

# Bridging from Atomistic to Continuum Modeling to Better Predict Nuclear Waste Glass Dissolution: Bulk glass, glass surfaces, and gels

**Louise J. Criscenti and Kideok Kwon**  
Sandia National Laboratories  
Albuquerque, NM

Collaborators: Peter Zapol and Haiying He (ANL)

May 24, 2012

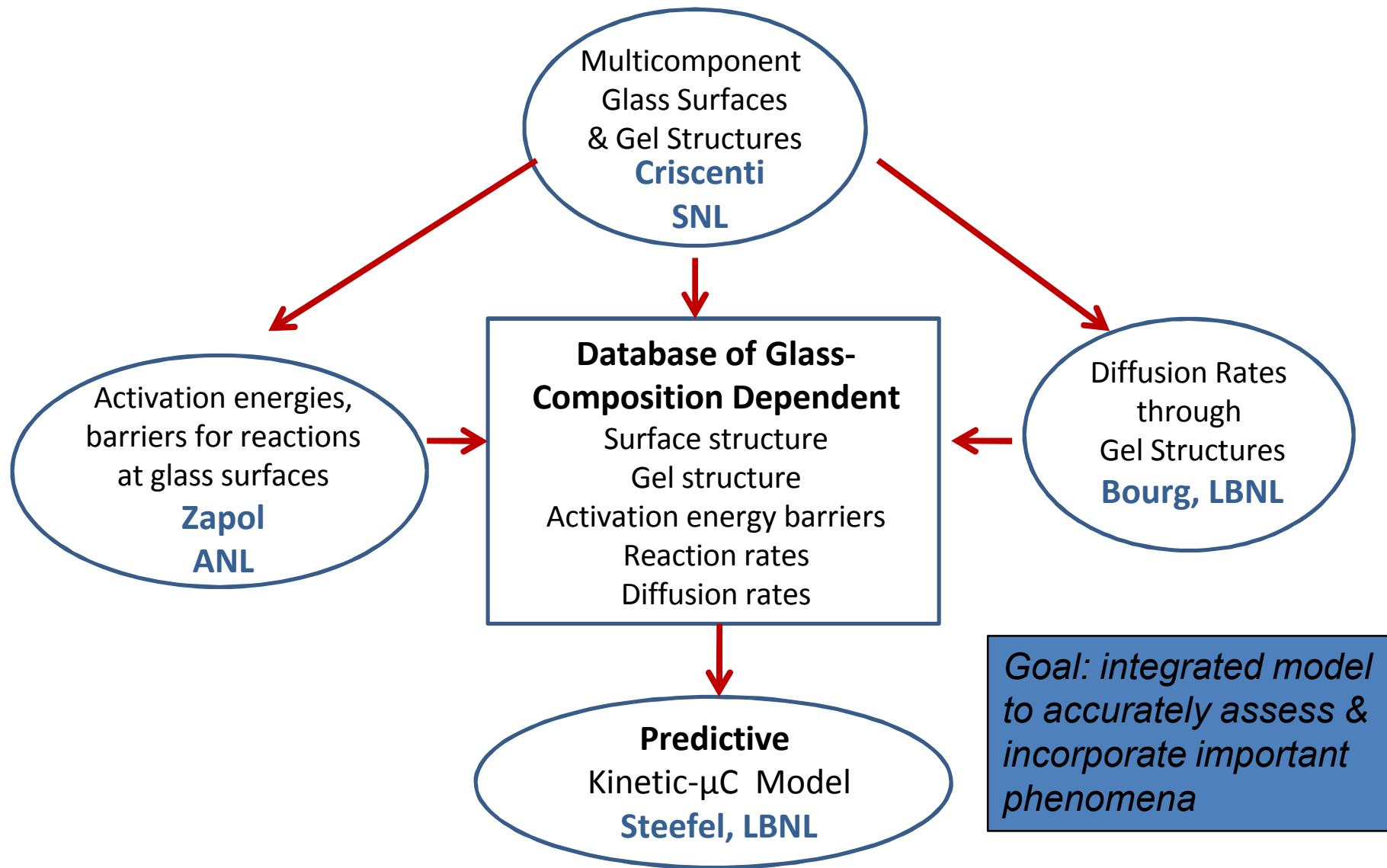
Sandia National Laboratories is a multi-program laboratory managed and operated by Sandia Corporation, a wholly owned subsidiary of Lockheed Martin Corporation, for the U.S. Department of Energy's National Nuclear Security Administration under contract DE-AC04-94AL85000.



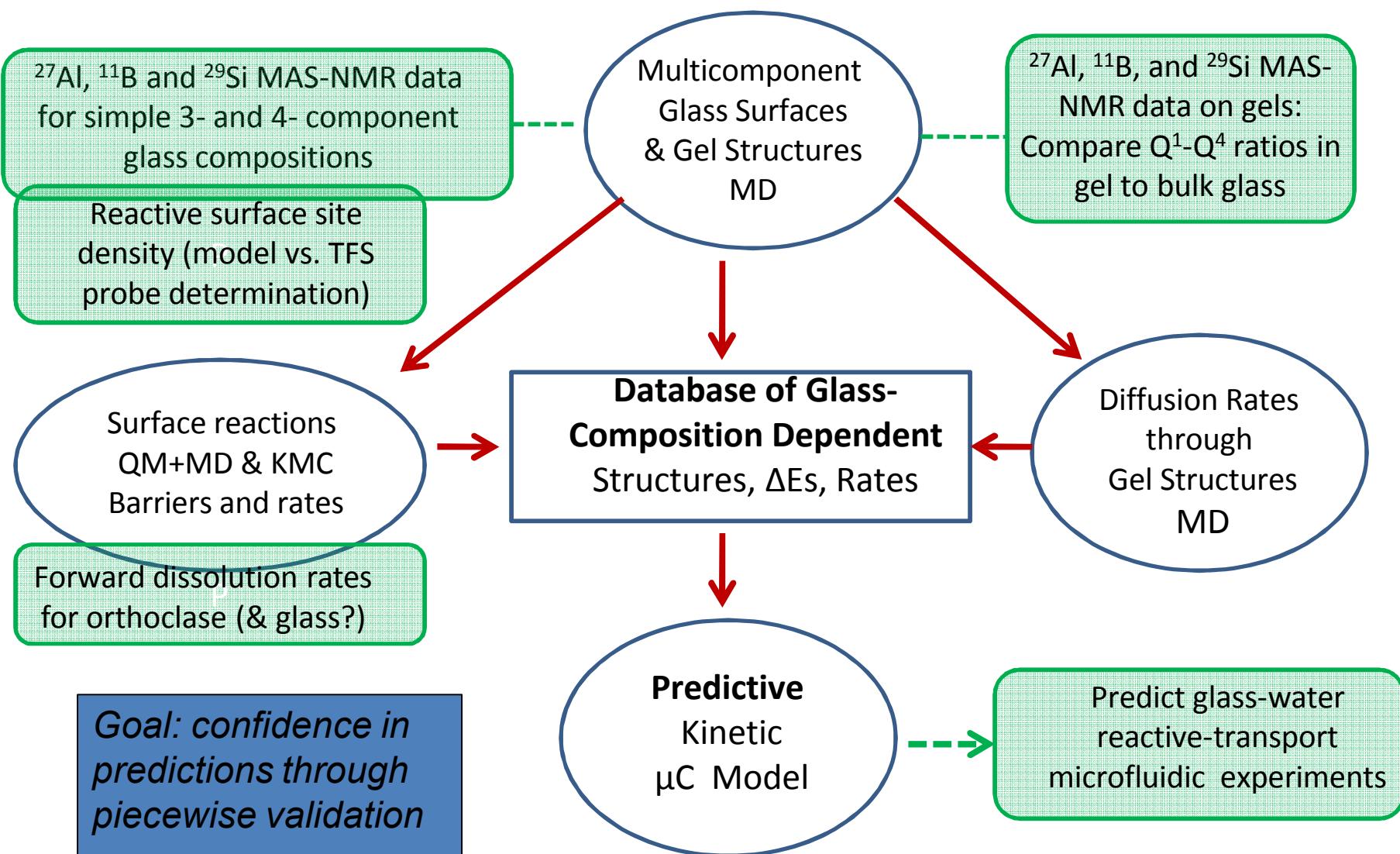
**Sandia National Laboratories**



# Integrated Multiscale Modeling Approach

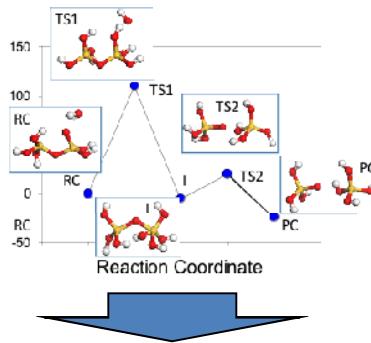


# Joint modeling-experiment efforts

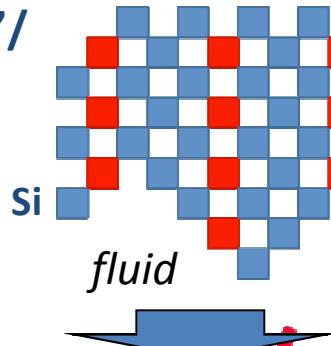


# Upscaling strategy: Assess

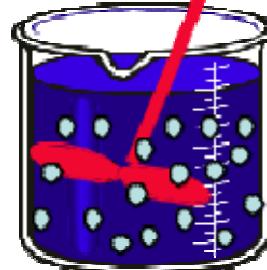
Atomic/  
Quantum:  
- Energies



“Mesoscale”/  
KMC:  
- Statistics  
& rates



Continuum:  
Reactive  
transport/  
Kinetic- $\mu$ C

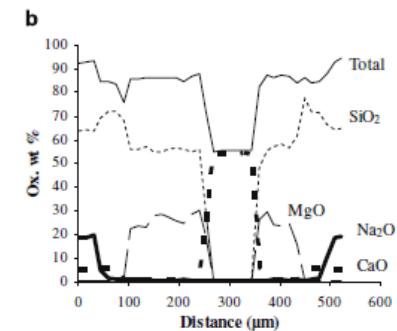


“Simple dissolution”: orthoclase  
+ amorphous glass  
+ multicomponent glass  
+ conformational (gels)  
+ secondary phases

...

