

**Chemistry GS visit, March 13, 2020**

**New *Methods* for de Novo Simulations  
and *Applications* to  
Catalysis, Materials, and Biochemistry**

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# To solve the most challenging problems, we develop methods and software simultaneously. Current Focus

**ARTIFICIAL PHOTOSYNTHESIS (JCAP)  $H_2O + hv \rightarrow H_2 + O_2$ ,: HER, OER**

**ARTIFICIAL PHOTOSYNTHESIS (JCAP):  $CO_2 + hv \rightarrow$  fuels: (MO/Cu); NP,**

**FUEL CELL CATALYST: Oxygen Reduction Reaction; Alkane fuels; Dealloyed**

**BATTERIES: Atomistic SEI-Li-IonicLiquid, anionic electrolytes, solid electrolytes**

**PEROVSKITES: MAPbI<sub>3</sub>, photoanodes, BaTiO<sub>3</sub>, Ferroelectrics,**

**CERAMICS: Ductile ceramics, FC electrodes, FC membranes, HiTc**

**POLYMERS: Batteries (PEO); Fuel Cells electrolytes (Nafion, Anionic) BIOTECHNOLOGY:**

**GPCR Membrane Proteins, Pharma, GP activation**

**2D MATERIALS: MBE-graphene; MoS<sub>2</sub>, CVD, ALD**

**CATALYSIS: NH<sub>3</sub> synthesis, selective ammoxidation and oxidation alkanes**

**ENERGETIC MATERIALS: PETN, RDX, HMX, TATB, TATP, Propellants**

**COMBUSTION: Kinetics from full reaction reactive simulations**

**THERMOELECTRICS: (mechanical properties (brittleness))**

**Less active**

**SOLAR ENERGY: dye sensitized solar cells, CuInGaSe (CIGS/CdS) cells, Ionic Liquids**

**GAS STORAGE (H<sub>2</sub>, CH<sub>4</sub>, CO<sub>2</sub>): MOFs, COFs, metal alloys, nanoclusters**

**SEMICONDUCTORS: damage free etching**

**MultiParadigm Strategy: apply 1<sup>st</sup> principles to complex systems**

# Materials Design Requires Improvements in Methods to Achieve Required Accuracy. Our Focus:

## 1: Quantum Mechanics

Challenge: increased accuracy

- New Functionals DFT (dispersion)
- Meta Dynamics QM (DG† 298K)
- Accurate implicit solvation (CANDLE)
- Grand Canonical QM (constant Potent)
- Accurate Band Gaps

## 2: Force Fields

Challenge: chemical reactions

- ReaxFF and RexPoN- Describe Chemical Reaction processes, Mixed Metals, Ceramics, Polymers
- Accelerated Reactive Dynamics
- NonEquil QM Dynamics (eFF)
- Hybrid QM-ReaxFF

## 3: Biological Predictions

1st principles structures GPCR-GP acti  
1st principles Ligand Binding  
Conditional siRNA therapeutics

## 4: Molecular Dynamics

Challenge: Extract properties essential to materials design

- Non-Equilibrium Dynamics
  - Viscosity, rheology
  - Thermal Conductivity
- Plasticity, Dislocations, Crack
- Interfacial Energies
- surface tension, contact angles
- Reaction Kinetics
- Entropies, Free energies
- surface tension, contact angles

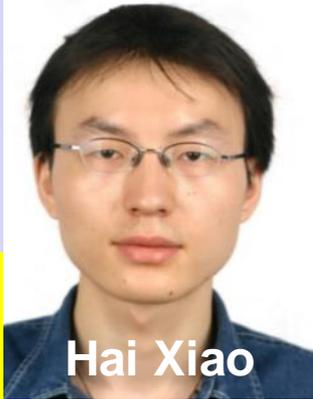
## 5: Coarse Grain Reactive MD

6: Integration: Computational Materials Design Facility (CMDF)

• Seamless across the hierarchies of simulations using Python-based scripts

Need new theory methods to solve key problems in energy and environment

# Grand Canonical QM (constant potential not constant electron as usual with QM)



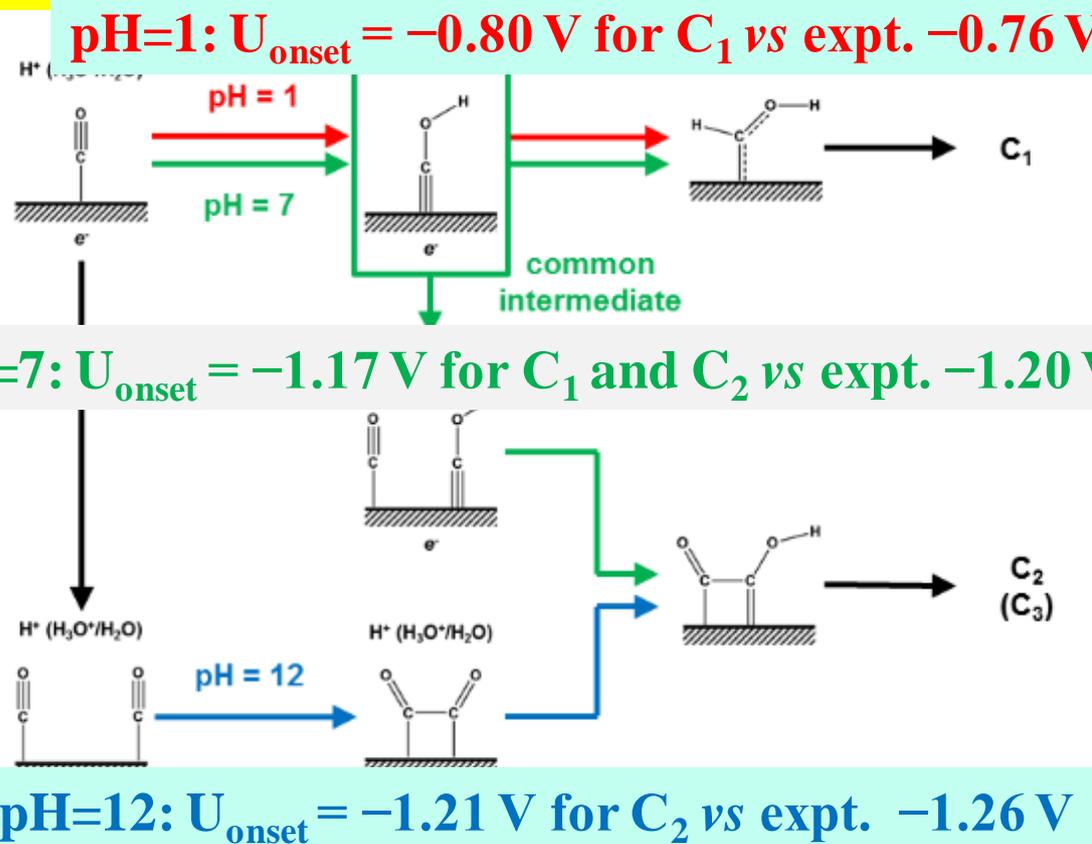
Hai Xiao

Traditional QM calculations use a fixed number of electrons. But in electrocatalysis, the potential at the electrode is kept constant during the electrocatalysis

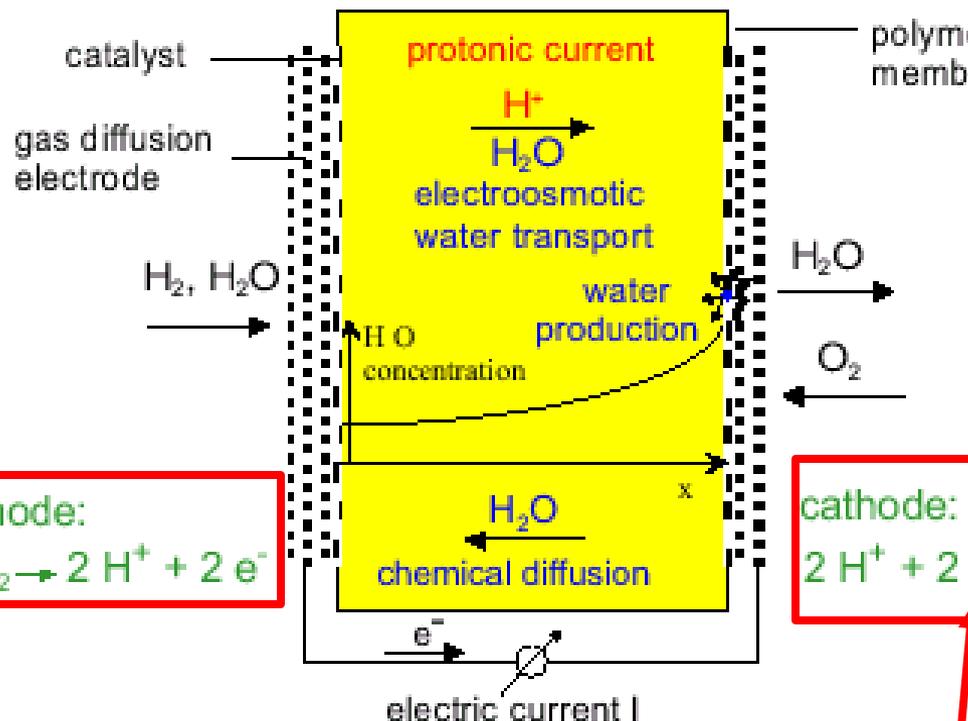
Methods for Grand Canonical QM now worked out. Onset potentials within 0.05V for Cu(111), all pH

**pH Dependence and Onset Potentials for Hydrocarbon Products from Electrochemical Reduction of CO on Cu (111);**  
Xiao, Cheng, Goddard, Sundararaman; J. Amer. Chem. Soc., 138 (2). pp. 483-486 (2016)

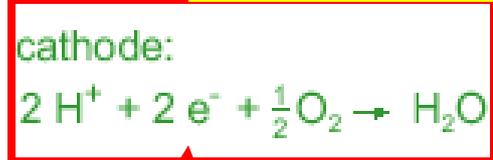
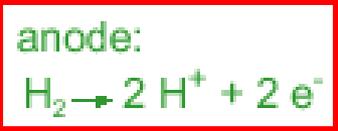
**Grand canonical electronic density-functional theory: Algorithms and applications to electrochemistry**  
Sundararaman, Goddard, Arias. J. Chem. Phys. (2017) 146 (11). Art. No. 114104.



# Kinetics of the Oxygen Reduction Reaction on Pt(111) from Ab initio Molecular Dynamics Free Energy Calculations at 298 K including explicit solvent



Hydrogen Fuel Cell  
 No CO<sub>2</sub> products  
 Need source of H<sub>2</sub>  
**Rate ORR too slow → Expense of Pt catalyst too high**

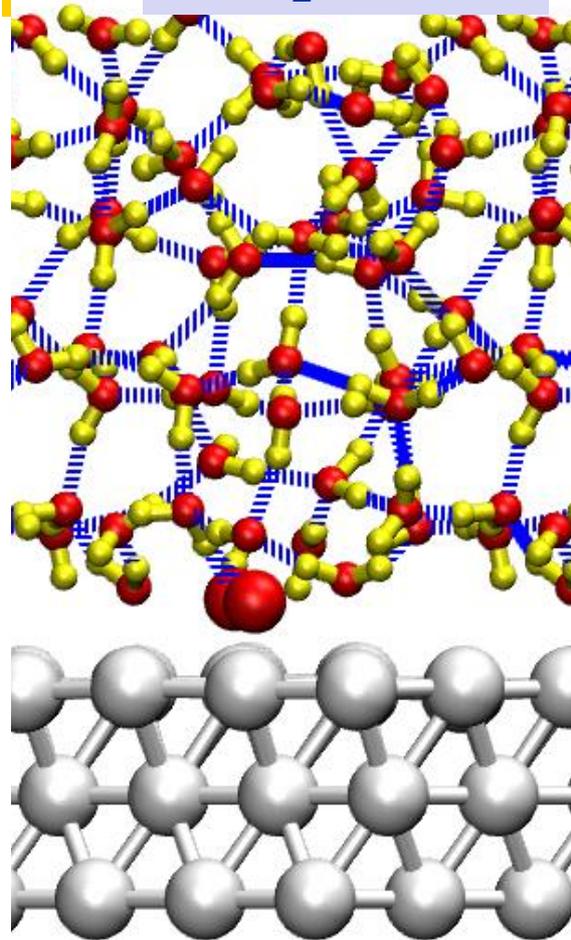


Oxygen Reduction Reaction (ORR)  
 Proton from anode + O<sub>2</sub> at cathode → H<sub>2</sub>O  
 Currently ~400 times slower than the anode reaction

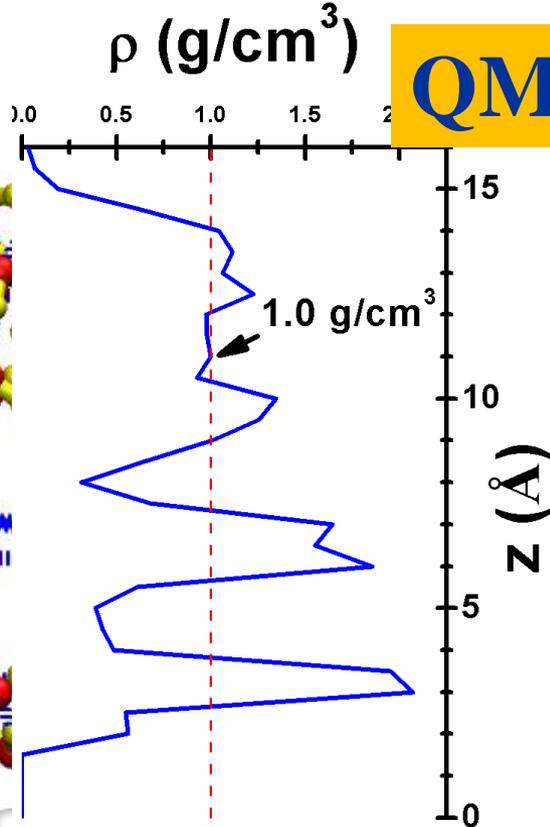


**A**

5 layers H<sub>2</sub>O  
50 H<sub>2</sub>O

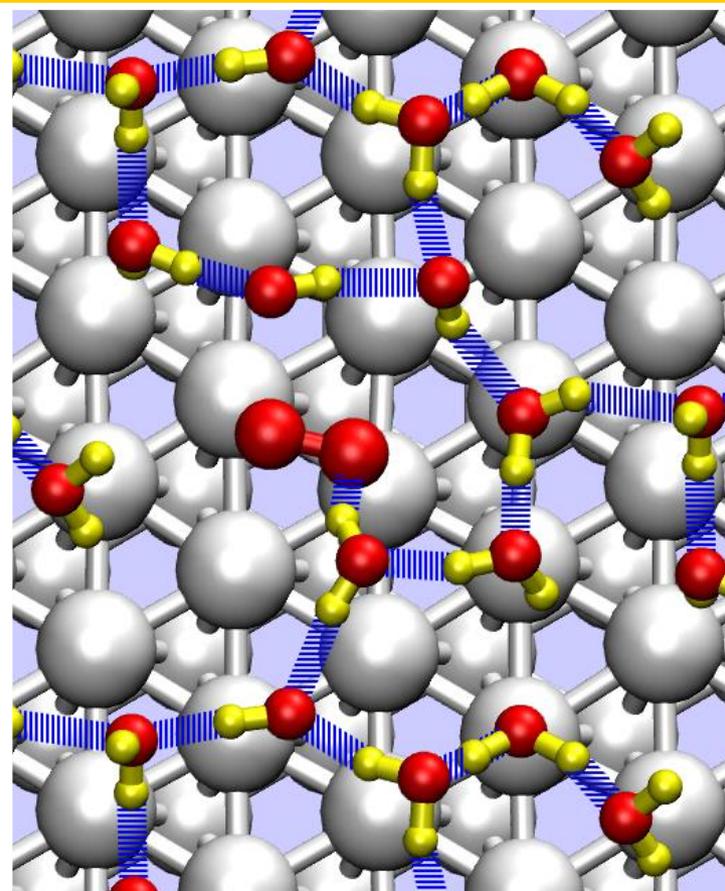


Side view of  
Surface structure  
of water/Pt(111)  
interface



water density  
distribution  
(center of mass)  
perpendicular  
to the surface (z  
direction),

QM metaMD full solvent



first contact layer of water  
from top view  
Pt(111) 4x4 unit cell

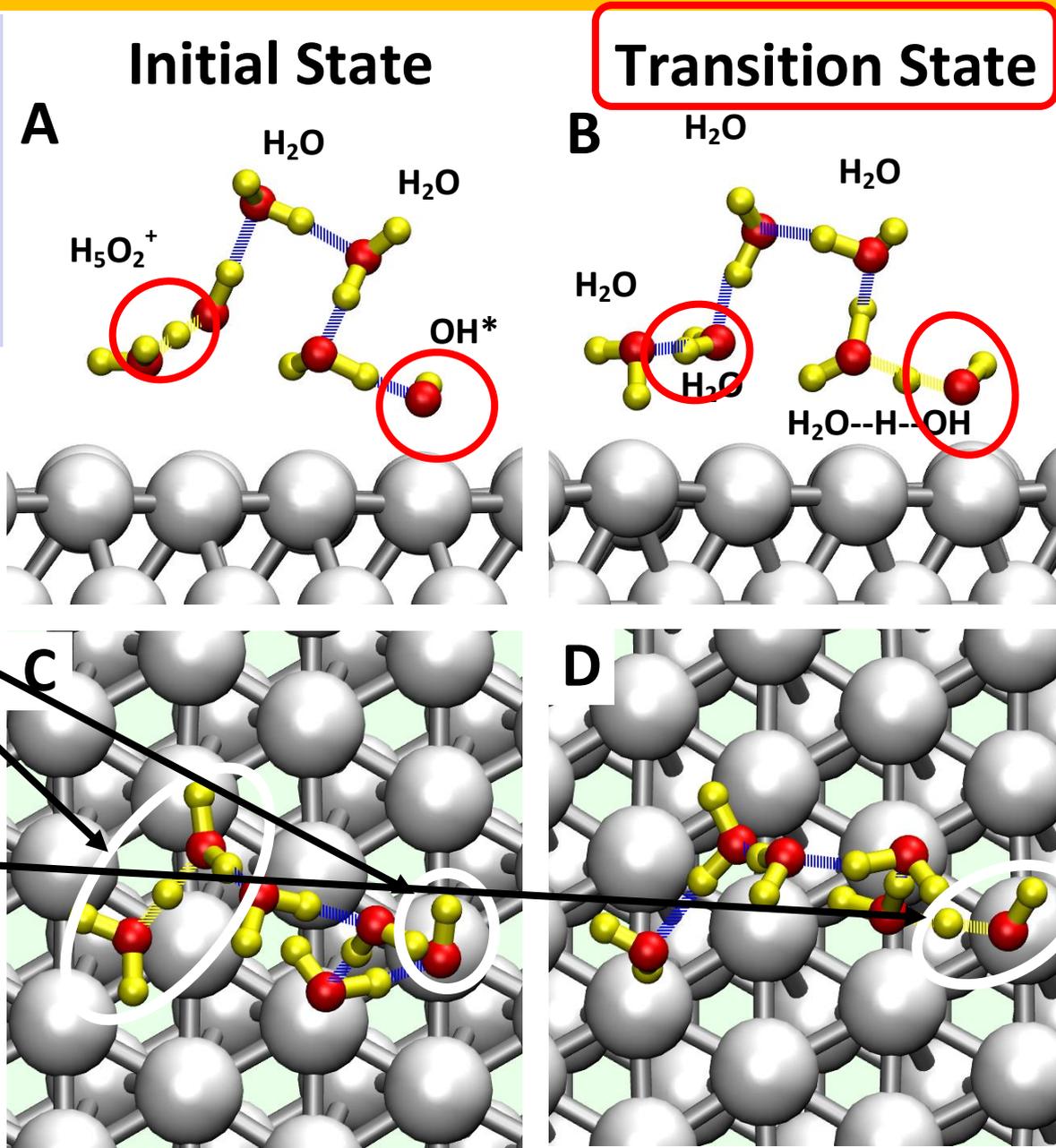
# Snapshots of structures during for water formation. $H_3O^+ + e^- + OH^* \rightarrow 2 \cdot H_2O$

0.6 V potential with one extra  $H_3O$  ( $H_3O^+ + e^-$ )  
 Initial state  
 Water molecules not involved are hidden, for viewing convenience.

