

The whole-block approach to measuring hydrogen diffusivity in nominally anhydrous minerals

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Abstract

A method is developed for determining the diffusivity of infrared-active species by transmission Fourier transform infrared spectroscopy (FTIR) in samples prepared as rectangular prisms *without* cutting the sample. The primary application of this “whole-block” or “3D-WB” method is in measuring the diffusion of hydrogen (colloquially referred to as “water”) in nominally anhydrous minerals, but the approach is applicable to any IR-active species. The whole-block method requires developing a 3-dimensional model that includes the integration of the beam signal through the sample, from rim to core to opposite rim. The analysis is carried out using both forward and tomographic inverse modeling techniques. Measurements collected from central slices cut from the whole block are simpler to interpret than whole-block measurements, but slicing requires destructive sample analysis. Because the whole-block method is nondestructive, this approach allows a time-series of diffusion experiments on the same sample.

The potential pitfalls of evaluating whole-block measurements without correcting for path integration effects are explored using simulations. The simulations demonstrate that diffusivities determined from whole-block measurements without considering path-averaging may be up to half an order of magnitude too fast. The largest errors are in fast and/or short directions, in which

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25 the diffusion profiles are best developed. A key characteristic of whole-block measurements is
26 that the central values in whole-block traverses always change before the concentration of the
27 IR-active species changes in the block's center because of signal integration that includes
28 concentrations in the sample rims. The resulting plateau in the measurements is difficult to fit
29 correctly without considering path integration effects, ideally by using 3D whole-block models.
30 However, for early stages of diffusion with less than 50% progress, diffusivities can be
31 accurately determined within 0.5 log units using a 1D approximation and the whole-block central
32 plateau values because diffusivities are more dependent on profile shape than absolute
33 concentrations.

34 To test the whole-block method, a dehydration experiment was performed on an oriented
35 piece of diopside from the Kunlun Mts with minimal zoning, cracks, or inclusions. The
36 experiment was performed in a gas mixing furnace for 3 days at a temperature of 1000 °C and
37 oxygen fugacity of $10^{-11.1}$ bar (QFM). First, whole-block analysis was performed by taking FTIR
38 traverses in three orthogonal directions. Then, a slice was cut from the center of the sample, and
39 hydrogen profiles were measured by FTIR and secondary ion mass spectrometry (SIMS). The
40 results of FTIR and SIMS measurements on the slice are in good agreement both with each other
41 and with diffusion profiles calculated based on the results of forward and inverse models of the
42 whole-block FTIR measurements. Finally, the new method is applied to previous whole-block
43 measurements of hydrogen diffusion in San Carlos olivine using both the forward and inverse
44 approaches.

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Keywords

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Transmission FTIR, diffusion, hydrogen, water in nominally anhydrous minerals, diopside,

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olivine

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Introduction

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Diffusion is an important transport mechanism and has numerous applications to the

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geological sciences (e.g., Watson and Baxter 2007). The mechanisms and rates of diffusion for

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many chemical species in geologically relevant systems are known (e.g., Zhang and Cherniak

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2010), but diffusivities, also called diffusion coefficients, are still poorly constrained for many

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systems. Determination of the relatively slow diffusivities in crystalline materials has been

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tackled with a variety of techniques (Ryerson 1987), some of which are still emerging (Watson

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and Dohmen 2010). All of these methods require both an experimental approach that results in

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diffusion and a way to measure the extent of that diffusion.

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Experimental methods may be divided into two broad categories: time-series or mass-loss

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experiments, which measure changes in concentration with time at a particular point, usually the

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center, of the sample, and diffusion profile experiments, which measure changes in concentration

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with distance at a given time. Sample geometry is a key factor in all diffusion experiments and is

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usually chosen in a way that simplifies the final diffusion analysis from Fick's 2nd Law, $\frac{\partial v}{\partial t} =$

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$D \frac{\partial^2 v}{\partial x^2}$, where D is the diffusivity, v is the concentration, t is the time, and x is the distance.

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Solutions are available for many geometries and boundary conditions in Carslaw and Jaeger

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1959 and Crank 1975.

65 Experiments that are designed to measure diffusion profiles have the significant advantage
66 that they can be directly compared to profiles calculated from the solution to Fick's 2nd Law, and
67 transport that is unrelated to diffusion is readily apparent (Ryerson 1987). Time series
68 experiments are, however, still useful to obtain an initial approximation for diffusivity and as a
69 complimentary technique for confirming interpretations based on diffusion profile measurements
70 (e.g., Woods et al. 2000). Other approaches useful for validation of diffusivity estimates include
71 cross-checking with multiple experimental methods, varying the sample geometry, and zero-time
72 experiments, in which the sample is removed from the experiment immediately after run
73 conditions are achieved (Hofmann 1980). Although diffusion profiles are preferable to time-
74 series measurements, profiles are more difficult to obtain because they traditionally require that
75 the sample be cut through the middle after the experiment, which requires careful manipulation
76 to avoid sample fracture or loss and also prevents the sample from being re-used in future
77 diffusion experiments, e.g., to generate time-series measurements or perform a reversal.

78 Hydrogen (H⁺; colloquially often referred to as "water") diffusivities in nominally
79 anhydrous minerals (NAMs) are most often determined using measurements from transmission
80 Fourier-transform infrared spectroscopy (FTIR) because this method is non-destructive,
81 commonly available, inexpensive, more sensitive to hydrogen (> ppm level) than most other
82 techniques, and can provide some structure and speciation information (see review by Rossman
83 2006). Reflectance FTIR may also be used to measure hydrogen and carbon speciation and
84 concentration in glasses, but the detection limits for this method are generally too high (~1000
85 ppm wt. H₂O) for measuring hydrogen or carbon in nominally anhydrous minerals (Grzechnik et
86 al. 1996; Hervig et al. 2003; King and Larsen 2013; Lowenstern and Pitcher 2013). Experimental
87 methods and measurement techniques for hydrogen diffusion experiments in NAMs are

88 reviewed by Ingrin and Blanchard 2006 and Farver 2010. In the standard approach, a slice is cut
89 from the center of a sample prior to diffusion profile measurements to minimize the effect of
90 profile overlap (also called signal accumulation or integration) along a transmission path (Movie
91 1). By cutting, profile overlap effects are eliminated if the hydrogen has not diffused into the
92 center of the slice and is of constant concentration in the thickness direction of the slice. Non-
93 constant slice profiles can still be evaluated, but non-path-integrated three dimensional models
94 must be used instead of relatively simple one dimensional models; see description below.

95 This paper describes how to acquire and interpret diffusion profile measurements for species
96 measured by transmission FTIR without cutting the sample after a diffusion experiment (Figure
97 1). This approach, which we refer to as the whole-block method, takes advantage of the fact that
98 the measured transmission FTIR concentration, normalized to thickness, represents a path-
99 averaged value in the sample. We focus on hydrogen in nominally anhydrous minerals cut in the
100 form of a rectangular parallelepiped, but other systems measured using a transmitted signal
101 moderated by absorption such as carbon by FTIR or a heavy element by X-rays, and other
102 sample geometries could also be evaluated using this approach. We explore this method using
103 simulated data, test it experimentally by dehydrating diopside, and then apply it to published
104 measurements on the hydration of olivine.

105 **The whole-block approach**

106 Signal accumulation in FTIR measurements provides an opportunity to obtain diffusion
107 profiles without cutting the sample, which typically is a rectangular parallelepiped oriented with
108 respect to crystallographic or morphologic axes of the material. These profiles (Figure 1) will be
109 called “whole-block” here to distinguish them from “slice” profiles measured on the central slice

110 of a sample cut after the diffusion anneal. In this section we describe the advantages of using
111 whole-block measurements as well as the details of the method. The whole-block method
112 involves (1) measurements of hydrogen concentration profiles through the entire, uncut sample
113 block, (2) forward modeling to obtain diffusivity estimates, and (3) inversion of the whole-block
114 measurements to provide an independent determination of the internal concentration structure of
115 the sample.

116 ***Advantages***

117 Whole-block measurements are useful for the following reasons: (1) whole-block
118 measurements can be obtained nondestructively and thus permit multiple experiments using the
119 same sample, e.g., a time series of progressively longer experiments or reversal of diffusion flux
120 by change of external reservoir concentration to promote diffusion into or out of a specimen. (2)
121 The cut slice may be too thin to provide a high quality FTIR signal for samples with low, ppm-
122 level hydrogen concentrations. (3) The diffusivities in three orthogonal directions can be
123 estimated from the same sample. Measurements for the third direction otherwise requires either a
124 second experiment or a slice taken from one of the butts left from the original slice. (4) In
125 dehydration experiments, the initial profile can be measured directly. (In hydration experiments
126 on pre-annealed samples, the initial concentration is 0 everywhere.)

127 The whole-block method also allows some independence from the following problems that
128 can sometimes result in high error in measurements taken near the edge of sample slices. The rim
129 hydrogen concentration may be very low after a dehydration experiment, which causes high
130 errors in FTIR calibrations. Rims are also less likely to be perfectly flat than the center of a
131 polished block because of chipping during cutting and/or polishing and/or quenching. If the

132 sample block shape is wedged and does not have near-perfect 90° angles, then the FTIR beam
133 does not sample the full thickness at some rims, resulting in artificially low values. Further, in
134 some thicker samples, scattering of the infrared beam near edges and cracks may result in
135 artificially low values.

136 Whole-block measurements taken in the center of the sample always show an apparent
137 decrease for dehydration experiments or increase for hydration in hydrogen content in the center
138 of a whole-block profile (see discussion and illustrations below) and therefore provide diffusion
139 information in the ray path direction without the problems associated with rim measurements.
140 While whole-block measurements also include rim effects, the central values, which include
141 integrated values from the rims, provide additional measurements to improve the overall
142 interpretation. Because the whole-block method is non-destructive, the sample can always be
143 sliced to confirm the final interpretation and resolve any ambiguity.

144 ***Whole-block profile measurements***

145 Whole-block concentration measurements are constructed by normalizing the peak area of
146 the FTIR absorbance spectra of interest obtained after the experiment to the maximum hydrogen
147 concentration, which is the initial concentration for diffusion out of the sample or the solubility
148 for diffusion into the sample. The relevant OH vibrations occur between wavenumbers 3000 and
149 4000 cm^{-1} . The resulting area ratios can be used directly in subsequent forward and inverse
150 analyses, or these values can be scaled up to the maximum hydrogen concentration and
151 expressed in concentration units such as ppm wt. H_2O or $\text{H}/10^6 \text{Si}$.

152 Diffusion studies have somewhat more leeway in how FTIR spectra are evaluated than
153 studies concerned with absolute hydrogen concentrations because the diffusivity estimate

154 depends on the overall shape, not the absolute values, of the concentration profiles. Thus,
155 approaches likely to introduce errors in absolute concentration values are often permissible as
156 long as they are applied uniformly to all spectra in the profile. For instance, linear baselines are
157 unusual in concentration studies but common in diffusion studies because they provide
158 consistency among measurements both along the profile and before and after the experiment
159 (e.g., Ingrin et al. 1995). The resulting concentration profiles may not be entirely accurate in their
160 absolute concentrations, but they still provide good estimates of the diffusivities, even after
161 diffusion has reached the center of the sample, in which case a 3-dimensional model such as one
162 described below is required.

163 Also unlike concentration studies, either polarized or unpolarized infrared radiation may be
164 used, even in highly birefringent or anisotropic samples. In the case of unpolarized radiation,
165 which can be slightly polarized due to prisms and mirrors along its optical path, the absorbance
166 spectrum changes with ray path R , which is also called the propagation direction or thickness
167 direction. Polarized radiation has the advantages of providing more meaningful additional
168 structural and concentration information and allowing consistent spectra between profiles if the
169 electric vector E is maintained in a constant direction, ideally the direction with the strongest
170 absorbance. However, unpolarized radiation typically has a higher signal-to-noise ratio and may
171 be necessary in samples with very low concentrations. Woods et al. (2000) measured hydrogen
172 diffusion profiles in diopside using both polarized and unpolarized radiation and obtained the
173 same diffusivity within error.