**Identify process**  
**Incorporate into model**  
**Validate model vs. data**  
**Assess impact of process**

? =



# Upscaling strategy - incremental *validated* complexity

- Start with feldspar/orthoclase ...
  - Simple crystalline structure - well controlled model system
  - Good experimental characterization - validate key elements of upscaling
- ...add amorphous character
  - Glass introduces structural complexity, statistical behavior
  - Similar chemistry, test-verify/validate another element of upscaling
- ... add multicomponent
  - Test-verify/validate another element of upscaling
- ... add conformational complexity
  - Bulk/surface/gel ... cracking, porosity contributions to surface area,
  - Test-verify/validate conformational models
- ... connect to downscaling continuum models
- ... develop/refine constitutive models for continuum scale dissolution

# Molecular Models of Bulk Glass, Glass Surfaces and Surface Gels

- How does the glass composition influence long-term glass dissolution rates?
  - How are changes in glass composition reflected in the glass surface?
  - How do changes in surface site types and densities influence glass dissolution rates?
- Does bulk glass composition influence gel structure?
  - What role does condensation play in gel formation?
  - How does this impact the transport of ions and water through the gel?

## Approach:

- Develop bulk and surface models for multicomponent (B, Al, Si, O, Na) glass compositions representative of nuclear waste glasses through molecular dynamics simulations
- Validate calculated glass structures and surfaces with experimental data such as  $^{27}\text{Al}$  and  $^{29}\text{Si}$  MAS-NMR data from Pierce et al. (2010).
- Compare NMR-determined and calculated surface density of reactive sites for different glass compositions.
- Use molecular dynamics simulation techniques to evaluate the impact of glass composition on gel layer formation and structure.

# Force Field Models For Glass Structures and Glass-Water Interaction

## State-of-the-art force field models for multi-component glasses:

- Pedone et al. (2006)
  - Self-consistent empirical interatomic potential model for silica-based glasses
  - Can model structures and mechanics of multicomponent glasses with different compositions
  - Published papers are on alkali silicate glasses (Li, Na, K) and Na-Ca silicate glasses.
- Garofalini (1990s)
  - Reactive force field model that provides good structural results
  - Dissociation of water
- Teter Potentials
- **Kieu, Delaye, Cormier, Stoltz (2011)**
  - Delaye and Galeb have been collecting NMR, WAXS, and other analytical data on nuclear waste glass for over 15 years.
  - Developed a force field to look at multicomponent nuclear waste glass compositions (Na, Ca, Al, Si, B, Zr, O.).
  - Kieu et al. (2011) is a new NaO-B2O3-SiO2 force field.

# Molecular Dynamics Simulations of Na-borosilicate Glass

Kieu et al. (2011)'s composition-dependent pair potential

$$U(r_{ij}) = \frac{q_i q_j e^2}{r} + A_{ij} \exp\left(-\frac{r_{ij}}{\rho_{ij}}\right) - \frac{C_{ij}}{r_{ij}^6}$$

Long-range Coulombic interaction

Short-range repulsion

Dispersion

\*  $q_i, A_{ij}, \rho_{ij}, C_{ij}$  depend on  $R$  &  $K$   
(i.e., composition dependent)

$$R = \frac{[Na_2O]}{[B_2O_3]}, \quad K = \frac{[SiO_2]}{[B_2O_3]}$$

# Molecular Dynamics Simulations of Na-borosilicate Glass

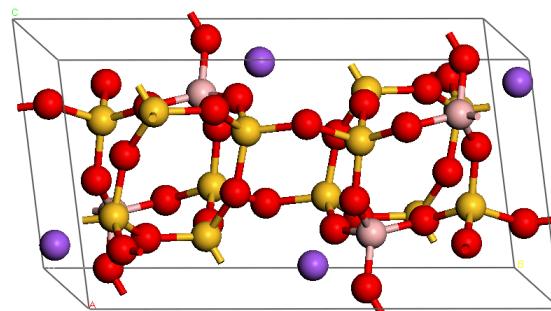
## Na-borosilicate crystal

Structural parameters	EXP <sup>a</sup>	Our MD results	DFT <sup>b</sup>
$a, b, c$ (Å)	7.84, 12.37, 6.81	7.77, 12.55, 6.83	7.89, 12.46, 6.84
$\alpha, \beta, \gamma$ (°)	93.3, 116.4, 92.0	93.7, 116.8, 91.9	93.4, 116.2, 91.8
density (g/cm <sup>3</sup> )	2.78	2.76	2.72
$d(\text{Si-O})$ (Å)	1.61	1.60	1.60
$d(\text{B-O})$ (Å)	1.47	1.47	1.49
$d(\text{Na-O})$ (Å)	2.54	2.60	2.53

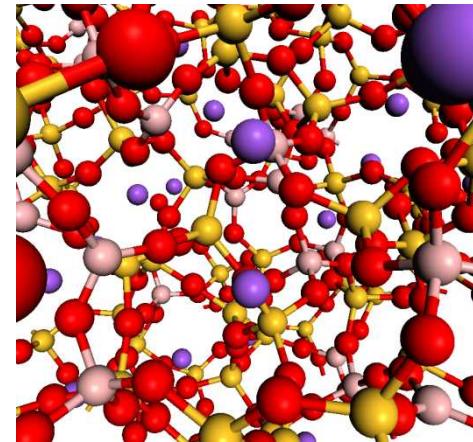
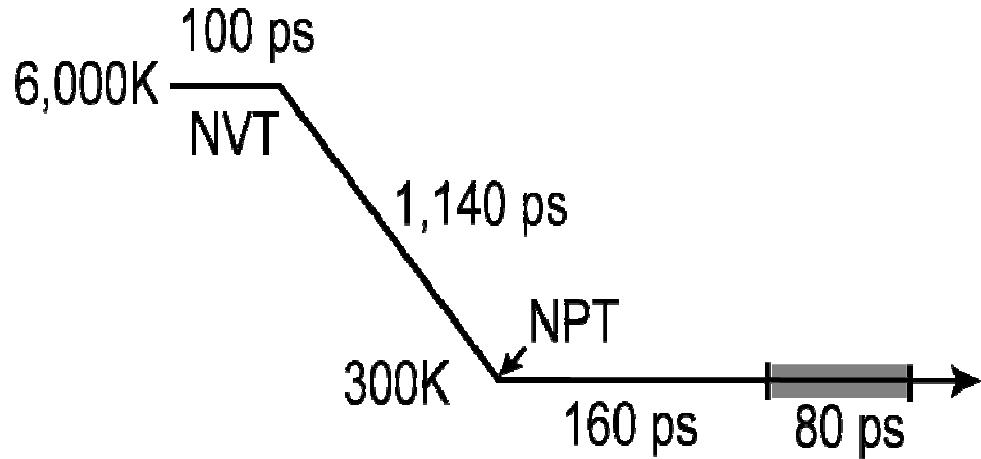
<sup>a</sup>Downs et al. (1999) Am. Mineral. 84, 333.

<sup>b</sup>GGA/PW91 (Argonne National Laboratory)

Reedmergnerite ( $\text{NaBSi}_3\text{O}_8$ )