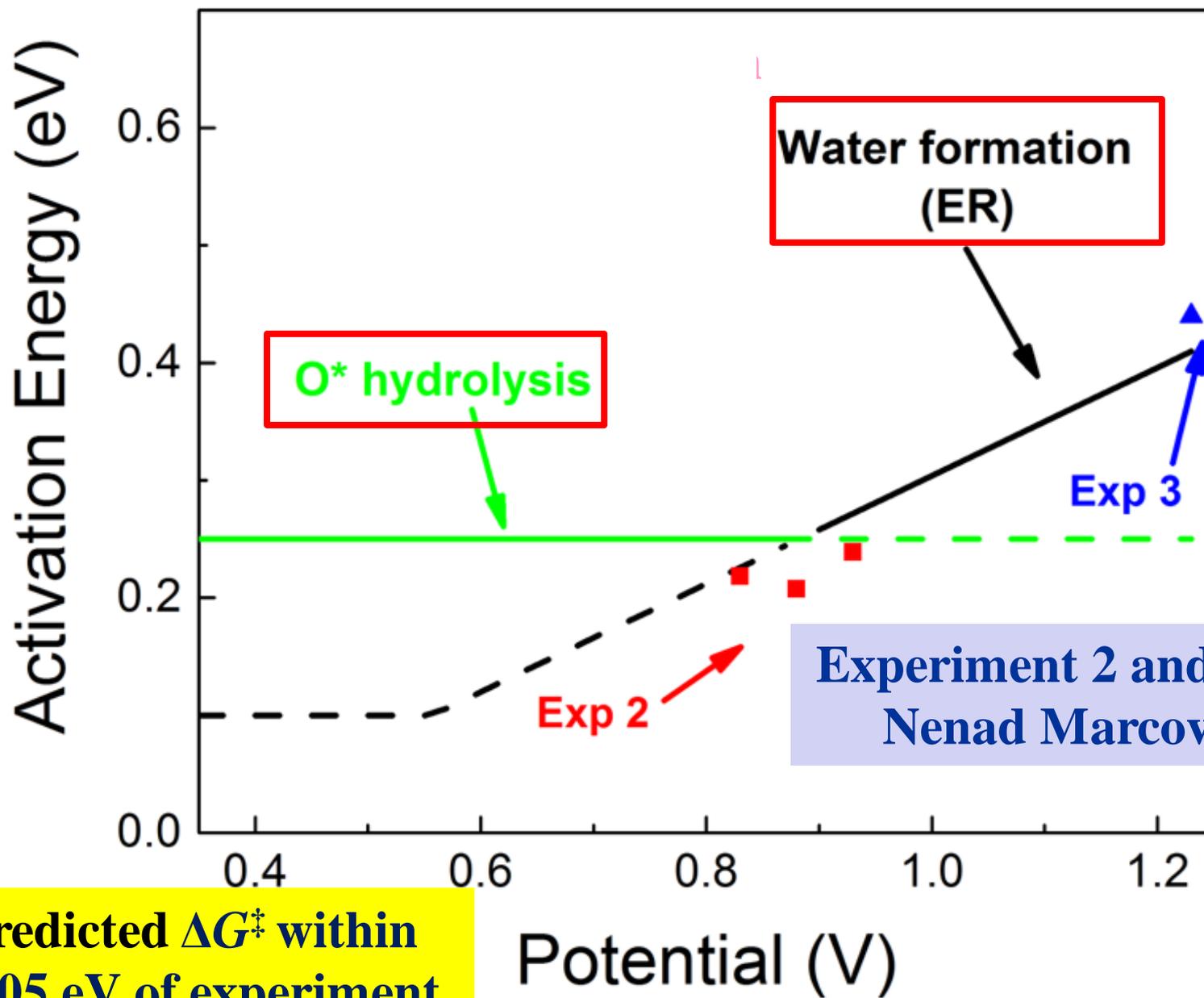
The reaction starts with one  $H_5O_2^+$  complex and one  $OH^*$ .  
 Three intermediate water molecules are involved in proton tunneling (Grotthuss mechanism)

Product  $H_2O$

The colors of atoms are:  
 Pt in silver, H in white, and O in white.



Predicted  $\Delta G^\ddagger$  in eV for the rate determining steps from full solvent DFT calculations at various applied potentials,



If we could decrease  $\Delta G^\ddagger$  from 0.42 to 0.25 at 1.2 V Rate would increase by factor of 600 Even factor of 6 make PEMH2 Fuel Cell for autos practical

Experiment 2 and 3 in water Nenad Marcovic, ANL

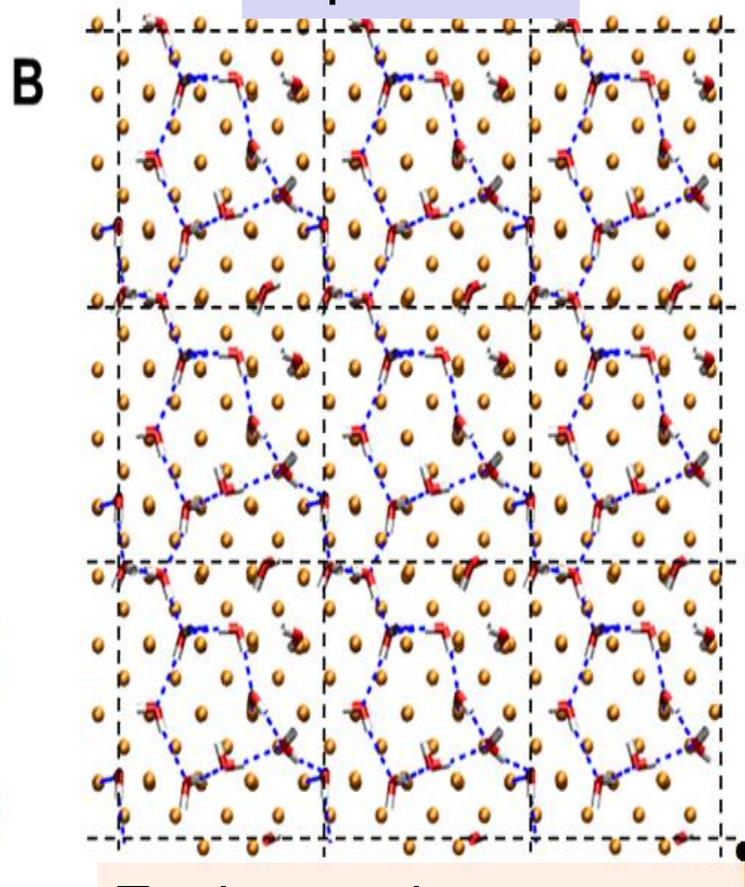
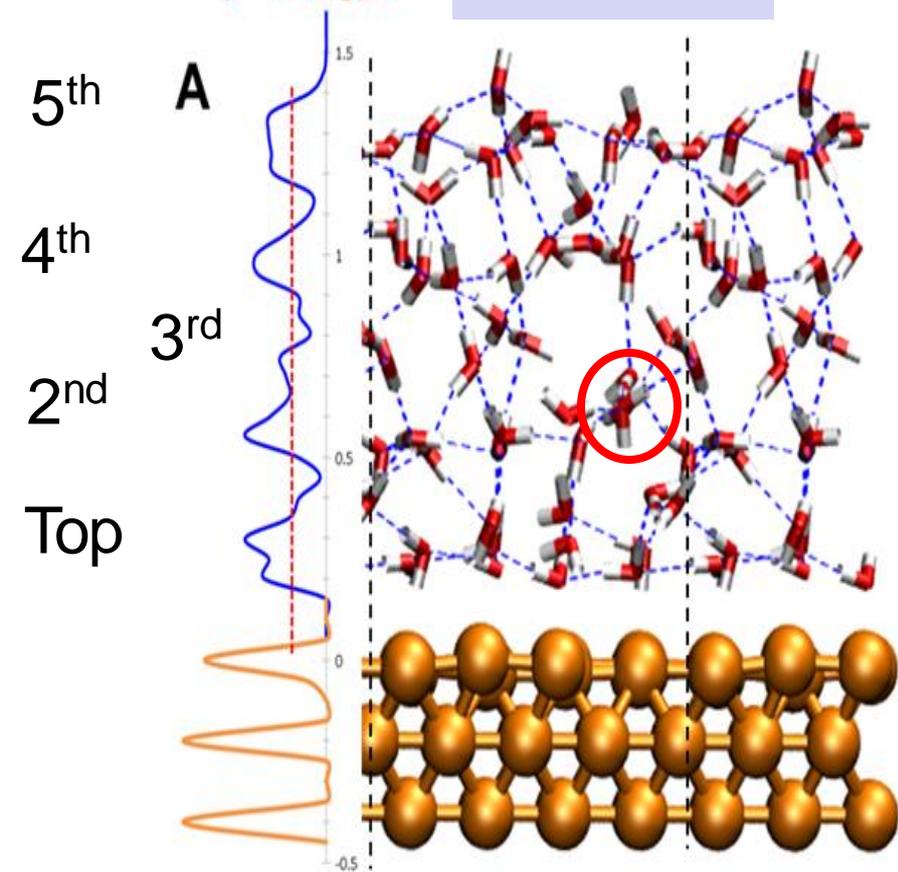
Predicted  $\Delta G^\ddagger$  within 0.05 eV of experiment

# activation of CO<sub>2</sub> with full solvent QM metadynamics on Cu(100) the stable surface under electrochemical conditions

$\rho = 1.0 \text{ kg/m}^3$

Side view

Top view



Cu 3 layers 4x4, 48 atoms  
49 H<sub>2</sub>O, one is H<sub>3</sub>O<sup>+</sup> → pH=0  
One H\* and 2 CO\*

Top layer only  
12 H<sub>2</sub>O; 16 sites;  $\frac{3}{4}$  ML

**Water/Cu(100) first layer**

**Observe a loosely packed hydrogen-bond network.**

# $\Delta G^\ddagger$ rates for QM-metaMD for CORR on Cu(100) at pH=7

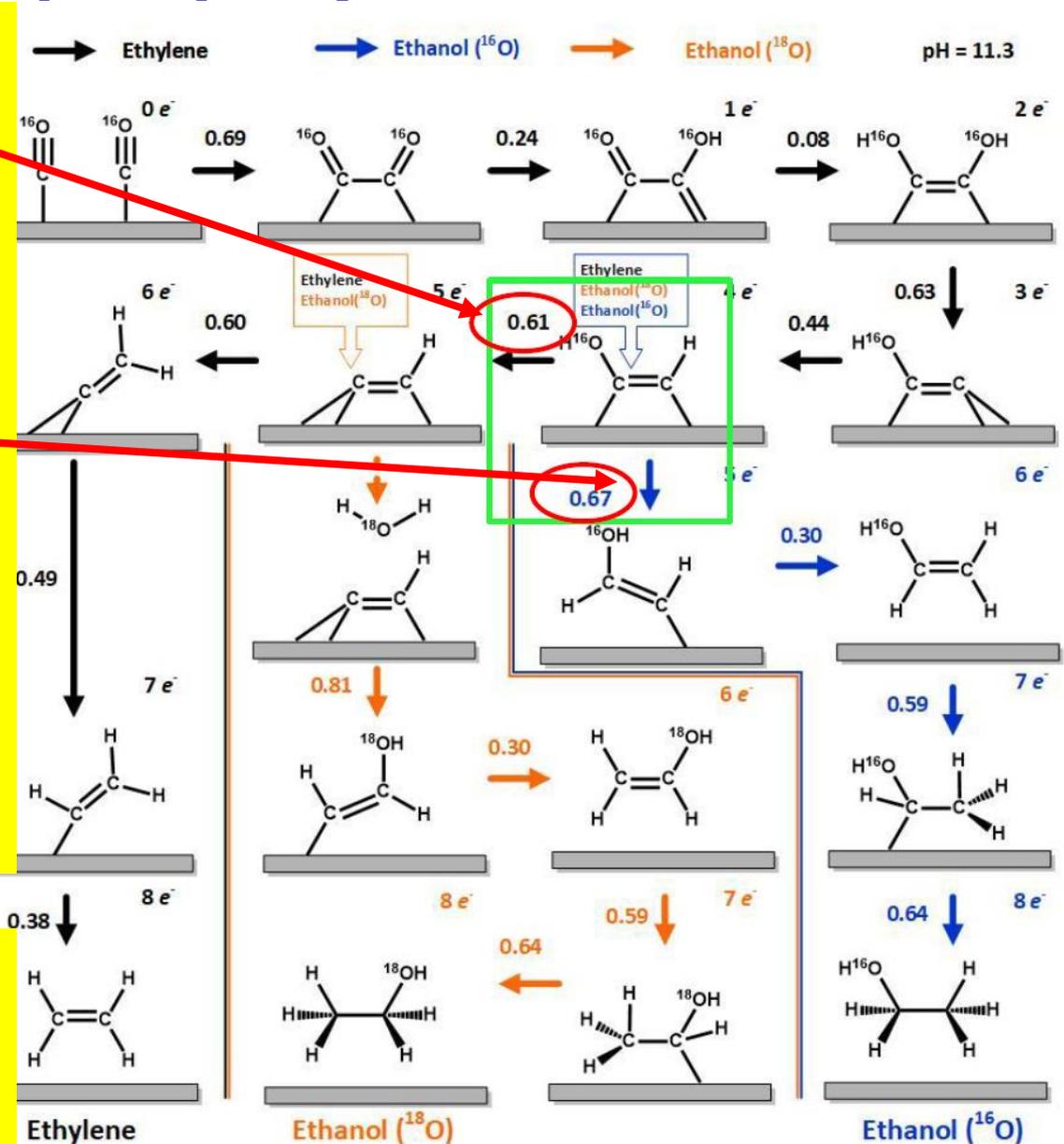
QM-metaMD barrier  $\Delta G^\ddagger=0.61$  eV to form  $^*C-CH$  is (red circle) leads to ethylene and  $^{18}O$  ethanol.

QM-metaMD barrier  $\Delta G^\ddagger=0.67$  eV to form  $^*H(^{16}OH)C-CH$  (red circle) leads to  $^{16}O$  ethanol.

This  $\Delta G^\ddagger=0.06$  eV  $\rightarrow$  ratio of 11:1

experimental ratio = 14  $\rightarrow$  0.066 eV (Lum, Agar)

Thus QM-metaMD distinguishes partitioning between pathways within 0.01 eV

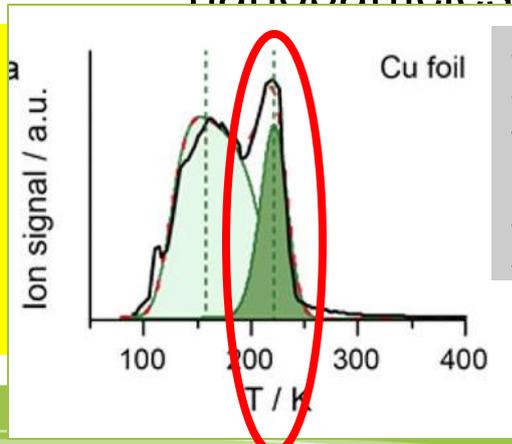


We now understand from QM on 200 atom systems the Reaction mechanism on low index surfaces  
But Nanoparticle and Nanowire catalysts often lead to dramatic improvements.

- Cu nanoparticles (NPs) can **reduce CO at potentials ranging from -0.3 V to -0.5 V (RHE)** reaching a mass activity of up to ~1.5A per gram of Cu and a Faradaic efficiency > 70% at -0.3 V;
  - CO reduction correlates with the **density of grain boundaries (GBs)** in Cu NPs.
- Concludes: Grain Boundaries are responsible for creating the vast majority of the active surfaces.

10-20 nm NP → 200,000 atoms  
How can we apply our now validated QM methods to these systems? Answer ReaxFF

TEM characterization of Cu nanoparticles in the Cu/CNT



**High CO binding → High performance**



# mechanism (100)

## Nanoparticle and Nanowire catalysts dramatic

Typical NP is 10-20nm → 200,000 atoms

How can we apply our now validated QM methods to these systems?