174 Forward modeling to obtain diffusivities requires whole-block profiles (Figure 1) measured
175 parallel to three orthogonal directions. The simplest geometry for obtaining these profiles is a
176 centered cross of measurements using one infrared ray path and a line of measurements through

177 the center of the sample using a second ray path. Profiles obtained using three separate ray paths,
178 additional off-center profiles, and/or full concentration maps are not necessary to produce
179 diffusivity estimates but would contribute to and support the final analysis, particularly for the
180 inverse approach described below.

181 ***Forward models***

182 **Whole-block profiles (3D-WB)**

183 Whole-block profiles can be calculated from a 3-dimensional (3D) array of hydrogen
184 concentrations determined from standard three-dimensional models (given below) by taking the
185 arithmetic mean of the concentrations in the direction parallel to the infrared beam ray path (R)
186 for each measurement. For example, for a whole-block point (x,y) measured with $R // [100]$, the
187 whole-block (3D-WB) concentration ν_{3D-WB} equals the average of all concentration points along
188 $[100]$ at (x,y). Unlike slice profiles, whole-block profiles can often be conveniently obtained with
189 R parallel to two different directions, e.g., a traverse $// [100]$ with $R // [010]$ and a traverse $//$
190 $[100]$ with $R // [001]$. Because they represent averages, these two profiles may differ
191 significantly, particularly when diffusion is strongly anisotropic.

192 This averaging must take place in the absorbance A domain, which is linearly related to the
193 concentration ν according to the Beer-Lambert law ($A = \nu \times$ molar absorption coefficient ε
194 \times thickness), rather than in the transmission domain. In FTIR, the transmission T is the variable
195 that is actually measured, and $T = I / I_0$ where I is the intensity of measured light and I_0 initial
196 intensity. The absorbance $A = -\log_{10} T$; although other log scales may also be used with an
197 appropriate conversion factor for ε when calculating ν from A . The absorbance is conventionally
198 reported by most commercial software for FTIR spectrometers. For each unit step through the

199 sample, the intensity is changed by a percentage of the initial value entering that unit, resulting in
200 an exponential change in the transmission.

201 The concentration of the diffusing species being measured, and hence the extent of that
202 exponential change, varies along each unit step-size in the integrated whole-block path after a
203 diffusion experiment. To understand the mathematics involved, consider the simplified example
204 of a zoned sample with three distinct parts, in which each part has a different concentration of the
205 IR-active species and hence different transmittances. The IR radiation first passes through Part 1
206 to produce transmittance $T_1=I_1/I_0$, then passes through Part 2 to produce transmittance $T_2=I_2/I_1$,
207 and finally through Part 3 to produce $T_3=I_3/I_2$. The generic sample in this example does not have
208 to be symmetric, although a symmetrical diffusion profile is assumed in the whole-block
209 diffusion forward models. The final transmittance, $T_{total} = T_1 \times T_2 \times T_3 = I_3/I_0$. (Following this
210 equation, whole-block path-integrated values may be calculated in the transmission domain by
211 using the product, not the average, of transmittance values along the ray path.) Because A is
212 related to T on a log scale ($A=-\log T$), the total absorbance A_{total} is determined from the
213 component A values as a sum rather than a product: $A_{total} = -\log(T_{total}) = -\log(T_1 \times T_2 \times T_3) =$
214 $-\log T_1 + -\log T_2 + -\log T_3 = A_1 + A_2 + A_3$. For a system with n concentration zones, $A_{total} =$
215 $\sum_{i=1}^n A_i$. Therefore, the integrated absorbance or concentration divided by the sample thickness
216 represents the arithmetic mean of absorbance or concentration values in the ray path of the
217 measurement, assuming that the grid size used to calculate the profile is evenly spaced and
218 sufficiently fine-scale that the concentration in each division is approximately uniform.

219 **3D non-path-integrated (3D-NPI) and 1D diffusion**

220 Production of the 3D-WB profiles described above assumes some knowledge of the internal
221 concentration structure of the sample, which can be determined using standard 3-dimensional
222 models for diffusion in a rectangular parallelepiped such as those found in Carslaw and Jaeger
223 (1959) Section 6.4, Crank (1975) Section 2.5.1, and Demouchy et al. (2006). These models are
224 based on either infinite sums of sines and cosines or error functions. In general, error functions
225 are preferred for the early stages of diffusion, and the infinite sum more accurately represents
226 diffusion at later stages (Crank 1975, Section 2.1). All models discussed here assume that the
227 diffusivity is independent of the concentration of the diffusing species. This assumption may not
228 hold for hydrogen diffusion in all NAMs (e.g., mantle-derived pyrope, Wang et al. 1996) and is
229 certainly incorrect for silicate glasses (e.g., Doremus 1975; Lanford et al. 1979; Zhang et al.
230 1991, Anovitz et al., 1999).

231 Examples for several boundary conditions are written out in Carslaw and Jaeger (1959)
232 section 6.4 for diffusion in an isotropic rectangular parallelepiped. Carslaw and Jaeger (1959)
233 further note that the same approach (i.e., the multiplication of known one-dimensional problems,
234 with a proof given in their Section 1.15) is also applicable to anisotropic media, given the critical
235 assumption that the main axes of diffusion are parallel to the coordinate planes of the
236 parallelepiped. For instance, for the region $L_a > x > 0$, $L_b > y > 0$, $L_c > z > 0$ with unit initial
237 concentration and zero surface concentration, the concentration at all positions in the crystal
238 $v(x,y,z)$ is as follows:

$$v(x, y, z) = \operatorname{erf}\left(\frac{x}{2\sqrt{D_a t}} + \operatorname{erf}\left(\frac{L_a - x}{2\sqrt{D_a t}}\right)\right) \operatorname{erf}\left(\frac{y}{2\sqrt{D_b t}} + \operatorname{erf}\left(\frac{L_b - y}{2\sqrt{D_b t}}\right)\right) \operatorname{erf}\left(\frac{z}{2\sqrt{D_c t}} + \operatorname{erf}\left(\frac{L_c - z}{2\sqrt{D_c t}}\right)\right) \quad \text{Eq. 1}$$

239 where t is time and D_a, D_b, D_c are the diffusivities parallel to each direction. Throughout this
240 manuscript, we refer to this model as three-dimensional non-path-integrated (3D-NPI).

241 For individual non-path-integrated concentration profiles (e.g., slice profile measurements
242 along a single dimension) in which the concentration of the diffusing species in the center of the
243 sample remains constant throughout the experiment, then the diffusivity can be determined more
244 simply using a 1-dimensional (1D) model, e.g.,

$$v(x)_{1D} = \left(\operatorname{erf}\left(\frac{x}{2\sqrt{D_a t}}\right) + \operatorname{erf}\left(\frac{L_a - x}{2\sqrt{D_a t}}\right) \right). \quad \text{Eq. 2}$$

245 ***Inverse models: Geochemical tomography***

246 Inversions similar to those used to evaluate geophysical tomography data constitute a
247 complementary analysis technique to the forward models described above. While inverse models
248 cannot by themselves provide diffusivity estimates, they produce independent estimates of the
249 internal concentration structure that provide a useful constraint and check on the concentration
250 profiles and true center concentration values predicted by forward models. Unlike forward
251 models, inverse models require no assumptions about diffusion or any concentration changes
252 over the course of the experiment. The data d constraining our model are the whole-block
253 estimates of path-averaged concentrations obtained as described above. As in all inverse
254 modeling, errors in the observations arising from the measurement technique and assumptions

255 will lead to some level of uncertainty in the final model. For example, in our case we assume that
256 our sample is a single crystal with uniform mineral composition to produce integrated
257 concentration estimates from FTIR.

258 A more complete treatment of discrete inverse theory and examples for geophysical data
259 may be found in Menke (1984). Here we include only a brief overview of the technique. The
260 sample, a rectangular prism, is subdivided into a series of regularly spaced miniature prisms or
261 blocks of equal volume, and each of these sub-blocks is assumed to have uniform concentration.
262 Each whole-block observation represents the average concentration through a subset of these
263 blocks. The problem is then formulated as the equation $G \cdot m = d^{obs}$, where d^{obs} is an $n \times 1$ matrix
264 containing n whole-block observations, m is a $k \times 1$ coefficient matrix containing k concentration
265 values (1 for each subdivision of the sample; these are the values we solve for), and G is an $n \times k$
266 sensitivity matrix that describes the contribution of each of the k sample subdivisions to the
267 whole-block measurements d^{obs} . The grid spacing, which determines k , is varied in each direction
268 with the goal of describing the measurements as simply as possible, i.e., with a smallest value of
269 k that adequately describes the measurements.

270 The internal concentration values are determined by minimizing a goodness of fit measure
271 related to the variance $\sigma^2 = \frac{\sum(d^{obs} - d^{pred})^2}{\overline{d^{obs}^2}} \left(\frac{1}{k}\right)$, where d^{pred} represents the whole-block values
272 predicted from the model results, k is number of degrees of freedom (i.e., the number of grid
273 cells used to model the measurements), and $\overline{d^{obs}}$ is the mean of the observed measurements,
274 which serves as a scaling factor to compare the goodness of fit between data sets. The error σ_i on
275 each concentration value i is evaluated as $\sigma_i = \overline{d^{obs}} \sqrt{C(i,i)}$, where C is the covariance matrix
276 $C = [G^T G] \cdot \sigma^2$, and G^T is the transpose of G . The formal error σ_i on each value is a minimum value

277 because it only includes the diagonal elements of the covariance matrix. To better assess the true
278 error, comparable whole-block data can be simulated and then inverted using the same approach
279 as that used to invert the measurements of interest.

280 Ideally, whole-block profiles would be obtained with at least three lines measured along
281 three different orthogonal planes of the sample block as well as additional, off-center profiles.
282 Newer FTIR spectrometers are equipped with digital stages and software that can rapidly
283 produce hydrogen measurement maps of entire surfaces with consequent improvement in the
284 accuracy of internal concentrations determined from the inversion. However, many laboratories
285 will be limited to the bare minimum required to apply the whole-block forward models: a cross
286 of measurements with one ray path and a single line of measurements with a different ray path.

287 A successful inversion is possible with a limited set of analyses by applying model
288 constraints to the rim values, symmetry, and/or smoothness of the model, following standard
289 practice in geophysical tomography. A rim value constraint encourages the model to match the
290 concentration in the subdivisions along the outermost rim of the sample to a known value. A
291 symmetry constraint gives preference to model outputs that are symmetrical along any
292 orthogonal line through the sample. A smoothness constraint causes the model to expect that
293 neighboring subdivisions contain similar values. Damping the model output through the
294 application of these constraints does not force the results (m) to comply exactly with the
295 constraints, and a weighting factor λ can be included and varied to evaluate the importance of
296 these constraints during the calculation. Each of the three constraints can be removed or
297 weighted separately to evaluate their effects on the fit individually. The value of λ should be kept
298 as low as possible so that the results are influenced primarily by the observed measurements.

299

Methods

300 Simulations were performed to provide more general insight into the potential for error
301 when applying 3D-NPI and 1D diffusion models to path integrated measurements. Then, the
302 whole-block approach was tested experimentally by dehydrating a diopside crystal and
303 comparing the resulting whole-block profiles with slice profiles. In the final section, we review
304 previously published whole-block profiles on hydrogen diffusion in olivine.

305 *Numerical simulations*

306 To investigate the usefulness and potential pitfalls of applying non-path-integrated diffusion
307 models to path-integrated measurements, 40,000 sets of 3D-WB data were simulated and then fit
308 using NPI models. In each simulation, 100 data points were generated along each of three
309 orthogonal whole-block profiles. Then, NPI profiles were fit to these simulated 3D-WB data
310 using the non-linear least squares fitting routine available in MATLAB curve-fitting toolbox.
311 This least squares method is objective but somewhat biased toward fitting points closer to the
312 sample rim where the points and the fit line can show the largest deviations from each other.