# MD Simulations of Na Borosilicate Glass

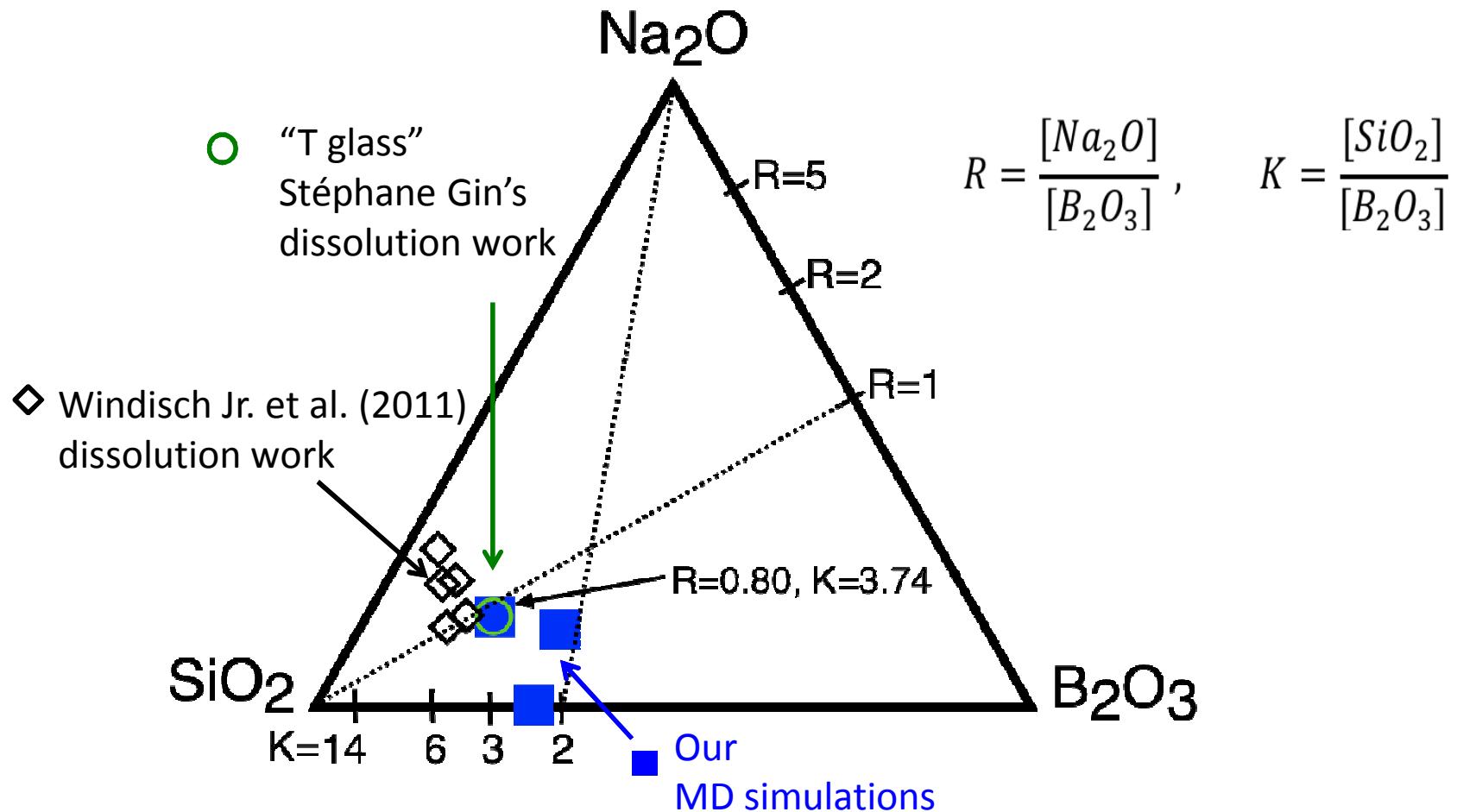


LAMMPS code with  
1,004 or 1,005 atoms

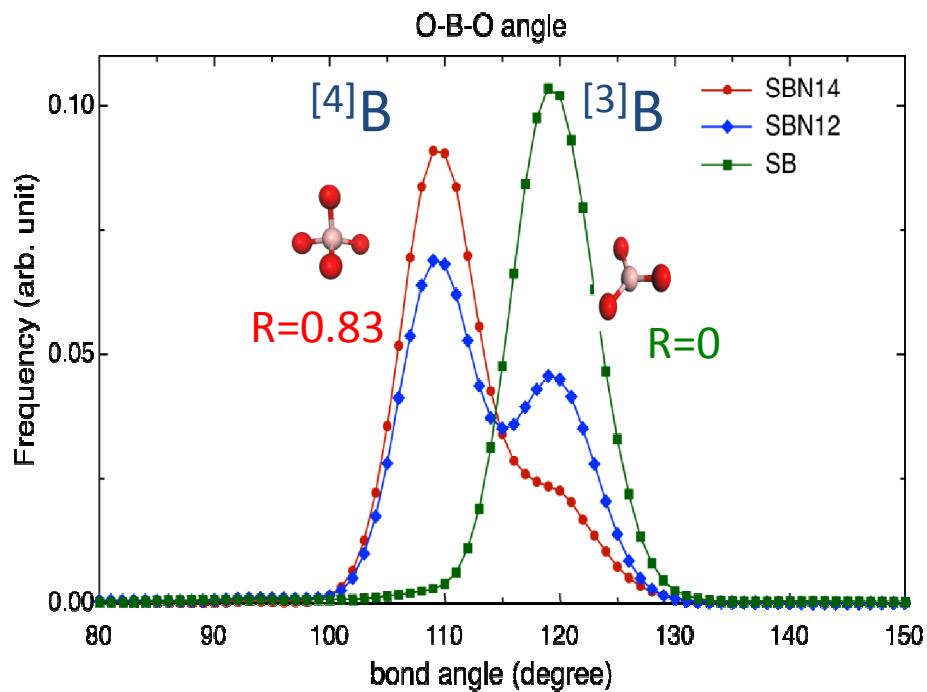
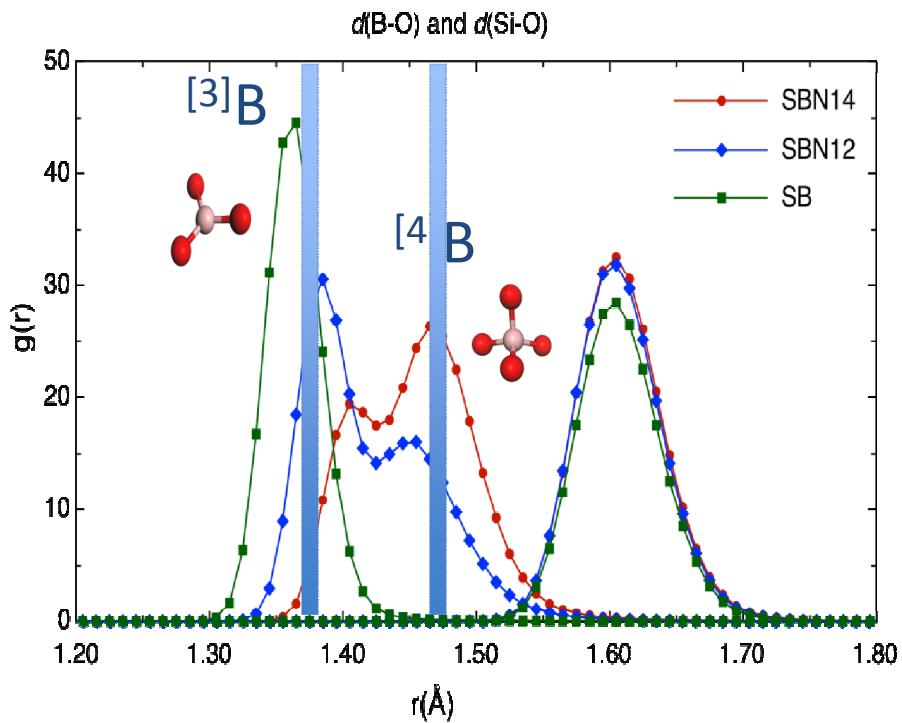
	SiO <sub>2</sub> (mol%)	B <sub>2</sub> O <sub>3</sub> (mol%)	Na <sub>2</sub> O (mol%)	R	K	density (g/cm <sup>3</sup> )	
						EXP.*	Our MD
SB	69.5	30.5	0	0	2.28	2.04	2.01
SBN12	59.66	28.14	12.20	0.43	2.11	2.37	2.39
SBN14	67.73	18.04	14.23	0.80	3.74	2.45	2.44

\*Kieu et al. (2011) and references therein.

# Composition of simulated glass models



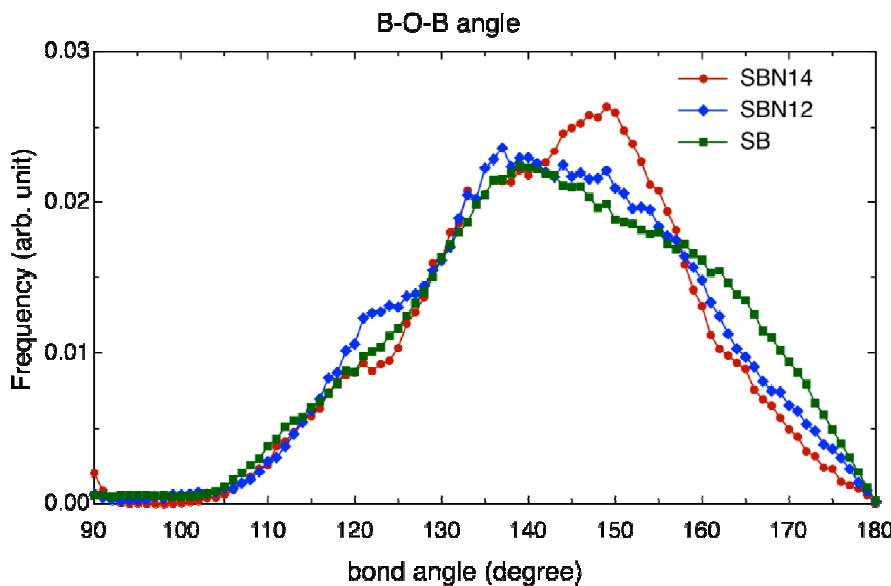
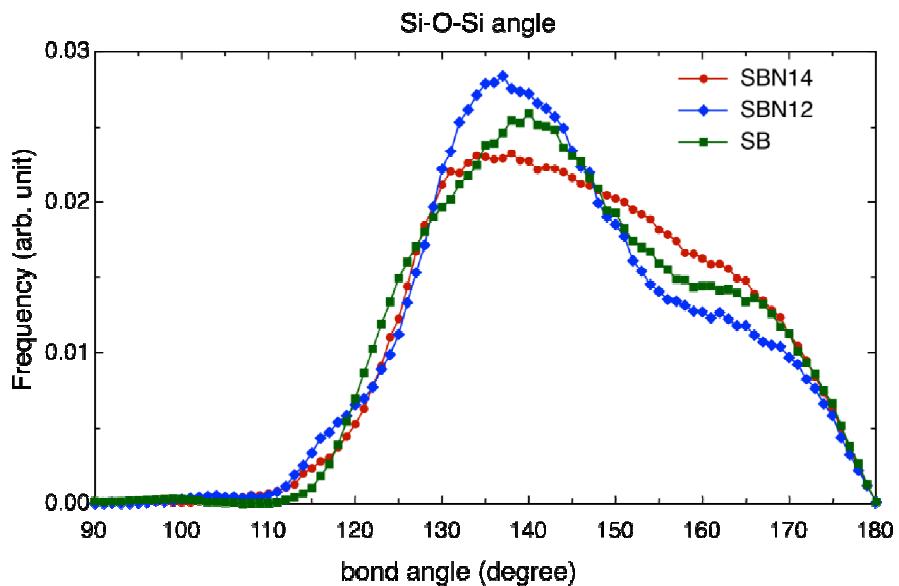
# Detailed structures



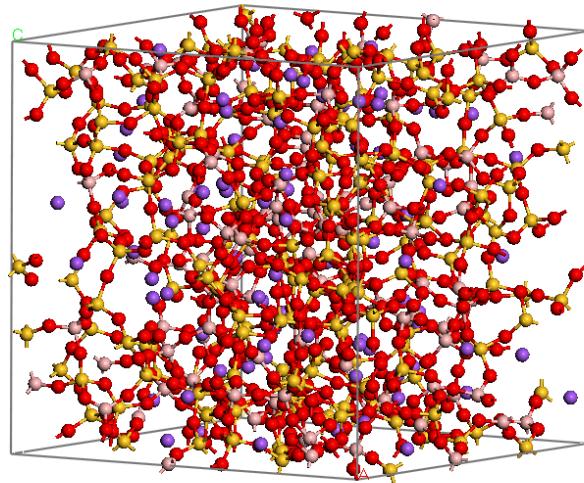
Glass	$d(\text{B-O})$		$CN_B$		$^{[4]}B$	NBO		$Q_3(\text{Si})$
	MD	EXP	MD	Theory	MD	MD	Theory	MD
SB	1.36	1.37	3.0	3.0	0 %			
SBN12	1.42	1.41	3.42	3.43	42 %	3 %	0 %	2 %
SBN14 ("T glass")	1.45	1.44	3.62	3.73	62 %	5 %	1 %	3 %