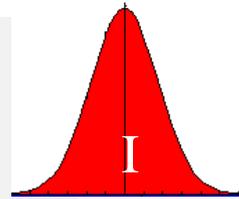
For systems too large for QM (>1000 atoms) ReaxFF reactive force field → energetics similar to QM but 10<sup>6</sup> times faster

$$E = E^{Val} + E^{Coul} + E^{VdW}$$

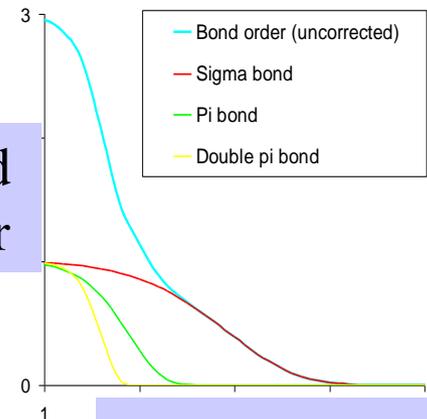
Valence energy

Electrostatic energy

- Self-consistent Charge Equilibration (QEq)
- Finite Gaussian shaped charge size of atom
- Allow charges to transfer as reactions occur

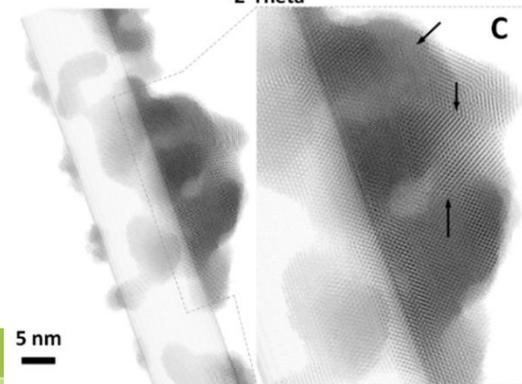
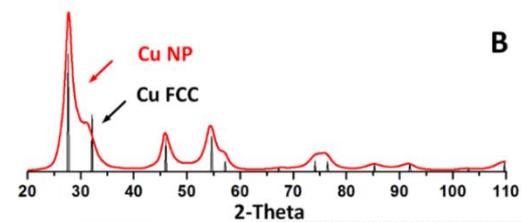
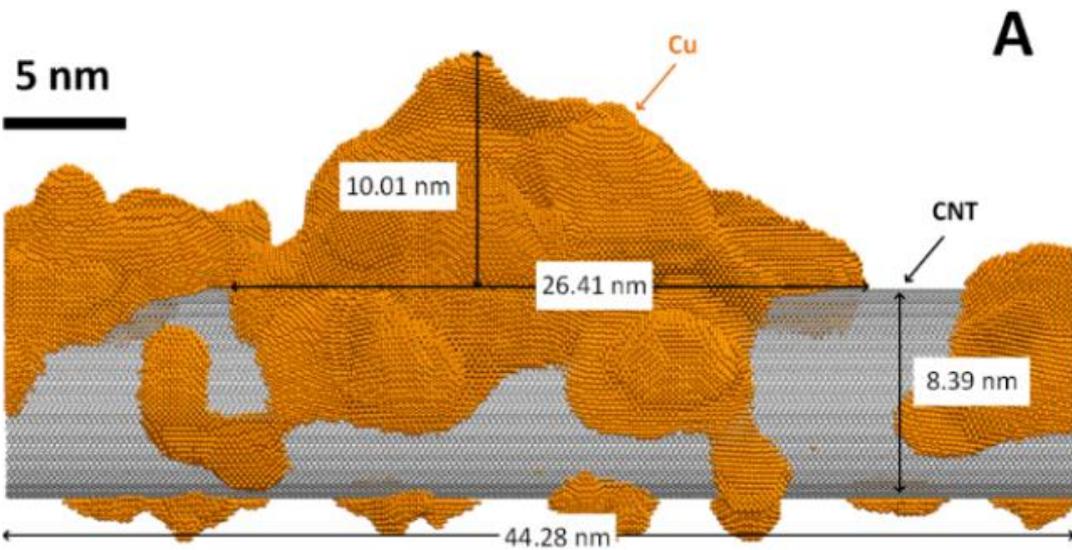
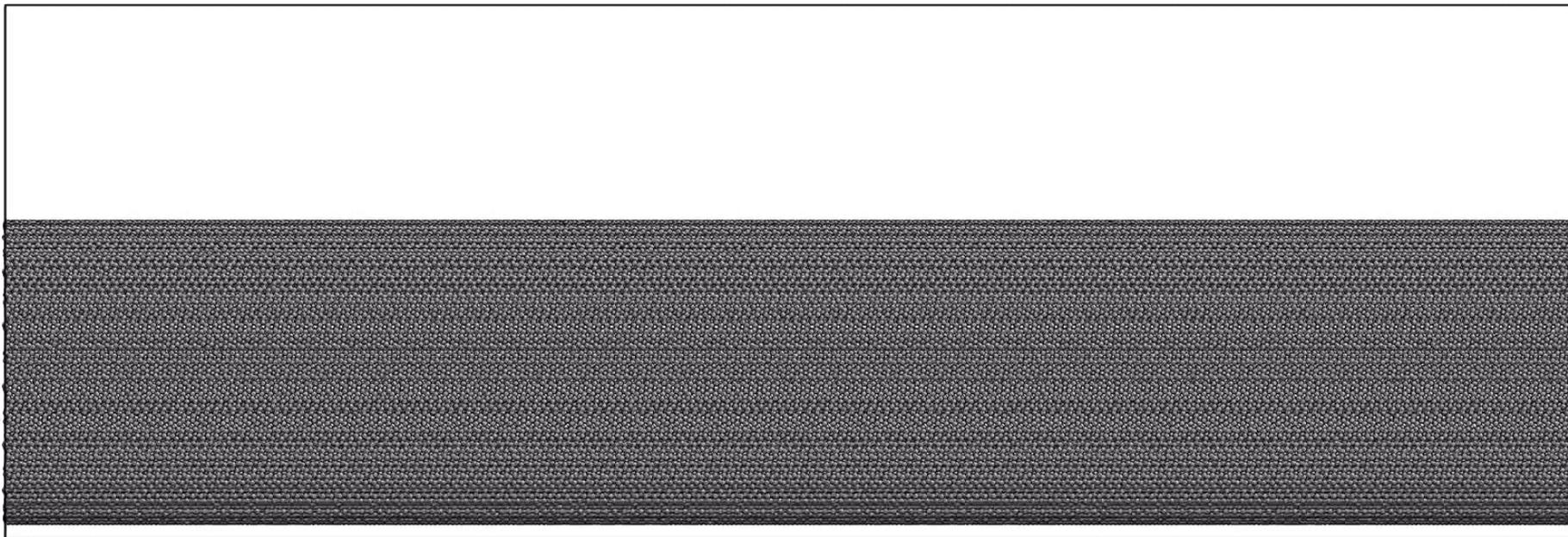


Bond order



Distance (Å)

# Computationally growth a Cu nanoparticle using Reactive force field

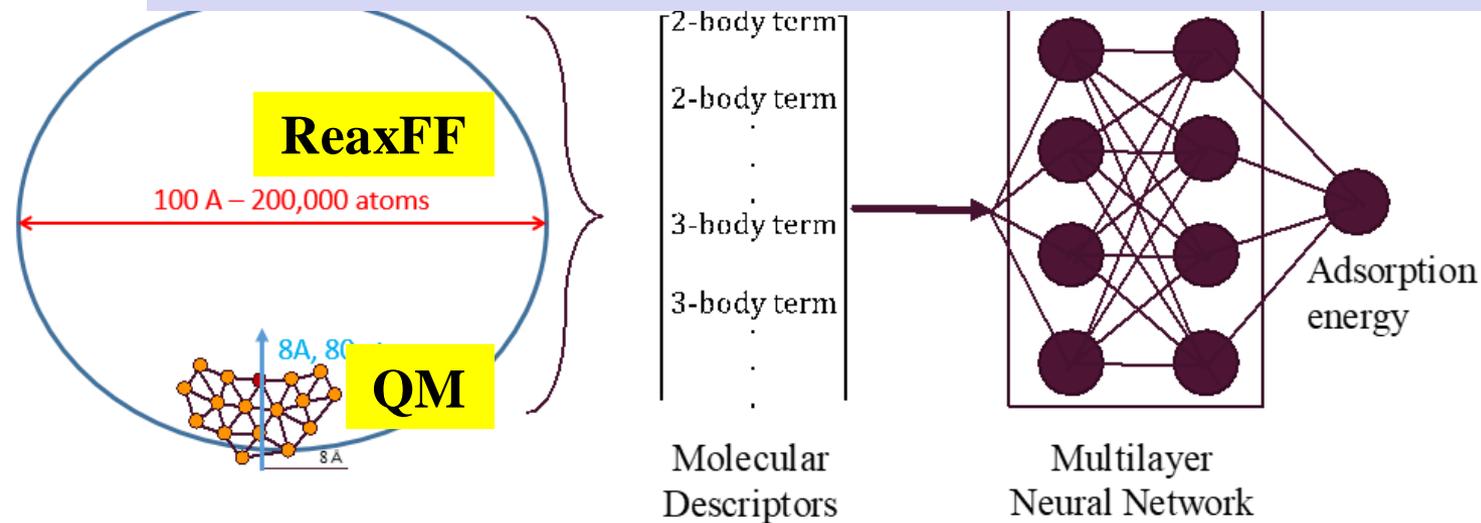


# Problem 10nm NP → 200,000 atoms 10,000 surface sites

How can we be sure to find best sites

Need fast way to predict binding at all 10,000 sites

Use *Neural Network Machine Learning Model*



Analyze 400 surface sites in terms of two-body and three-body neighbors

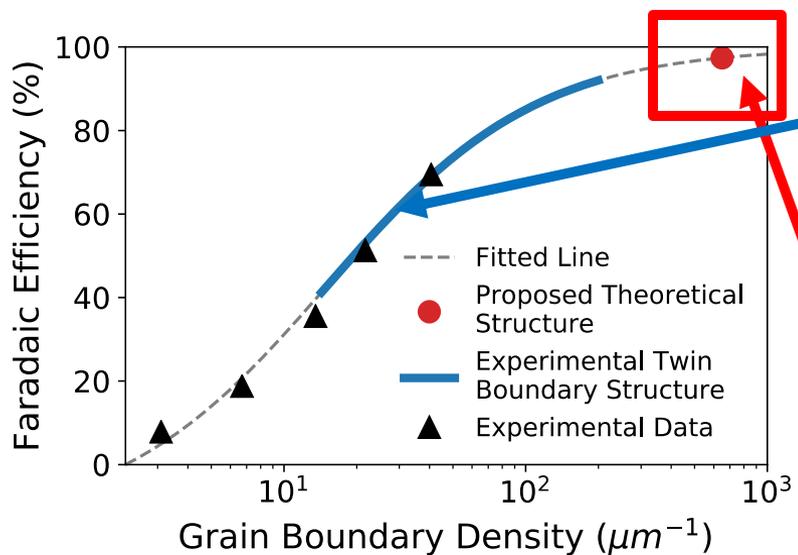
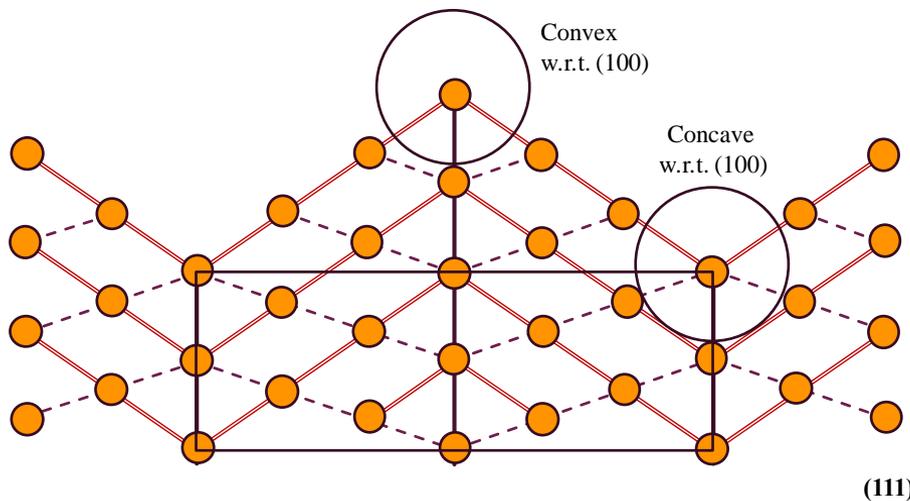
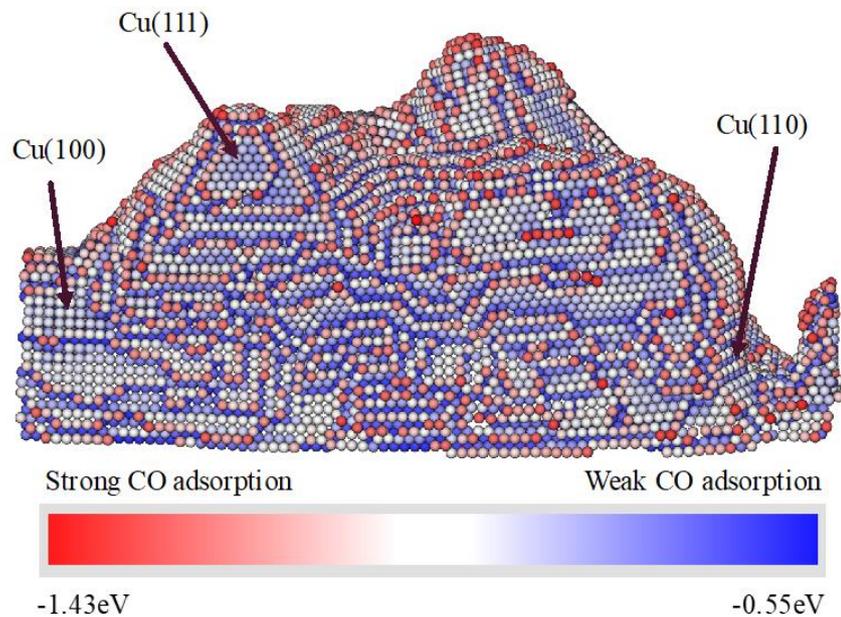
$$G_{i\alpha}^{(2)} = \sum_j \phi_{\alpha}^{(2)}(\mathbf{p}_{ij}), \quad G_{i\beta}^{(3)} = \sum_{jk} \phi_{\beta}^{(3)}(\mathbf{p}_{ij}, \mathbf{q}_{ijk})$$

$$E_{CO} = F_{NN}(\{G_{i\alpha}^{(2)}, G_{i\beta}^{(3)}\}; \mathbf{w}, \mathbf{b})$$

$E_{CO}$  predicted adsorption energy  
network network transforms input descriptors  $G$  into single values & weights  $w$ , biases  $b$ .

2-body descriptors    3-body descriptors

# Designed periodic structure with optimum twin boundary



Extrapolated faradaic efficiency (FE) of the concave site on the minimal periodic structure using experimental data.

Predicted copper structure with maximized twin boundaries extrapolated based on the densities of the boundaries.

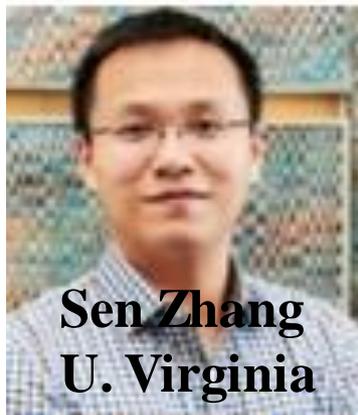
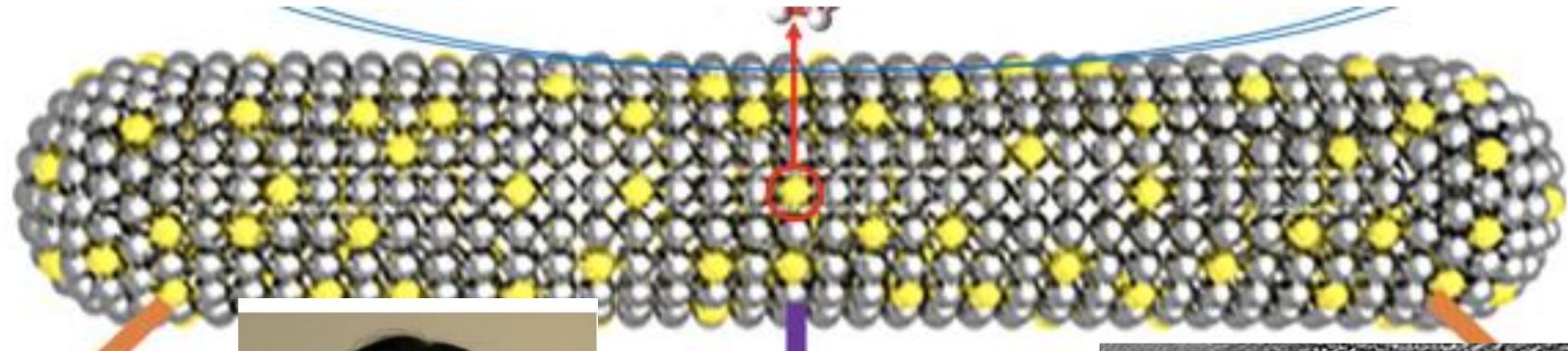
Next do full solvent QM on predicted structure to validate.

Then use ReaxFF to develop same structure with Atomic Layer Epitaxy with Au and Fe lines

# Problem with previous OER predictions

## Structure is not known, we assume $\gamma$ -NiOOH

Sen Zhang U. Virginia: Single Atom Doped TiO<sub>2</sub> uses organic solution colloidal synthesis (**TiCl<sub>4</sub> precursor, C<sub>18</sub> surfactants (oleylamine, OAm; oleic acid, OAc) and 1-octadecenorganic solvent**) to **Get single crystal Brookite structure with {210} surface square cross-section**. EXAFS, XAS → no Co-Co neighbors

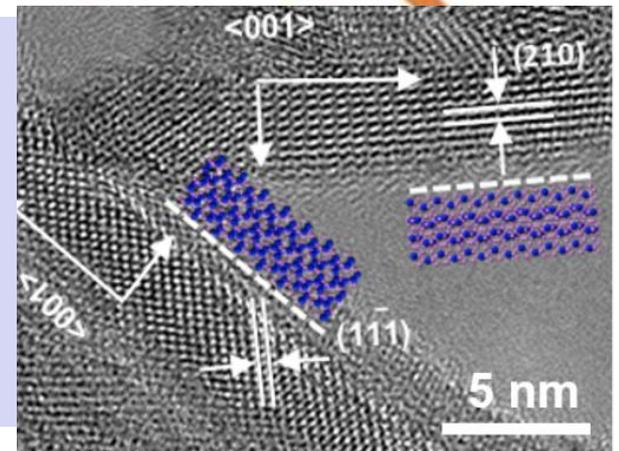


Sen Zhang  
U. Virginia



Jin Qian

For the first time  
we know exactly  
the surface  
structure  
Each Co is isolated

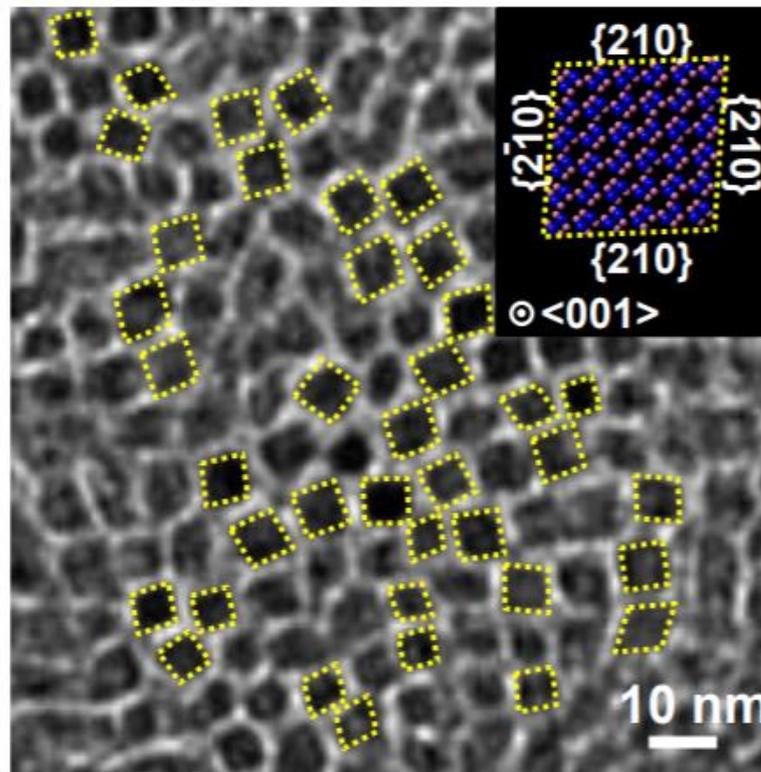
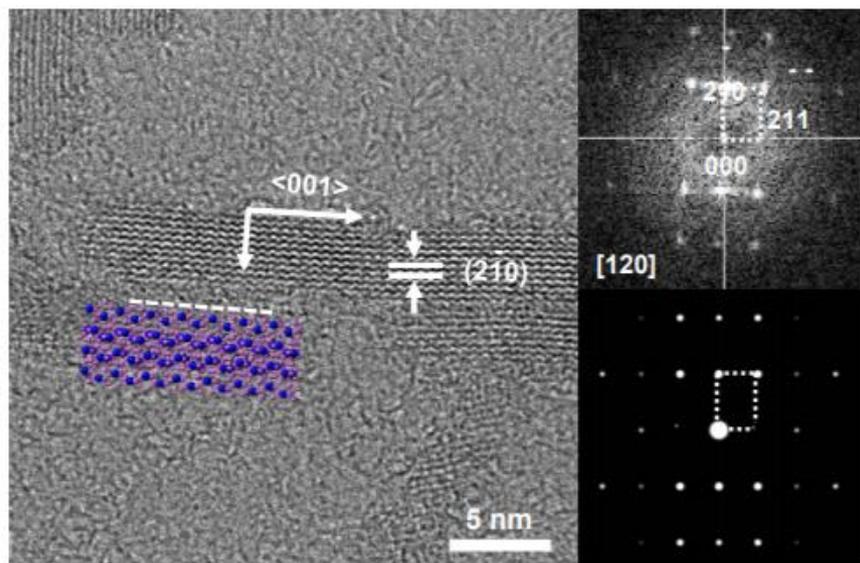