313 Simulations were performed assuming diffusion into the sample for isotropic and anisotropic
314 materials, equant and inequant samples, and early to later stages of diffusion (up to 80%
315 saturation, at which point the profiles are approaching a straight line, and error functions no
316 longer produce good fits). Two different NPI fitting approaches were tested using different initial
317 concentrations during the NPI fit: a 3D-NPI fit that uses the true, known initial value of zero, and
318 a set of three 1D-NPI fits (one for each profile direction) that use the plateau in the 3D-WB data
319 in place of the initial value during the fit.

320 ***Experiments***

321 **Sample preparation**

322 A dehydrogenation experiment was performed on a diopside from the Kunlun Mountains,
323 China, similar samples to which were used in previous diffusion experiments (Van Orman et al.
324 1998, 2001). The sample was prepared as a 3.5 mm × 1.6 mm × 1.8 mm orthogonal prism with
325 faces perpendicular to [100]* (which is ~16° from [100]), [010], and [001], and polished using 1
326 μm diamond paste.

327 **3D-WB measurements**

328 Unpolarized FTIR measurements were obtained before and after the experiment using a spot
329 size of 100 μm, resolution of 4 cm⁻¹, and averaged over 200 scans on the Nicolet 20 XSB with IR
330 microscope at the American Museum of Natural History. To aid in FTIR spot location in the
331 absence of a digital sample holder, pits were placed along each of the three orthogonal profiles in
332 100 μm intervals using a laser ablation microprobe. Initial integrated peak areas for unpolarized
333 FTIR spectra with linear baselines range between 20 and 80 cm⁻², depending on the ray path
334 direction. The precision error for the FTIR areas is 2%, based on repeated measurements at the
335 same location according to the following equation: $100 \times \sigma / (m \times \sqrt{n})$, where σ is a single standard
336 deviation and m is the mean area for $n=3$ measurements.

337 **Dehydration**

338 The prepared sample block was heated in a gold-calibrated (error ± 3°) 1-atm vertical
339 alumina tube furnace for 75 hours at 1000 °C and 10^{-11.1} bar oxygen fugacity (at the quartz-
340 fayalite-magnetite buffer) using CO/CO₂ mixes of 97.0% CO₂ (Deines et al. 1974). The resulting
341 absorbance values were normalized to the initial profile measurements to generate whole-block

342 diffusion profiles in three orthogonal directions with concentration units of final absorbance (A) /
343 initial absorbance (A_0). The precision error for A/A_0 concentration values of 3% was determined
344 by propagation of the 2% precision error on each measurement.

345 **Slice measurements by FTIR and SIMS**

346 After whole-block profiles were obtained, the sample was sliced perpendicular to [100]*,
347 and slice measurements were obtained parallel to [010] and [001] by unpolarized FTIR and
348 secondary ion mass spectrometry (SIMS). Concentrations profiles for the slice FTIR
349 measurements were generated by normalizing to the initial FTIR absorbance values adjusted for
350 thickness. One potential concern with this approach is that absorbance measurements obtained
351 using unpolarized radiation may not be directly proportional to thickness, at least for grains with
352 unpolarized absorbance greater than 0.15 and linear polarized absorbance greater than 0.3
353 (Libowitzky and Rossman 1996; Kovacs et al. 2008; Sambridge et al. 2008; Withers 2013; Xia et
354 al. 2013). All initial absorbance values used here are less than 0.13, and therefore should scale
355 directly with thickness.

356 SIMS measurements were carried out at the Carnegie Institution of Washington (DTM)
357 using the Cameca 6f ion microprobe. The primary beam of Cs^+ was 5-10 nA. The accelerating
358 voltage was 10 kV, and the final spot size was 20-40 μm . Hydrogen was measured as $^{16}\text{O}^1\text{H}$. The
359 ratio $^{16}\text{O}/^{30}\text{Si}$ was monitored, and concentrations were obtained following the procedures in
360 Wade et al. (2008). The SIMS measurements were calibrated using clinopyroxene standards
361 PMR-53, ROM271-10, -16 and -21. Standard concentrations are given in Aubaud et al. (2007).
362 The calibration factor for SIMS was 4540 ± 560 ppm wt. $\text{H}_2\text{O}/(\text{counts}/\text{s})$, which corresponds to
363 an accuracy or calibration error of 12%. The SIMS precision errors were determined similarly to

364 those for FTIR using $n=4$ to 5 measurements of each data point, and these errors ranged from 1.6
365 to 7.9 %.

366 **Initial concentration**

367 Initial concentration estimates through the untreated whole block of the Kunlun diopside
368 were complicated by the lack of an infrared polarizer on the FTIR at the time this work was
369 begun. An initial concentration of 33 ppm wt. H_2O was calculated from the average unpolarized
370 spectra measured along three orthogonal ray paths and using polynomial baselines with the
371 clinopyroxene calibration of Bell et al 1995. The error associated with estimating the total
372 absorbance in this way may be considerable but cannot be estimated accurately (Withers 2013).
373 Nonetheless, this value is nearly identical to the average of the eight SIMS measurements that
374 create a plateau of 34.3 ± 0.6 ppm H_2O in the central slice of the heated whole block (shown in
375 Figure 5). When FTIR measurements are scaled to this SIMS value, they overlap in
376 concentrations, supporting both the initial FTIR estimate, and the initial value still preserved in
377 the core of the block. Thus to facilitate comparison with the SIMS measurements, the FTIR
378 concentrations were scaled up from A/A_0 to units of ppm wt. H_2O using a concentration of 34
379 ppm wt. H_2O .

380 **Forward models**

381 Diffusivities were determined from the whole-block diffusion profiles and each of the two
382 sets of slice profiles (FTIR and SIMS measurements) by least-square fits assuming a starting
383 concentration of 34 ppm wt. H_2O and an equilibrium concentration of 0 ppm and using both 3D-
384 WB and 3D-NPI approaches. Fits using 1D models are identical to 3D-NPI because diffusion has
385 not reached the center of the sample.

386 **Inverse models**

387 Inverse analysis of the 3D-WB measurements were performed for both the Kunlun diopside
388 and previously published San Carlos olivine hydration measurements (Demouchy and Mackwell
389 2006) using comparable parameters, with the main focus on San Carlos olivine. The ray paths for
390 the San Carlos olivine measurements are as follows: for the traverse // [100], R is // [010]; for the
391 traverse // [010], R is // [100]; and for the traverse // [001], R is // [010] (Sylvie Demouchy,
392 personal communication). To determine the best set of parameters, the number of grid cells in
393 each direction was systematically increased, and λ was varied to minimize the variance while
394 fitting the olivine profiles as simply as possible (Table 1). Preference was given to odd-
395 numbered grid cell spacing to reduce ambiguity about which values in the central slices were
396 sampled by observations. The maximum number of grid cells was restricted to 5 // [100], 11 //
397 [010], and 9 // [001] to avoid including parameterized sections unconstrained by observations.
398 Final grids were $9 \times 9 \times 9$ in diopside and $5 \times 9 \times 9$ for San Carlos olivine. A constraint
399 weighting factor λ of 0.05 was applied for both smoothness and symmetry to both sets of
400 measurements, and the rim value was not constrained for the diopside.

401 In addition to calculating the formal error σ_i , which is a minimum value, we assessed the
402 true error on the inversion by simulating a comparable number and distribution of noise-free
403 whole-block hydration profiles followed by inversions. The resulting calculated central value
404 estimate was then compared to the known central value as a function of time and diffusivity //
405 [100] to assess the error. For simplicity, the simulated profiles were produced using only a
406 single-mechanism model of diffusion starting at 0 and increasing to a final solubility of 10 ppm
407 wt. H₂O. The diffusivities // [010] and [001] were held constant at $10^{-13.4}$ and $10^{-12.2}$ m²/s, and the
408 diffusivity // [100] was varied from 10^{-11} to $10^{-12.5}$ m²/s. Removing the rim value constraint while

409 retaining smoothness and symmetry constraints decreased the error in the simulated profile
410 inversions considerably, e.g., from 0.7 below the true value to near 0 ppm wt. H₂O error for
411 simulated profiles after 20 hours and $D_{// [100]}=10^{-12}$ m²/s.

412 **Results and Discussion**

413 In the following sections, we first use numerical simulations to determine the errors that
414 might occur in calculated diffusivities if 3D-whole block (3D-WB) data are interpreted using a
415 3D-non-path-integrated (3D-NPI) model. We then verify both the forward and inverse 3D-WB
416 methods by obtaining both 3D-WB and slice data on a natural diopside sample that has been
417 dehydrated in the laboratory. Finally, we consider some pitfalls that may occur using published
418 3D-WB data for olivine diffusion.

419 There are three general issues at stake in interpreting 3D-WB data. One is the effect on
420 calculated diffusivities; another is the effect on calculated concentrations. These are separate
421 issues, and we demonstrate that it is possible to calculate accurate diffusivities from profile
422 shapes, without knowing absolute or even central value concentrations. The final issue is how
423 one interprets 3D-WB data in terms of equilibrium partitioning or solubility. Here accurate
424 determination of absolute concentrations from core to rim is essential.

425 ***Numerical simulations***

426 Simulation results are illustrated for 3D-NPI models used to fit simulated 3D-WB data in
427 Figure 2 and for the 1D-NPI models that use the plateau in place of the initial value in Figure 3.
428 The extent of diffusion progress with time is represented by percent saturation of the sample. The
429 absolute values of the errors in the best-fit diffusivity on a log scale were added together across
430 all three orthogonal directions to produce a single summed error for each simulation. The effect

431 of changing ray paths was also investigated for samples with varying levels of diffusive
432 anisotropy. The results differ for individual directions, but the final summed errors are the same
433 as in Figure 2 and Figure 3.

434 The diffusivities estimated using 3D-NPI applied to 3D-WB simulated data are generally too
435 large, which corresponds to faster diffusion. This error is concentrated on well-developed
436 profiles in fast or short directions and results in errors in both the absolute value of the diffusivity
437 and the apparent anisotropy. This error arises primarily because the 3D-NPI fit requires at least
438 one diffusion direction fast enough for diffusion to reach the center of the sample to account for
439 relatively rapid changes in the central value of 3D-WB data.

440 **Changes in the apparent central value**

441 An important potential source of confusion when interpreting 3D-WB measurements comes
442 from differences between the central values of 3D-WB data and those of slice data. The central
443 value in a given whole-block profile always changes from the initial value before the
444 concentration in the actual center of the sample changes because the central measurement
445 includes the concentration at the rims (Movies 2 and 3; still image in Figure 4). If, for example,
446 the diffusing species begins to diffuse into a sample block with initial concentration $C_i=0$ from
447 all sides, then the average value of the concentration measured through the sample thickness in
448 any given direction necessarily increases from 0. The whole-block profile will then show an
449 apparent increase in concentration due to contributions from the sample rim in the FTIR
450 integrated value, *but the concentration in the center of the sample may still be 0.*

451 During 3D-NPI analysis of slice measurements, the concentration measured at the center of
452 each profile must be the same in all three directions because only one central point exists. In
453 contrast, the central values measured in whole-block profiles, because they represent averages,

454 may differ for each of the six possible orientations: the profile // [100] with ray path R // [010]; //
455 [100] with R // [010]; // [010] with R // [100]; // [010] with R // [001]; // [001] with R // [100];
456 and // [001] with R // [010]. This effect is more pronounced when R is parallel to a faster
457 direction.

458 **Effect of diffusion progress**

459 The summed error for the 3D-NPI fit tends to first increase as the difference increases
460 between the central values in the whole-block profiles, which change immediately, and the slice
461 profiles, which remains constant at $C_i=0$ until the total concentration change in the sample
462 reaches ~17% saturation. The error then decreases after diffusion reaches the center of the
463 sample. After hydrogen reaches the center, the shape of the best-fit 3D-NPI profiles are able to
464 match the 3D-WB simulated data much better than in previous stages of diffusion, resulting in
465 improved fits (lower r^2 values) and lower summed error (e.g., Figure 2-A1 versus -A2). This
466 improved fit likely results because once hydrogen has reached the center of the sample, the shape
467 of all three orthogonal profiles for both slice and 3D-WB simulated data are controlled by
468 diffusion only in the fast and/or short direction(s) from which hydrogen is infiltrating rather than
469 in all three directions. After the center is reached, diffusivities in the long and/or slow-direction
470 are usually modeled correctly (e.g., // [001] in Figure 2-A1 and -B1), and the error is entirely in
471 the fast or short directions(s). That error in the fast or short direction also tends to decrease with
472 diffusion progress as the system approaches saturation.

