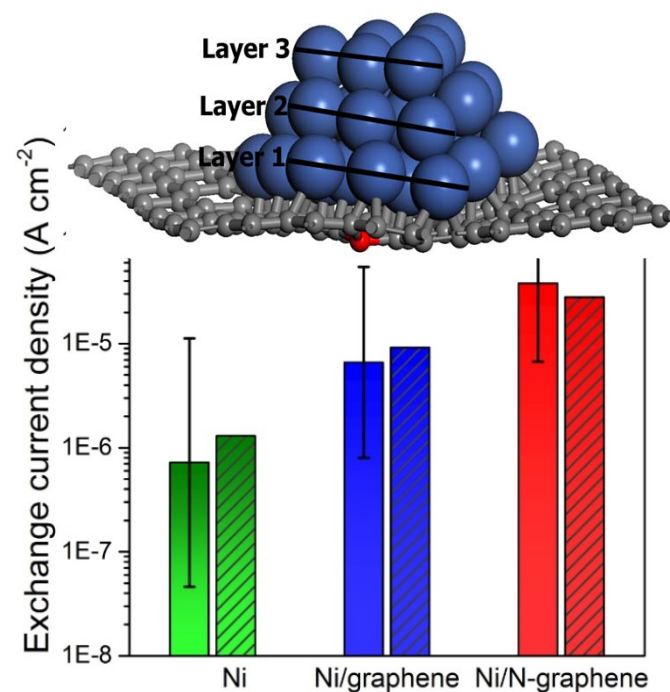
Multiscaling

The Multiscale Challenge[^]



9 orders disparity in scales
Quantum modeling inadequate

The Mesoscale Challenge*



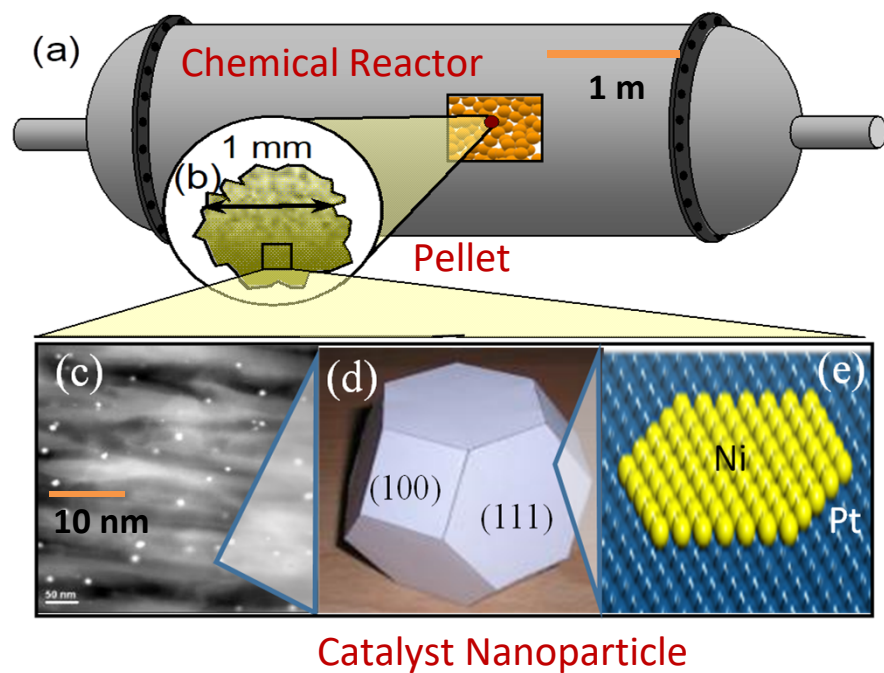
1.5 orders performance enhancement
Support and dopant effects important

[^]Vlachos, *Adv. Chem. Eng.* **30**, 1 (2005);

*Zhuang et al., *Nature Comm.* **7** (2016); 10.1038/ncomms10141

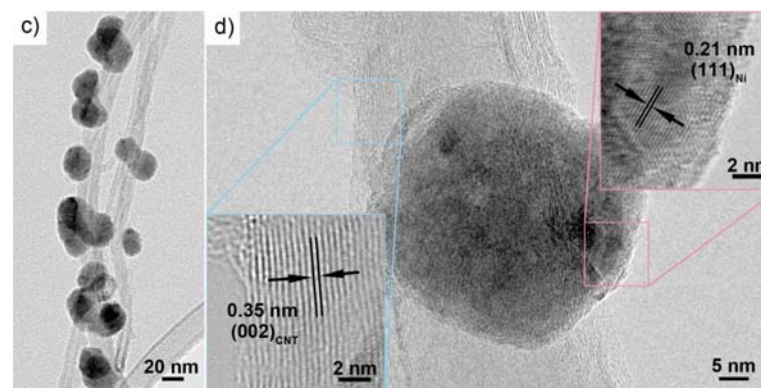
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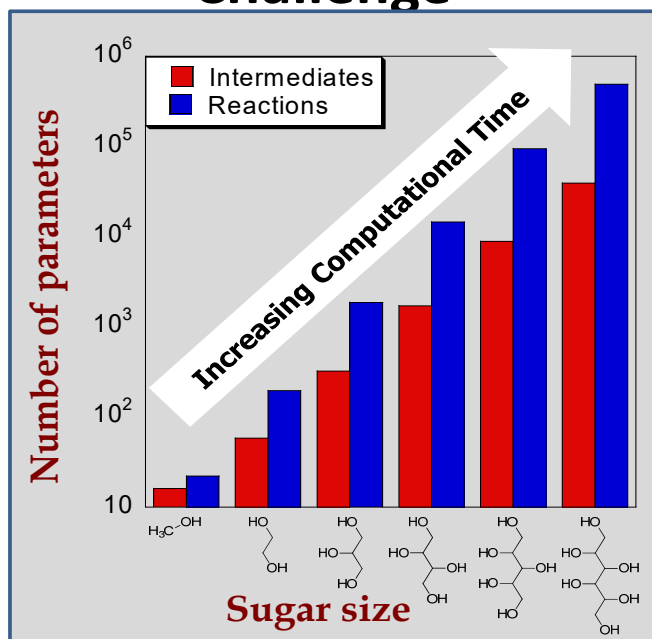
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Parametric Complexity

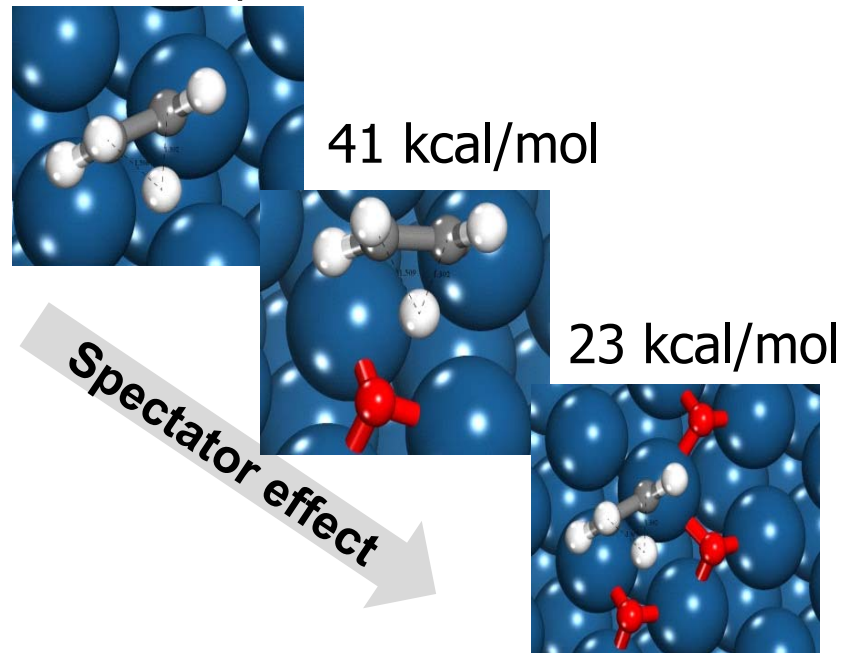
Explosion in Parameters Challenge



1 million chemical reactions
Quantum calculations infeasible

Coverage-Effect Challenge

54 kcal/mol



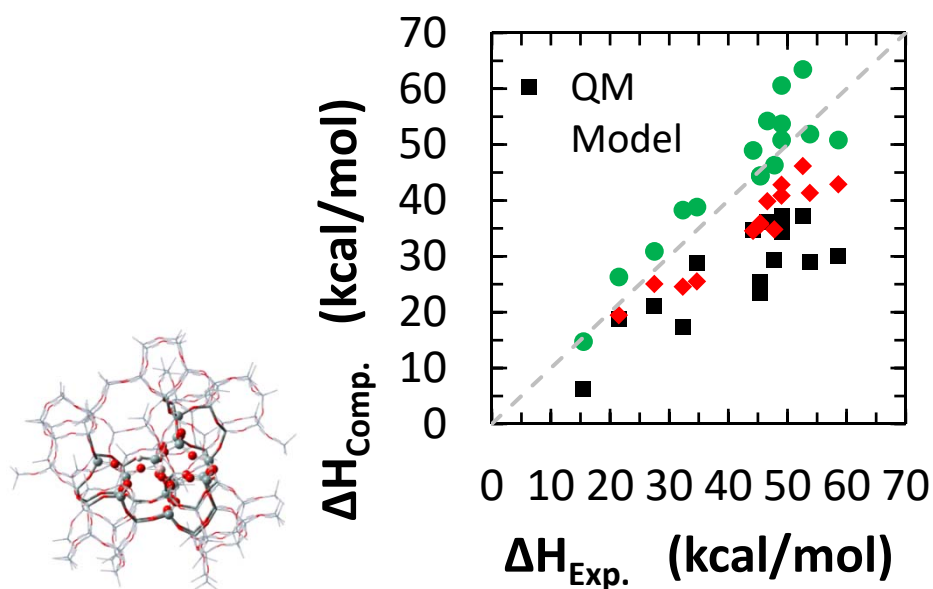
Combinatorial no. of configurations
Quantum calculations infeasible



The **Predictive** Ability of Models Is Unclear

Parametric Uncertainty

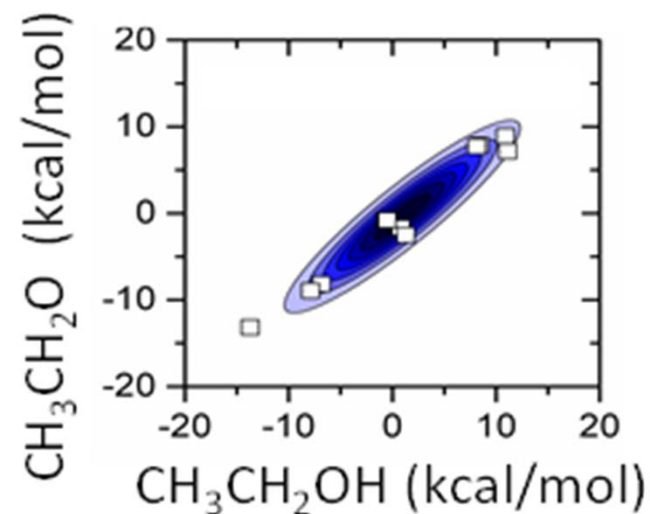
The DFT-Error Challenge[^]



Adsorption in H-MFI zeolite

DFT errors are not well understood
How good are our models?