# Co Catalytic Single-Site in Well-Defined Brookite $\text{TiO}_2$ Nanorod Surface

XRD shows brookite single crystal

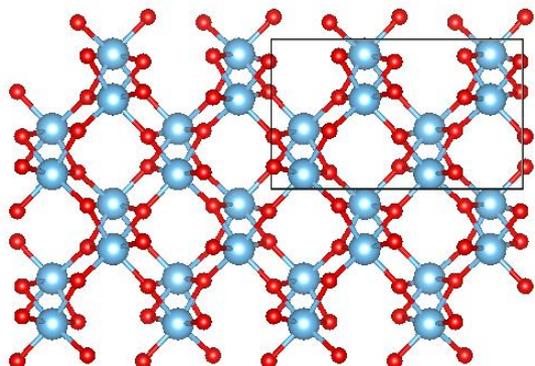
EXAFS no Co-Co neighbors

Well defined  $\{210\}$  surfaces

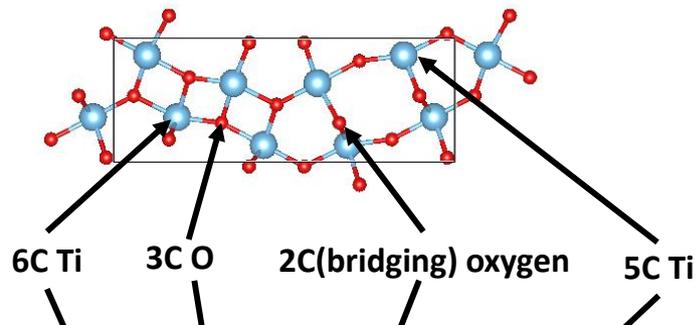


# QM calculations on (210) Brookite surface, 12% Co

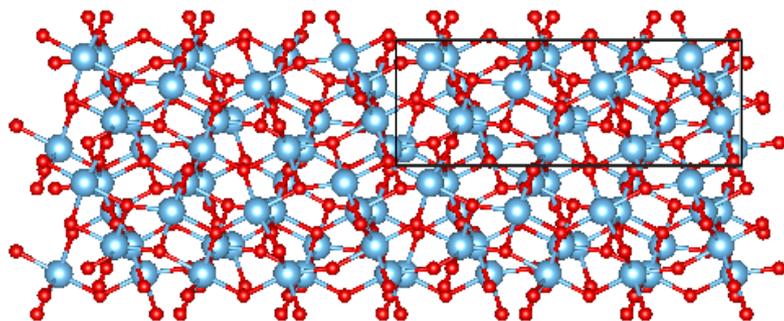
a). Bulk Brookite  $\text{TiO}_2$



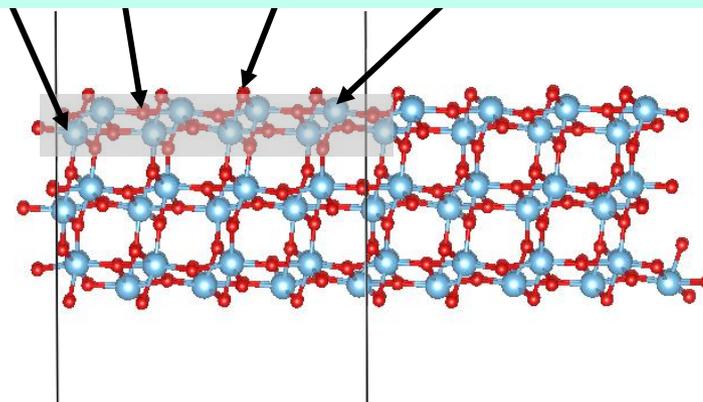
top view of  $\text{TiO}_2(210)$  surface  
grey shaded area in d).



There are two types of Ti on this surface: 6 coordinate Ti and 5 coordinate Ti,



c). Top View  $\text{TiO}_2(210)$



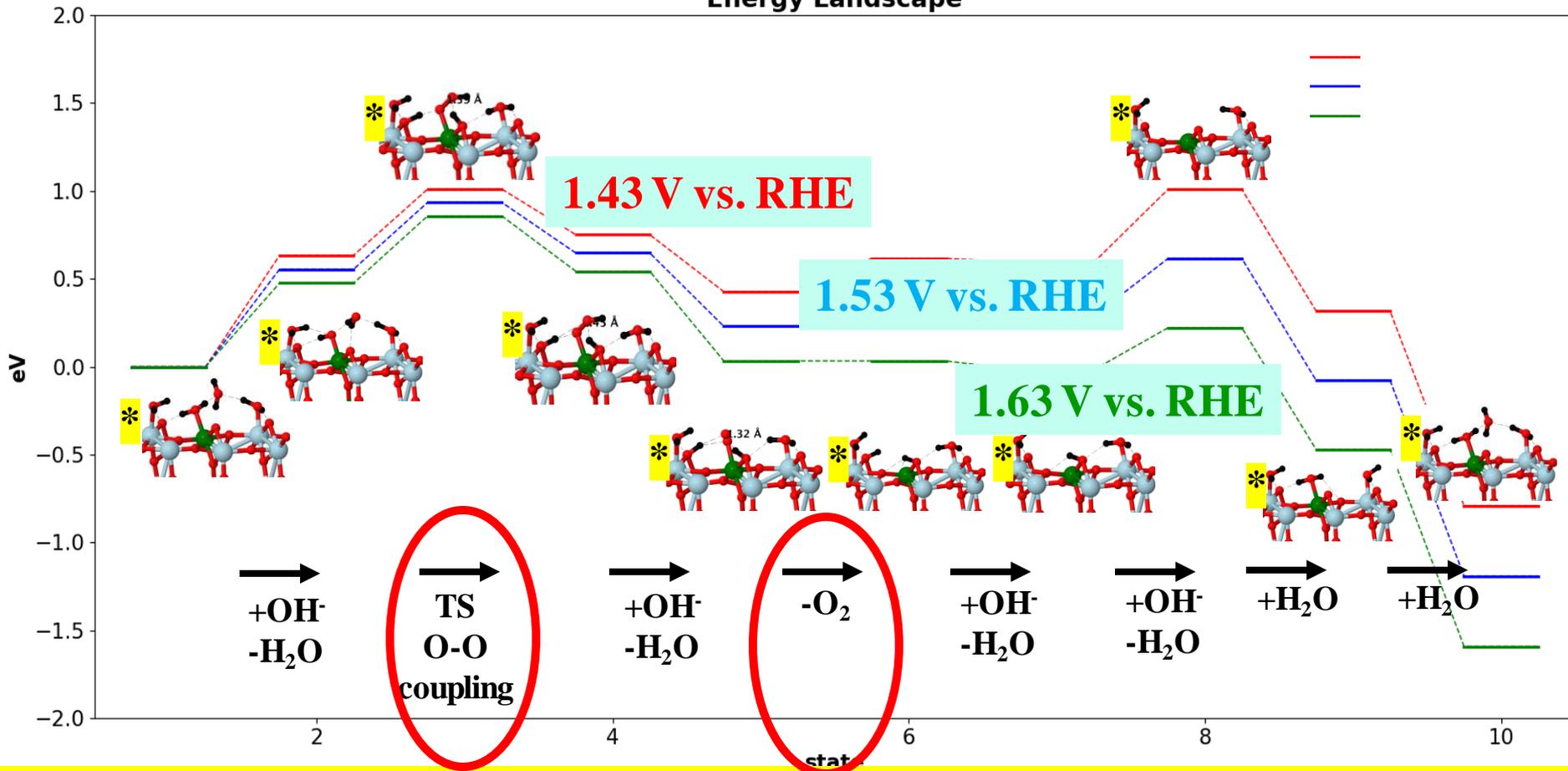
d). Side View  $\text{TiO}_2(210)$

there are two types of O on this surface, 3 coordinate O and bridging O,  
where 3C O sits lower on z axis, and 2C (bridging) O sits higher on z axis

# Grand Canonical QM: full QM reaction barriers at constant potential

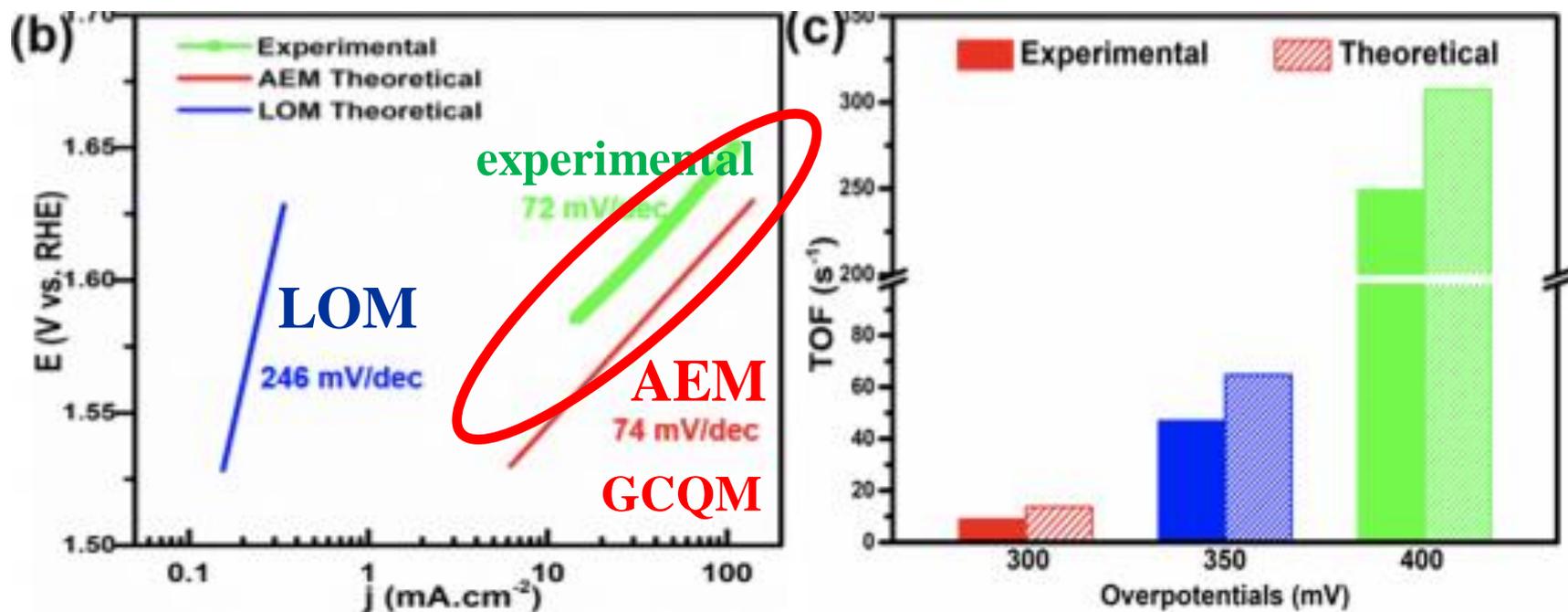


Energy Landscape



The OER mechanism involves a M=O bond with radical character on the O atom forming a bond to an OH derived from surface water (referred to as Adsorbate Evolution Mechanism (AEM))

# Comparison of GCQM predictions of kinetics with experiment Co-TiO<sub>2</sub> SA electrocatalyst



Extremely good comparison for fully characterized surface

GCQM: TOF = 13.1 s<sup>-1</sup> at  $\eta=300$  mV per Co SA; TOF = 307.4 s<sup>-1</sup> at  $\eta=400$  mV

experiment: TOF = 9.1 s<sup>-1</sup> at  $\eta=300$  mV per Co; TOF = 249.2 s<sup>-1</sup>, at  $\eta=400$  mV

GCQM: Tafel slope= 74 mV/dec; experimental result of 72 mV/dec

This is the first application of QM based OER kinetics for a system in which the catalyst surface structure is known, validating the very high accuracy for GCQM

# Recent breakthroughs

**Discover strong anchors between G-Protein and 3 IC loops of GPCR**  
**New mechanism of activation: GP binds first then agonist binding causes activation. Leads to new site for design of agonists**

**Discover strong anchors between beta arrestin and 3 IC loops of GPCR**

**Find mechanism of activation for nonbiased ligands**

**Hope to use in designing biased ligands treat pain and Parkinson's without side effects**

**New RexPoN reactive FF, more accurate than DFT. First FF to give correct description of 1<sup>st</sup> shell. Find Strong HB lifetime 90 fs (new experiment 100 fs). Leads to dynamic 1D polymer network.**

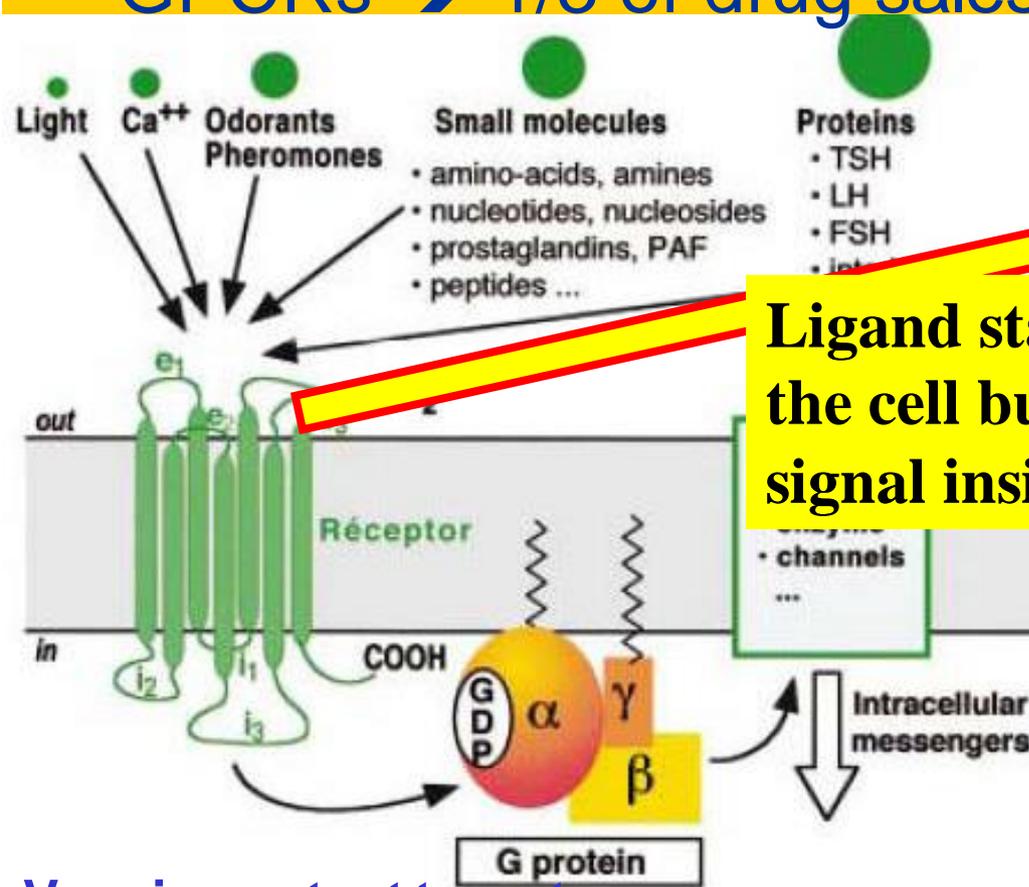
**Explain anomalous properties water at 230K**

**Developed Machine learning model to predict the activity of the 11,000 surface sites of 10nm nanoparticle. Discover active sites responsible for greatly improved Faraday Efficiency ethanol production.**

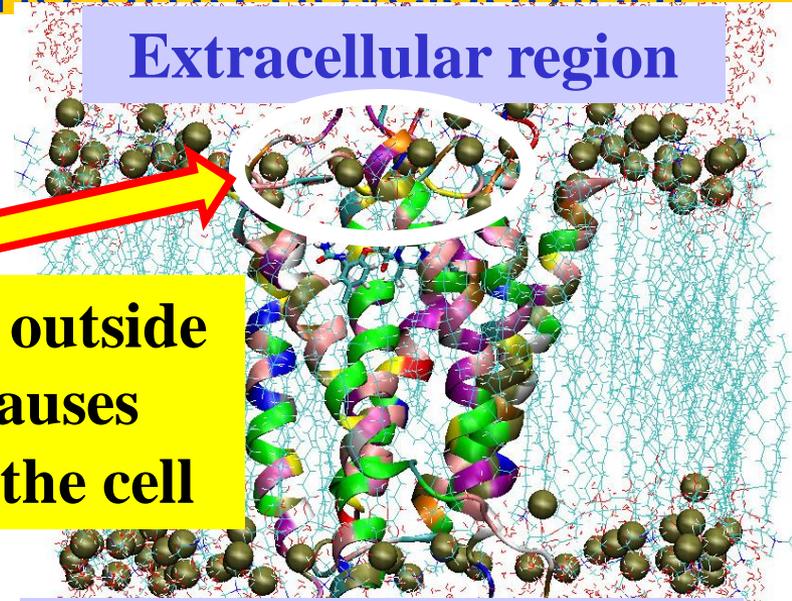
# G-Protein Coupled Receptors (GPCRs)