473 **Effect of sample geometry and anisotropy**

474 The effect of crystal shape was investigated by holding two sides of a rectangular
475 parallelepiped sample constant at 2.2 mm and varying the length of the third side in a sample

476 with isotropic diffusion ($\log_{10}D=-14$ in m^2/s). The effect of changing anisotropy was investigated
477 in the same way for an equant crystal (each side 2.2 mm) with diffusivities held constant at
478 $\log_{10}D=-14$ m^2/s in two orthogonal directions and varied in the third direction (Figure 2B).

479 The 3D-NPI models typically provide the poorest fit and highest errors in short and/or fast
480 direction(s), which have the most well developed profiles, whereas profiles parallel to slow
481 and/or long directions, which are largely controlled by diffusion from other directions, are
482 relatively easy to fit and do not significantly contribute to the overall error. Equant, isotropic
483 samples have higher error than those with any deviation, perhaps because they lack any direction
484 that is largely controlled by another and so have relatively high errors in all directions. In
485 contrast, 3D-WB data with a single fast direction are modeled remarkably well with 3D-NPI
486 models (right side of Figure 2B) likely because of this low error associated with long/slow
487 directions and because the signal integration, which causes 3D-WB profiles to deviate from slice
488 profiles, is minimized when the ray path is parallel to a slow direction.

489 In these simulations, the maximum summed error in the diffusivities is around 10% on a log
490 scale. For diffusivities on the order of 10^{-14} m^2/s , 10% summed log error corresponds to
491 estimated diffusivities in each direction that are about a factor of 5 too fast. This error is
492 concentrated on the fast and short direction(s) of a rectangular parallelepiped, resulting in errors
493 in the calculated extent of anisotropy, and is most pronounced when the concentration in the
494 center of the sample first changes significantly from the initial concentration. These factor of 5
495 errors are associated with simulated, and therefore perfect, data. As shown in the next section,
496 the error associated with fitting a 3D-NPI model to real 3D-WB measurements may be closer to
497 a full order of magnitude.

498 **A 1D approximation using central plateau values**

499 If the goal is only to extract diffusivities, as opposed to true concentrations, it is possible to
500 obtain better fits using NPI models than those shown Figure 2 by assuming that the central
501 plateau in the 3D-WB data is also the initial value during fitting. This assumption is clearly not
502 the case because the central concentration measured through the whole block is neither the initial
503 nor the true center value, but a path integrated value. Nonetheless, by taking the measured 3D-
504 WB central value as both the initial and final concentration, the fitting procedure becomes
505 entirely dependent on the curvature of the profile, which has enough information to predict a
506 diffusivity that is remarkably accurate, especially for early stages of diffusion. For diffusion
507 progress less than 50%, the total summed error on the three diffusivities is less than 0.5 log units
508 (Figure 3). As diffusion progresses and the shape of the profiles change to become less steep, the
509 assumption of no change in the true central concentration (essentially the approximation here)
510 becomes worse, and the corresponding fits also worsen. A set of three 1D profiles works better
511 than 3D-NPI (which we explored but do not show here) for use in this approximation because it
512 allows greater flexibility to accommodate differences in the 3D-WB profile plateau values
513 (Figure 4). Moreover, with a 1D approximation, profiles are not needed in all three directions to
514 provide a diffusivity estimate. Thus, while the full 3D-WB approach is certainly preferable for
515 obtaining accurate diffusivities and concentrations, the 1D approximation using the central
516 plateau values can provide a reasonable estimate of the diffusivities based on 3D-WB
517 measurements. Interpretations that rely on an understanding of true concentrations, however, still
518 require forward or inverse methods that take path integrations effects into account. Examples of
519 such applications include partitioning, equilibrium solubility, and “metastable equilibrium”, a
520 concept we discuss in the final section on hydrogen diffusion in olivine.

521 ***Experiments***

522 After the 3D-WB measurements were obtained for heat-treated Kunlun diopside by FTIR,
523 the central slice of the treated block was cut perpendicular to [100]* and analyzed by both FTIR
524 and SIMS. The resulting hydrogen concentration profiles are reported in Supplementary Table 1
525 and plotted in Figure 5. When the unpolarized A/A_0 FTIR measurements are scaled to a 34 ppm
526 wt. H₂O initial value, the central slice zonation and central plateau are very similar to those
527 measured by SIMS, with concentrations overlapping within the analytical precision for many
528 points. Both the SIMS and FTIR measurements show a sudden drop of ~10 ppm wt. H₂O near
529 the center of the slice in the profile parallel to [001]. This drop suggests non-diffusive behavior
530 unrelated to measurement precision, although no nearby cracks were visible. The concentrations
531 measured through the treated whole block are lower than the initial value because of path
532 integration through the lower-concentration rims.

533 We applied the inverse tomographic method to these whole block measurements to solve for
534 the 3-D concentration structure using a $9 \times 9 \times 9$ grid and symmetry and smoothness constraints
535 with weighting parameter $\lambda=0.05$. The fit to the whole block data is shown Figure 6A, and the
536 predicted profiles in the central slices based on the inversion is shown in Figure 6B. The formal
537 error on the inversion σ^2 is extremely small, only 5×10^{-4} , although this value is a minimum and
538 does not include the 2% error from the FTIR area determination. The formal error σ^2 is not
539 always so small; see Table 1 for σ^2 as a function of grid spacing and λ for the San Carlos olivine
540 measurements discussed below and the Methods section for a description of simulations to assess
541 the true error in these simulations.

542 Results for the inverse model of the whole-block measurements produced a good fit to both
543 the whole-block and slice measurements with central values around the initial concentration, 34

544 ppm wt. H₂O, which is identical to the plateau in the SIMS measurements (Figure 5B). Thus,
545 there is strong coherence among the SIMS and FTIR measurements and the inverse approach for
546 solving for 3D concentration zonation within the whole block. These results agree with the
547 simulations showing that the central value of the 3D-WB measurements decreases from the
548 initial value of as a result of signal integration, while the true central value of the slice
549 measurements remains at the initial value.

550 The hydrogen diffusivities were then calculated by forward models from the whole-block
551 measurements (Table 2) using both 3D-WB and 3D-NPI approaches and from both the FTIR and
552 SIMS slice data using 1D models, which are adequate given that diffusion did not reach the
553 block center. The FTIR and SIMS slice data yield diffusivities that are within 0.1-0.2 log units of
554 each other. The 3D-WB analysis produced diffusivities that are within error but 0.1 to 0.5 orders
555 of magnitude slower than those determined from the slice measurements. In contrast to the
556 excellent agreement and fits of the whole-block and slice measurements analyses, 3D-NPI fits to
557 the whole-block measurements fail to accurately match the measurements, with a correlation
558 coefficient r^2 only 0.21 versus 0.64 for 3D-WB, and produce diffusivities almost an order of
559 magnitude too fast // [001] and about a factor of five too fast // [010]. Thus, given real
560 uncertainties in analytical measurements, the whole block effects can lead to order of magnitude
561 summed error in diffusivities when path integration effects are not taken into account.

562 Attempts to improve the fits by adjusting the diffusivities manually did not yield any clear
563 improvement because in 3D-WB analysis, the profiles are highly interdependent and often a
564 fundamentally different shape than in 3D-NPI. For instance, increasing $D_{[001]}$ to closely fit the
565 shape of the 3D-WB profile // [001] results in calculated profiles that are lower than most of the
566 central measurements // [010]. Also, the 3D-NPI fits cannot easily match the squat 3D-WB

567 profiles, and the resulting error is concentrated in the direction with the most well-developed
568 profile, in agreement with the results of the numerical simulations.

569 An important source of uncertainty in these FTIR measurements arises from a change in the
570 relative heights of the individual O-H stretching peaks, which are grouped together in this
571 analysis to produce a single bulk hydrogen diffusivity. In some dehydration experiments, the
572 final bulk absorbance area is as large or larger than the initial area measured at the same position
573 and orientation. We attribute this change to a re-distribution of the hydrogen among different
574 crystallographic environments similar to the changes first described in clinopyroxene by Skogby
575 and Rossman (1989). A full peak analysis is beyond the scope of this paper, but this result
576 emphasizes both the need for future work that considers O-H peaks individually and the potential
577 importance of a pre-annealing (heating to dryness) and/or a hydrogen saturation step to allow the
578 point defects to reach equilibrium prior to the main experiment.

579 ***Hydrogen diffusion in olivine***

580 One potential application of the whole-block method is the diffusion of hydrogen in olivine,
581 which is critical to how the mantle melts and deforms. Experimental measurements of hydrogen
582 diffusion have been performed on San Carlos olivine (Mackwell and Kohlstedt 1990; Kohlstedt
583 and Mackwell 1998; Demouchy and Mackwell 2006; Du Frane and Tyburczy 2012) and
584 synthetic forsterite (Demouchy and Mackwell 2003; Padrón-Navarta et al. 2014). However,
585 significant uncertainty remains about how best to interpret and apply these studies to obtain
586 ascent rates based on olivine dehydration profiles (Le Voyer et al. 2014; Thoraval and
587 Demouchy 2014). Here we apply the whole-block method to the hydration profiles of Demouchy

588 and Mackwell (2006) both to gain insight into the system and to provide an example of the
589 whole-block approach.

590 Although Demouchy and Mackwell (2006) interpreted their measurements using non-path-
591 integrated models, their concentration profiles were obtained through the whole block and must,
592 therefore, be revisited using models that take path integration effects into account. This task is
593 significantly complicated by the proposed existence of two separate hydrogen diffusion
594 mechanisms (Kohlstedt and Mackwell 1998): one (“proton-polaron”) that is fast but short-lived
595 and dependent on ferric iron content and one (“proton-vacancy”) that is relatively slow and
596 modeled as beginning once the fast mechanism saturates at a concentration value that Kohlstedt
597 and Mackwell (1998) call “metastable equilibrium”. Isolating and quantifying diffusivities in this
598 sequential two mechanism model requires using the concentration of the final value in profiles
599 measured over short time scales as the initial profile for diffusion profiles measured over longer
600 times scales. Curiously, the slice profiles of Kohlstedt and Mackwell (1998) are not entirely
601 consistent with this approach (Figure 7), possibly as a result of changes in experimental oxygen
602 fugacity or differences in ferric iron content between their 1 hour and 8 hour samples. Demouchy
603 and Mackwell (2006) demonstrate the principle using a consistent set of experimental parameters
604 and 4 ppm wt. H₂O for “metastable equilibrium” (Figure 8A – DM06). However, 4 ppm is not a
605 reasonable initial value for fitting their 20 hour hydration profiles if the minimum concentration
606 in the sample is actually lower.

607 We applied the inverse tomographic method to these whole block measurements to solve for
608 the true central concentration using a $5 \times 9 \times 9$ grid, symmetry and smoothness model constraints
609 with a weighting parameter $\lambda=0.05$. Tomographic inverse modeling provides a good fit to these
610 measurements (Figure 6B) and indicates that the minimum value in this sample is 3.0 ± 0.1 ppm

611 wt. H₂O. This value is not in the exact center of the sample, consistent with higher measured
612 whole-block values in the center of the profile // [010] and perhaps indicating some
613 heterogeneity within the sample similar to that observed in the slice measurements of Kunlun
614 diopside. Note that the errors reported here include only the formal error on the inversion, σ_i , not
615 the calibration error on the original concentration estimates based on FTIR, which is on the order
616 of a few ppm wt. H₂O.