The Correlative Data and Uncertainty Challenge^{*}



Energies in ethanol mechanism

DFT data is correlated
What's the role in predictions?

[^]Patet, Caratzoulas, and Vlachos, *Phys. Chem. Chem. Phys.* **18**(37), 26094 (2016).

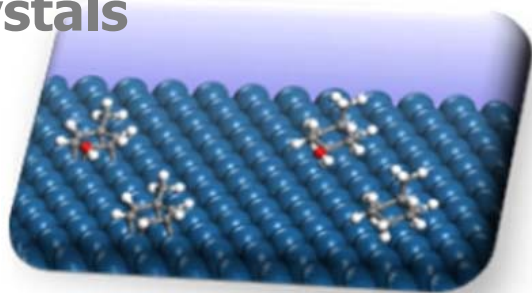
^{*}Sutton, Guo, Katsoulakis, and Vlachos, *Nat. Chem.* **8**(4), 331 (2016).



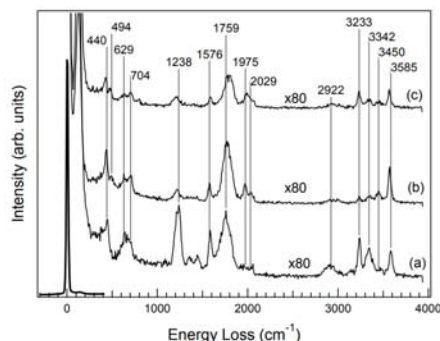
Materials Gap

Model Form Uncertainty

Models Catalysts Single Crystals



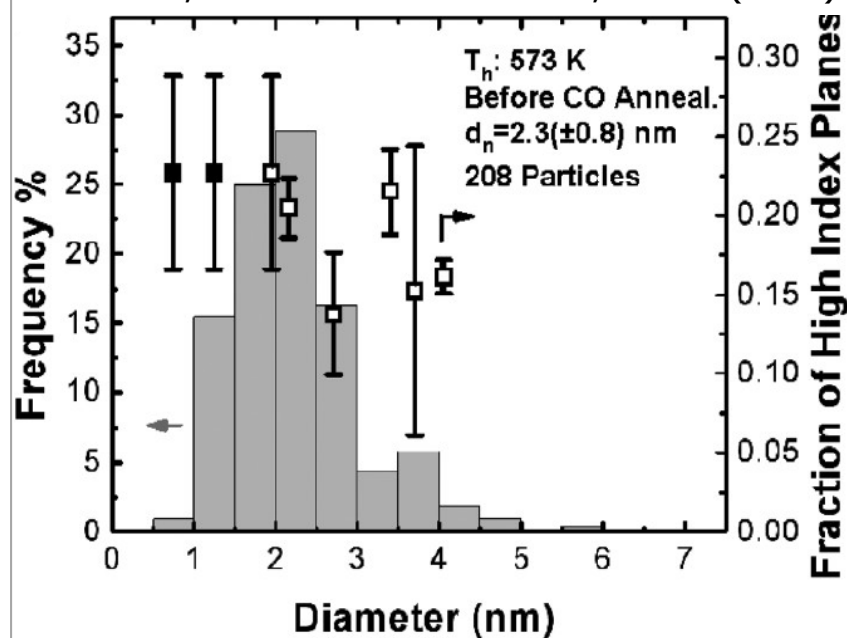
UHV
STM
TPD
HREELS



Model studies on ideal catalysts and far from realistic conditions

Real Catalysts Are Complex

Lee et al., *J. Am. Chem. Soc.* **131**, 15669 (2009)



Multicomponent, heterogeneous, unobserved, multi-oxidation states

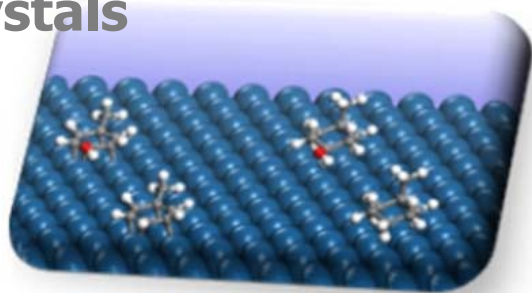
Ni/Pt: Tupy et al., *ACS Catal.* **2**, 2290 (2012); Ru/RuO_x: Jae et al., *ChemSusChem* **6**, 1158 (2013); Pt/WO_x: Wang et al., *Catal. Lett.* **148**, 1047 (2018); Pt SAC/TiO₂: In preparation.



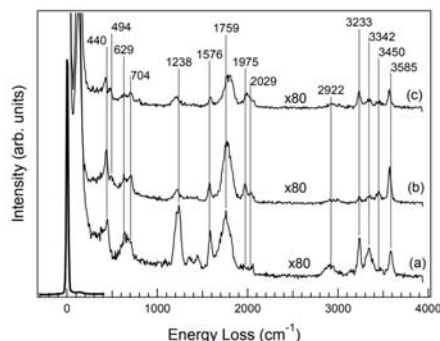
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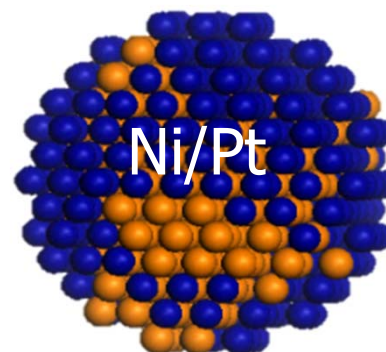
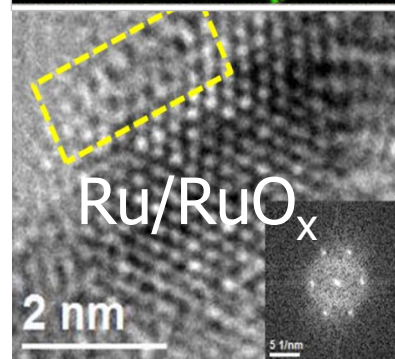
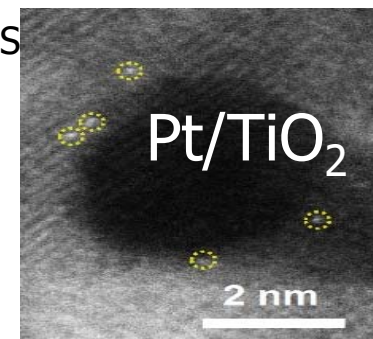
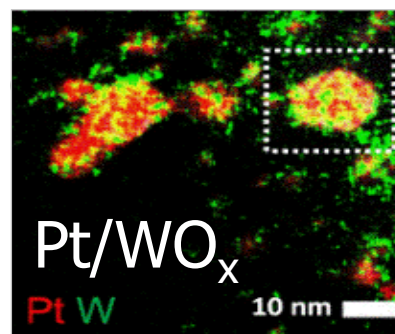


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Predicting Novel Catalytic Materials

- First principles methods (DFT) promise to deliver strategies for rational catalyst design
- Current studies limited to cases when:
 - Thermodynamics (heat of adsorption) dominates^{1,2}
 - Linear interpolation is employed³
 - Sabatier's principle optimizes catalyst activity (no selectivity)

Sabatier's Principle

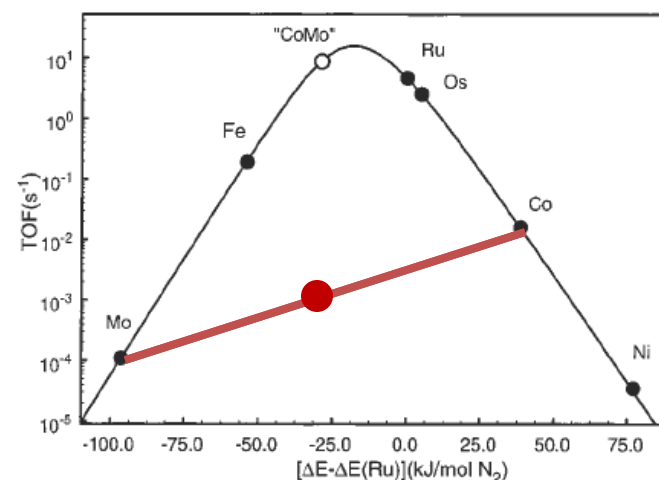
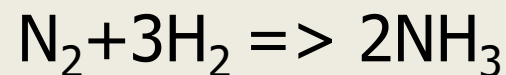


Figure 1. Calculated turnover frequencies for ammonia synthesis as a function of the adsorption energy of nitrogen. The synthesis conditions are 400 °C, 50 bar, gas composition H₂:N₂ = 3:1 containing 5% NH₃.