Critical to signaling between cells. 800 human GPCRs

GPCRs → 1/3 of drug sales and drug development



Ligand stays outside the cell but causes signal inside the cell

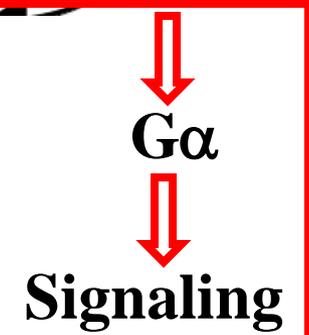


Extracellular region

Intracellular region



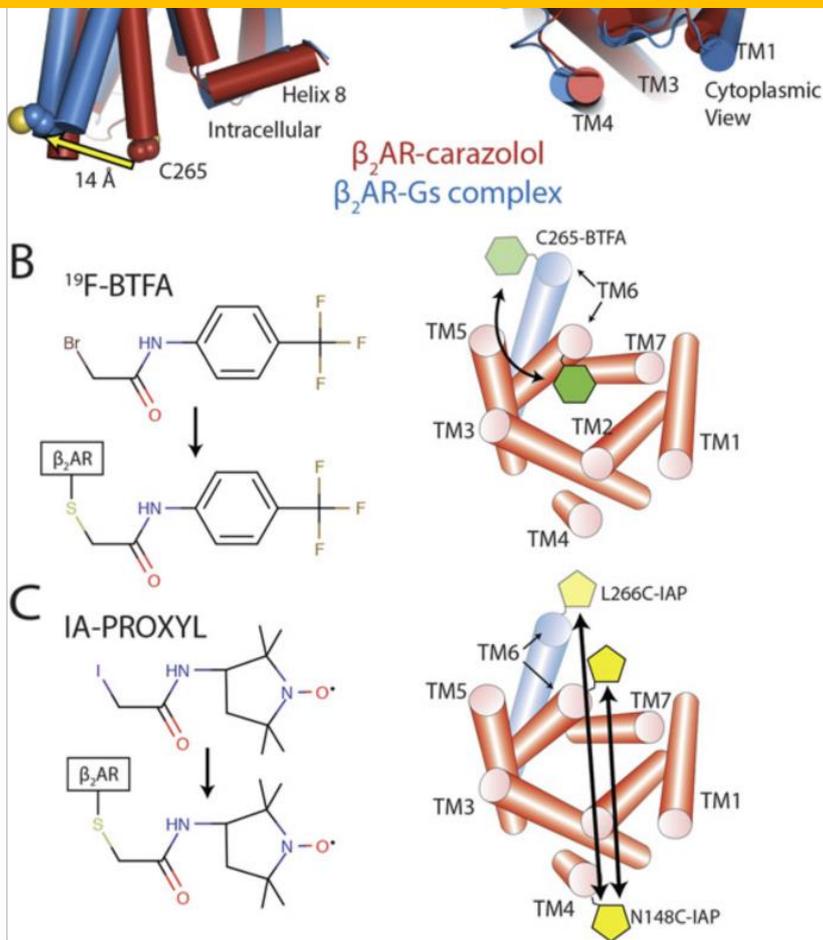
G-Protein



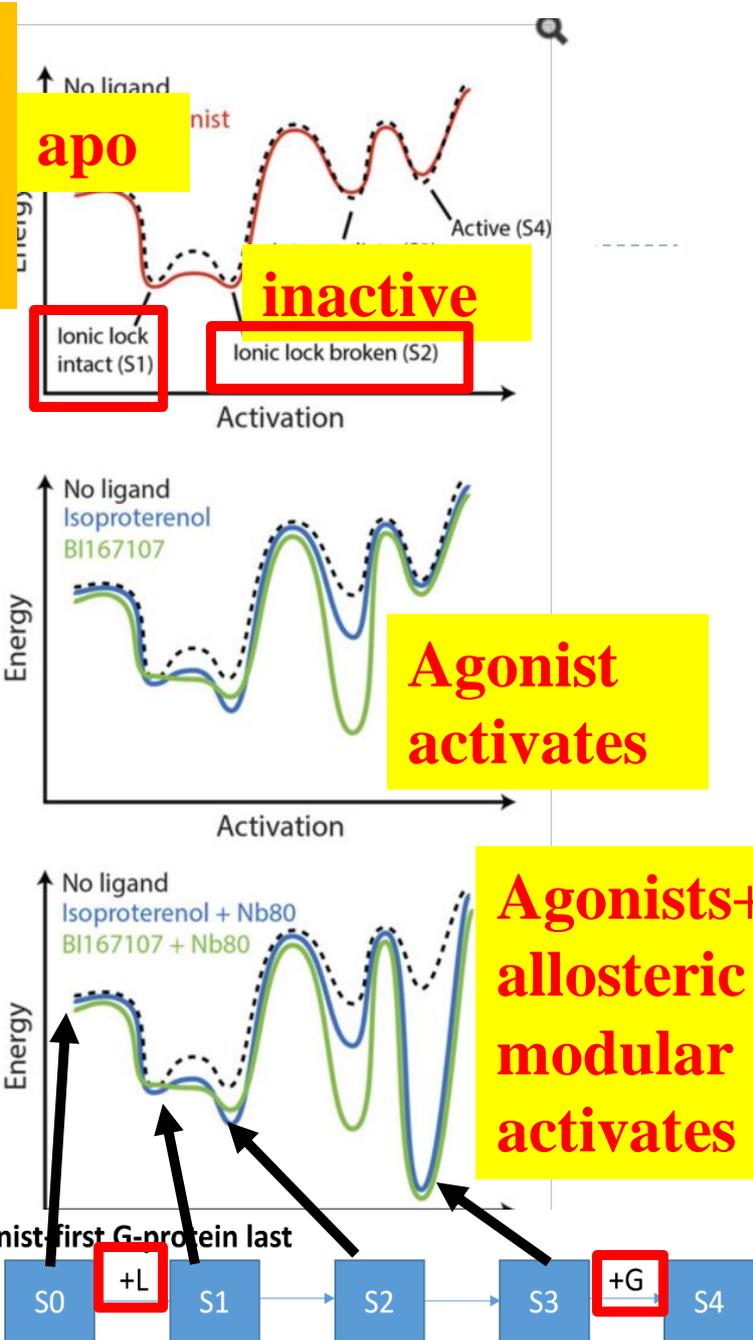
Gβγ

Very important target  
GPCR Sensors (smell, taste, vision, Pain)  
GPCR signaling (acetylcholine, serotonin, bradykinin, adrenoceptors, LPA, S1P1, chemokine Dopamine)

# Accepted mechanism of how Ligand/GPCR → activates G-Protein



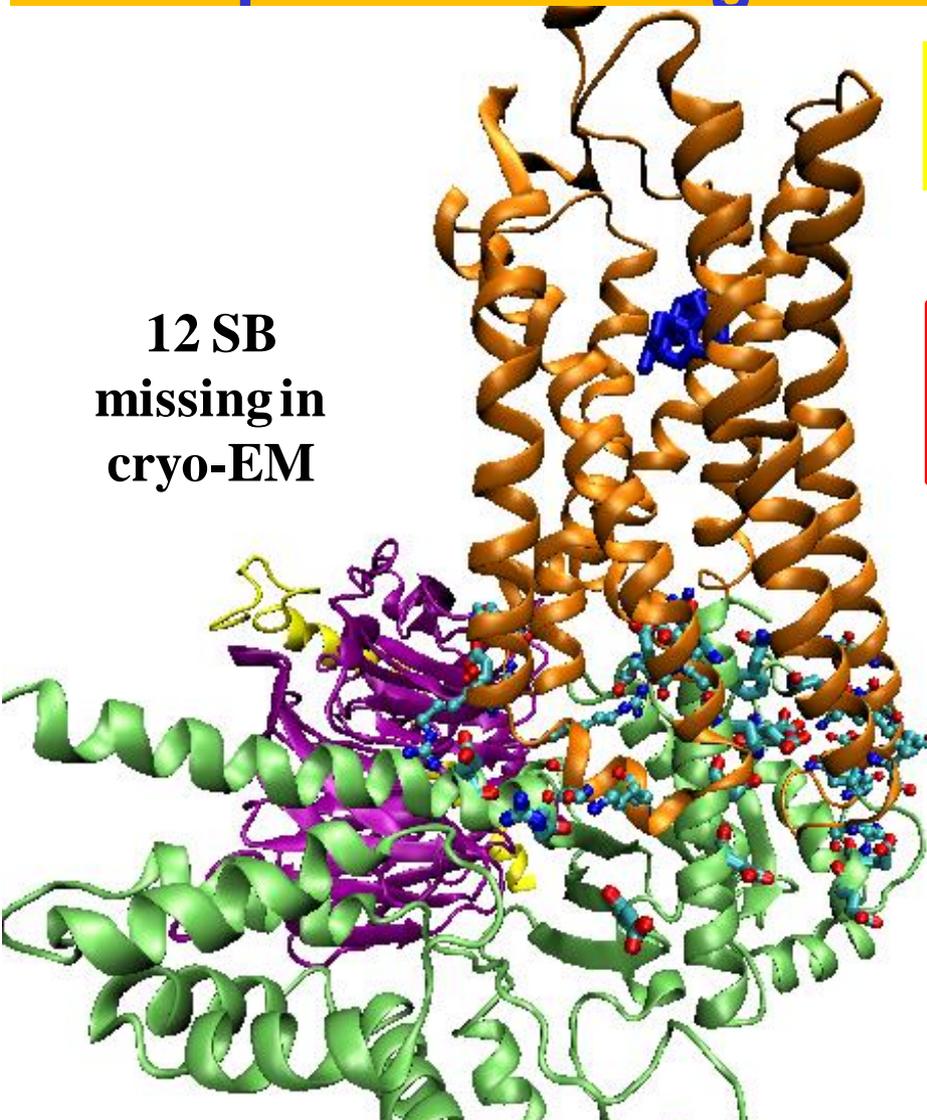
Spectroscopic methods for detecting conformational changes of  $\beta_2$ AR.



# Cryo-EM (2018) → structure of full GPCR with Gprotein and agonist – but not resolve SB,HB

We used SCREAM and metaMD to find the missing SB and HB

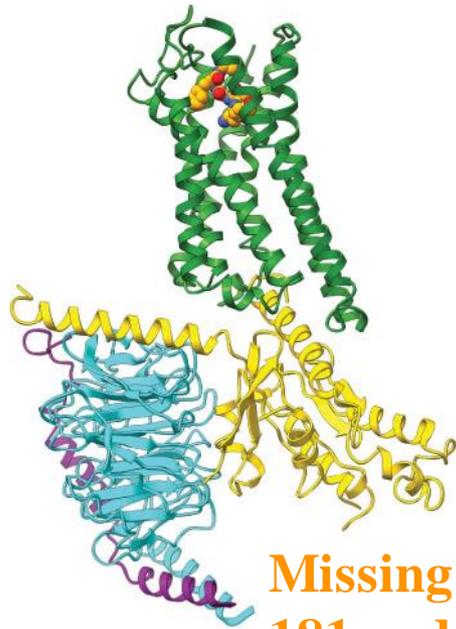
12 SB missing in cryo-EM



Mouse-μOR		Human-μOR	
m-MOR	G-alpha	h-MOR	G-alpha
R179(IL2)	D350	R181(IL2)	D350
R182(4.40)	E28	R184(4.40)	E28
D177(IL2)	R32	D179(IL2)	R32
R263(IL3)	D337	R265(IL3)	E298
D270(IL3)	K314	D272(IL3)	K314
K271(6.26)	D315	K273(6.26)	D315
R277(6.32)	F354(COO-)	R279(6.32)	F354(COO-)
K269(IL3)	E289	K271(IL3)	E289
K271(IL3)	D341	K273(IL3)	D341
m-MOR	G-beta	h-MOR	G-beta
K98(IL1)	D312	K100(IL1)	D312
m-MOR	m-MOR	h-MOR	h-MOR
R179(IL2)	D166(3.49)	R181(IL2)	D166(3.49)

Discovered 3 strong anchors: to ICL1, ICL2, ICL3  
Not previously found in xray or cryo-EM

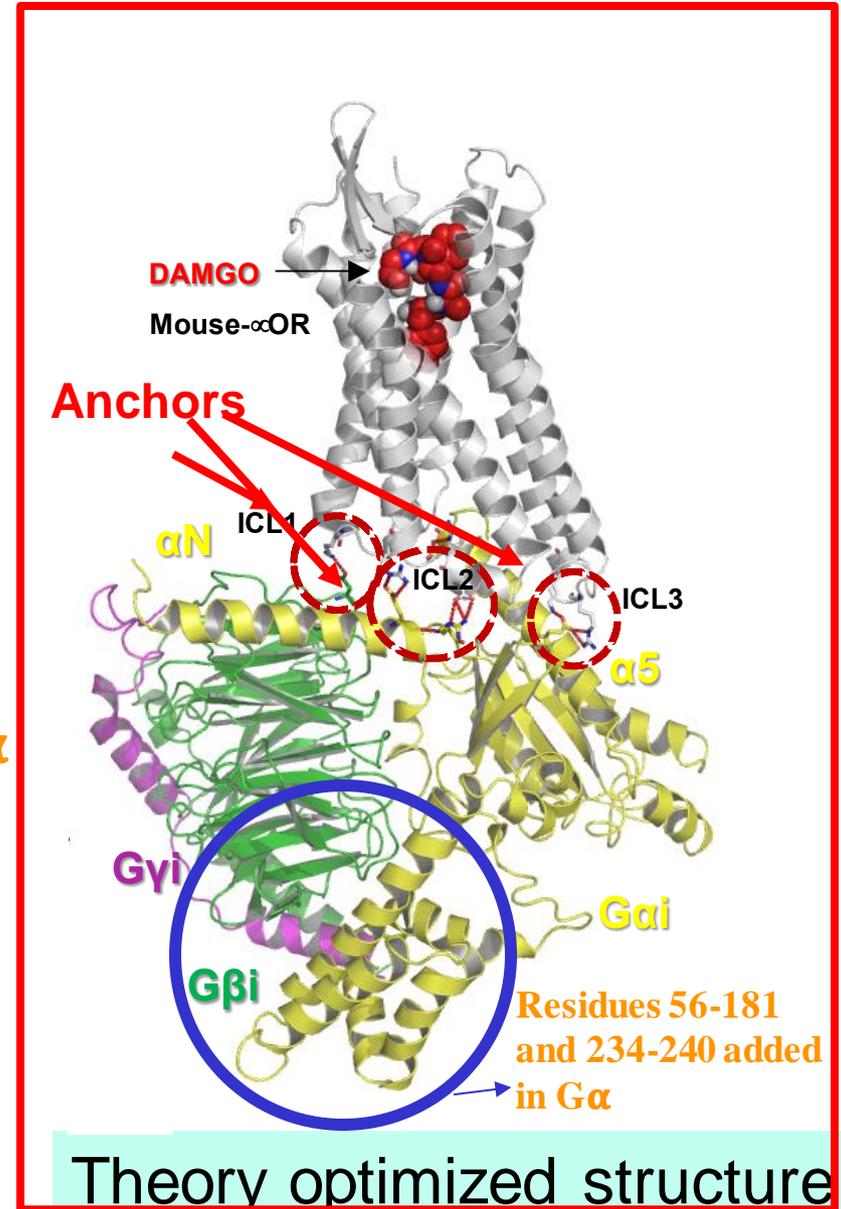
# Compare cryo-EM and theory $\mu$ -OR-GiProtein



Missing Residues 56-181 and 234-240 in  $G\alpha$

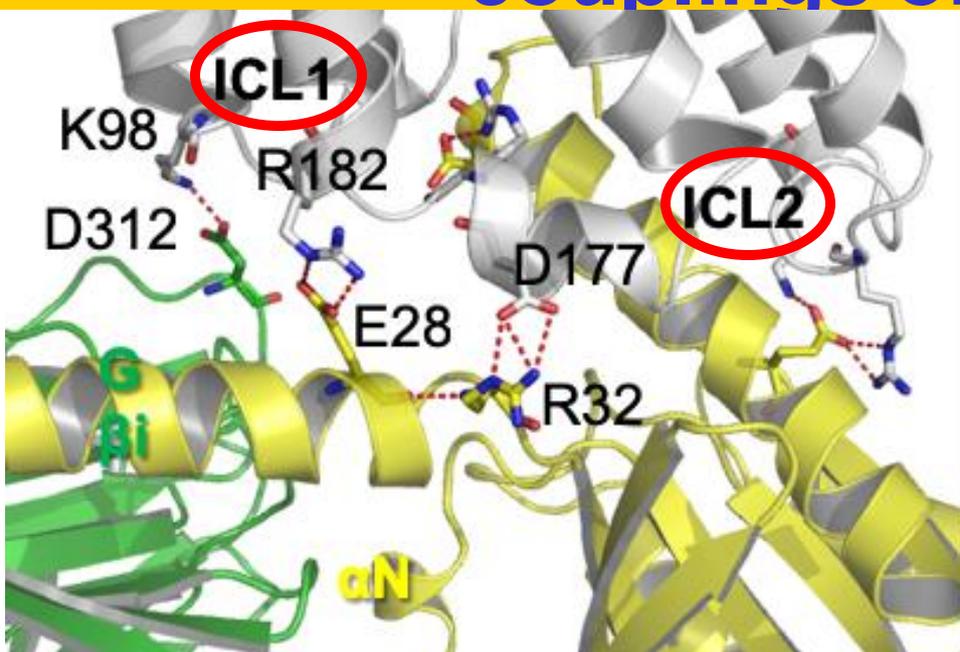
Cryo-EM mouse  $\mu$  opioid receptor-Gi complex (Resolution 3.2Å)

Keohl et al. *Nature* 2018, 558, 547

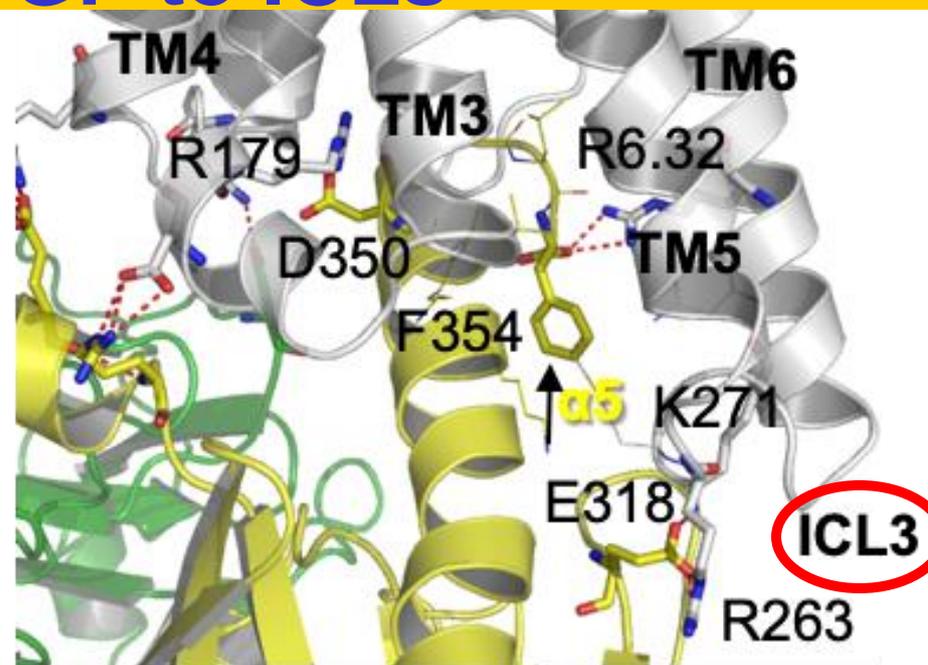


Theory optimized structure

# From full structure discovered strong couplings of GP to ICLs



anchors between ICL1 and ICL2  
of  $\mu$ OR and GiProtein



anchors between ICL3 of  $\mu$ OR  
and GiProtein

The 3 anchors from G-Protein to 3 Intracellular loops of GPCR  
Lines up the  $\alpha 5$  helix just right to insert into GPCR upon agonist  
activation