617 Approximately 25% of the observed 4 ppm wt. H₂O plateau value in the 20 hour San Carlos
618 olivine hydration profile of Demouchy and Mackwell (2006) is the result of signal integration
619 through the whole block. This difference, though trivial in absolute terms (3 ppm vs. 4 ppm wt.
620 H₂O, within calibration error for FTIR measurements), is important for two reasons. First, when
621 applying whole-block forward models and the sequential two mechanism model, 3 ppm wt. H₂O
622 is more appropriate than 4 ppm wt. H₂O as a transitional “metastable equilibrium” value. The
623 error in the central value translates directly into an error in the “metastable equilibrium” state,
624 and this error is preventable by applying the 3D-WB method. Second, there is actually an
625 inconsistency in the olivine measurements and their interpretation for a central value of 3 ppm
626 wt. H₂O. The metastable equilibrium model dictates that the olivine reaches a metastable
627 saturation value after 1 hour. This value is 3 and not 4 ppm H₂O, based on the inverse modeling
628 of the 3D-WB measurements. This interpretation is inconsistent with the maximum
629 concentrations in the 1 hour profiles (4 to 5 ppm wt. H₂O at the rims), which are certainly greater
630 than the central minimum concentration in the 20 hour sample (3 ppm wt. H₂O). This difference
631 cannot be dismissed as simply within error for FTIR because calibration error generally affects
632 all values the same way, i.e., all reported values are wrong by approximately the same amount in
633 only one direction. Thus, the model in which two diffusion mechanisms operate sequentially and

634 have some “metastable equilibrium” concentration is likely in error. At stake is not just an
635 estimate of a particular diffusivity, but a fundamental understanding of how hydrogen diffuses in
636 Fe-bearing olivine. Does diffusion proceed primarily by one dominant mechanism or multiple?
637 Assuming multiple mechanisms are important, under what time scales do they operate, and to
638 what extent can they be separated and individually quantified?

639 These questions are difficult to answer unambiguously with the available published profiles.
640 Figure 8 shows a series of possible fits that all take path integration into account. In Figure 8A
641 we show 3D-WB profiles that use the revised “metastable equilibrium” value of 3 ppm H₂O and
642 the diffusivities from both the original interpretation of Demouchy and Mackwell (2006; labeled
643 DM06 3D-WB) and Kohlstedt and Mackwell (1998; labeled KM98 3D-WB). The diffusivities
644 from Demouchy and Mackwell (2006) provide a much better fit to this set of measurements
645 ($r^2=0.87$) than the diffusivities of Kohlstedt and Mackwell (1998) ($r^2=0.61$). Two examples of
646 possible alternative fits to the profiles from Demouchy and Mackwell (2006) are also shown in
647 Figure 8: a single diffusion mechanism and two mechanisms that operate simultaneously and
648 assume a fast mechanism that has diffusivities taken from Kohlstedt and Mackwell (1998) and
649 saturates at 3 ppm wt. H₂O. These models provide slightly worse r^2 values than the 3D-WB
650 revised sequential model (0.79 for single mechanism and 0.83 for simultaneous vs. 0.87), but
651 they have the virtue of allowing the concentration at the rims in the 1 hour experiment (4-5 ppm
652 wt. H₂O) to be greater than a “metastable equilibrium” value of 3 ppm H₂O. The single
653 mechanism model is also much simpler to apply. Other fits are certainly possible, e.g., using
654 different values at which the fast mechanism saturates or by varying the diffusivities, particularly
655 for the rough 20 hour profiles // [100]. The original fits of Demouchy and Mackwell (2006) are
656 very similar to many other models, including the one that applies the true central concentration

657 of 3 ppm H₂O as the “metastable equilibrium” starting concentration in the 20 hour experiments
658 and the model using simultaneously operating mechanisms. This agreement arises because of the
659 success of the 1D approximation using the central 3D-WB plateau value (Figure 3), which is in
660 essence what Demouchy and Mackwell (2006) did. Experiments that show greater diffusion
661 progress would better enable distinction among the different models.

662 Two conclusions can be drawn from this exercise with regard to hydrogen diffusion in
663 olivine. First, taken in isolation, the hydration profiles reported by Demouchy and Mackwell
664 (2006) can be adequately modeled by a relatively simple (and therefore preferable) single
665 mechanism model. Second, none of the modeling approaches discussed here can satisfactorily
666 explain or reconcile the large (0.9 to 1.8 orders of magnitude) discrepancy between the slow
667 mechanism hydrogen diffusivity in San Carlos olivine observed by Demouchy and Mackwell
668 (2006) and that observed by Kohlstedt and Mackwell (1998) and in forsterite (Figure 10 and
669 Table 3). In particular, Kohlstedt and Mackwell (1998) show a clear change in anisotropy in San
670 Carlos olivine from a fast direction // [100] at 1 hour to a fast direction // [001] at 8 hours that is
671 not as readily apparent in the Demouchy and Mackwell (2006) profiles. Additional experiments,
672 especially those that consider site-specific diffusivity as in Padrón-Navarta et al. (2014), are
673 required to fully understand this system.

674 **Implications**

675 A new treatment of whole-block measurements successfully yields hydrogen diffusivities in
676 three orthogonal directions of a rectangular prism without cutting the sample. This approach will
677 simplify future diffusion experiments that rely on FTIR and be bolstered by the results of
678 measurements from cut samples. Multiple hydrogen diffusion experiments are now possible

679 using the same sample (e.g., a time series using profiles), which will minimize both the amount
680 of sample material needed and the error associated with heterogeneity among samples from the
681 same locale. The efficacy of the method has been demonstrated experimentally for hydrogen
682 diffusion in diopside. Because the differences between whole-block and slice measurements are
683 often significant, path integration effects must be considered when evaluating whole-block
684 measurements. Failure to include these path integration effects may result in errors of half an
685 order of magnitude or more in fast and/or short directions. In cases where diffusion has not
686 progressed significantly, it is possible to use a 1D approximation with the whole-block central
687 plateau to accurately obtain diffusivities with errors less than 0.5 log units because diffusivities
688 are more sensitive to profile shape than the absolute concentration of the central plateau.
689 However, in cases where absolute concentrations are required, forward or inverse 3D-WB
690 models are necessary to recover core concentrations. An important application of these effects is
691 in the interpretation of olivine diffusion experiments. We use the new whole block forward and
692 inverse methods to re-evaluate whole block measurements of olivine dehydration (Demouchy
693 and Mackwell 2006). The resulting diffusivities are within the range of those originally
694 published for the slower “proton-vacancy” mechanism, but could also be consistent with a
695 simpler single diffusive mechanism. The newly determined central value concentrations create
696 inconsistencies with existing interpretations of dual mechanisms and “metastable equilibrium”
697 concentrations. The whole-block approach is likely to prove very useful for future work on
698 olivine, pyroxene, and other nominally anhydrous minerals to illuminate the parameters that
699 define different hydrogen diffusion mechanisms (particularly the role of differing
700 crystallographic environments as in Padrón-Navarta et al. 2014) for use in broader Earth science
701 applications such as determining crystal residence times (Danyushevsky et al. 2002), magma

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702 degassing rates (Hauri 2002), and ascent times (Demouchy et al. 2006, Peslier and Luhr 2006,
703 Denis et al. 2013, Chen et al. 2013); interpreting hydrogen concentrations in melt inclusions
704 (Portnyagin et al. 2008, Chen et al. 2011, Gaetani et al. 2012, Bucholz et al. 2013, Lloyd et al.
705 2013) and mantle xenoliths (review by Peslier 2010); and electrical conductivity (e.g., Karato
706 1990).

707

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714

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- 886

887

List of figure captions

888 Figure 1. Sketch showing representative depth slices of an initially dry rectangular prism
889 that has been partially hydrated during a diffusion experiment. The whole-block approach
890 developed here evaluates measurements derived from signals that are transmitted and absorbed
891 by all the layers of the entire sample along the transmission direction rather than cutting a slice
892 from the center of the sample.

893

894 Figure 2. Sensitivity maps showing summed error in the diffusivities on a log scale for fits to
895 simulated 3D-WB data using a diffusion model that does not take path integration into account
896 (the 3D-NPI model) (A) for isotropic diffusion as a function of sample shape and diffusion
897 progress and (B) in an equant crystal as a function of anisotropy and diffusion progress. Example
898 profiles are labeled and shown below each map. In all simulations, the diffusing species is
899 entering the sample, the initial concentration is 0, and the equilibrium concentration is S . D
900 values are the initial diffusivities in m^2/s used to simulate the 3D-WB data (only 20 simulated
901 data points out of 100 are shown for clarity) in each orthogonal direction ($[100]$, $[010]$, and
902 $[001]$; length of crystal along $[100]$ is 2.2 mm in all examples), and D_{fit} is the diffusivity in each
903 direction in m^2/s determined using a least-squares best-fit to the 3D-NPI model. Ray paths R are
904 $// [010]$ for profiles $// [100]$, $R // [100]$ for profiles $// [010]$, and $R // [100]$ for profiles $// [001]$.
905 For comparison, simulated slice profiles (from 3D-NPI equations) are shown for each example.

906

907 Figure 3. Sensitivity maps calculated using the same numerical simulation approach as in
908 Figure 2 showing summed error in the diffusivities for a set of three 1D-NPI models fit to 3D-

909 WB data using the the 3D-WB plateau concentration in place of the true initial value (zero), thus
910 ignoring the apparent drop in the 3D-WB central concentration and emphasizing the profile
911 curvature, which is what primarily determines the calculated diffusivities.

912

913 Figure 4. Simulated diffusion profiles for the hydration of an initially dry cube in which
914 diffusion is strongly anisotropic. D is diffusivity in each direction, and R is the direction of the
915 ray path of the infrared beam. See Movies 2 and 3 for changes in diffusion profiles with time.

916

917 Figure 5. Concentration profiles in diopside after partial dehydration for 75 hours at 1000 °C
918 measured using unpolarized FTIR and SIMS for (A) the uncut block (3D-WB) and (B) a slice cut
919 from the center of the sample. Diffusivities (D) were determined by least square fits in each
920 direction for the 3D-WB FTIR measurements using both the 3D-WB model and the non-path-
921 integrated 3-dimensional model (3D-NPI) and for the slice FTIR and SIMS measurements using
922 3D-NPI models. The initial concentration is assumed to be 34 ppm wt. H₂O and uniform, and the
923 equilibrium concentration is assumed to be 0. The measurements are provided in Supplementary
924 Table 1, and error estimates and correlation coefficients are listed in Table 2.

925

926 Figure 6. (A) Results of inverse analyses showing FTIR 3D-WB measurements, predicted
927 values based on inverse model results, and associated residuals for the dehydrated Kunlun
928 diopside from this study. (B) FTIR measurements of a slice cut from the center of the sample and
929 predicted values based on inverse analysis. Light grey dashed lines show the sub-divisions in the
930 $9 \times 9 \times 9$ grid used to model the system, where each block is assumed to contain a uniform
931 concentration. Propagated errors plot within the symbols.

932

933 Figure 7. Hydrogen profiles // [100] from central slice FTIR measurements reproduced from
934 Kohlstedt and Mackwell (1998). The original paper used a value of 7 H/10⁶Si (red line) as the
935 initial value when fitting the 8 hour data, but this “metastable equilibrium” value cannot also be
936 used as the final value to fit their 1 hour data. Note the different temperatures, pressures, and
937 oxygen fugacities of the two experiments.

938

939 Figure 8. Whole-block FTIR measurements (squares) for hydrogen diffusion experiments at
940 1 and 20 hours on San Carlos olivine at 200 MPa, the Ni/NiO buffer, 900 °C, and observed
941 initial concentration $C_i=0$ reproduced from Demouchy and Mackwell (2006), and five different
942 approaches to modeling diffusivities based on these profiles. Error bars are shown in the original
943 text and are omitted here for clarity. (A) “DM06” is the original fit by Demouchy and Mackwell
944 (2006) that does not consider path integration effects and uses 4 ppm wt. H₂O (dotted grid line)
945 as a transitional state between two mechanisms. “DM06 3D-WB” uses the diffusivities estimated
946 from the original fit with the whole-block forward model and a transitional state of 3 ppm wt.
947 H₂O (the central value in the sample determined by tomography; see text). “KM98 3D-WB” uses
948 an identical approach as “DM06 3D-WB” except the diffusivities are those expected from
949 Kohlstedt and Mackwell (1998). “Single mech.” illustrates a possible fit to both the 1 hour and
950 20 hour profiles using the same diffusivity, initial concentration, and solubility for both sets of
951 concentration profiles. (B) Shows the two mechanisms described by Demouchy and Mackwell
952 (2006) modeled as operating simultaneously rather than sequentially. For simplicity and
953 consistency with previous work, the fast mechanism diffusivities (mech. 1) were held constant at
954 the values expected based on Kohlstedt and Mackwell (1998), and best fit slow diffusivities were

955 obtained by fitting to both sets of profiles simultaneously. The diffusivity estimates for all
956 models are given in Table 3 and plotted on an Arrhenius diagram in Figure 10.