# Detailed structures

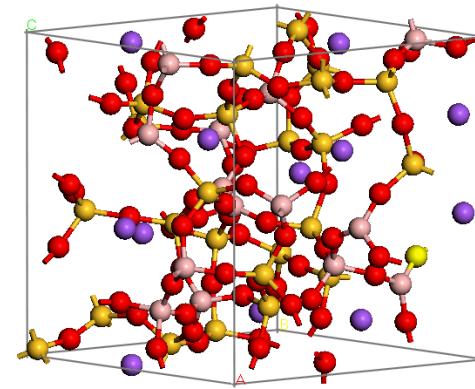
Glass	$d(\text{Si-O})$	$d(\text{Na-O})$	$d(\text{Si-Si})$	$d(\text{B-B})$	$d(\text{Si-B})$	$\angle(\text{O-Si-O})$	$\angle(\text{Si-O-Si})$	$\angle(\text{O-B-O})$	$\angle(\text{B-O-B})$
SB						109.3	145.3	119.2	143.9
SBN12	1.61	2.54	3.03	2.66	2.86	109.5	144.0	113.5	142.6
SBN14	1.61	2.51	3.05	2.71	2.88	109.4	145.8	111.7	142.8
EXP	1.61	2.29-2.62	3.08			109.5-109.7	144-147	111.6-118.6	129.4-143



# Glass structure for Argonne's DFT work



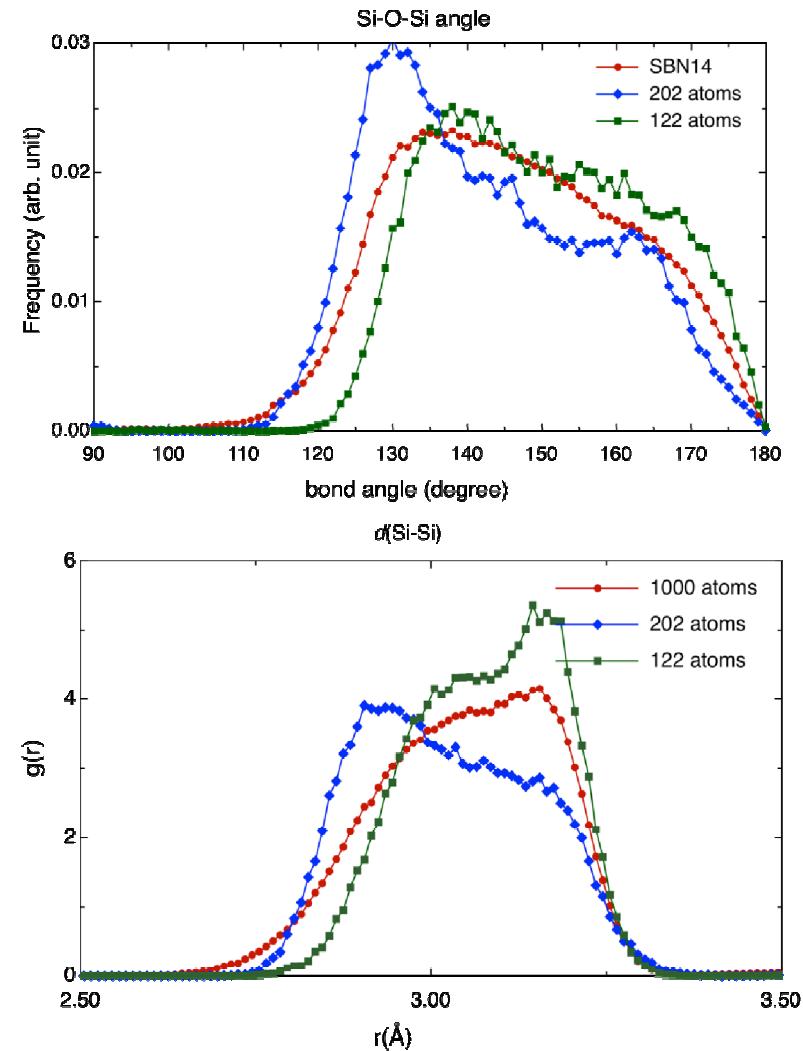
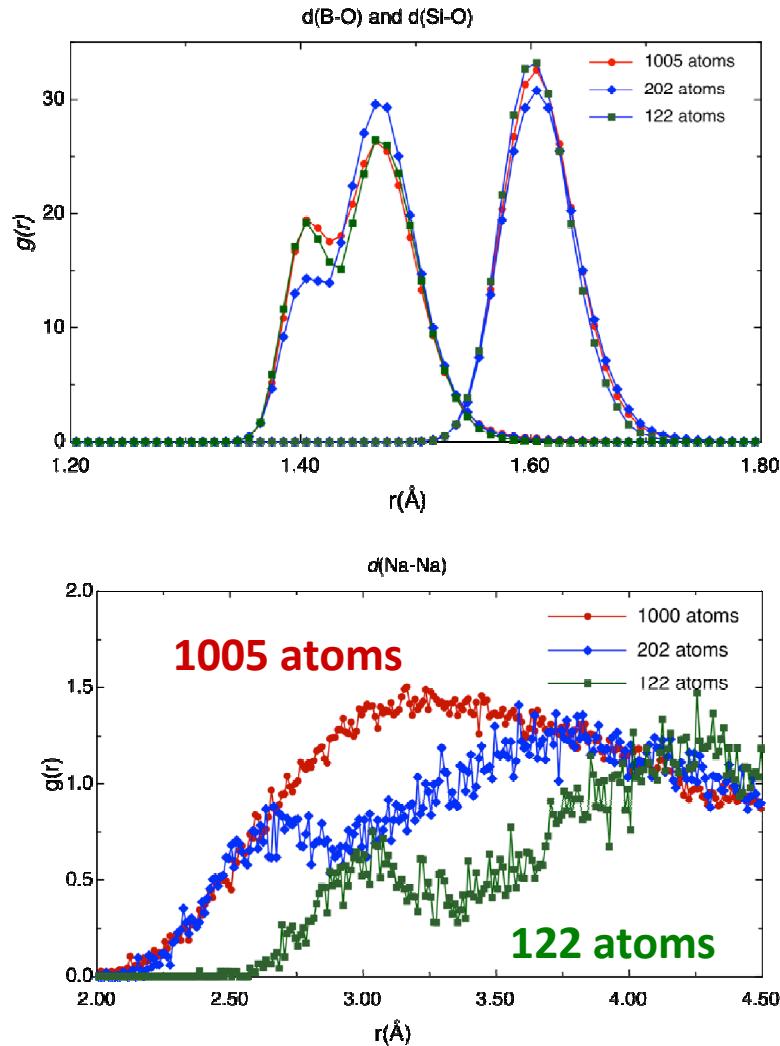
$T_{\text{glass}}$   
(1005 atoms)



$T_{\text{glass}}_{122}$   
(122 atoms)

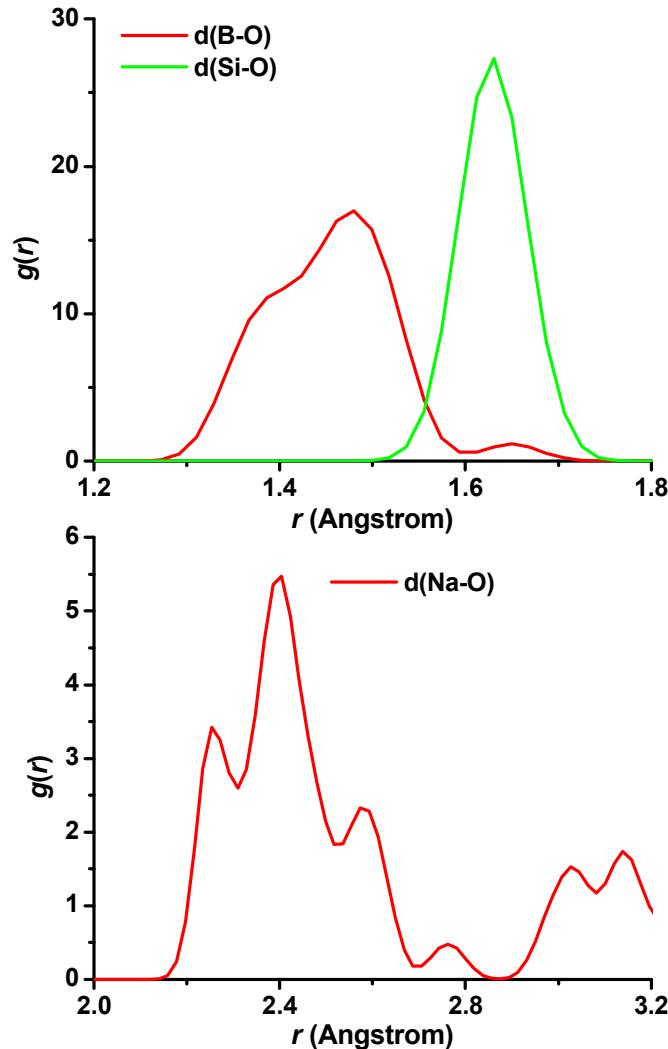
Glass	density		$d(\text{B-O})$		$CN_{\text{B}}$		$^{[4]}\text{B}$	NBO	
	MD	Exp	MD	Exp	MD	Theory	MD	MD	Theory
$T_{\text{glass}}$	2.44	2.45	1.45	1.44	3.62	3.73	62 %	4.7 %	1 %
$T_{\text{glass}}_{122}$	2.46	2.45	1.45	1.44	3.64	3.73	64 %	1.4 %	1 %