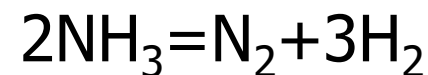
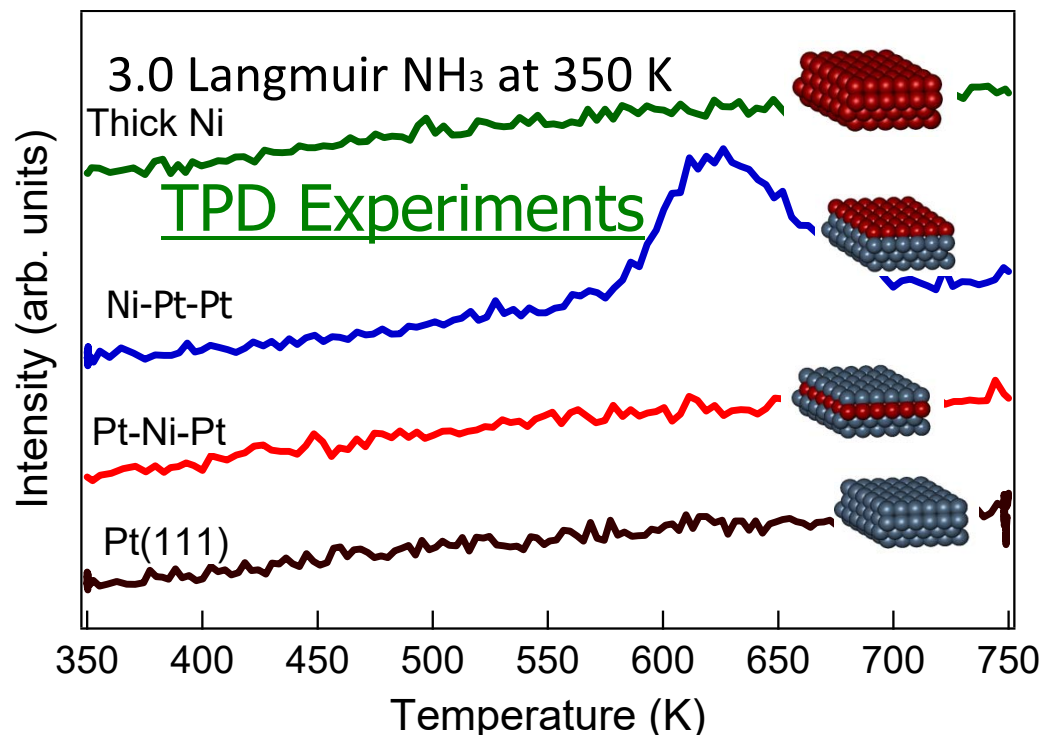
¹ Strasser et al., *J. Phys. Chem. B* **107**(40), 11013 (2003)

² Greeley and Mavrikakis, *Nat. Materials* **3**(11), 810 (2004)

³ Jacobsen et al., *J. Am. Chem. Soc.* **123**, 8404 (2001)



Bimetallics Exhibit Emergent Behavior



- Ni-Pt-Pt is the most active single crystal catalyst known today!
- Interpolation principle is inadequate
- Core-shell bimetallics provide a 'new class of materials' that 'expands the periodic table'

ARTICLES

PUBLISHED ONLINE: 25 APRIL 2010 | DOI: 10.1038/NCHEM.626

nature
chemistry

Using first principles to predict bimetallic catalysts for the ammonia decomposition reaction

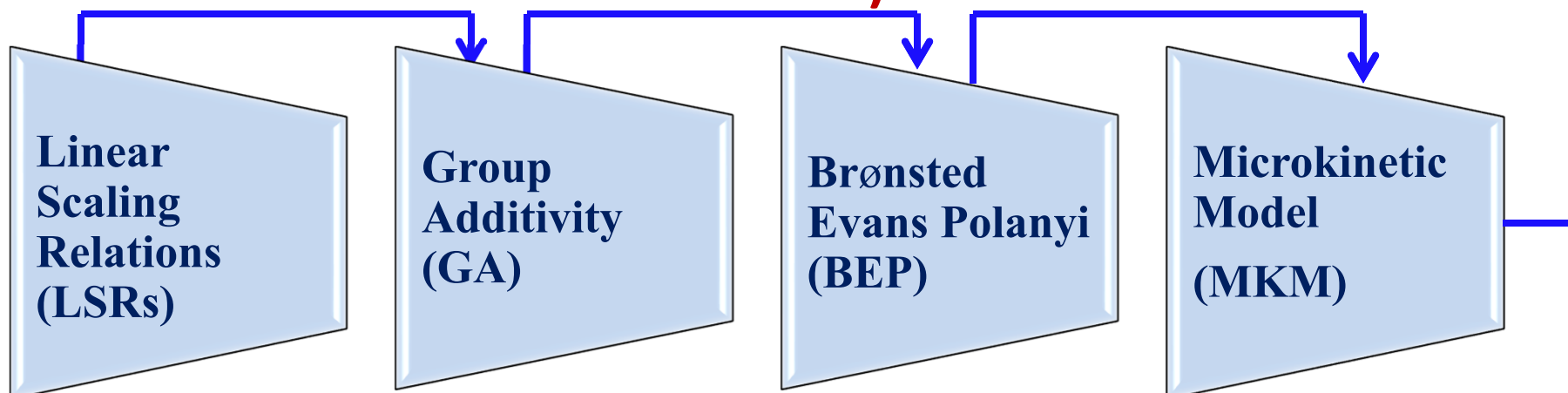
Danielle A. Hansgen, Dionisios G. Vlachos* and Jingguang G. Chen*

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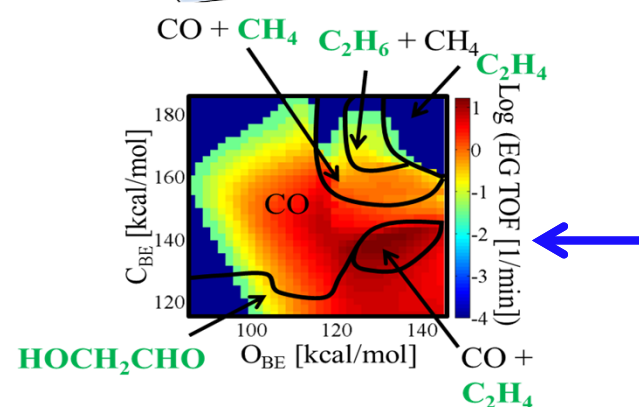


Toward High-throughput Computing:

Extendibility (Periodic Table) and Scalability (Molecular & Network)



- **Hierarchical refinement** allows first-principles accuracy at minimal computation;
- **Machine learning** to develop **correlations** (surrogates) and **descriptor-based models**
- Descriptor-based modeling allows high throughput computing and solving the **inverse problem**

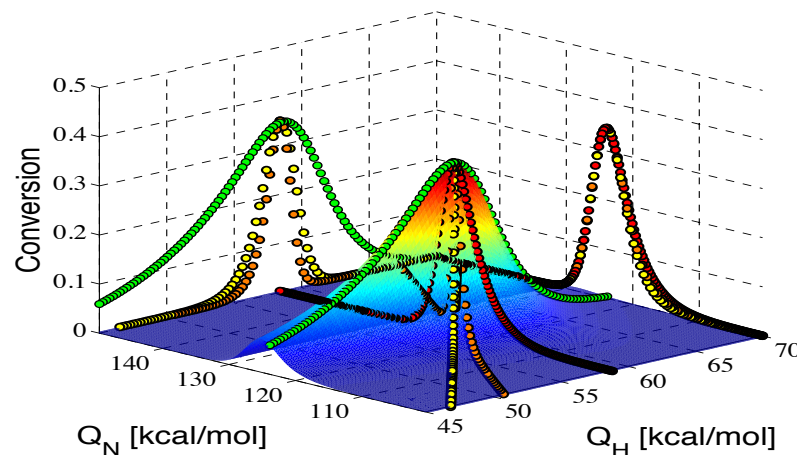


Review: Saliccioli et al., *Chem. Eng. Sci.* **66**, 4319 (2011);

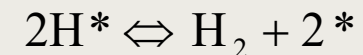
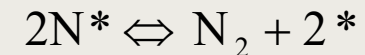
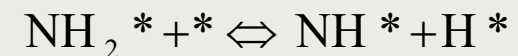
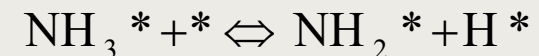
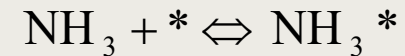
Saliccioli et al., *J. Phys. Chem. C* **114**, 20155 (2010); *J. Phys. Chem. C* **116**, 1873 (2012); Sutton and Vlachos, *ACS Catal.* **2**, 1624 (2012); *J. Catal.* **297**, 202 (2013)

High-throughput Multiscale Model-based Optimization for Catalyst Discovery

350 °C
1 atm



NH₃ decomposition



- Search is done on atomic descriptors while running the full chemistry and reactor models
- Optimal catalyst properties are identified



Toward Predictive Models

Errors, Correlations, **Small Data**,...

Correlative Global Uncertainty Quantification – Ethanol steam reforming

nature
chemistry

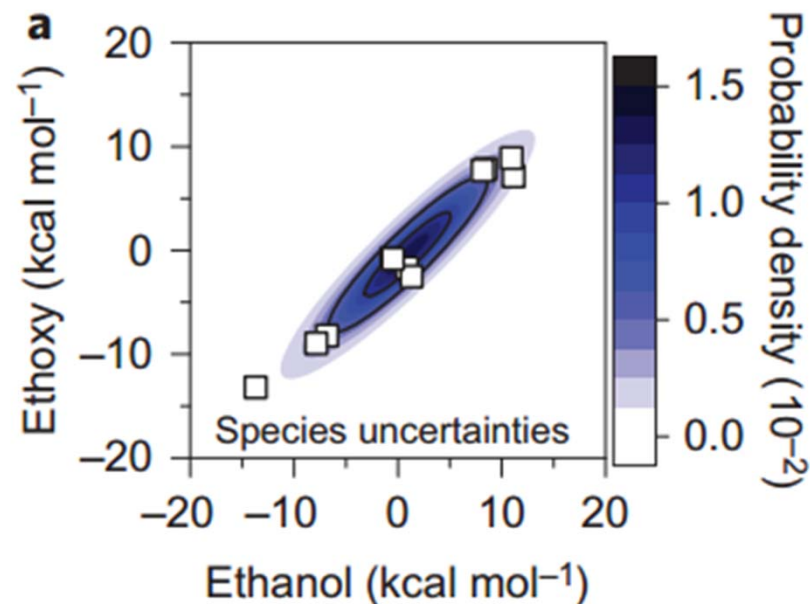
ARTICLES

PUBLISHED ONLINE: 22 FEBRUARY 2016 | DOI: 10.1038/NCHEM.2454

Effects of correlated parameters and uncertainty in electronic-structure-based chemical kinetic modelling

Jonathan E. Sutton¹, Wei Guo^{1,2}, Markos A. Katsoulakis³ and Dionisios G. Vlachos^{1*}

Kinetic models based on first principles are becoming common place in heterogeneous catalysis because of their ability to interpret experimental data, identify the rate-controlling step, guide experiments and predict novel materials. To overcome the tremendous computational cost of estimating parameters of complex networks on metal catalysts, approximate quantum mechanical calculations are employed that render models potentially inaccurate. Here, by introducing correlative global sensitivity analysis and uncertainty quantification, we show that neglecting correlations in the energies of species and reactions can lead to an incorrect identification of influential parameters and key reaction intermediates and reactions. We rationalize why models often underpredict reaction rates and show that, despite the uncertainty being large, the method can, in conjunction with experimental data, identify influential missing reaction pathways and provide insights into the catalyst active site and the kinetic reliability of a model. The method is demonstrated in ethanol steam reforming for hydrogen production for fuel cells.