Removing agonist  $\rightarrow$   $\alpha 5$  helix retreats back toward GP but 3 anchors  
still couple strongly  $\rightarrow$  GProtein binds first NOT the agonist

# Summarize stages of activation

$\Sigma 0$ -apo

$\beta 2$  adrenergic receptor

M1

$\Sigma 4'$ -bind ligand  $\rightarrow$   
GP fully activated  
G $\alpha$  opens up loses  
GDP

$\Sigma 3$ -relax GP

$\Sigma 2$ -add GP

10 Å

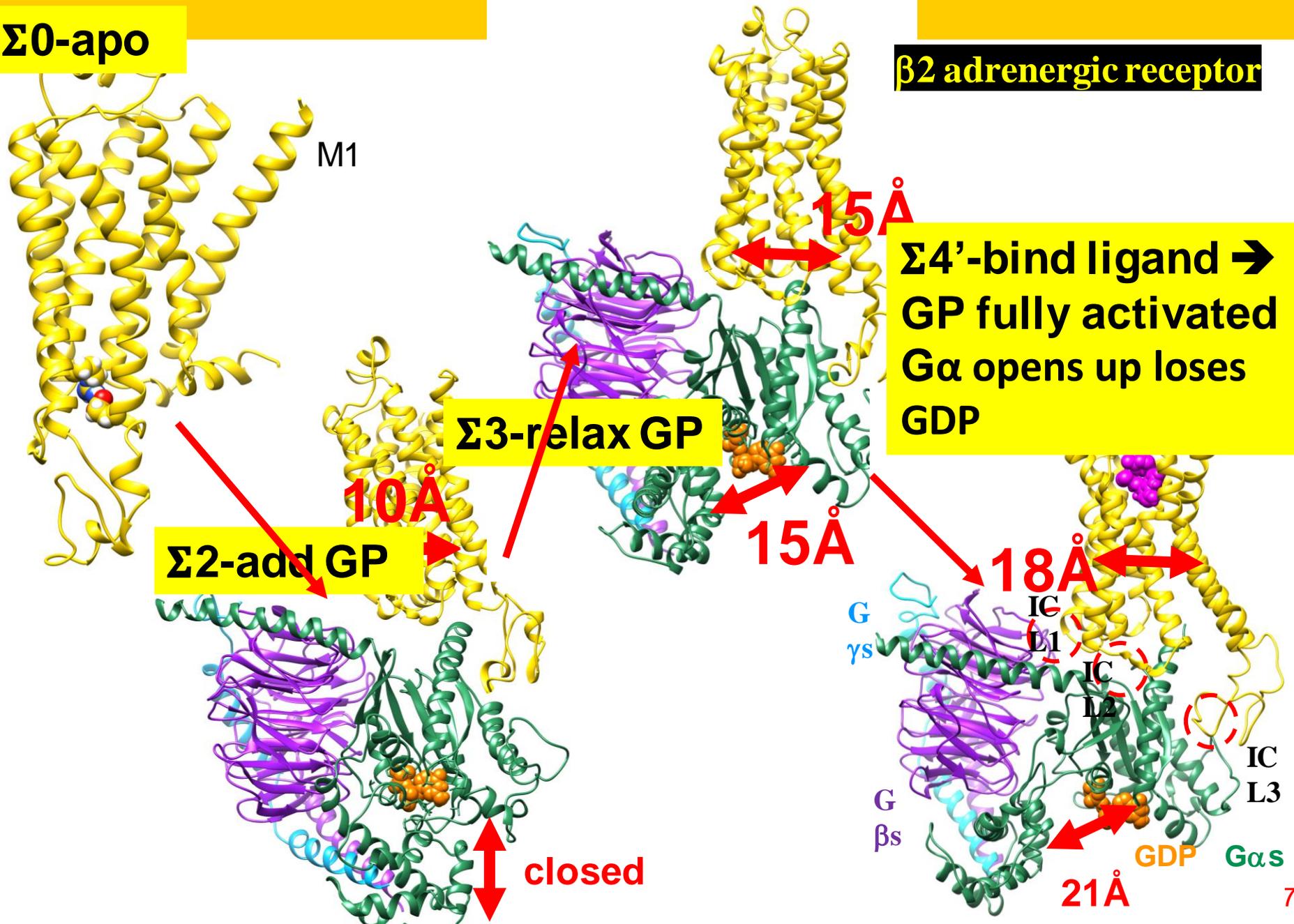
15 Å

15 Å

18 Å

closed

21 Å



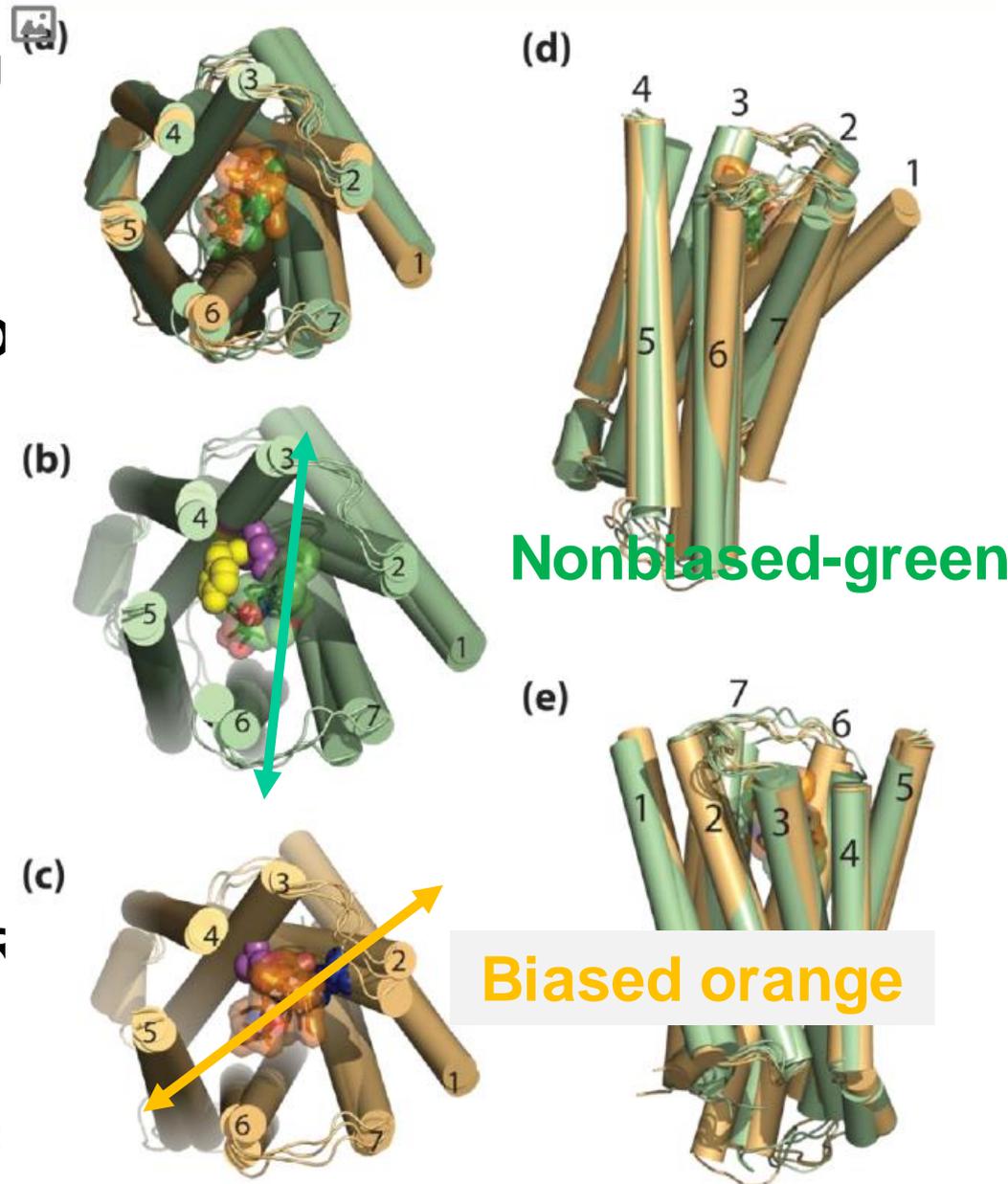
# Biased agonists for pain ( $\mu$ -opioid Receptor)

Morphine is a strong analgesic or painkiller through GProtein signaling from activation of opioid receptors (OR) in the central nervous system.

But morphine leads to potentially lethal **side effects mediated by  $\mu$ -O signaling through the  $\beta$ -arrestin pathway** or by actions at other receptors.

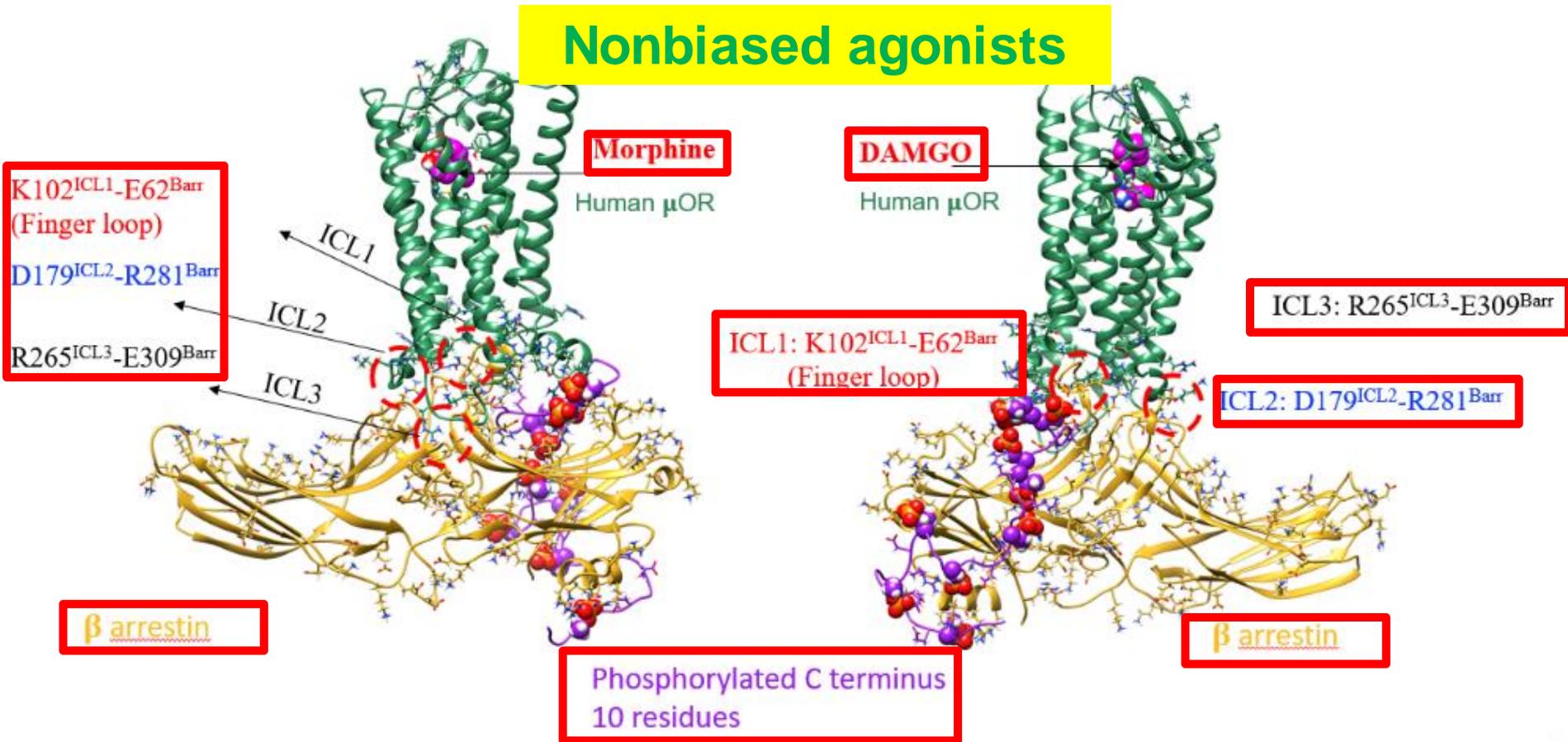
Biased Gi-protein activation with minimal  $\beta$ -arrestin recruitment has been shown alternative efficient and safe pain treatment to replace traditional narcotic analgesics.

There is **strong evidence that  $\mu$ -OR biased ligands that induce GP signaling but *not*  $\beta$ -arrestin signaling can relieve pain without the side effects**



# $\beta$ -arrestin-GPCR complex $\mu$ -Opioid Receptor

## Nonbiased agonists



Anchors to ICL2 and ICL3 play important roles in stabilizing  $\beta$ -arrestin-GPCR complex

# New Strategy for very accurate reactive force field

1. **PQEq Charge and Polarization model.** Tested against QM. Predicts QM polarization as point dipoles are brought into various molecules.
2. Determine **long range non-bond parameters** from PBE-D3 or P3LYP-D3 QM that describe Equation of State for molecular solids up to 100 GPa. Extract pure Pauli Principle repulsion (PR) and pure London dispersion (LD) attraction.
3. Given PR and LD, **extract valence parameters to fit exactly highest quality ab initio QM (CCSDT)** for bond breaking and reactions. Describe by ReaxFF formalism
4. Add in **Hydrogen Bonding corrections** from QM on dimers at CCSDT. 30,000 points fitted to analytic functions to  $\sim 4$  cm<sup>-1</sup> (Joel Bowman, Emory)

# Properties of water

	Expt.	RexPoN	TIP3P	TIP4P-2005	SPC/E	PBE	SCAN	MB-pol	CC-pol
$T_{melt}$	273.15	273.3	146	252	215	420	-	263.5	-
$S^0$	69.9	68.43	72.51	57.47	60.30	51.32	-	-	-
$\rho$	0.9965	0.9965	0.98	0.993	0.994	0.944	1.050	1.007	-
$\epsilon$	78.4	76.1	94	58	68	112	-	68.4	-
$\Delta H_v$	10.52	10.36	10.05	11.99	11.79	6.20	-	10.93	10.89
$R_1$	2.86	2.84	2.79	2.77	2.75	2.71	2.74	2.81	2.79
$g_{oo}(R_1)$	2.50	2.34	2.79	3.22	3.05	3.69	3.17	2.76	2.77

melting temperature  $T_{melt}$  (K) at 1 atm pressure.

standard molar entropy  $S^0$  (J/mol/K)<sup>27</sup>,

density  $\rho$  (g/cm<sup>3</sup>),

static dielectric constant  $\epsilon$ ,

heat of vaporization  $\Delta H_v$  (kcal/mol),

$R_1$  (Å) the position of the first peak of  $g_{oo}$

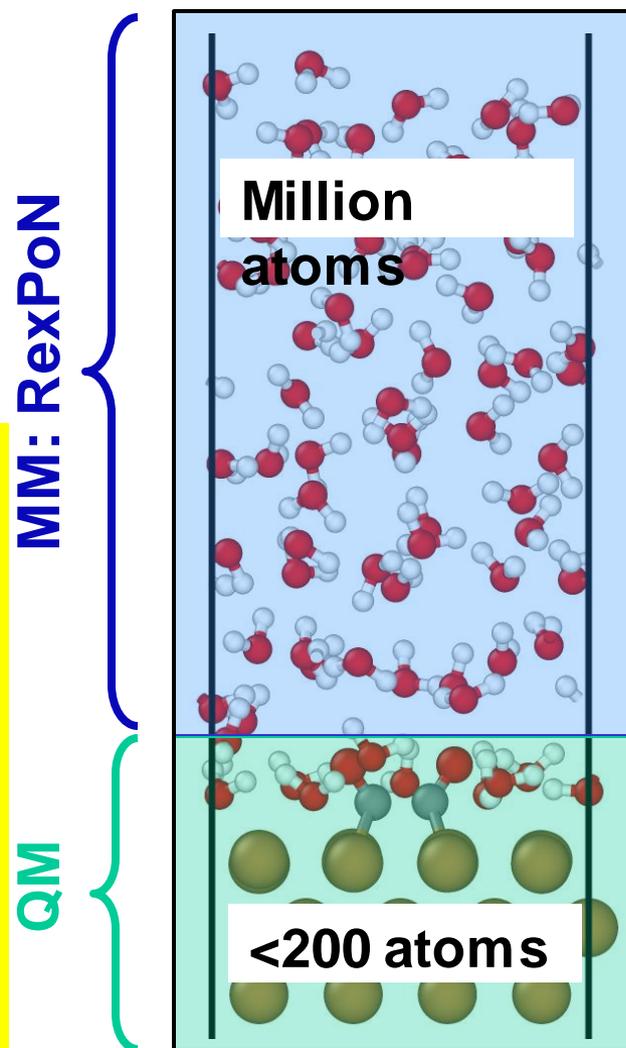
at T= 298 K, p=1 atm.