# SBN14 glass 1005 atoms vs. 122 atoms

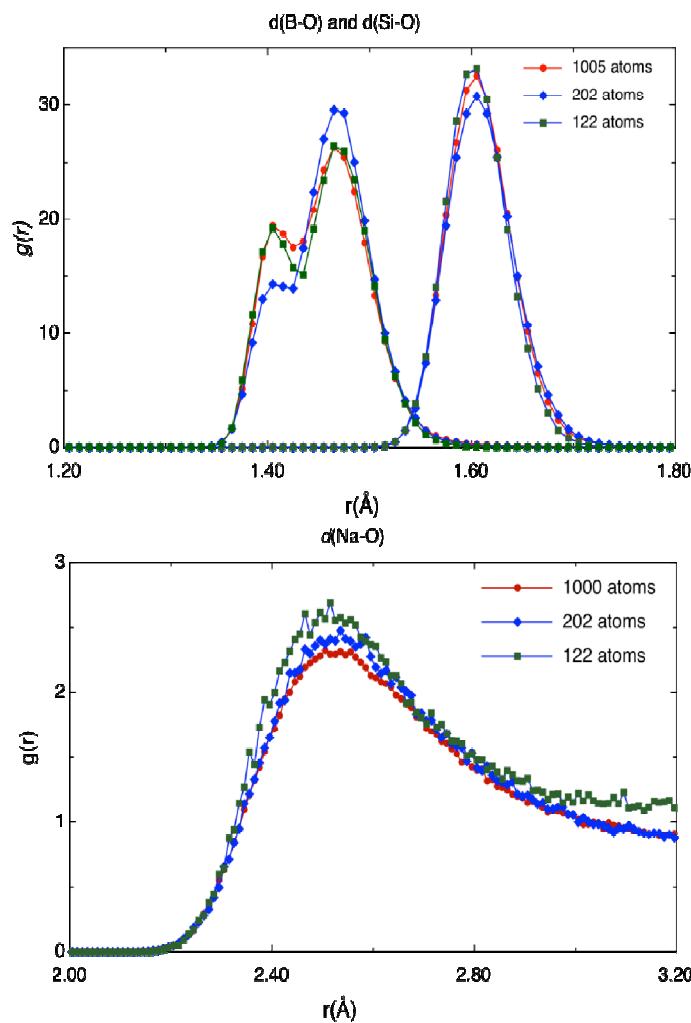


# DFT geometry optimization of T-glass (ANL)

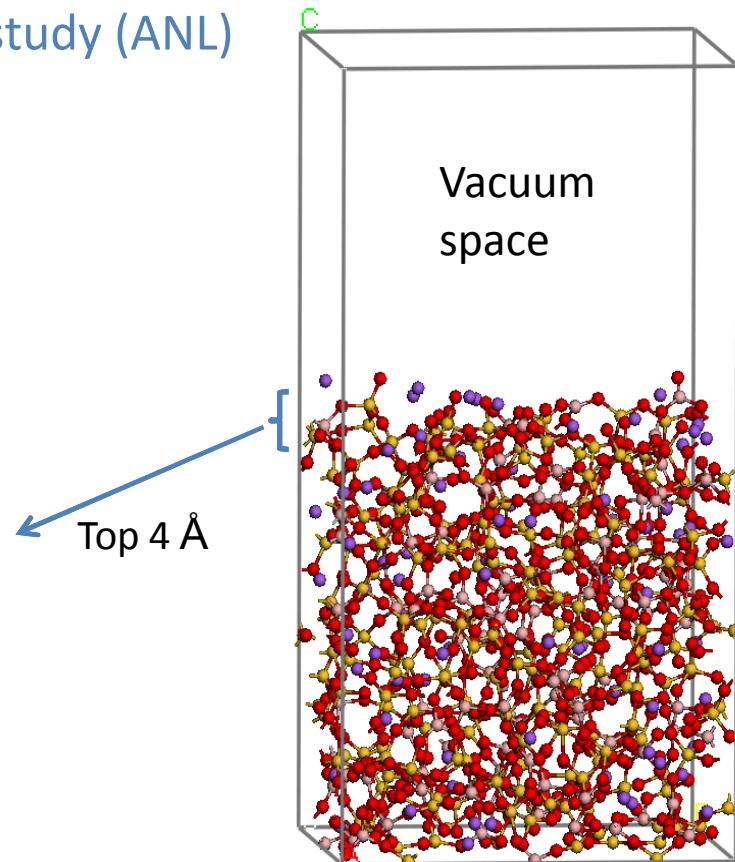
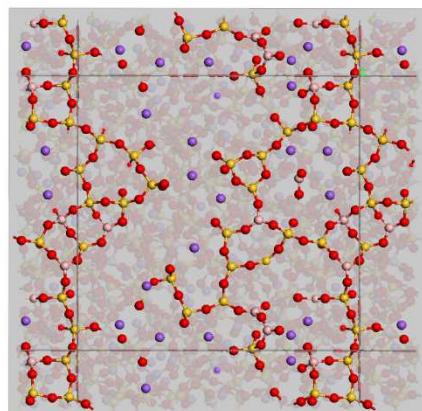
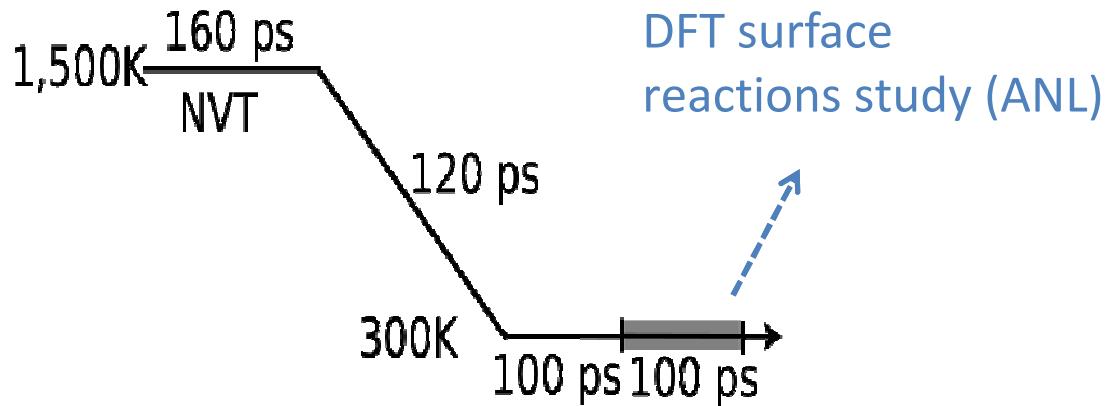
DFT (Argonne) of 1 configuration



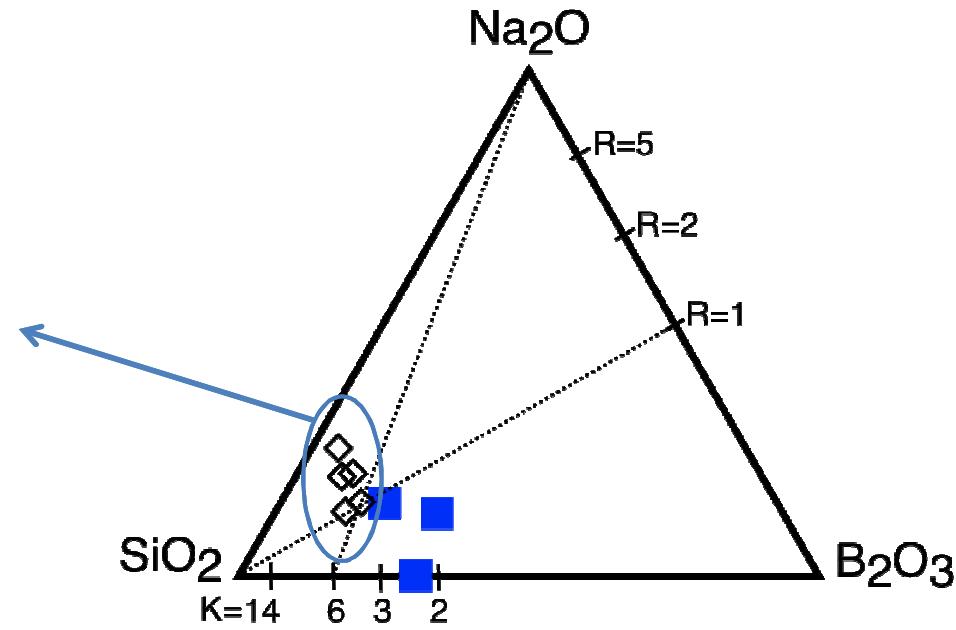
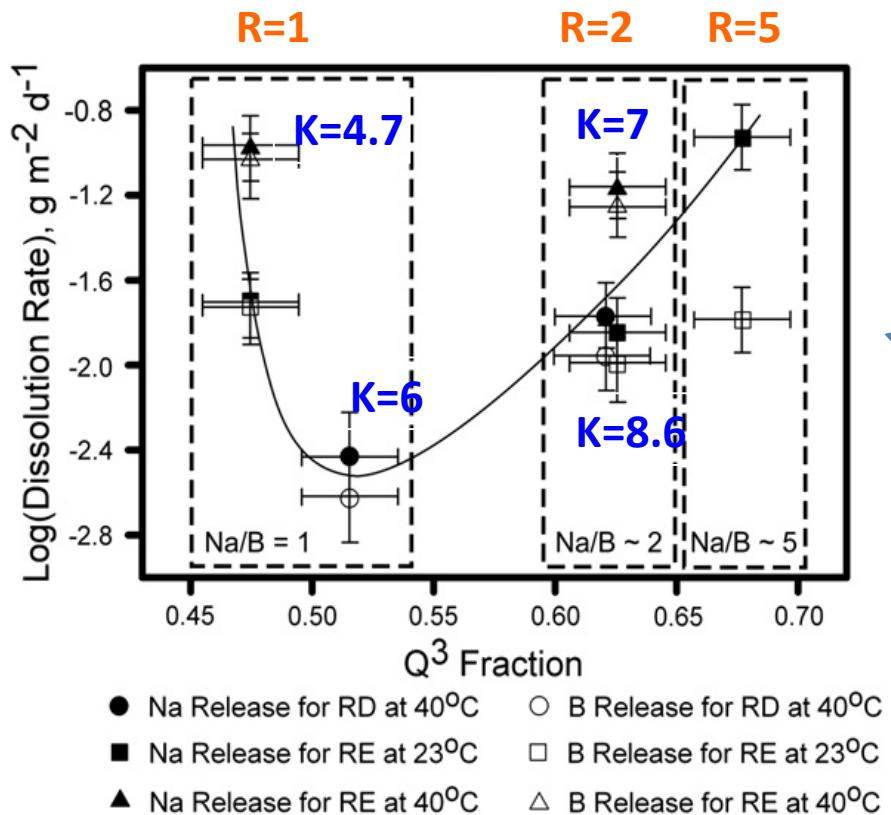
MD simulations (Sandia) of 1600 conf.