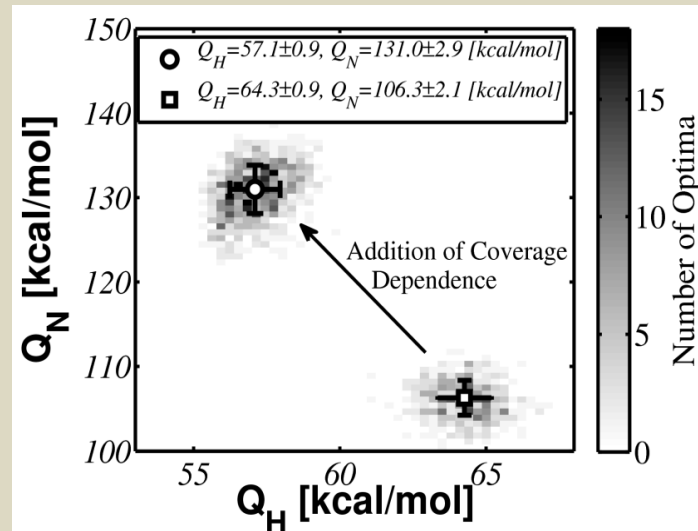




Identifying Bimetallic Catalysts: **Uncertainty**

- Predictions are robust!
- Adsorbate-adsorbate interactions are crucial

Activity (& Selectivity) Maps

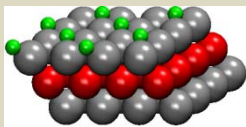
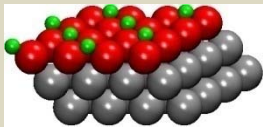


Variation in pre-exponentials

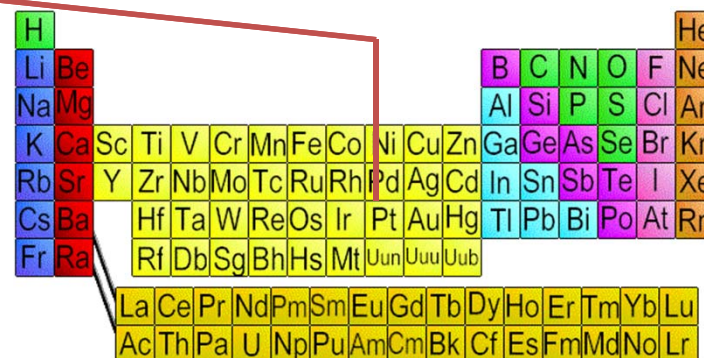
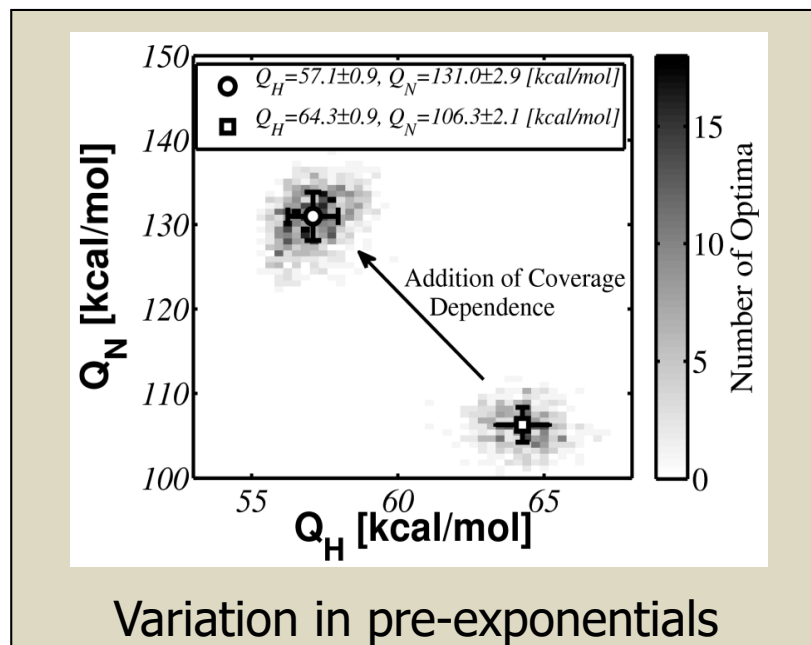


Identifying Bimetallic Catalysts: Uncertainty & Informatatics

Informatatics Databases

| Structure | Metals | BE _N (kcal/mol) |
|---|--------|----------------------------|
| Subsurface Core-shell- shell  | PtTiPt | 56.5 |
| | PtVPt | 59.5 |
| | PtCrPt | 72.6 |
| | PtMnPt | 84.9 |
| | PtFePt | 83.9 |
| | PtCoPt | 87.0 |
| | PtNiPt | 89.8 |
| Surface Core-shell  | CoPtPt | 127.8 |
| | NiPtPt | 130.5 |
| | FePtPt | 137.7 |
| | MnPtPt | 162.2 |
| | CrPtPt | 166.5 |
| | VPtPt | 184.1 |
| | TiPtPt | 191.5 |

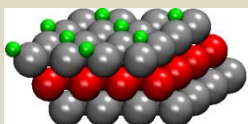
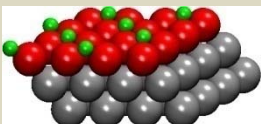
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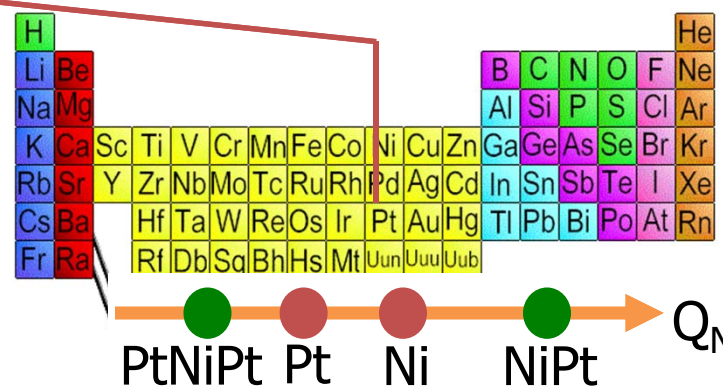
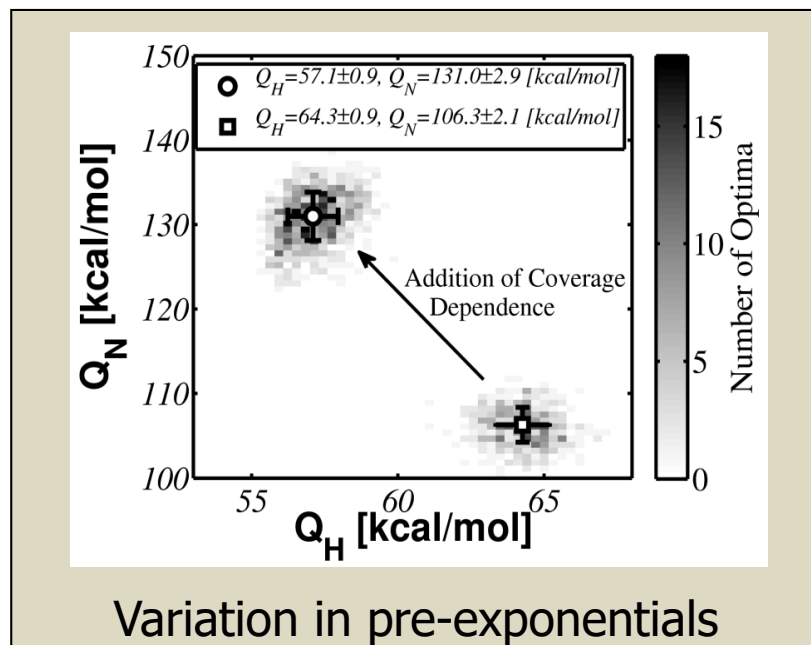


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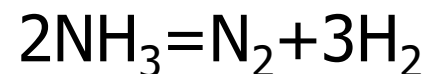
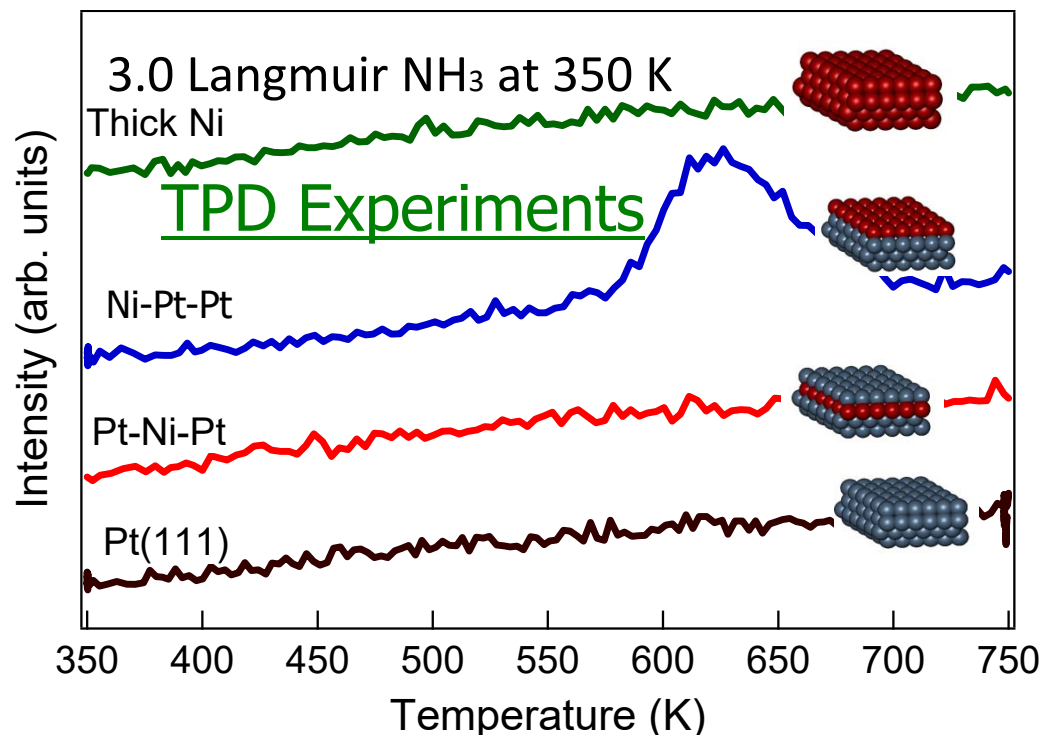
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| | MnPtPt | 162.2 |
| | CrPtPt | 166.5 |
| | VPtPt | 184.1 |
| | TiPtPt | 191.5 |
| | Pt | 102.1 |
| | Ni | 113.8 |