957

958 Figure 9. Observed whole-block measurements for hydrogen diffusion in San Carlos olivine
959 (Demouchy and Mackwell 2006) after 20 hours at 900 °C, predicted whole-block measurements
960 based on inverse model results, and associated residuals. Light grey dashed lines show the sub-
961 divisions in the $5 \times 9 \times 9$ grid used to model the profiles, where each block is assumed to contain
962 a uniform concentration. Propagated errors on the predicted values plot within the symbols.

963

964 Figure 10. Arrhenius diagram of hydrogen diffusivity measurements in olivine. Self-
965 diffusivity is from Du Frane and Tyburczy (2012). New diffusivity estimates from this study
966 (Figure 8) based on the measurements of Demouchy and Mackwell (2006; labeled DM06) are
967 shown for both the assumption of a single diffusion mechanism and for a slow mechanism
968 assumed to operate simultaneously with the fast mechanism of Kohlstedt and Mackwell (1998;
969 labeled KM98). PN is Padrón-Navarta et al. (2014), and bracketed elements in PN refer to
970 different H defects inferred from FTIR absorption bands. The five studies in which orientation
971 was not determined (Padrón-Navarta et al. 2014, Gaetani et al. 2012, Hauri 2002, Chen et al.
972 2011, and Portnyagin et al. 2008, who report a minimum diffusivity) are plotted only on the first
973 panel for comparison.

974

List of table headings

975 Table 1. The number of grid cells parallel to each direction and constraint weighting factor λ
976 used to model the 20 hour San Carlos olivine hydrogen profiles from Demouchy and Mackwell

977 (2006), resulting goodness of fit parameter σ^2 , and estimated minimum concentration in the
978 center in ppm wt. H₂O.

979

980 Table 2. Diffusivities (D in m²/s), associated errors, and correlation coefficients (r^2)
981 determined for Kunlun diopside at 1000°C using four different approaches. “3D-WB” is the
982 whole-block path-integrated forward model. “1D” and “3D-NPI” are 1- and 3-dimensional non-
983 path-integrated models. The reported errors are based on 95% confidence intervals in the fitting
984 procedure. Diffusivity estimates // [100]* from slice measurements are not available and are
985 labeled “N/A”.

986

987 Table 3. Diffusivity estimates in m²/s for various forward models (Figure 8) to the whole-
988 block measurements of Demouchy and Mackwell (2006). Diffusivities reported by Kohlstedt and
989 Mackwell (1998) are provided for comparison.

990

991 Supplementary Table 1. Hydrogen concentration as a function of position in heat-treated
992 Kunlun diopside. Total lengths are 3450±10 μm // [100]*, 1609±5 μm // [010], and 1757±6 μm //
993 [001]. FTIR measurements are reported assuming a uniform initial concentration of 34 ppm wt.
994 H₂O. Slice data were not available // [100]* and are labeled “N/A”.

995

Fig 1

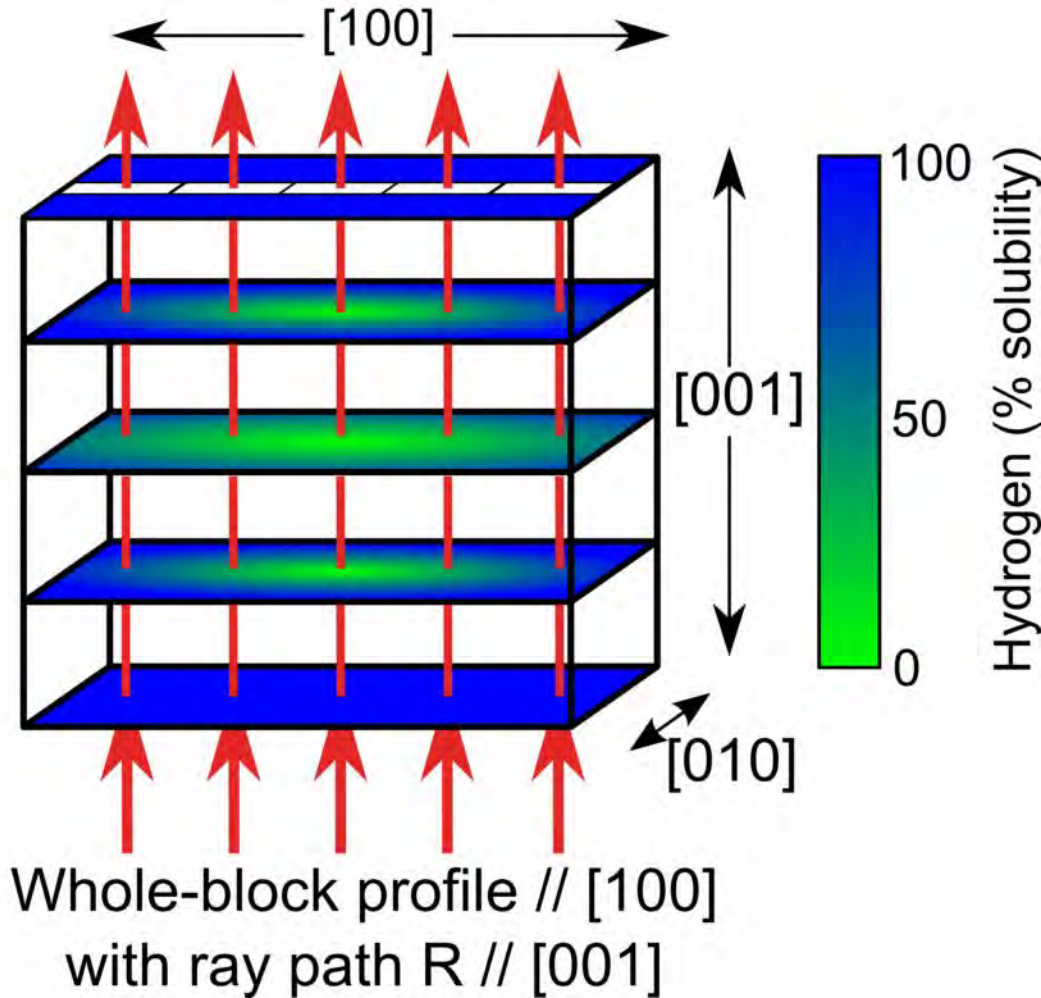


fig 2

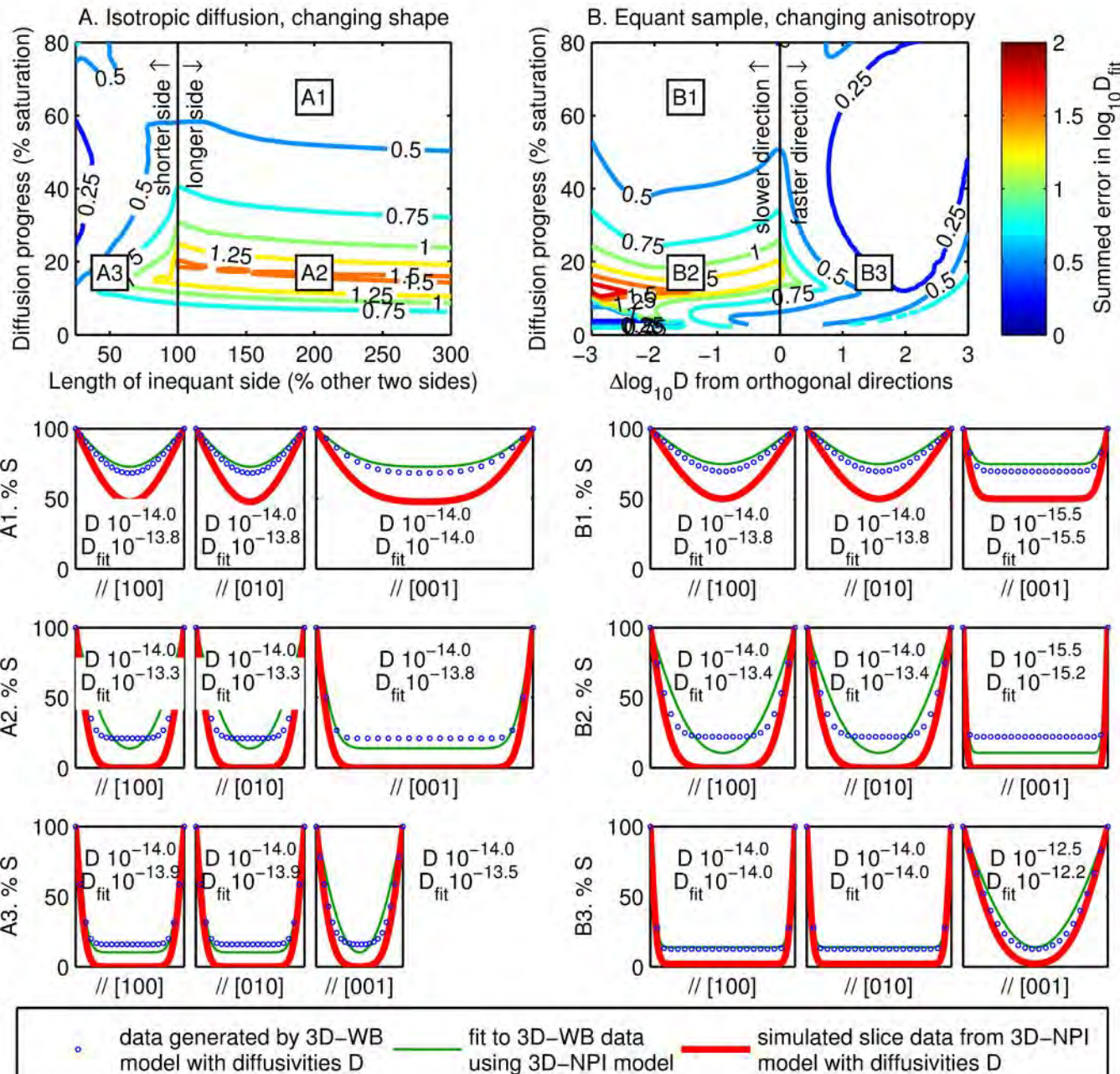


Fig 3 (1D plateau approximation)