# MD simulations of glass surfaces

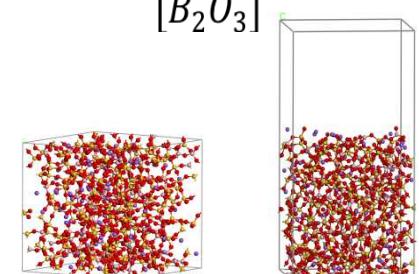


# Composition, dissolution, and structures



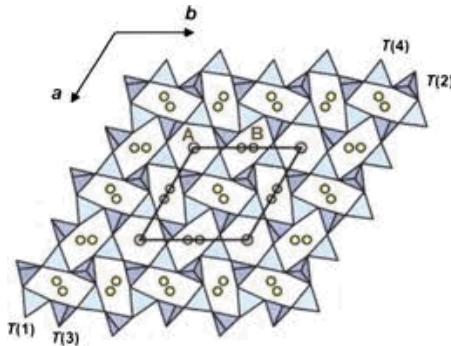
$$R = \frac{[\text{Na}_2\text{O}]}{[\text{B}_2\text{O}_3]}, \quad K = \frac{[\text{SiO}_2]}{[\text{B}_2\text{O}_3]}$$

Windisch Jr. et al. (2011) *J. Non-Cryst. Solids*, 2170.



# Future Work: Add $\text{Al}_2\text{O}_3$ to Force Field

- Add additional chemical components to Kieu et al. force field.
  - $\text{NaO-B}_2\text{O}_3\text{-SiO}_2$  force field was derived from a multi-component potential for silicate melts (*Guillot and Sator, 2007*) that includes:
    - $\text{Al}_2\text{O}_3$
    - $\text{CaO}$ ,  $\text{K}_2\text{O}$ ,  $\text{MgO}$ ,  $\text{FeO}$ ,  $\text{Fe}_2\text{O}_3$ ,  $\text{TiO}_2$
  - Provides starting point for adding these elements to the  $\text{NaO-B}_2\text{O}_3\text{-SiO}_2$  force field.
- Adjust force field parameters to fit:
  - Crystal structure data for nepheline ( $\text{NaAlSiO}_4$ ) and malinkoite ( $\text{NaBSiO}_4$ ) to prepare to study Pierce et al.  $\text{NaO-Al}_2\text{O}_3\text{-B}_2\text{O}_3\text{-SiO}_2$  glasses.
  - Experimental structure factors of aluminosilicate glass compositions obtained by different spectroscopic techniques (e.g., WAXS, Delaye et al., 2001).

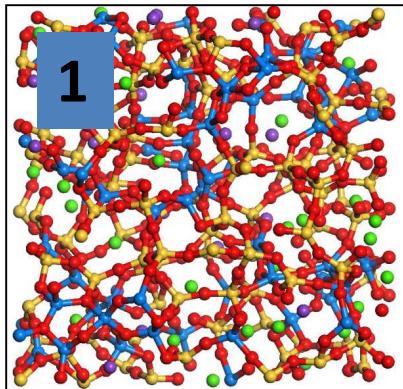


## Future Work: Add $\text{Al}_2\text{O}_3$ to Force Field

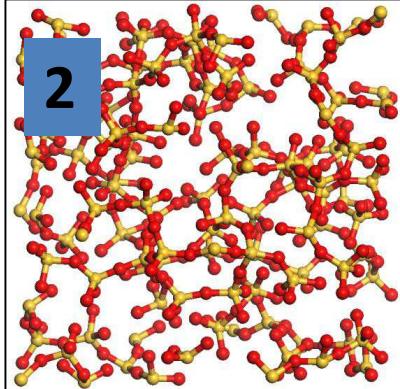
- Simulate  $\text{NaO-Al}_2\text{O}_3-\text{B}_2\text{O}_3-\text{SiO}_2$  glasses along nepheline-malinkoite join studied by *Pierce et al.*
- Examine simulated bulk and surface structures, and determine relationships between structure, composition, and dissolution.
- Evaluate hypothesis based on NMR TFS probe determinations (*Washton*) that the non-H-bonded  $\text{Q}^3$  groups are the most reactive by examining accessibility and calculated activation energy barriers.
- Provide surface structures to *Zapol* for DFT calculations of glass dissolution (i.e., bond-breaking at glass surfaces).
- DFT results will inform Kinetic Monte Carlo models (*Kerisit*) to calculate overall rate without prior assumption of the dissolution reaction mechanisms.

# Model (?) Gel Structure from MD Simulations

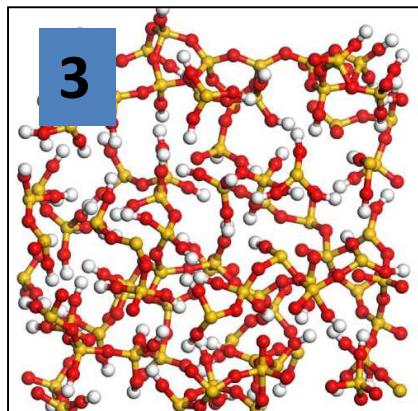
Mellott and Pantano, 2003



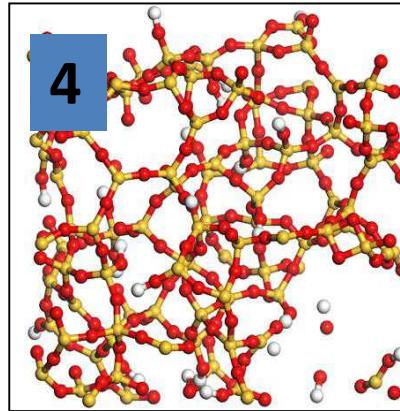
Fully polymerized  
glass structure



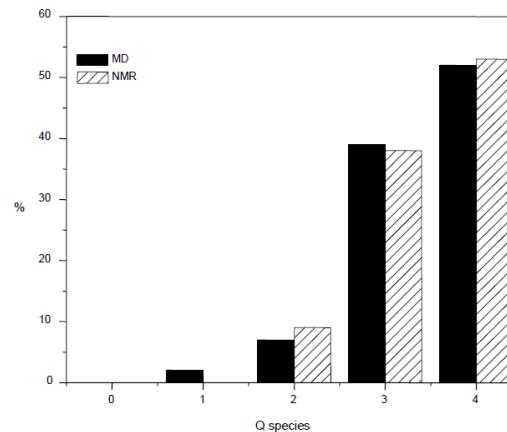
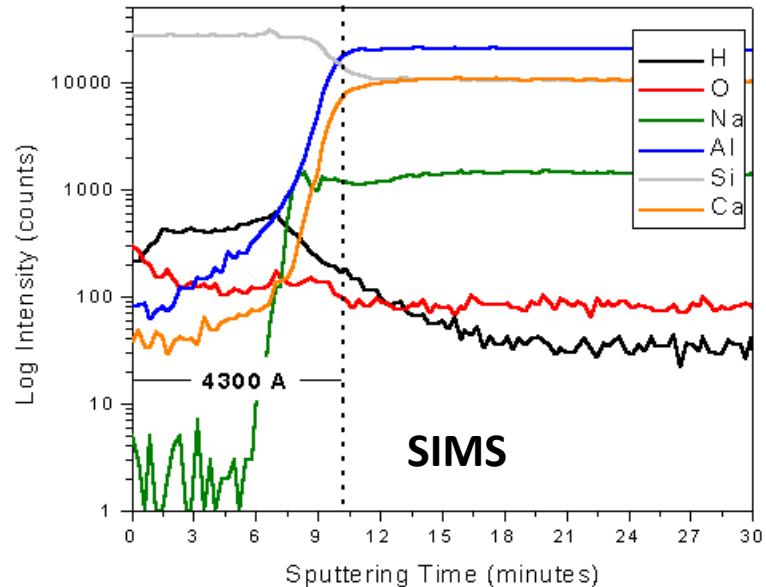
Remove Al, Na, Ca



Hydroxylate NBOs



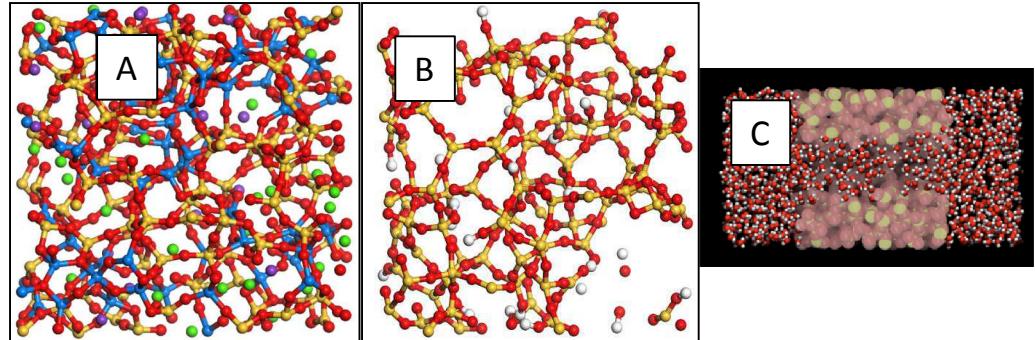
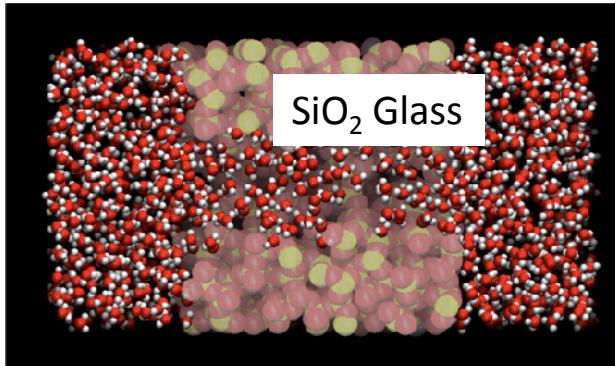
Condensation to form  
 $\text{SiO}_2$  gel layer



Figures fr. Mellott, 2003

# Gel Porosity and Diffusion

Bourg and Steefel, in press



Gel Characterization by *Cailleteau et al.* (2011)

- Composition determined by ToF-SIMS: Sharp compositional boundary analogous to that of Mellott
- Morphology determined by SAXS.
- Pore diameters from 10 Å– 45 Å in diameter
- Pore sizes increase over time

Diffusion Rates of  $\text{H}_2\text{O}$  through pores (B & S):  $0.28 - 1.54 \times 10^{-9} \text{ m}^2 \text{ s}^{-1}$

Diffusion is not rate-limiting in pores  $> 1 \text{ nm}$  diameter

Distance from glass → bulk solution  
Time zero → aged gel

How should I characterize simulated gel structures (B) to compare with experimental analyses?