**RexPoN make NO use of empirical data**  
**Only high level QM**  
**But it leads to spectacular agreement with experiment**

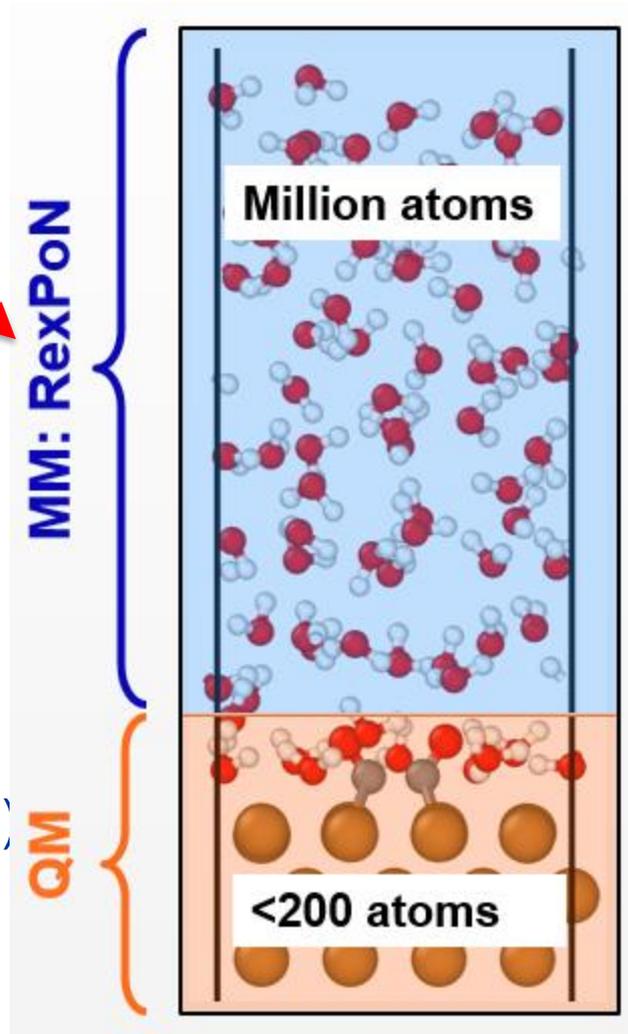
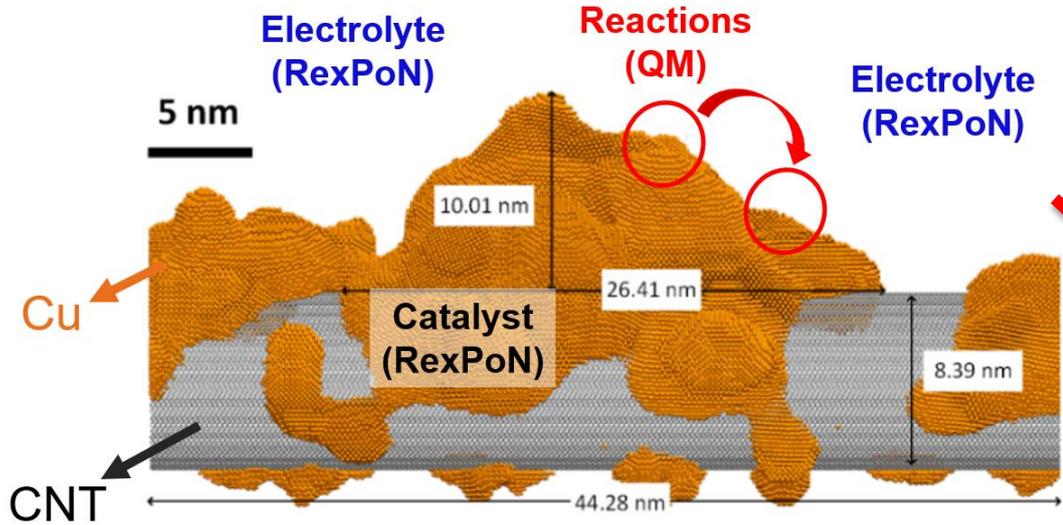
**skip the details, jump to the summary**

# RexPoN Embedded QM (REQM)

- QM/MM: a great idea (Noble prize in 2013) but many challenges!
- Explicit coupling is often needed; cannot include polarization effects of MM in the QM region.
- RexPoN accurately describes the long-range polarization effects (of full solvent) in QM region.
- RexPoN accurately describes the long-range van der Waal attraction from QM
- In addition, RexPoN bond breaking curve fits QM (CCS) exactly. no discontinuity in forces



RexPoN embedded Quantum Mechanics (REQM) describe full millions atoms with RexPoN QM based reactive FF  
**Select 200 atom region for QM, focus QM on reactions at the interface. Move QM region to examine different spots**

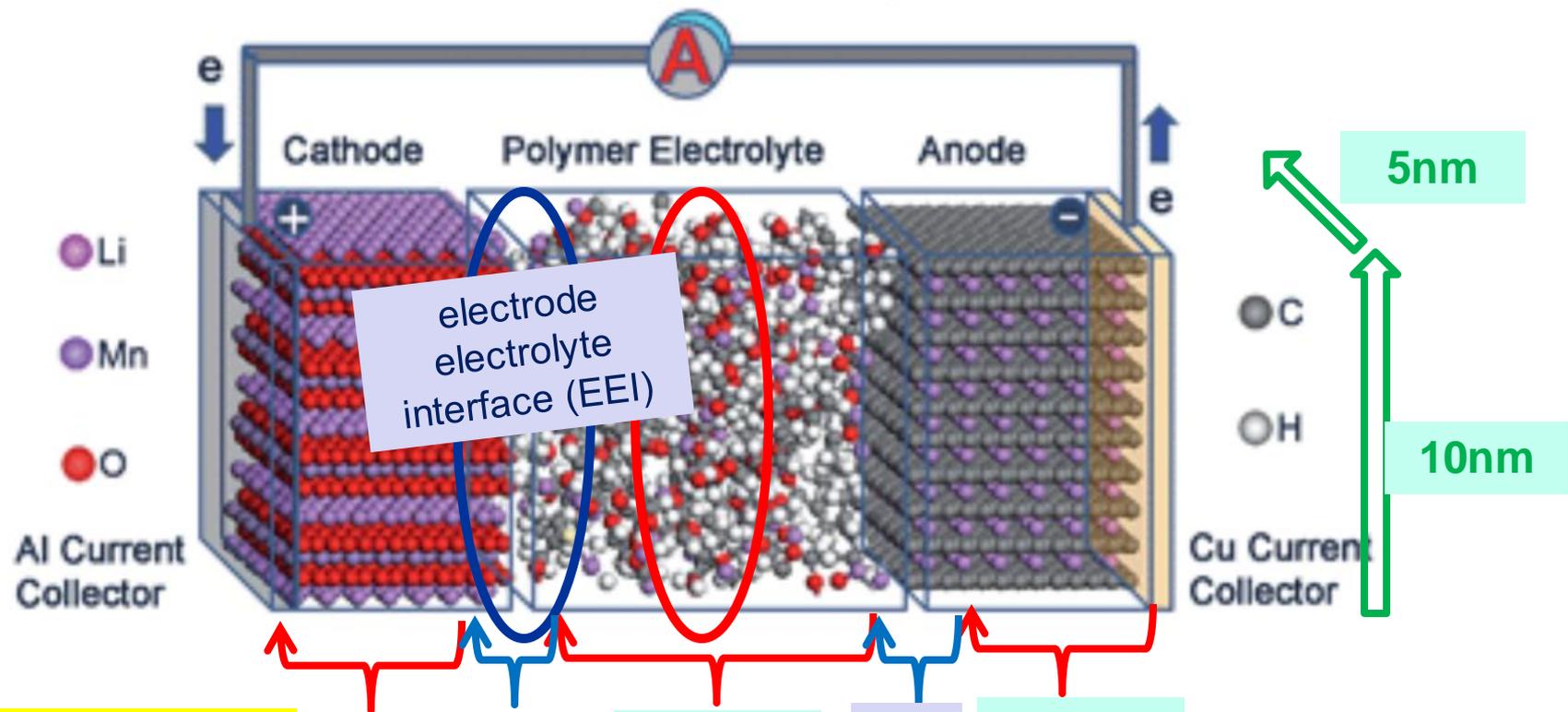


Periodic box 44nm x 22 nm x 22 nm (7 million atoms)  
 Carbon support (here carbon nanotube)  
 Catalysts deposit (200,000) atoms  
 Solvent: H<sub>2</sub>O, pH=1, positive and negative atoms

# Application new generations batteries fuel cells: predict chemistry & charge flow at electrode electrolyte interface (EEI) WHILE describing diffusion, ions etc throughout the device

Target: QM/RaxPoN on full device

Lithium Polymer Battery



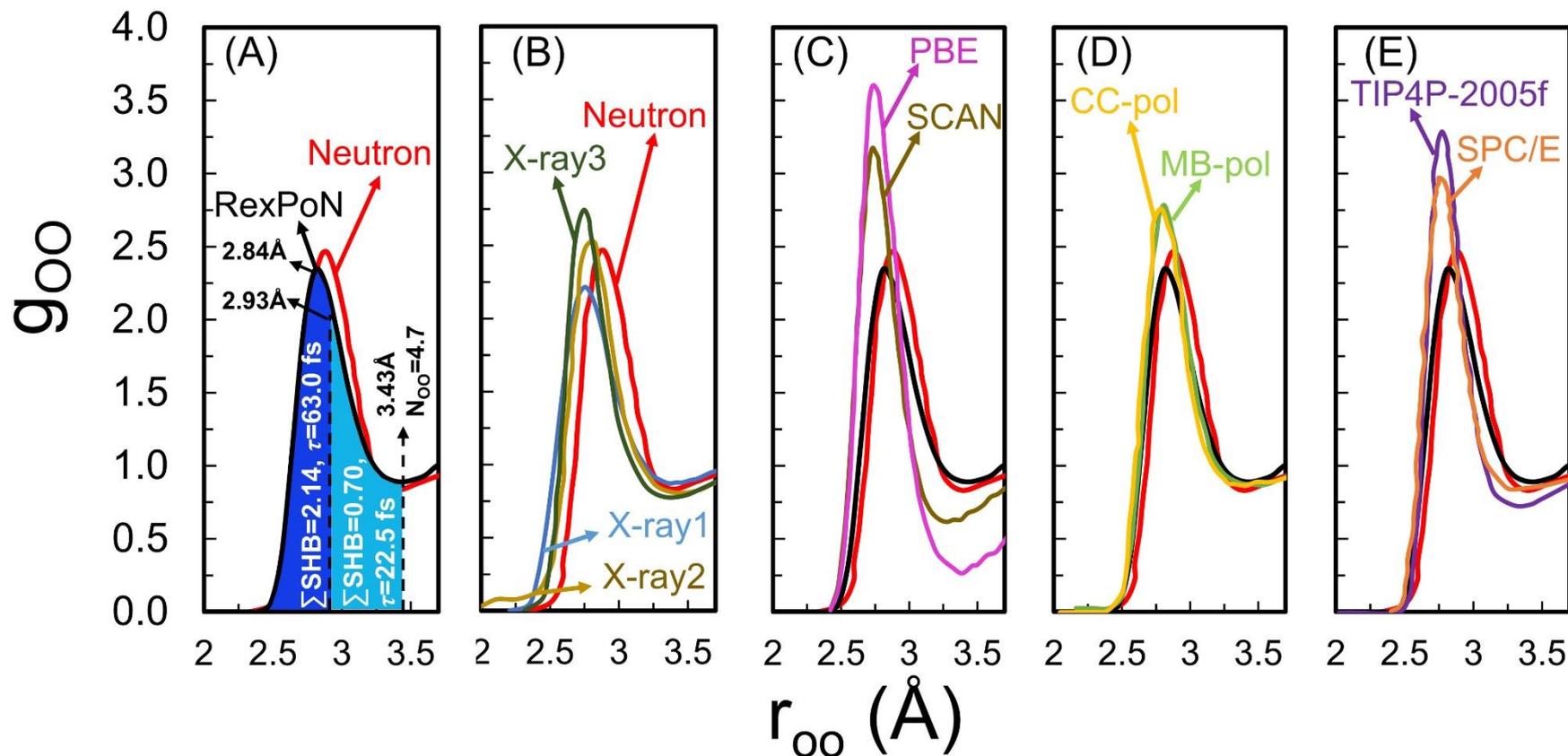
10 ps QM metadynamics  
Each point 1ns  
RexPoN



1.2 million atoms  
RexPoN  
300 atoms QM

# OO radial distribution

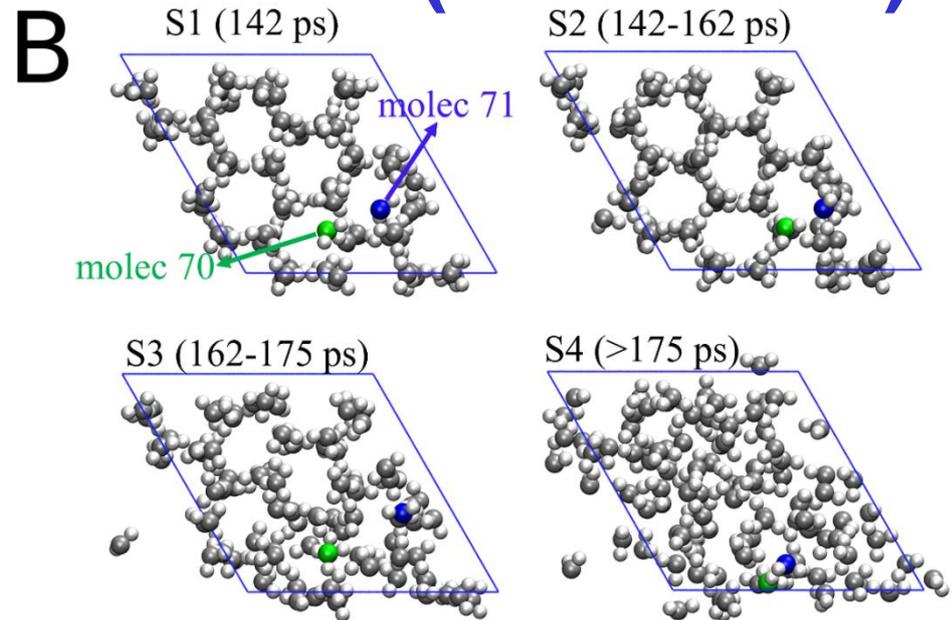
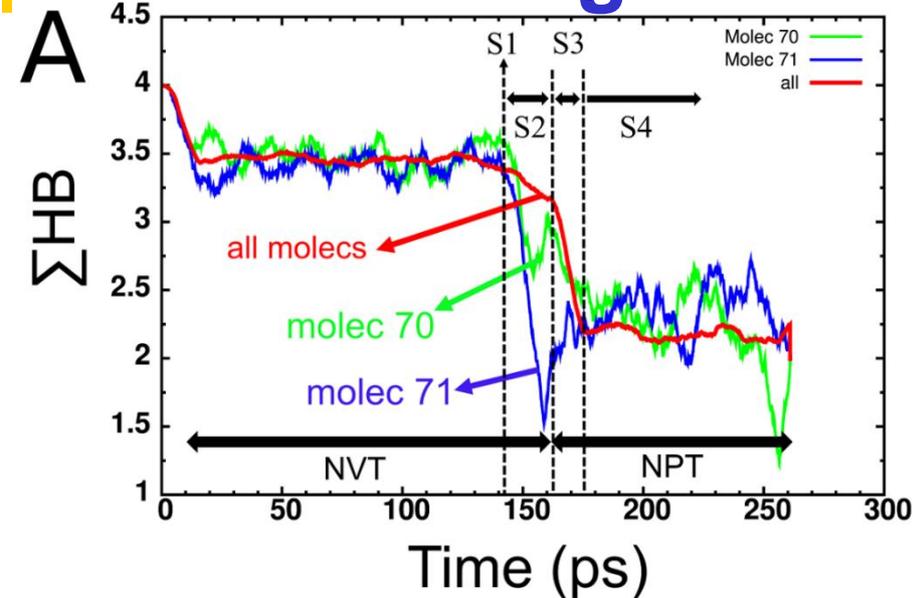
Only direct experimental property related to structure



Only RexPoN captures the OO radial distribution  
Even DFT QM (PBE and SCAN do NOT get it

# Initial steps of melting at 273.5K

define strong HB as  $OO < 2.93 \text{ \AA}$  (H<sub>2</sub>O dimer)



Water heated 0 K to 273.5 K over 10 ps followed by MD-NVT for 150 ps.

Volume allowed to adjust to 1 atm during 100 ps of MD-NPT simulations.

**Initiation melting dominated by blue and green waters in same hexagonal ring of ice** at 142 ps adjacent blue and green molecules start to break out of its hexagonal ring, sum SHB drops from 3.5 to 2.7 at 153 ps.

**The sum SHB of blue goes down to 1.5 at 158 ps Green goes back to 3 at 162ps**

162 to 175 ps the blue and green molecules in, out of the hexagonal rings repeatedly whole structure melts at 175ps → sum SHB=2.18

**Ice has 4 strong HBs, just after melting, water has only 2.2 strong HBs**

# Now we have the most accurate ever QM based description of water

Define Strong HB as  $R=2.93$  (H<sub>2</sub>O dimer)

sum of SHB on any H<sub>2</sub>O = 4.0 for ice

we find a sharp drop from sum SHB=4.0 for 273 K to 2.28 at 273K, which decreases to sum SHB=2.14 at 298K and 1.85 at 350K.

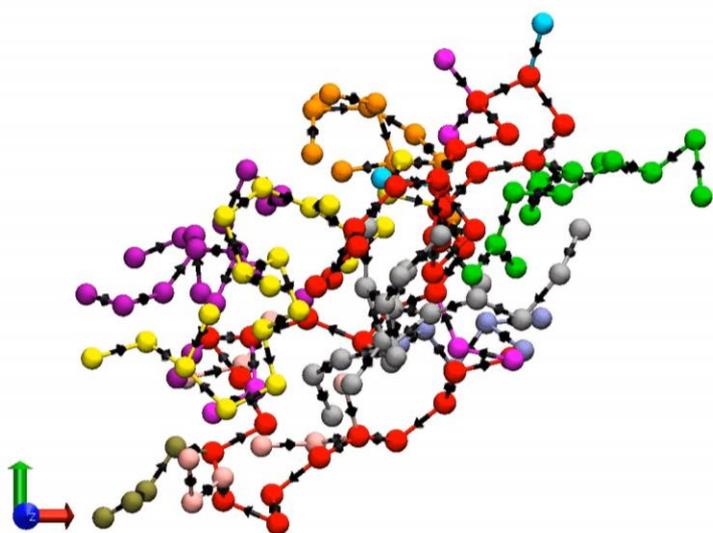
**We were shocked by these unexpected results.**

T (K)	$M_c$	$\sum$ HB	SD	SHB lifetime (fs)	Density (gr/cm <sup>3</sup> )
150	216 ( $\infty$ )	3.14	1.55	175	-
273.5	177 ( $\infty$ )	2.28	1.08	101	-
277	168 ( $\infty$ )	2.22	1.05	103	1.000
298	148 ( $\infty$ )	2.14	1.00	90	0.9965
350	72	1.85	0.84	78	0.960
400	36	1.63	0.71	68	-

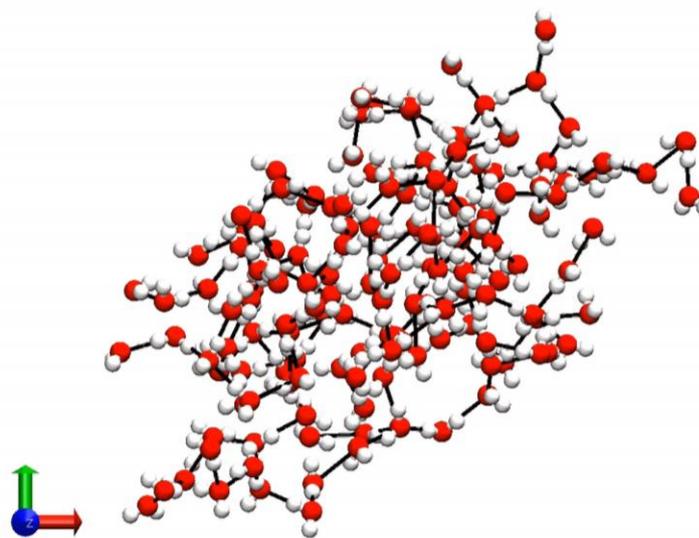
**Liquid water at 298K has ~2 Strong HB,  
with lifetime ~90 femtoseconds**

# Liquid Water is a Polymer

- The SHBs connect to form a **dynamic high molecular weight branched polymer**.
- Average number of strong HB at 298K  $\approx 2.0$ . That is a 1 dimensional chain structure.



(A) The polymer backbone (longest chain) is shown in red.