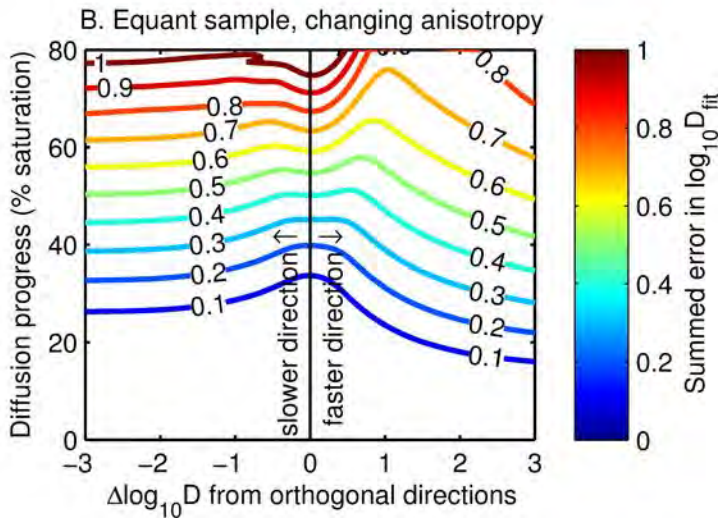
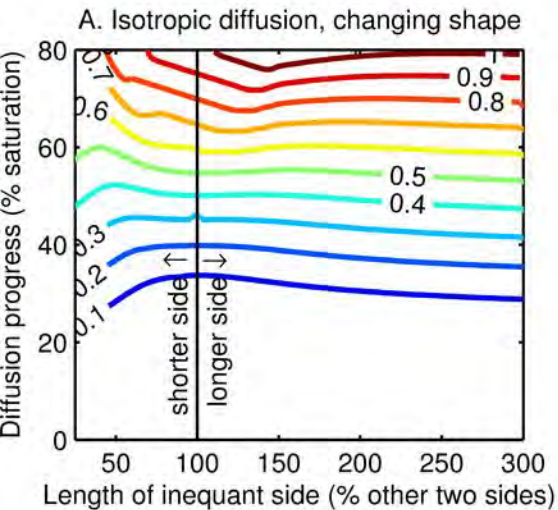


fig 4

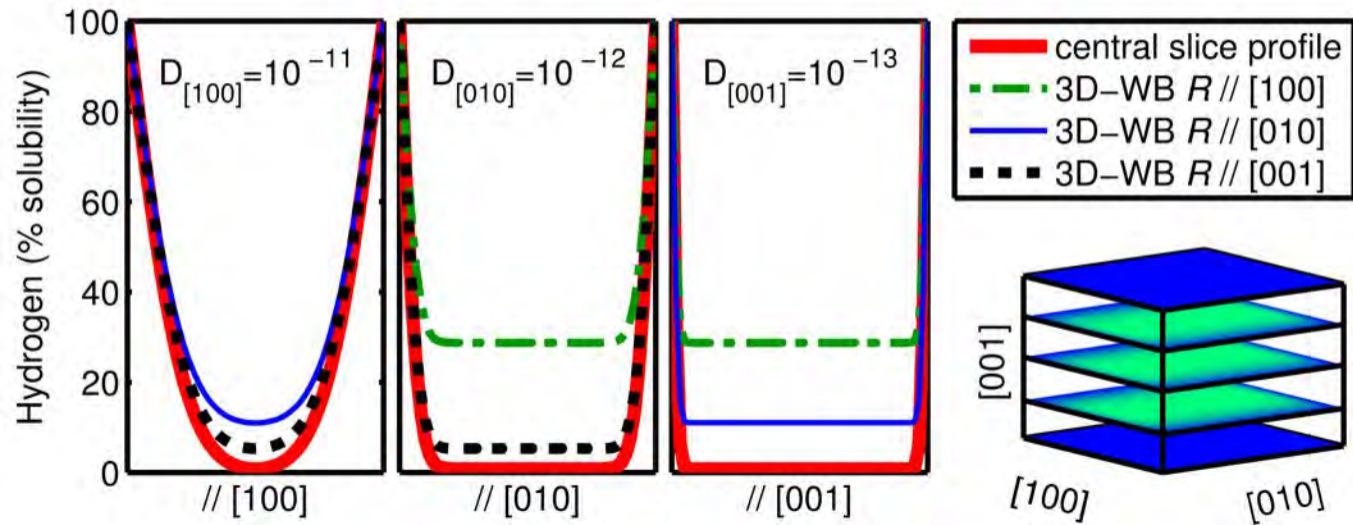


Fig 5

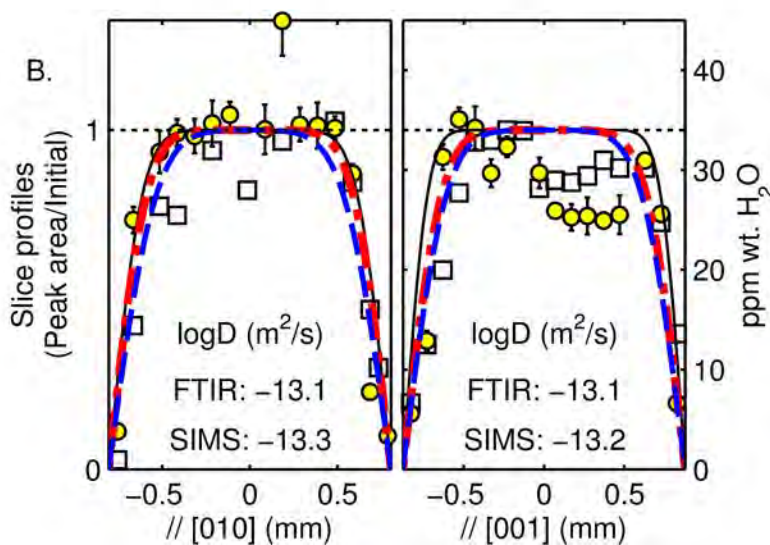
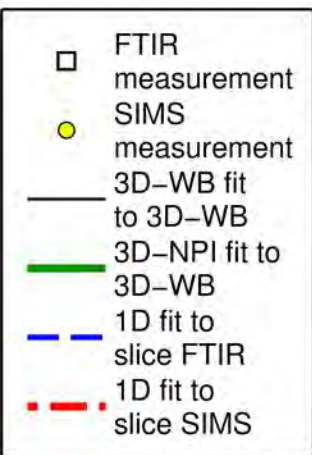
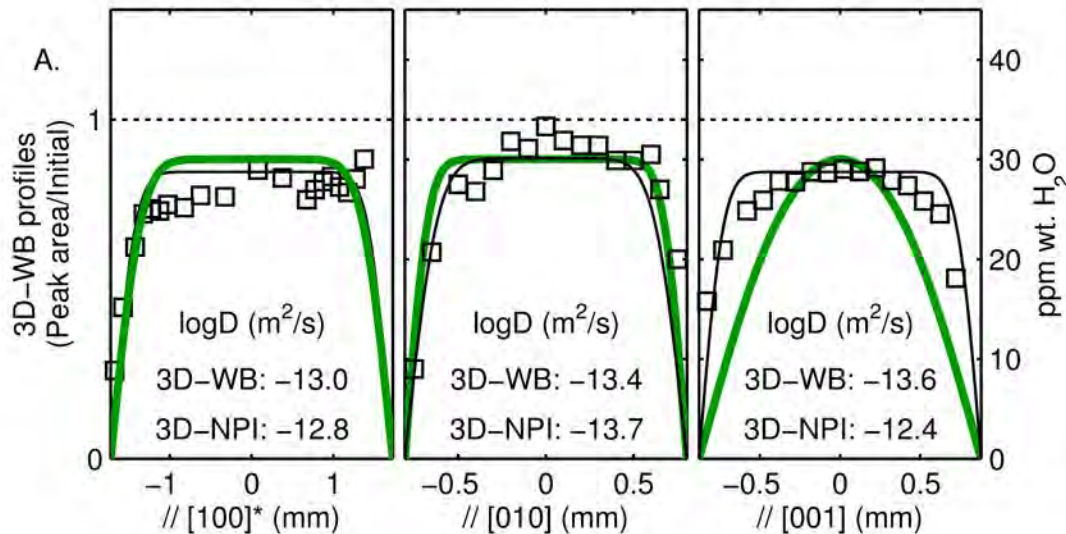


Fig 6

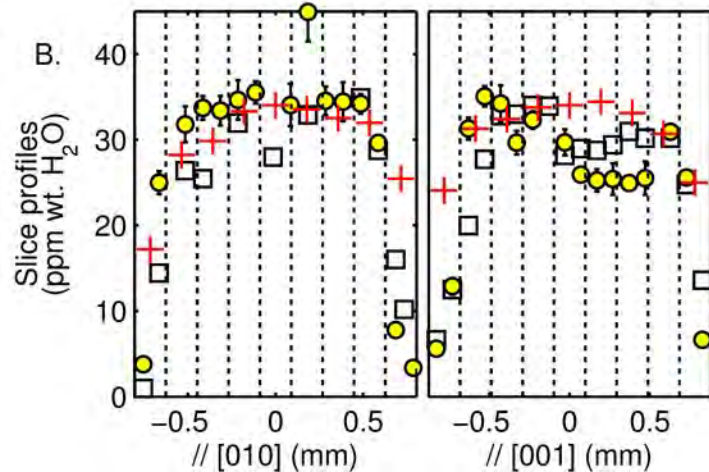
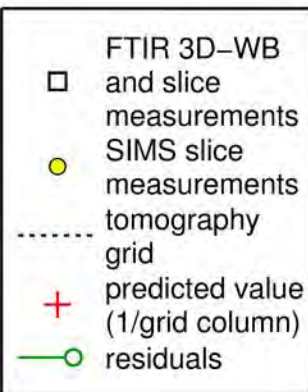
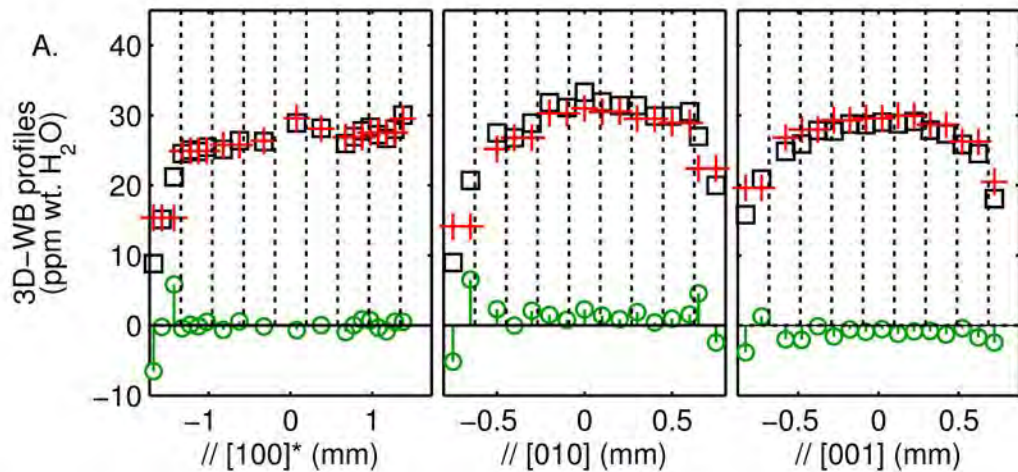