Activity (& Selectivity) Maps





Emergent Behavior



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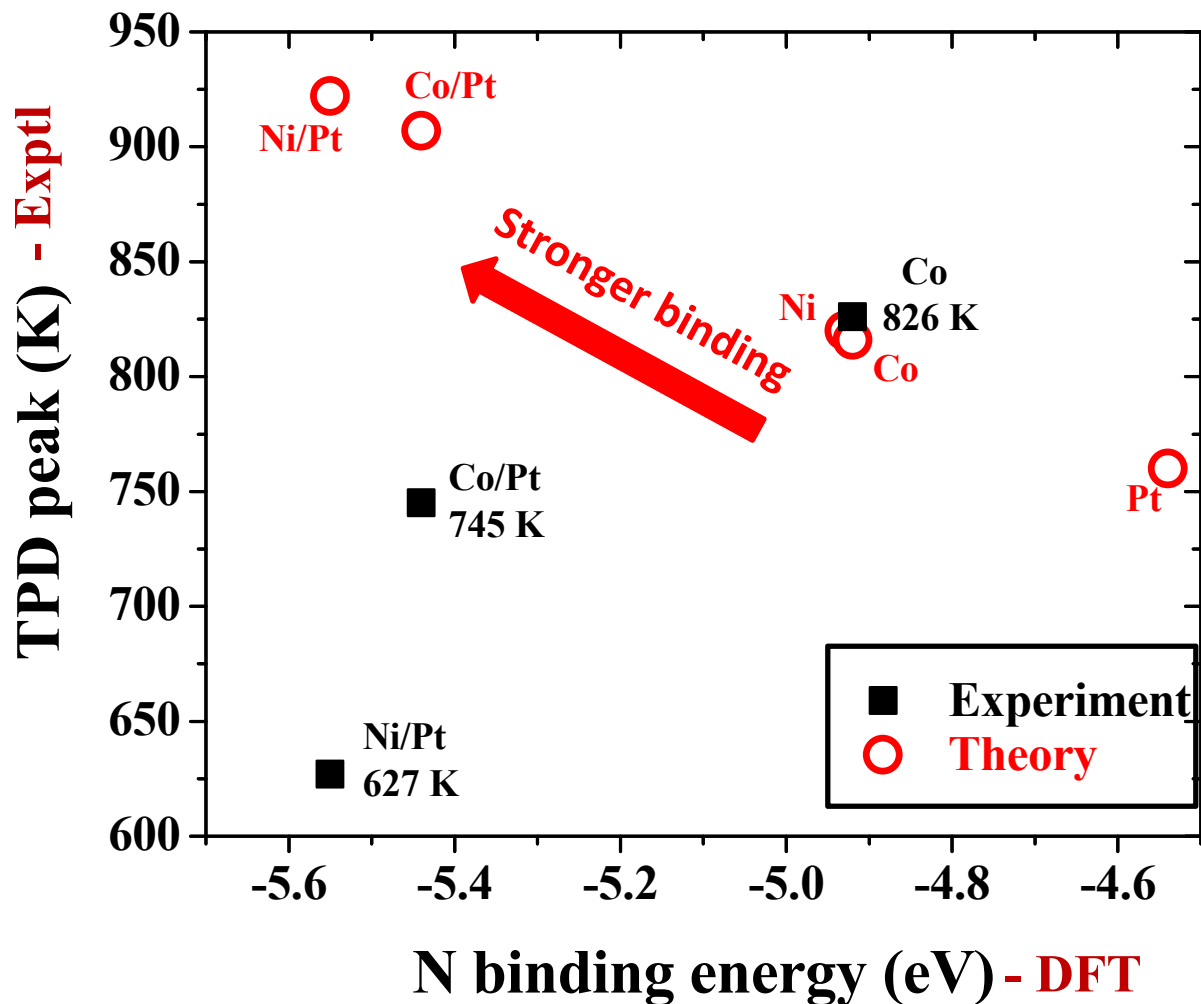
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Typical Descriptor Fails to Describe Experiments



$$-\frac{d\theta_N}{dT} = \frac{k_0}{\beta_H} \theta_N^2 \exp(-E_a(\theta_N)/(k_B T))$$

- **Binding energy** on terrace sites is **not a good descriptor**, especially for bimetallic surfaces



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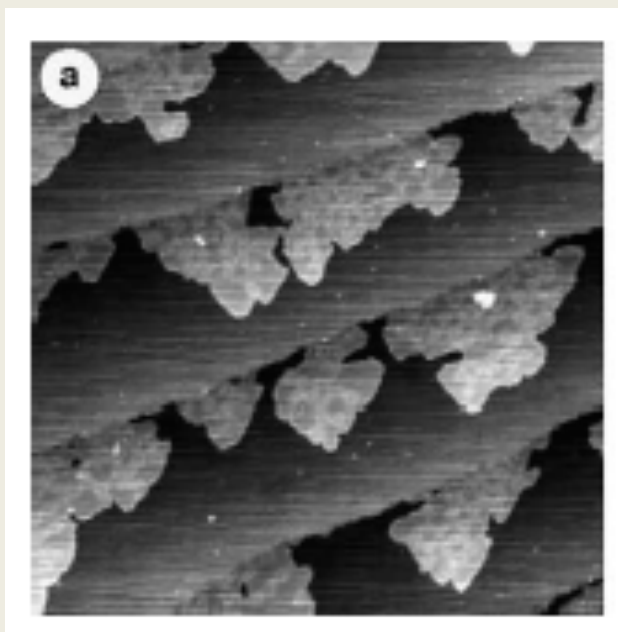
catalytic science



Microstructure of Core-Shell Catalysts

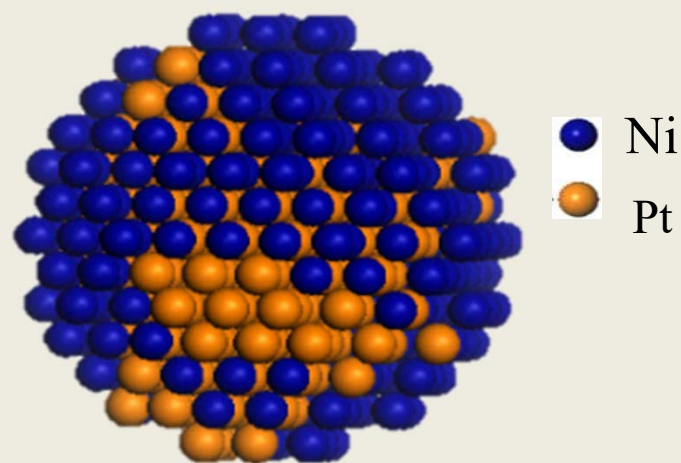
➤ **Microstructure** consists of a defected monolayer

STM image



Kitchin et al., *Surf. Sci.* **544**, 295,
(2003)

Reconstruction (in situ
EXAFS data)

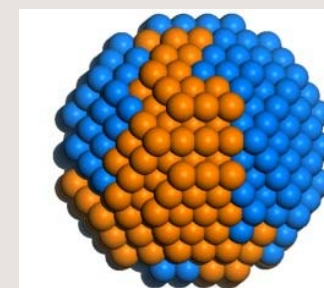
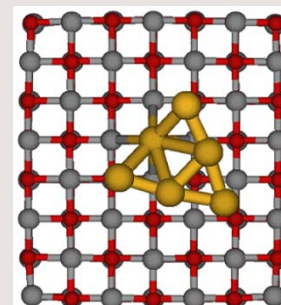
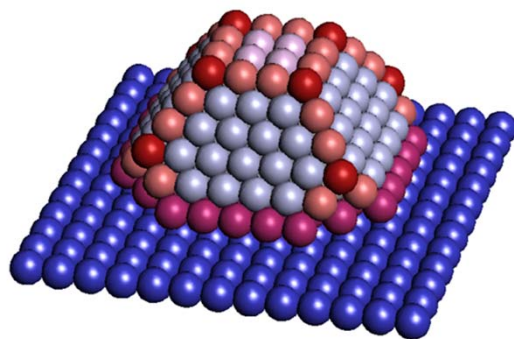
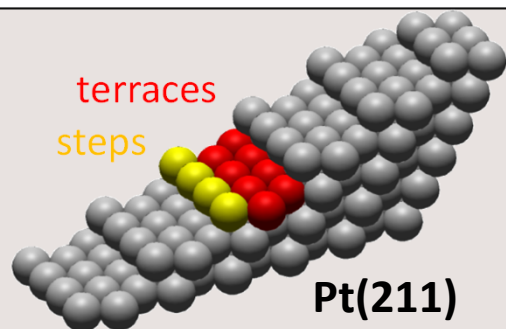
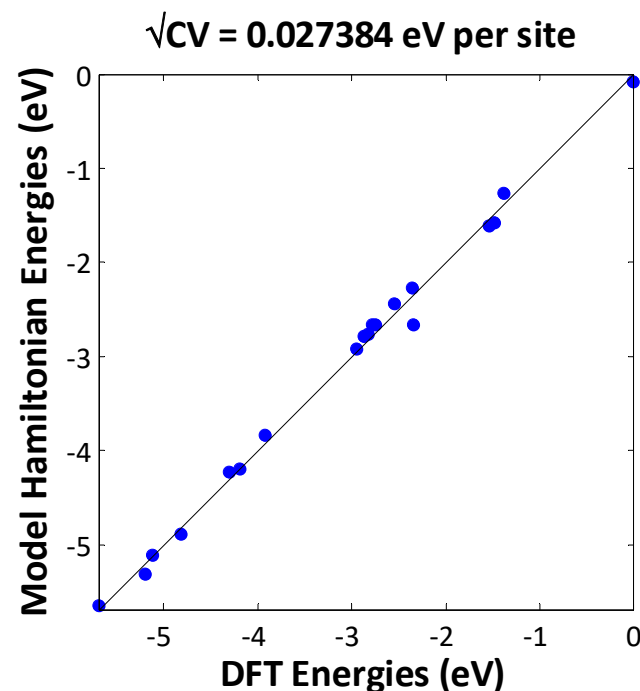


Tupy et al., *ACS Catal.* **2**, 2290
(2012)



Features of Graph-Theoretical KMC

- Complex chemistries
- Accurate lateral interactions – Cluster expansion
- **Microscopic reversibility and thermodynamic consistency**
- **Fast algorithms**
- **Multiple facets, nanoparticles, clusters, multifunctional materials**