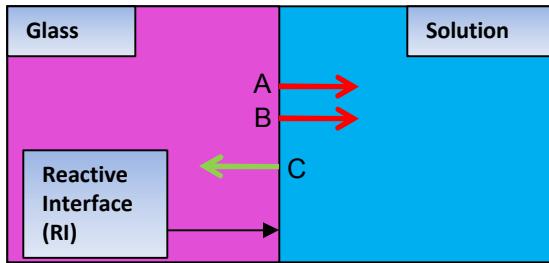
Pores  $< 1 \text{ nm}$  diameter

Fractal dimension

Pore Volume

Skeletal Structure: (polymerization?)

# Incorporating Molecular-Level Information into K<sub>μ</sub>C for Na<sub>2</sub>O-B<sub>2</sub>O<sub>3</sub>-SiO<sub>2</sub> Glass Dissolution



Assume same general form of constitutive expression for glass

## Questions:

1. Can we incorporate this expression into K<sub>μ</sub>C?
2. With continued dissolution do we multiply this expression by (1-Q/K) to account for SiO<sub>2</sub> saturation?
3. This is a surface-site specific model for dissolution. Analogous to MUSIC surface complexation model?

## Constitutive Expression from 1<sup>st</sup> Principles For Orthoclase (Zapol, ANL)

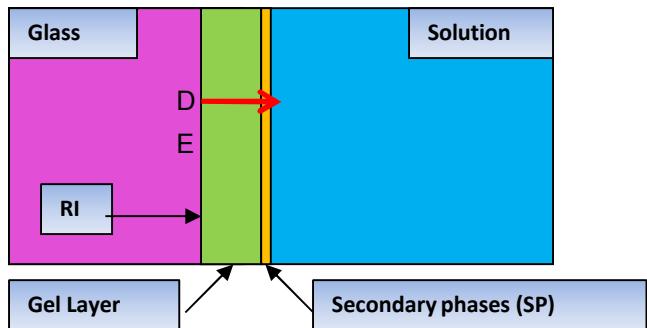
$$k = \sum_{\text{structure}(Q)} \sum_{\text{protonation}(i)} n_Q f_{Qi} k_{Qi}$$
$$f_{Qi} = f(Q, i, \text{aq}, \text{pH})$$

$$k_{Qi} = A_{Qi} e^{-E_a(Q,i)/RT}$$

$i = \text{SiOH}_2^+, \text{SiOH}, \text{SiO}^-$

- Overall rate  $k$  is a linear combination of rates for all exposed sites
- Fractional distribution of three sites  $f_{Qi}$  varies with pH
- Each different type of sites ( $Q, i$ ) has different rate constant  $k_{Qi}$

# Gel Layer Formation and Evolution



- Dissolution (i.e., bond-breaking) at glass surface
- Diffusion of glass components through gel layer (limited?)

- Pore size and connectivity for diffusion of chemical components into glass surface and out into bulk solution
- $\text{H}^+$  migration in gel – through large pores or silica matrix?  $\text{Na}^+$  can also “hop” in silica glass.
- Activation energy barriers within gel structure for dissolution (bond-breaking) and recondensation (bond-making)?
- Dual porosity transport model?

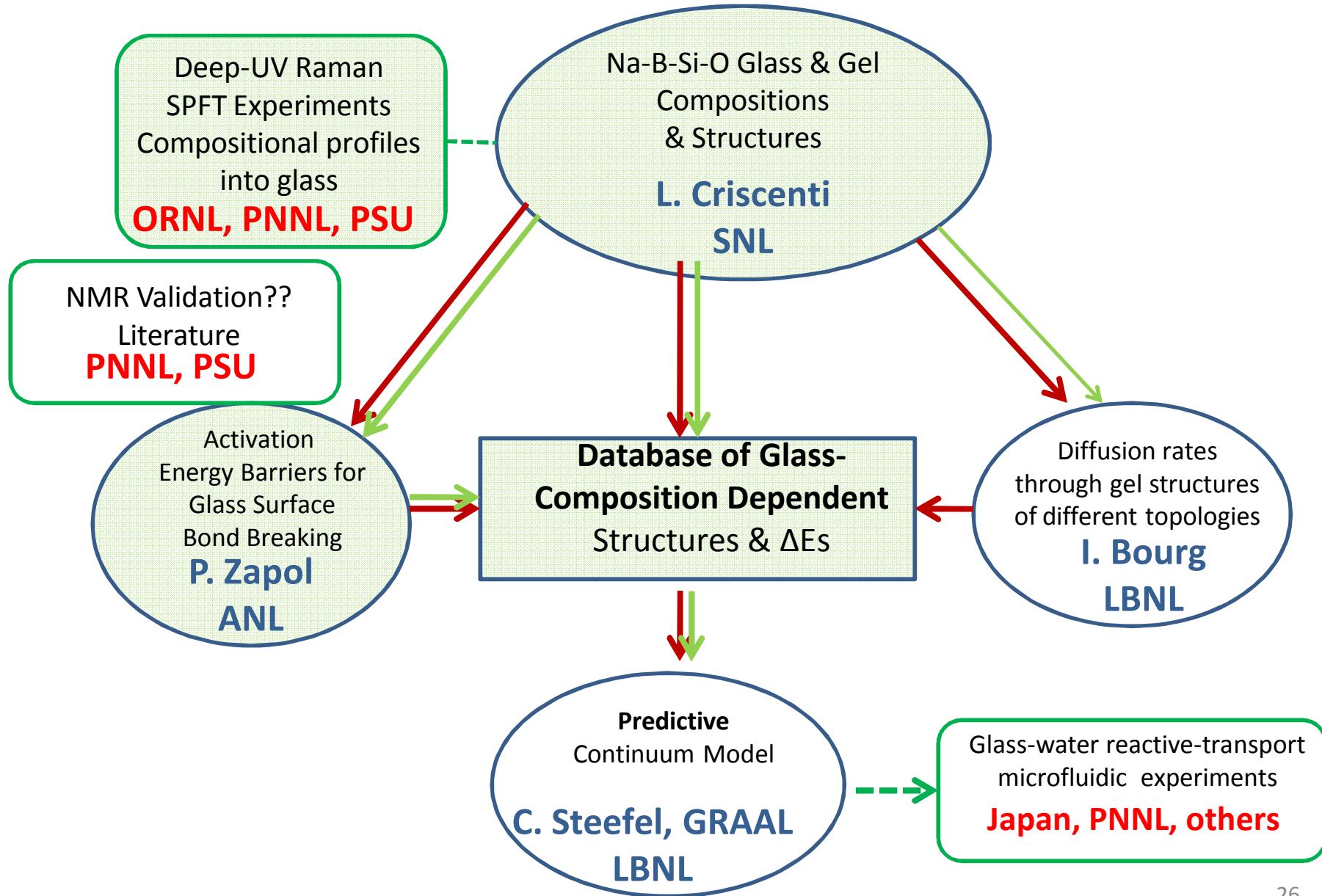
$\text{H}_4\text{SiO}_4 \text{ (aq)}$ ?

Recondensation of gel layer

# Incorporate Molecular-Level Information into K $\mu$ C for NaO-B<sub>2</sub>O<sub>3</sub>-SiO<sub>2</sub> Glass Dissolution

- Complete the “upscaling” between from first principles to micro-continuum K $\mu$ C calculations for a simple 3-component glass.
  - Provide constitutive equations for activation energy barriers as a function of surface composition
  - Provide diffusion rates for solutes through gel structures with < 1 nm pores
- Use K $\mu$ C in both standard and new predictive approaches to model experimental data from ORNL (Pierce) for 3-component glasses (and Gin’s T-glass)
  - SPFT leachates for NaO-B<sub>2</sub>O<sub>3</sub>-SiO<sub>2</sub> glass dissolution
  - Leached glass surface to depth profiles
  - UV-Raman Spectroscopy on glass structure
- Determine strengths and weaknesses of 1<sup>st</sup> principles informed model.
- Reiterate.

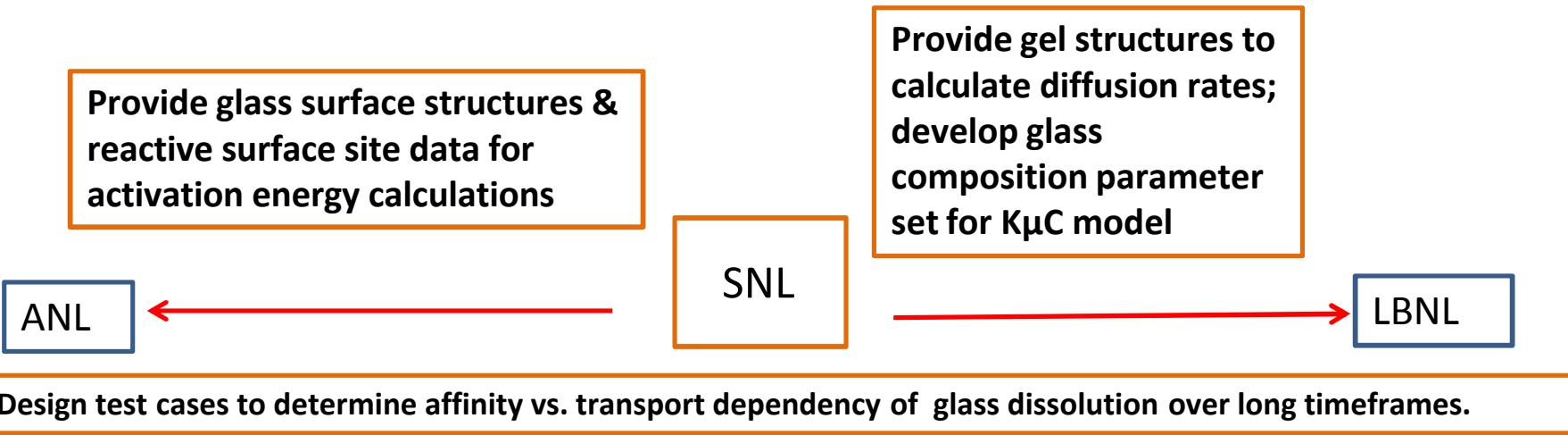
# Integrated Modeling Plan



# EXTRAS

# Next Steps Continued

- Examine how Kieu et al. (2011) reproduces surface structure of Na-borosilicate glasses
  - Changes in bond length and bond angle distribution
  - Changes in Si and B coordination
  - Depth of variation due to surface
- Validate Na-borosilicate glass bulk and surface structures with neutron diffraction and NMR data where possible.
- Create representative 100-atom glass surface structures to hand off to Peter Zapol (ANL) for activation energy barrier calculations.