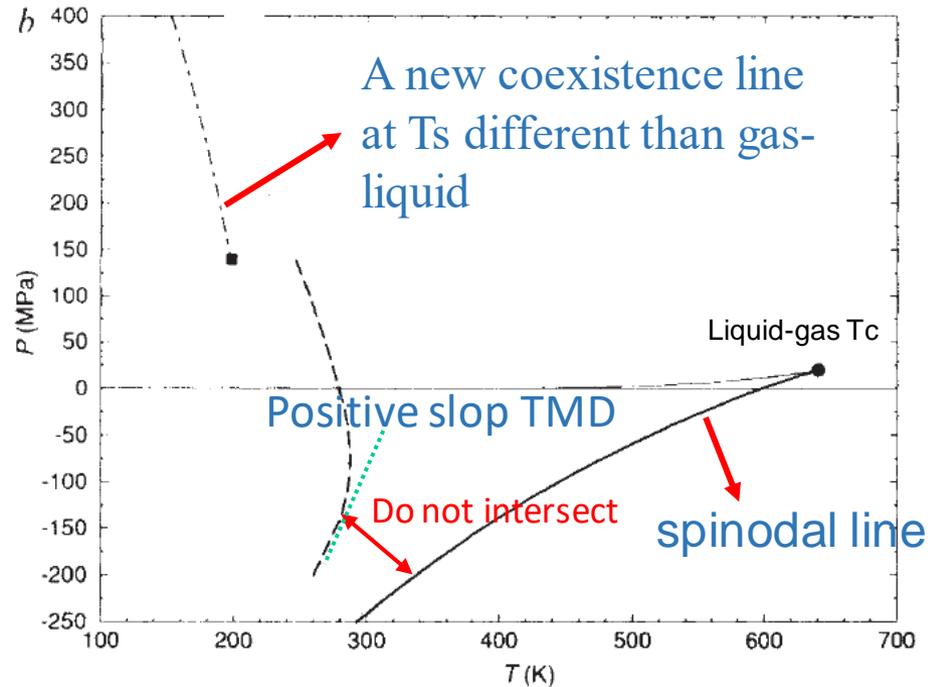
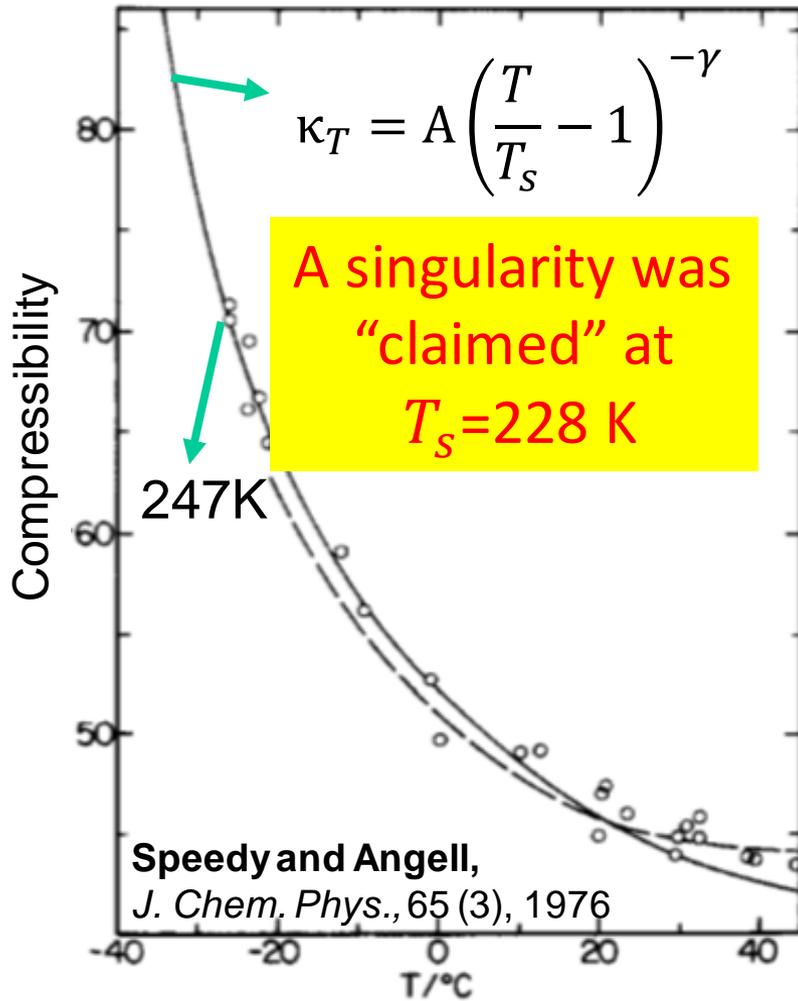
(B) Same structure as in (A) but showing the OH bonds.

# Anomalies of Deeply Supercooled Liquid Water

## Liquid-Liquid Critical Point (LLCP) Hypothesis -1992

2 Phases Liquid

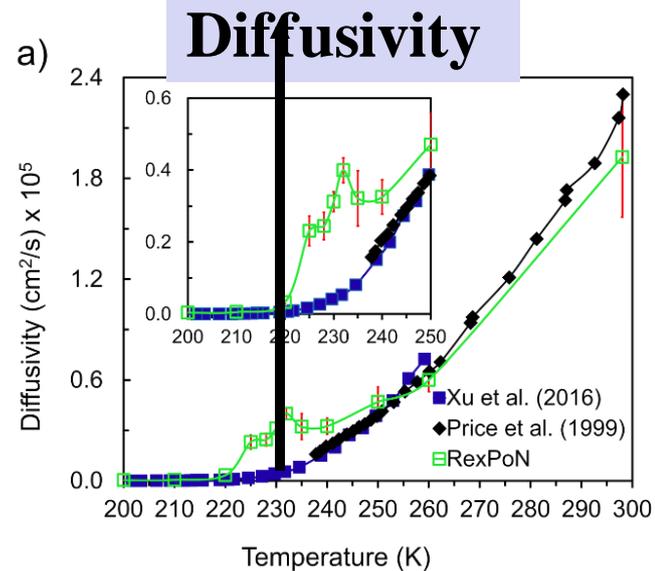
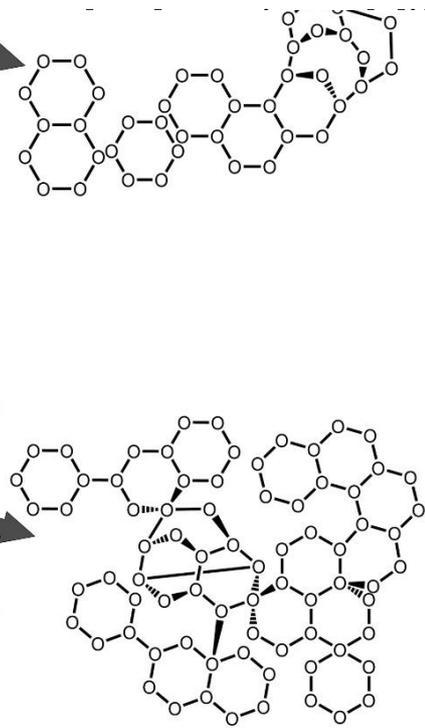
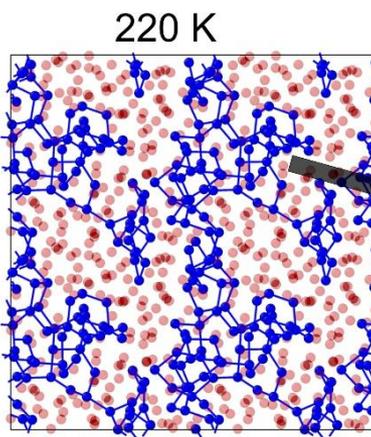
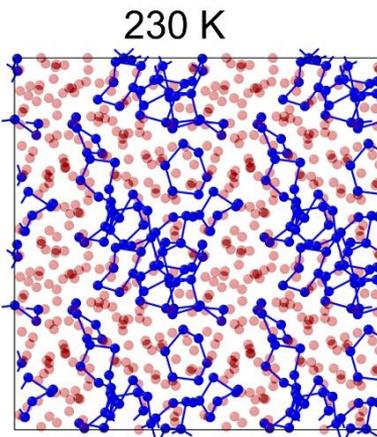
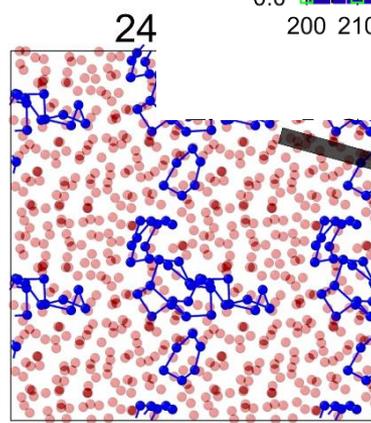
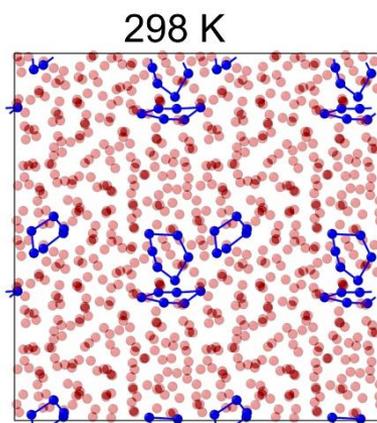
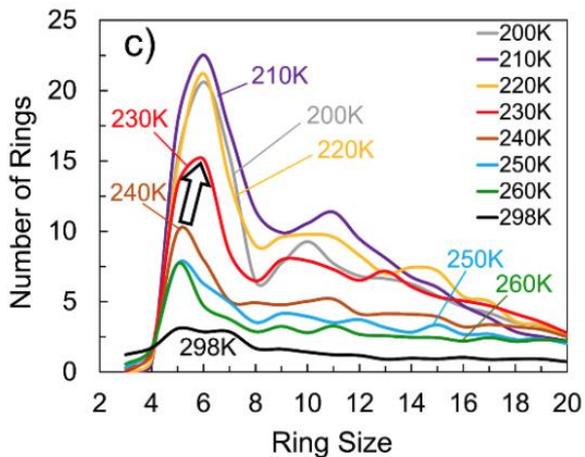
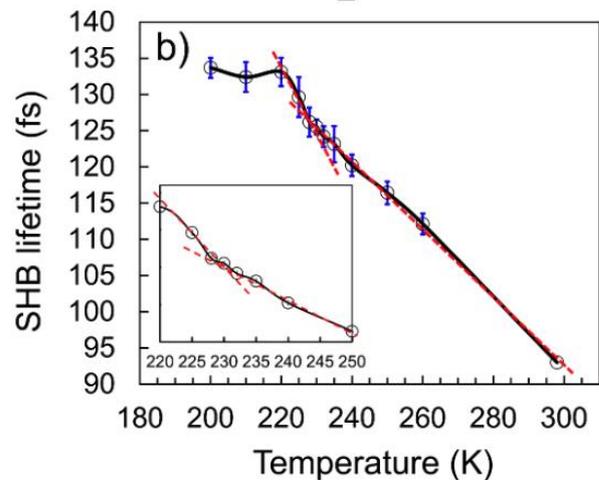
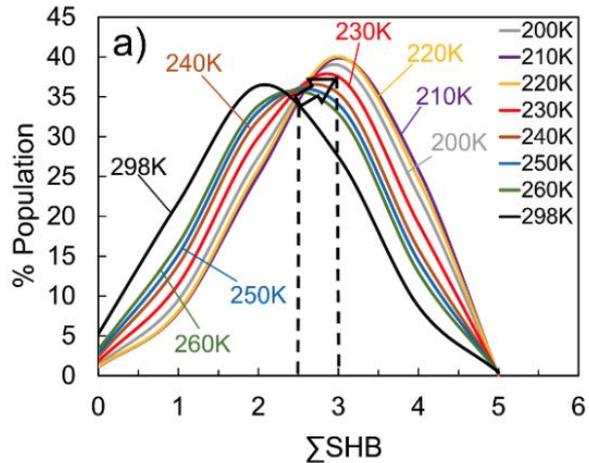
ST2 Model Simulations



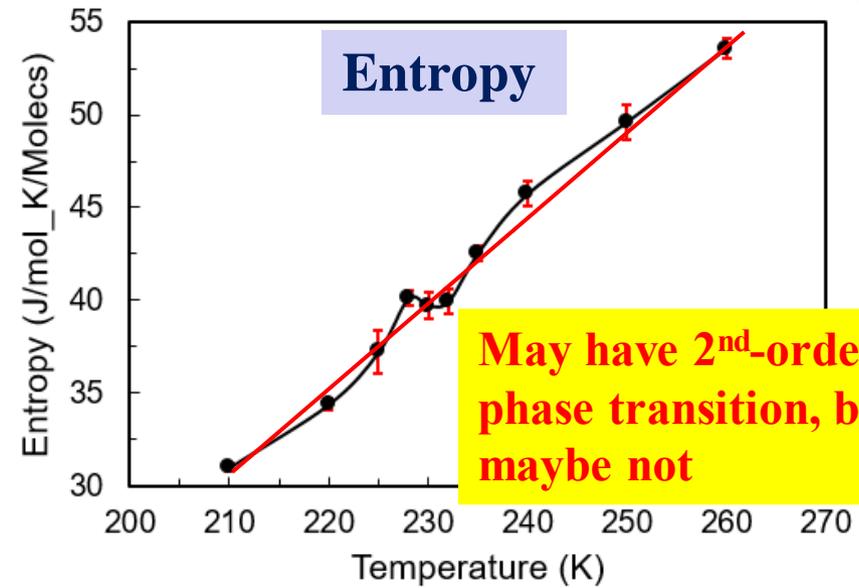
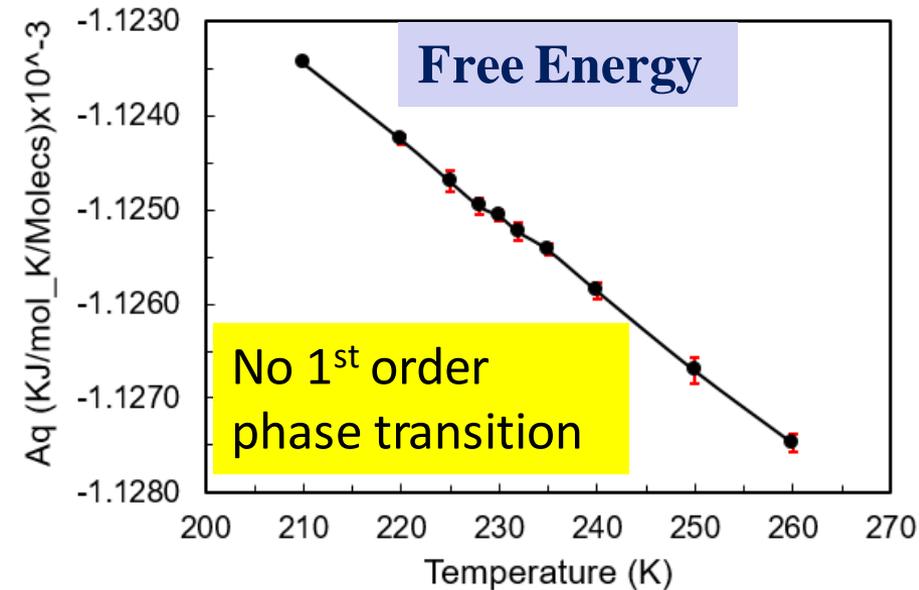
- Non-continuous spinodal line with positive slope.
- Coexistence of high and low density liquids at  $T_s$ .

1 at

Stanley et al , *Nature* 360, 6402 (1992)

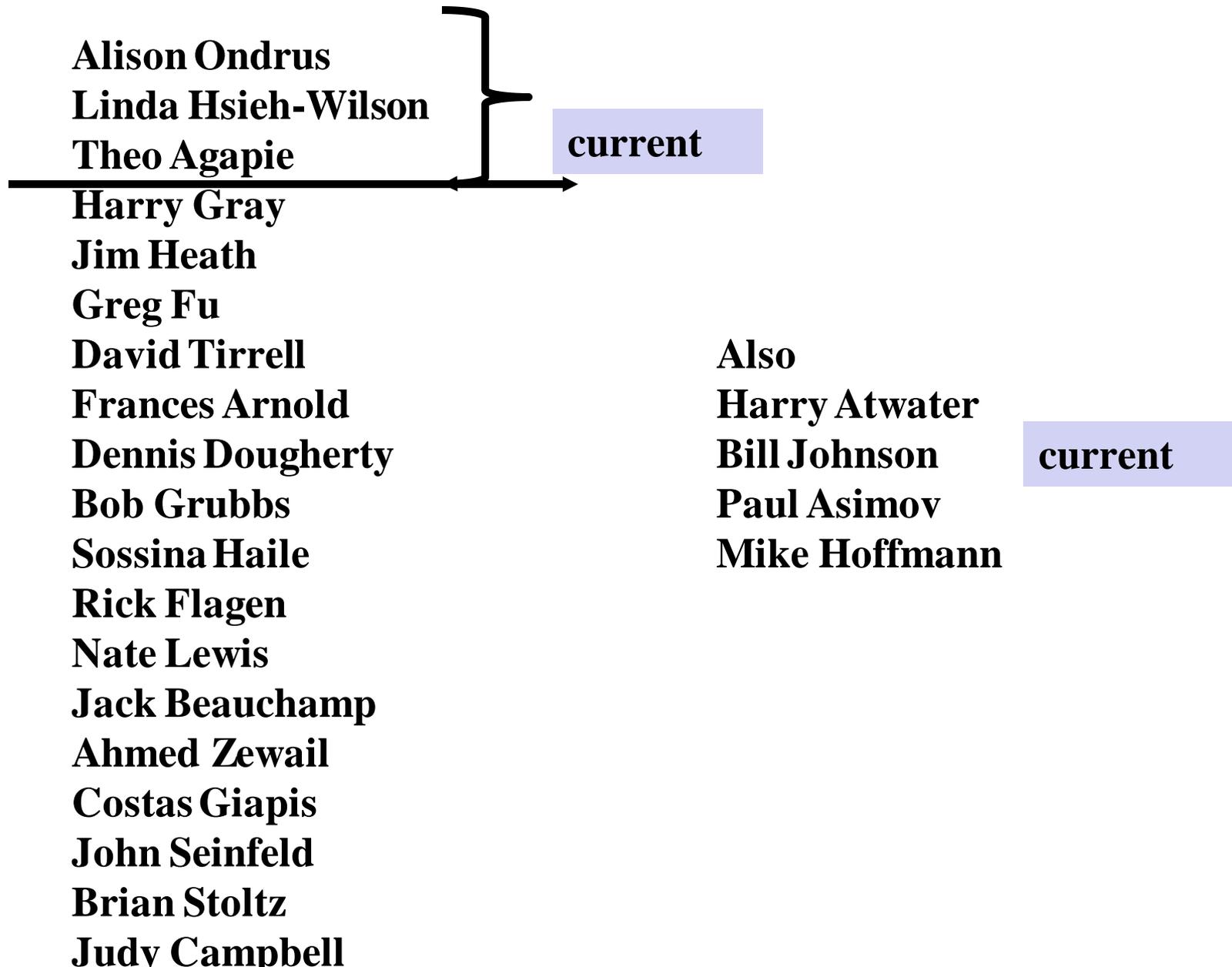


# Free Energy, Entropy, Diffusivity of H2O (using 2PT Method)



- Therefore, only one phase at each temperature.
- No Liq-Liq Crit point
- No phase transition.
- Instead have a topology transition, which leads to dramatic change in dynamical properties

# Experimental-Theory collaborations over the years in CCE



# First principles theory and simulation are now at the point where it can drive the design and development of new materials

Artificial photosynthesis

Photo CO<sub>2</sub> reduction

Fuel Cell,

Battery,

Solar cells

Water Purification,

CO<sub>2</sub> Sequestration

Energy

Hydrogen

**BIZARRO** By Dan Piraro

I want a computer that will do what  
I WANT it to do, not what I  
TELL it to do.