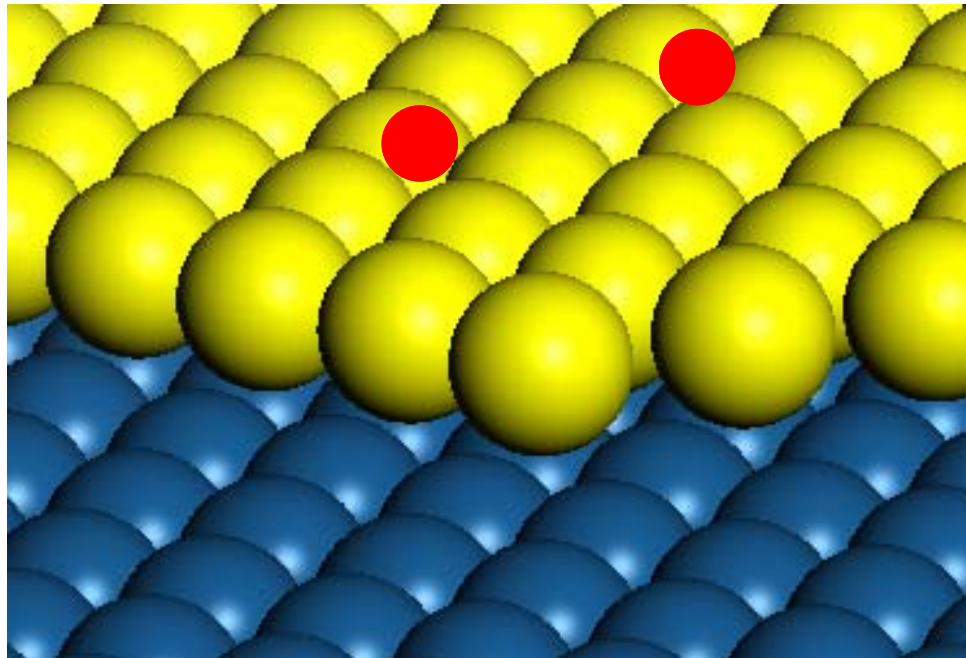


Ni/Pt - EXAFS

Reviews: Chatterjee and Vlachos, *J. Comp.-Aided Mat. Design* **14**, 253 (2007); Stamatakis and Vlachos, *ACS Catal.* **2**, 2648 (2012).

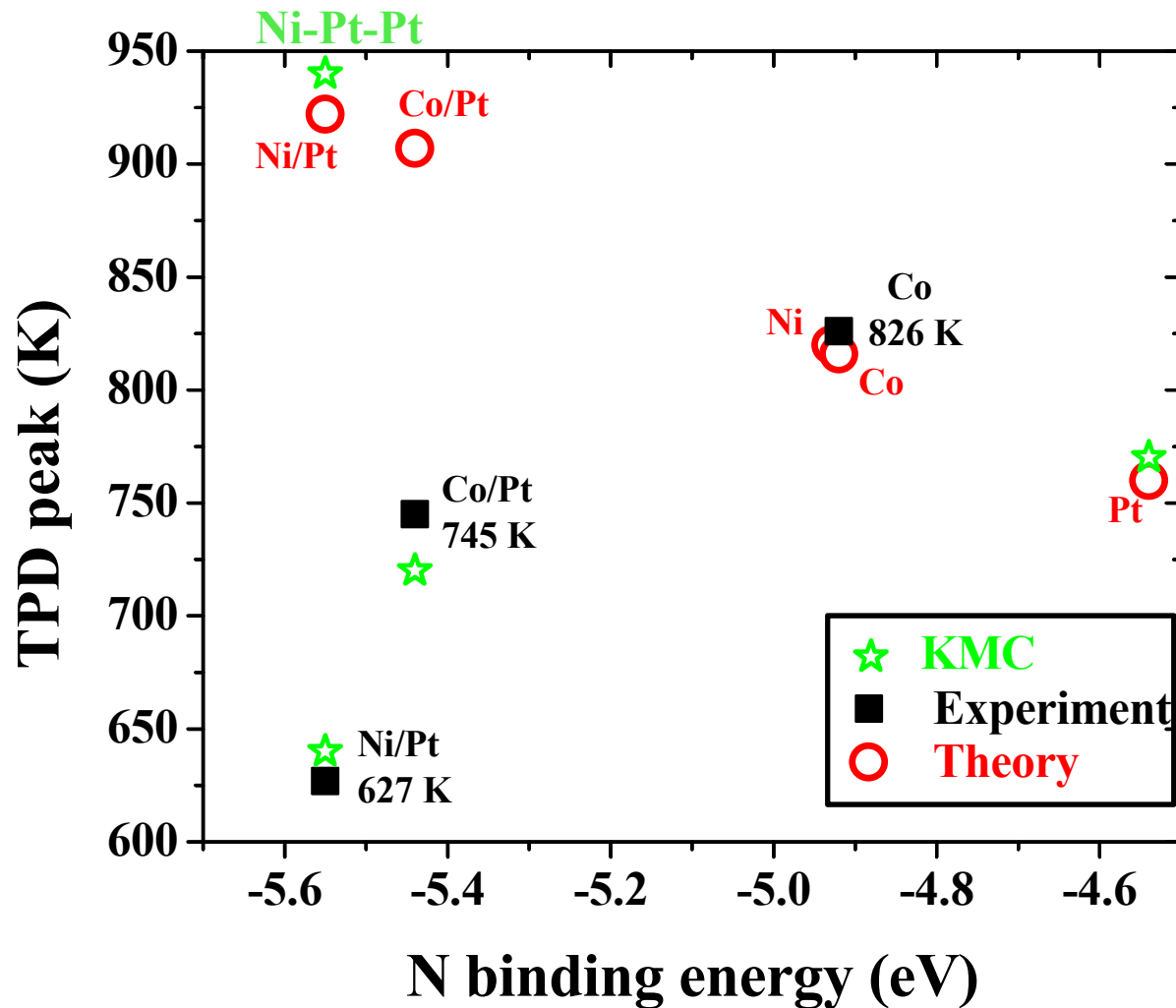


A Minimum Energy Path: Defected Bimetallics



- Ni/Pt serves as a reservoir of N due to strong binding
- N diffuses at interfacial sites invoking steps and Pt terraces and associates and desorbs from there

First-Principles KMC Resolves Structure Sensitivity



- Flat surfaces: strong N binding results in high desorption temperature, well described by Redhead theory
- Steps are responsible for the low N₂ desorption temperature on Ni/Pt and Co/Pt surfaces