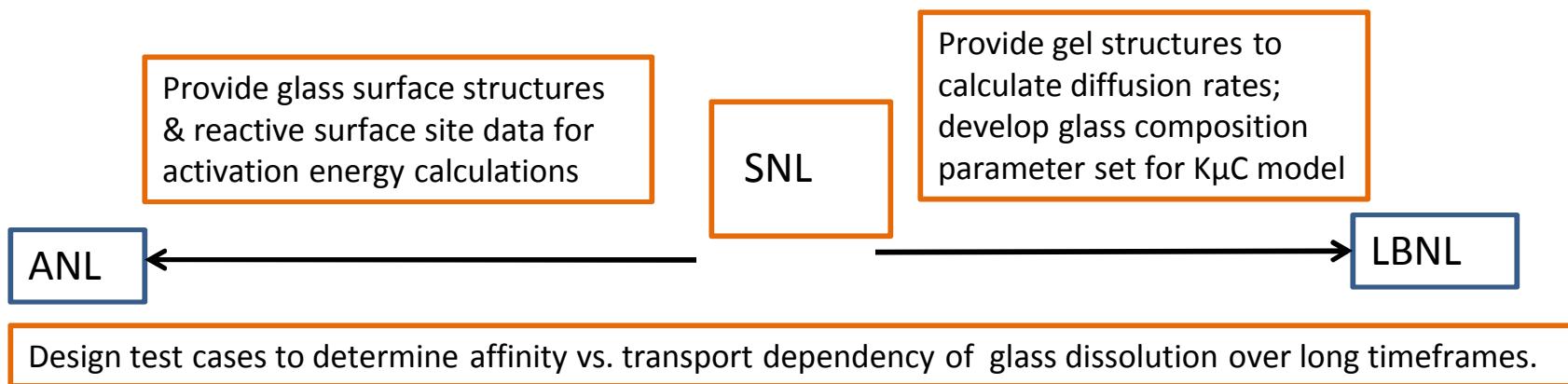


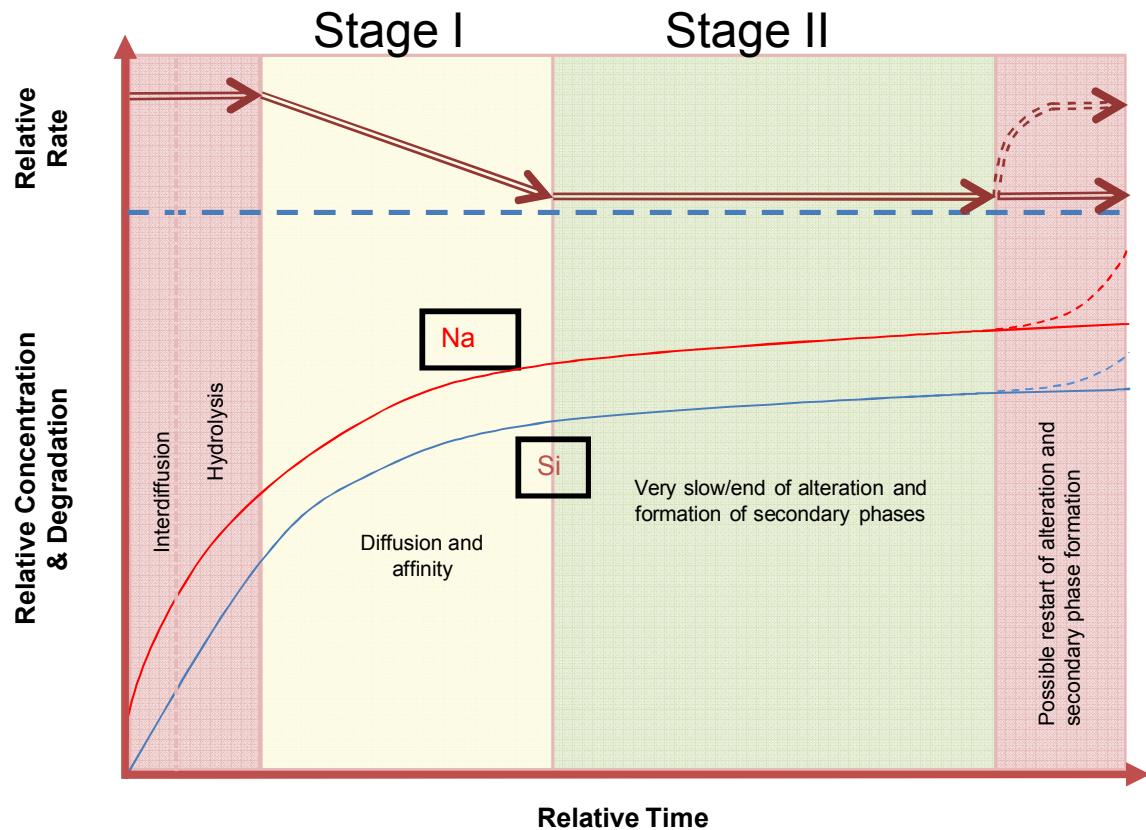
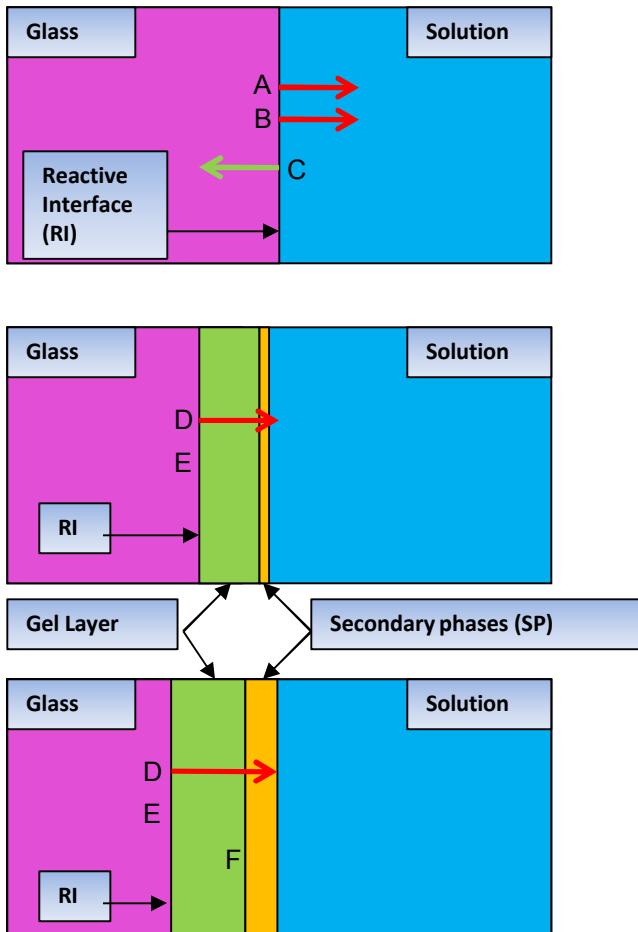
# Incorporating Molecular-Level Information into K $\mu$ C for Na<sub>2</sub>O-B<sub>2</sub>O<sub>3</sub>-SiO<sub>2</sub> Glass Dissolution

- *Windisch et al. (2011)*: Na<sub>2</sub>O-xB<sub>2</sub>O<sub>3</sub>-(3-x)SiO<sub>2</sub> glass structure correlated to dissolution rates in SPFT testing.
- SPFT leach tests and ToF SIMMS data collection for Na<sub>2</sub>O-xB<sub>2</sub>O<sub>3</sub>-(3-x)SiO<sub>2</sub> currently underway at ORNL (*Pierce*).
- Can we reproduce ORNL data using constitutive expression for glass dissolution obtained by SNL/ANL?
- By using K $\mu$ C using both empirical and predictive approaches, we will be able to assess and improve upscaling approach before moving on to more complex glasses.
- This would be the first time an atomistic-based constitutive equation for dissolution kinetics would be incorporated into a reactive-transport model.

# Interactions

- ❖ *Collaboration* with K. Murphy, C. Pantano, and K. Mueller to coordinate the simulation of glass structures that are simultaneously being investigated by NMR spectroscopy.
- ❖ *Handoff* to P. Zapol of glass surface structures for 1<sup>st</sup> principles calculations of activation energy barriers.
- ❖ *Collaboration* with J. Icenhower to investigate the importance of ion exchange on the long-term dissolution rate of glasses.
- ❖ *Handoff* to I. Bourg of gel surface structures for diffusion studies.
- ❖ *Collaboration* with C. Steefel to create a database of kinetic reaction and transport parameters that can be used with K $\mu$ C. Further collaboration to design test cases for K $\mu$ C that can be validated against experiment.





A – rate of alkali surface release

B – rate of Si, Al surface detachment (tends to be limiting)

C – rate of  $H_2O$  diffusion into glass

D – rate of silica diffusion through the gel ( $\pm$  SP) layer (at least partially limiting). Other constituents mass transport rates may also be important.

E – the aqueous composition at the RI is not the same as the Solution (aqueous silica concentration especially) and the surface area of the RI may be reduced by glass-gel contact area.

F – the gel and secondary phase layers may be acting as a mantle, in part isolating the fresh glass from the Solution.

- Fraction of dissolved boron, which is the yardstick of the degree of glass corrosion, decreased as the Ca content increased in the glass.

**Long-term Behavior Science: The cornerstone approach for reliably assessing the long-term performance of nuclear waste (Gin et al., 2011)** A large multi-scale approach is required and involves a mechanistic understanding of the key phenomena controlling the source term (i.e. the flux of radionuclides released from the waste as a function of time), as well as parametric studies, integrated and in situ tests. As a result, it is eventually possible to develop an operational model based on clever simplifications of a very complex reality, ensuring that predictions will always remain conservative despite conceptual and numerical uncertainties. Finally, predictive models must be validated based on the study of natural or archaeological analogues.