Microstructure Can Profoundly Impact Activity



ARTICLE

Received 17 Jan 2015 | Accepted 12 Sep 2015 | Published 7 Oct 2015

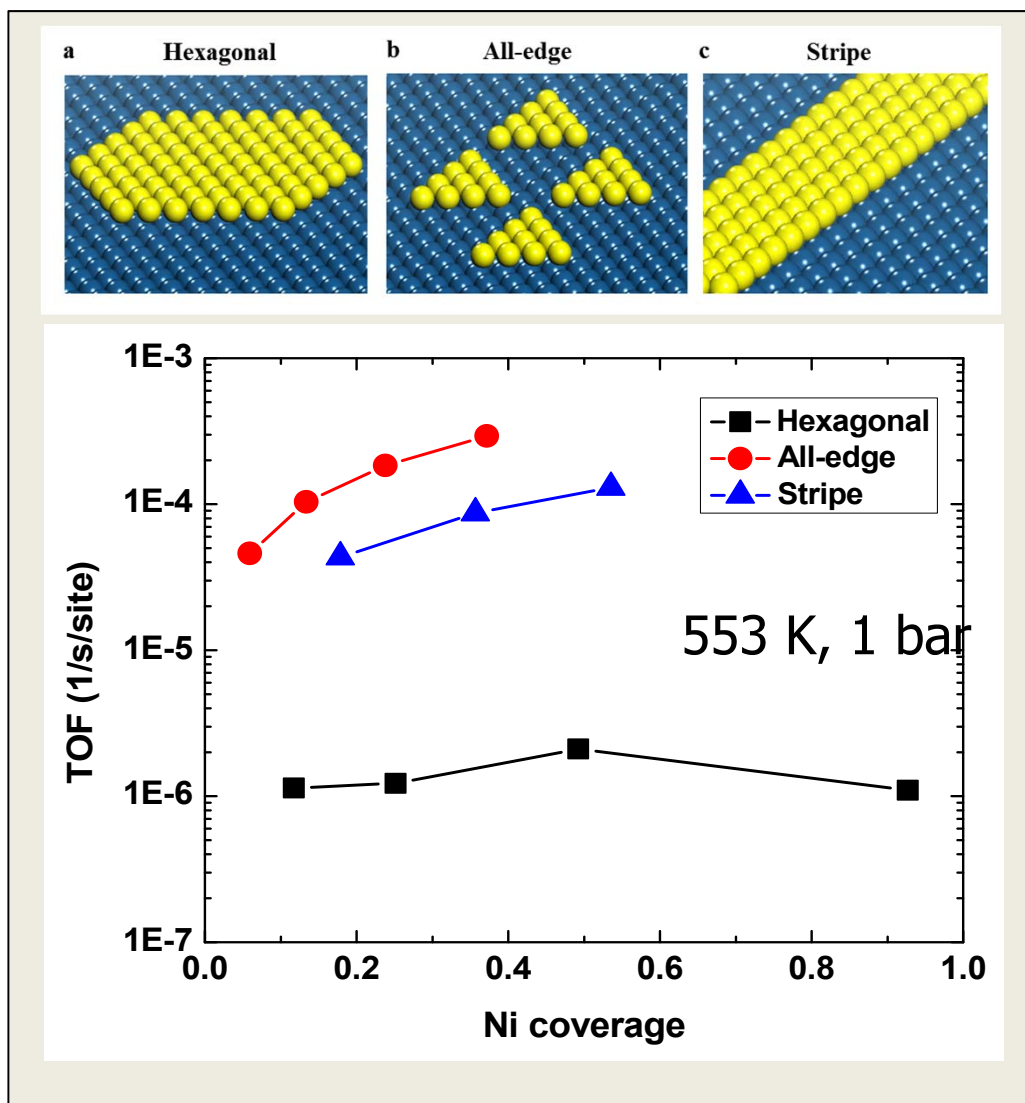
DOI: 10.1038/ncomms9619

OPEN

Patched bimetallic surfaces are active catalysts for ammonia decomposition

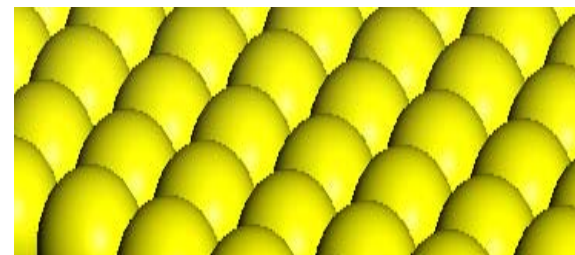
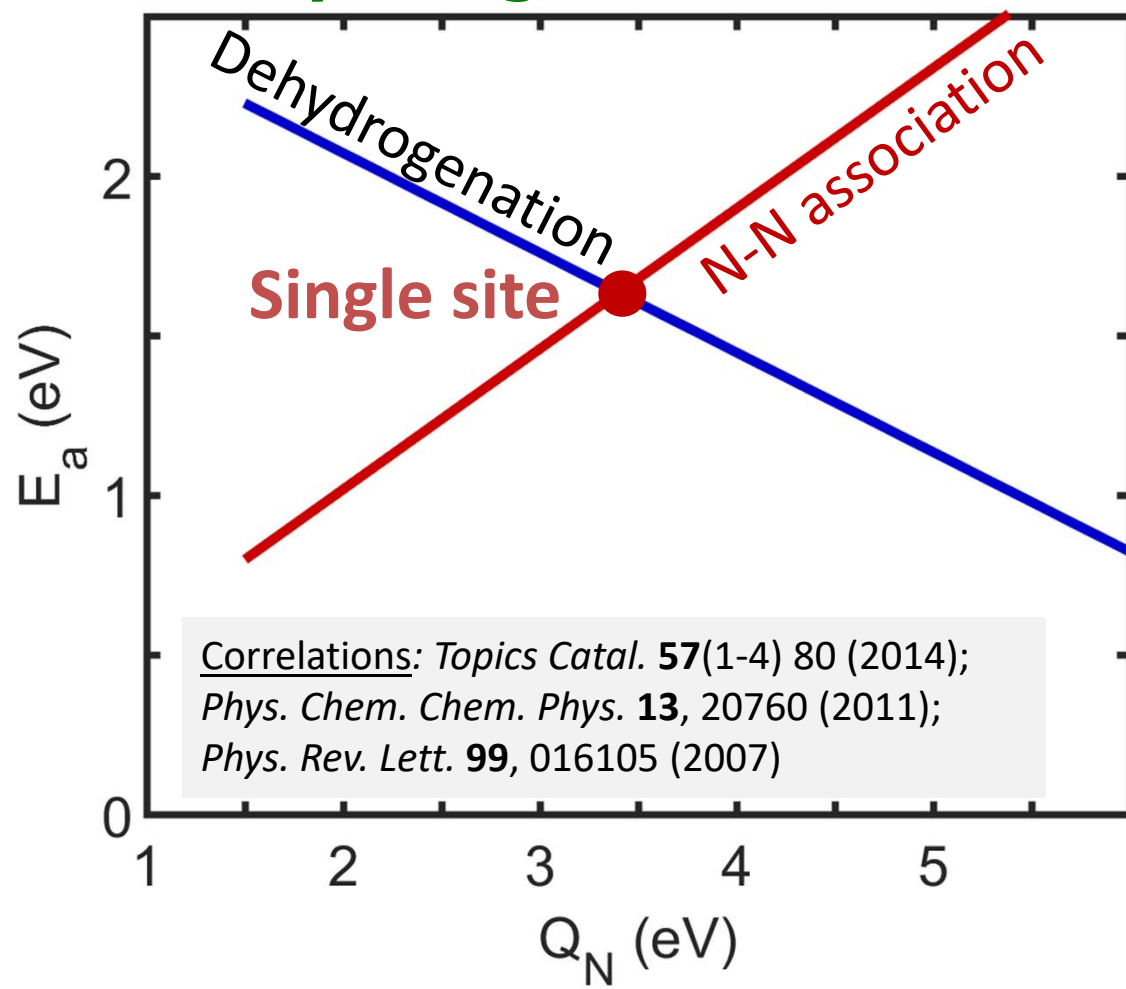
Wei Guo^{1,2,3} & Dionisios G. Vlachos¹

- How do we determine the active site?
- How do we find the optimal catalyst structure?
 - Shape, size, spatial arrangement in multifunctional materials



Controlling Multiple (Tandem) Reactions

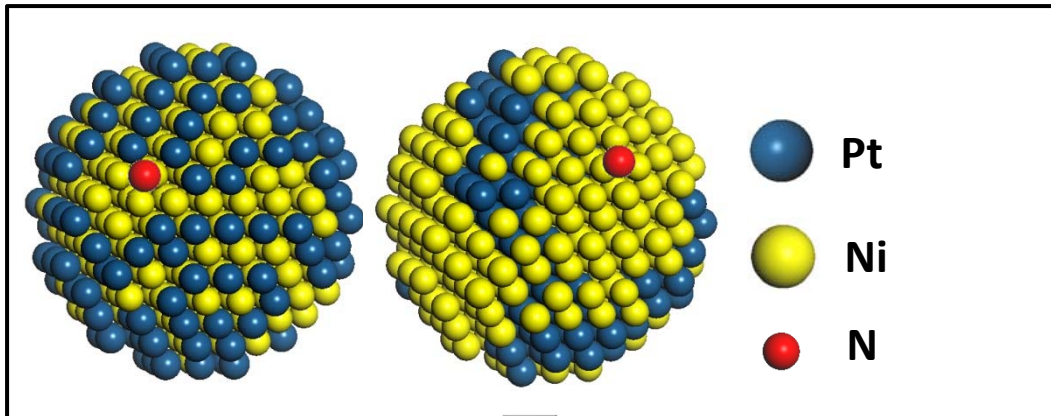
Re-interpreting the Volcano curve



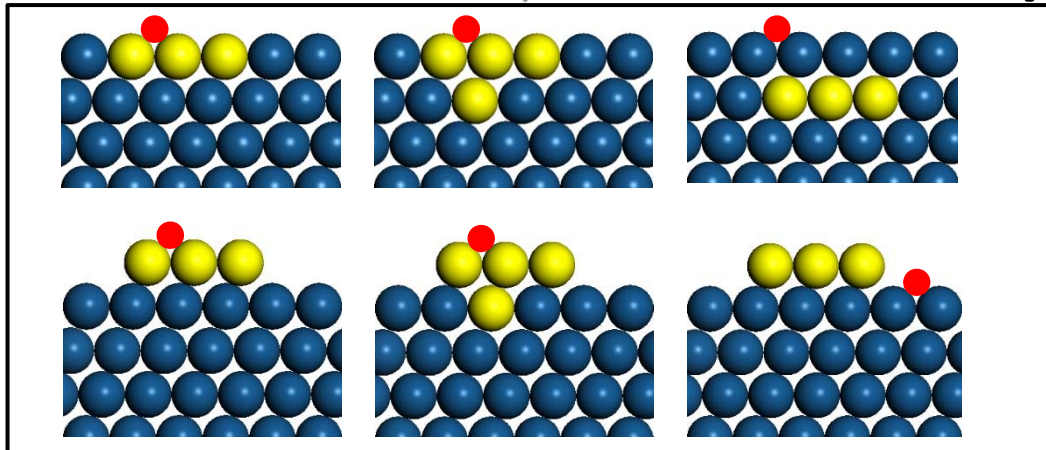
Single site leads to a volcano due to competing rate determining steps

Highest rate is controlled by thermodynamics and kinetics

Energetics of Complex Microstructures



Structure Sampling



DFT



Hamiltonian

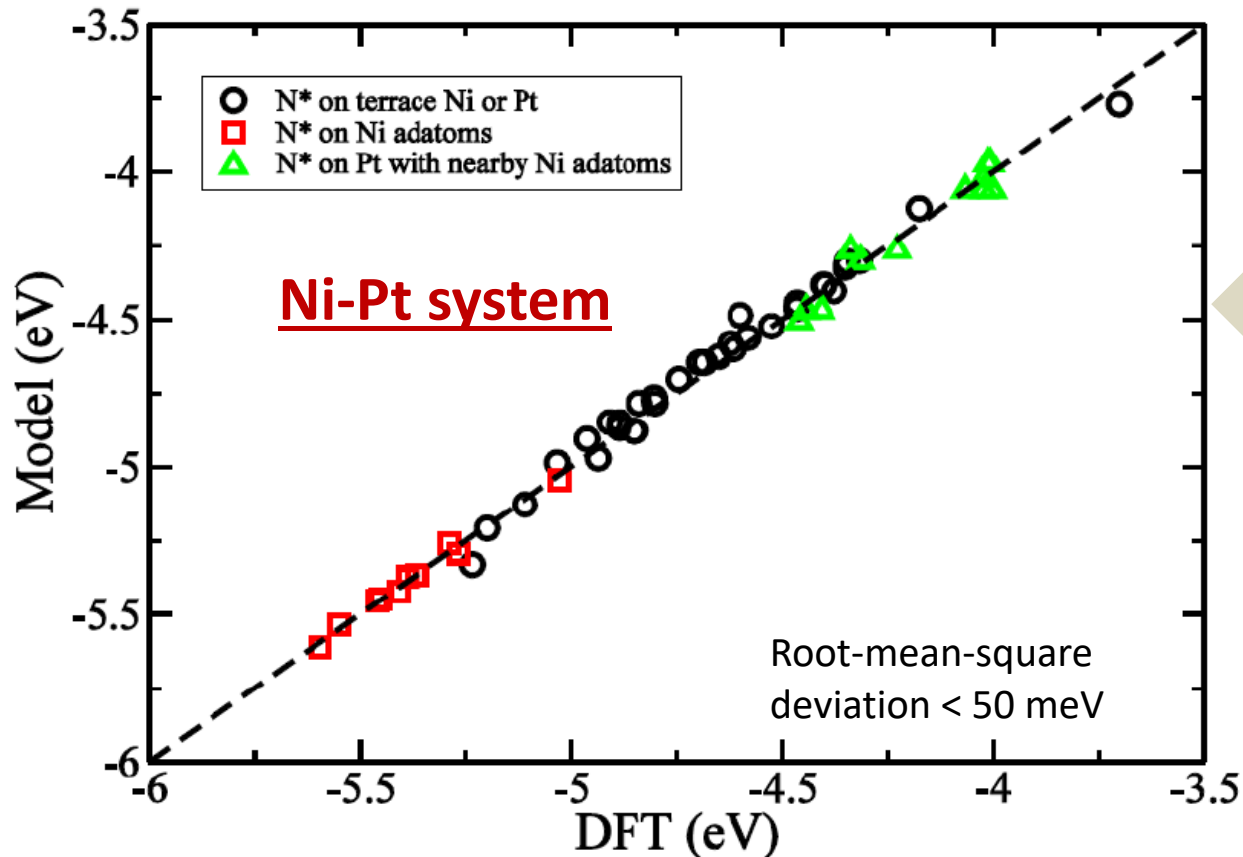
- Effect of Ni coordination
- Effect of lateral interactions

$$\Delta E(N^*) = \Delta E_{Pt}(N^*) + tr(\bar{E} \cdot \bar{n}_{Ni})$$

— # of i^{th} ($i=1, 2, 3$) Ni
nearest neighbors
around an N atom

- A bilinear model is postulated with respect to microstructure (Ni coordination) and N coverage

Designing Active Sites with Dialed-in Properties

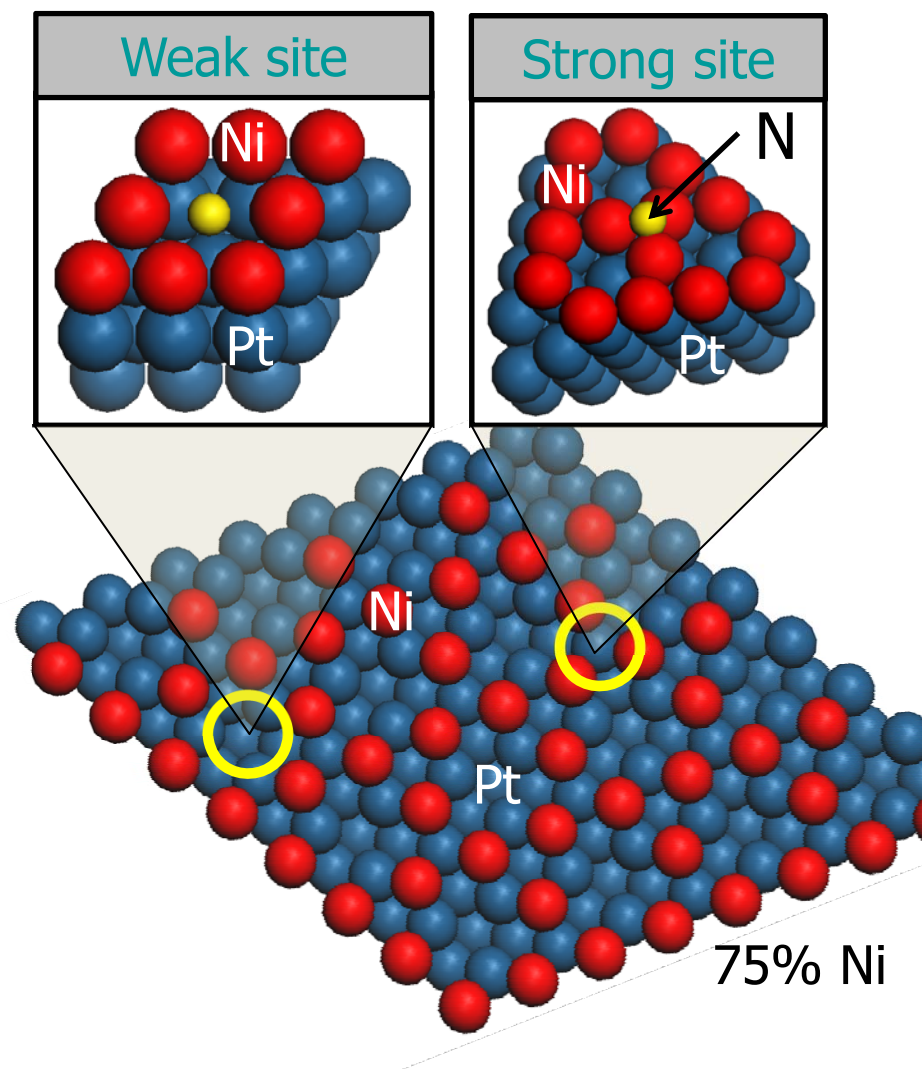
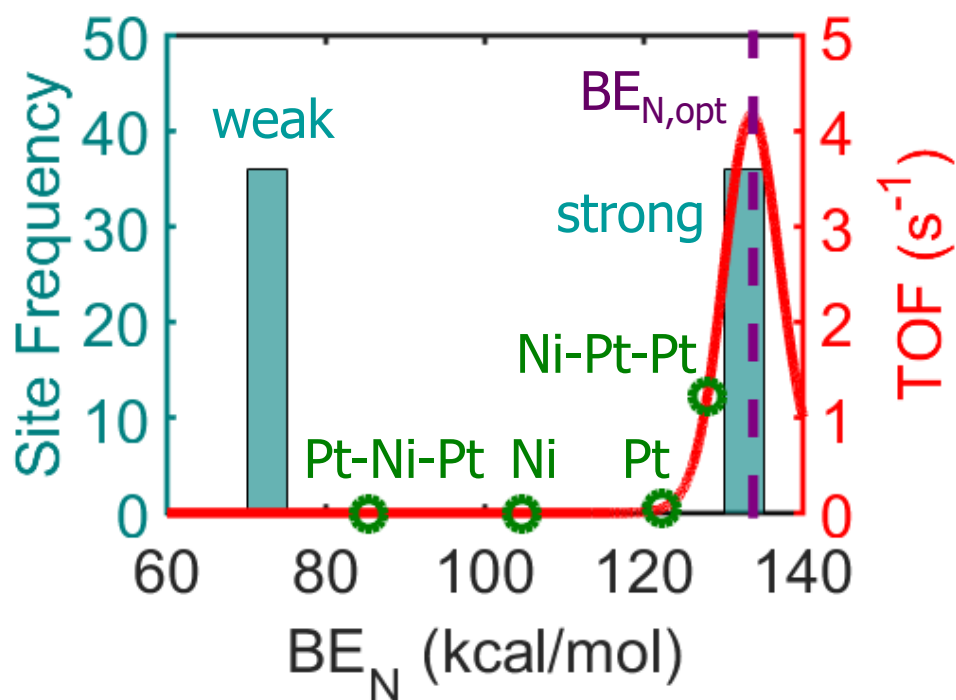


Hamiltonian Model
describes energies well

- **Energies can be tuned over a wide range of $\sim 2 \text{ eV}$ even for a single bimetallic** (loading and location dependent)
- **Multimetals provide an uncharted territory**

Determining the Active Site(s)

- A genetic algorithm automatically identifies active site(s) with machine learning for rate





Outline

- Energy, sustainability, and technological needs
- Grand computational challenges
- Topics
 - Catalyst prediction
 - Uncertainty
 - Design of active site
 - **Materials gap**
- Outlook

CATALYSIS

Synthesis,
multiscale modeling and
***in situ* characterization**
form the pillars of modern
catalytic science



Nat. Comm. 8, 1842 (2017)

ARTICLE

DOI: 10.1038/s41467-017-01983-6

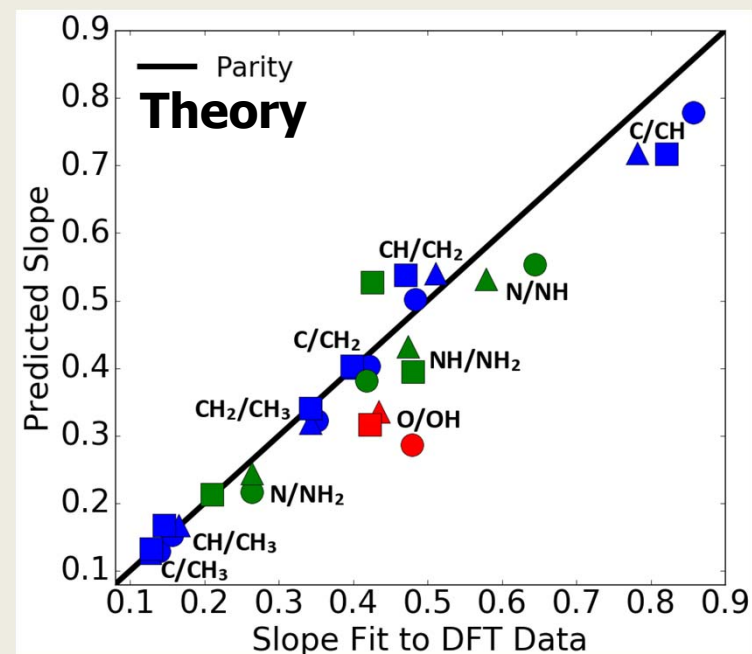
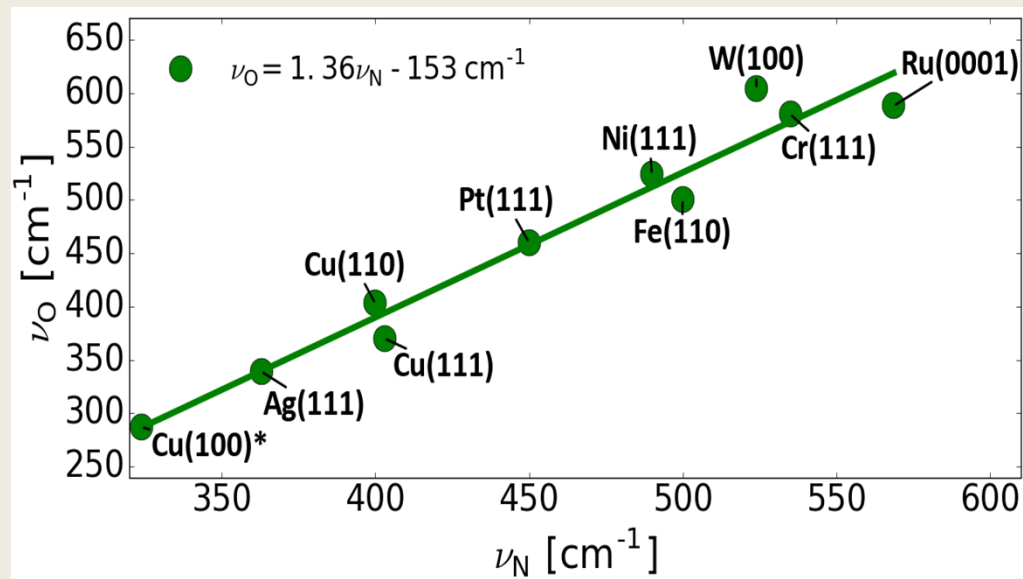
OPEN

Scaling relationships and theory for vibrational frequencies of adsorbates on transition metal surfaces

Joshua L. Lansford¹, Alexander V. Mironenko^{1,2} & Dionisios G. Vlachos^{1,2}

Can vibrational signatures close the materials gap?

Data from 40+ different HREEL studies





Outline

- Energy, sustainability, and technological needs
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CATALYSIS

Synthesis,

multiscale modeling and

***in situ* characterization**

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Summary and Outlook

Drivers & Current Status

- Sustainability and breakthroughs are necessary
- Materials are key to energy applications
- Tremendous advances in synthesis & operando characterization
- Models provide unprecedented opportunity for atom-by-atom design of materials and for process design
- Models for *in silico* materials prediction are not currently adequate

Three Grand Challenges

- **Materials gap:** optimization of microstructure, synthesis, and stability
- **Predictive modeling:** Parametric and model form uncertainty
 - Can **data science** make models more predictive?
 - How do we design experiments?
- **Process and product design vs. materials/media design**
 - Should catalyst and media (solvent) around the active site be optimized simultaneously with process and product design?



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