Fig 7

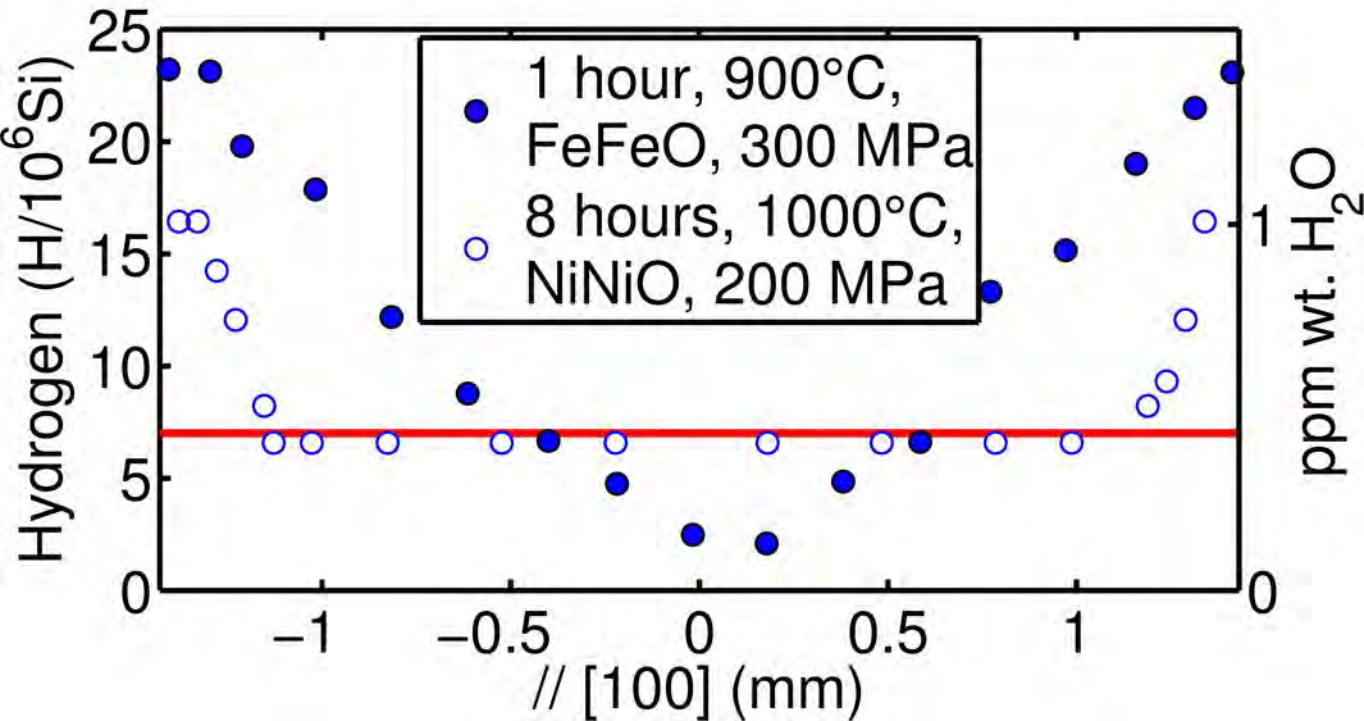


Fig 8

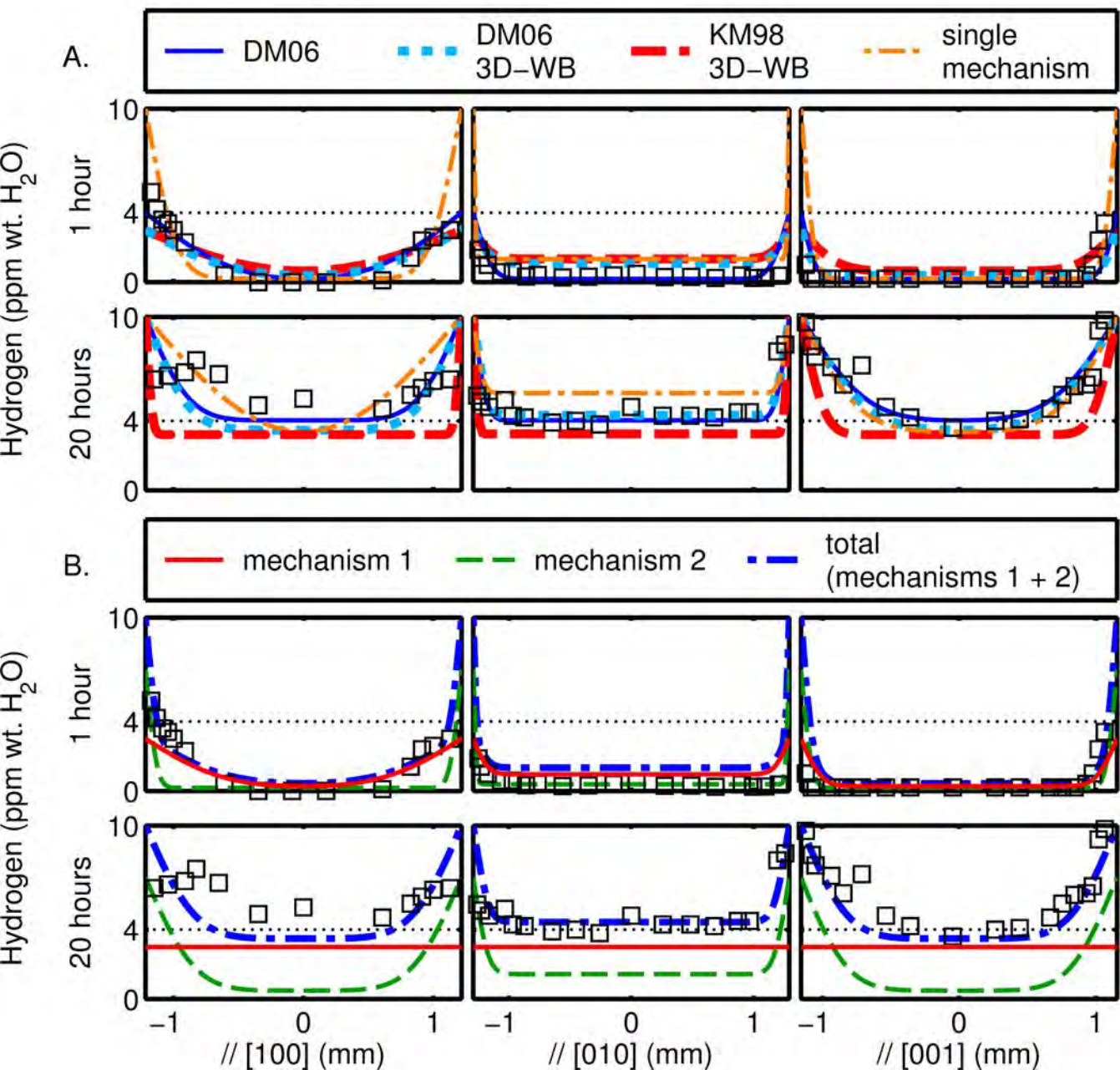


Fig 9

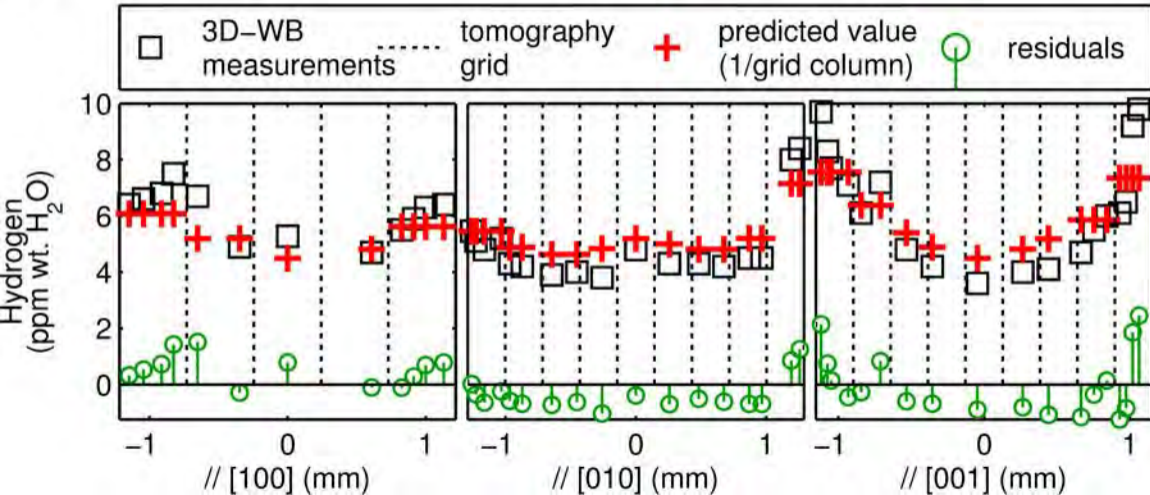
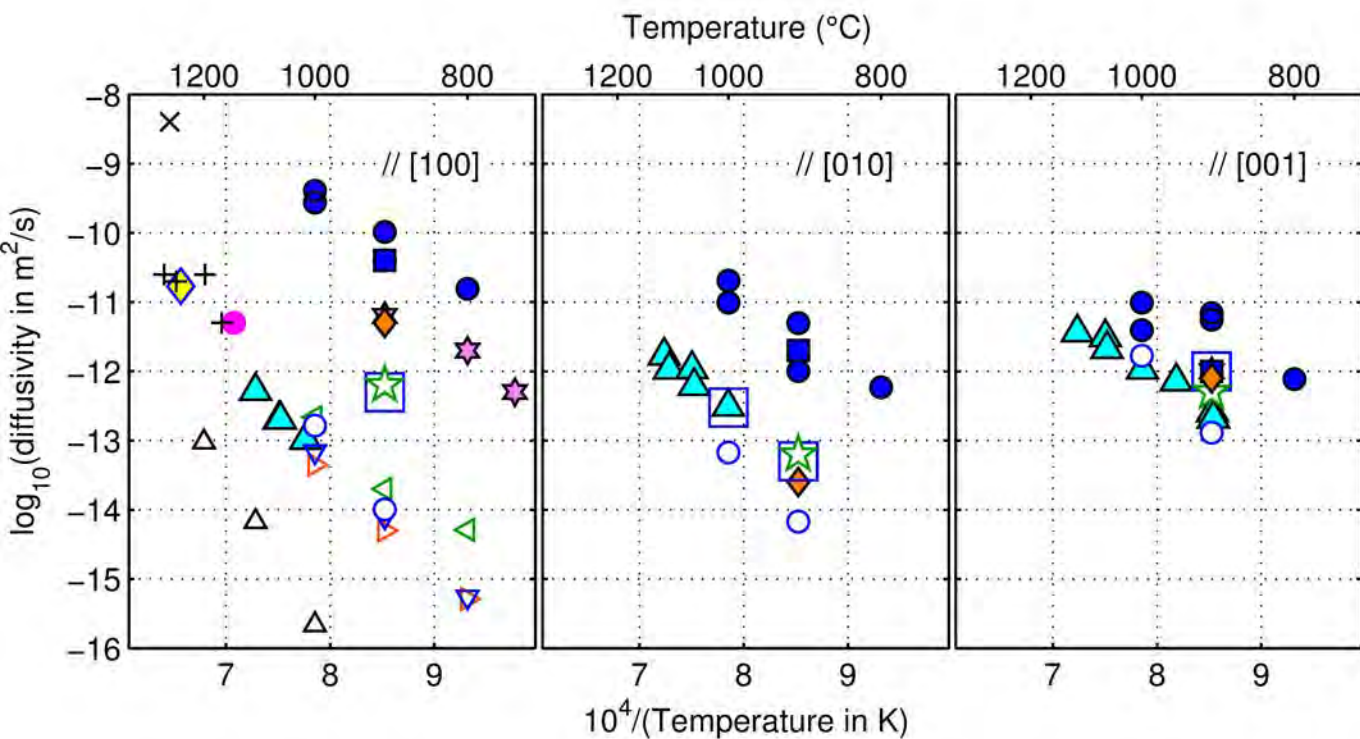


Fig 10



San Carlos olivine

- ☆ Self-diffusivity
- ◇ Single mechanism

Fast mechanism

- KM98
- DM06

Slow mechanism

- KM98: Kohlstedt & Mackwell 1998
- DM06: Demouchy & Mackwell 2006
- ☆ Slow mech. 2 in simultaneous model

Other olivines

- ◇ Gaetani et al. 2012
- × Hauri, 2002
- + Chen et al. 2011
- Portnyagin et al. 2008 (minimum)

Forsterite

- ▲ Demouchy & Mackwell, 2003

Padrón-Navarta et al. 2014

- ◁ PN [Mg]
- △ PN [Si]
- ▽ PN [Ti]
- ▷ PN [Si]-[Ti]

grid // [100]	grid // [010]	grid // [001]	λ	σ^2	concentration at the center (ppm wt. H2O)
1	1	1	0.05	3.7	5.8
2	2	2	0.05	0.4	4.4
3	3	3	0.05	0.08	3.6
5	5	5	0.05	0.01	2.9
5	7	7	0.05	0.005	2.9
5	9	9	0.05	0.002	2.9
5	11	9	0.05	0.002	2.9
5	9	9	0.2	0.003	3.2
5	9	9	0.5	0.01	5.2

<i>Hydrogen in diopside</i>	$\log_{10}D // [100]^*$	$\log_{10}D // [010]$	$\log_{10}D // [001]$	r^2
3D-WB fit to 3D-WB FTIR measurements	-13.0±0.2	-13.4±0.2	-13.6±0.3	0.64
3D-NPI fit to 3D-WB FTIR measurements	-12.8±0.3	-13.7±0.3	-12.4±0.1	0.21
3D-NPI fit to slice FTIR measurements	N/A	-13.1±0.3	-13.1±0.2	0.87 // [010] 0.76 // [001]
3D-NPI fit to slice SIMS measurements	N/A	-13.3±0.4	-13.2±0.4	0.90 // [010] 0.76 // [001]

Fitting approach	Geometry of measurements	Measurements used in fit (experimental time in hours)	Diffusion mechanism	Initial concentration used in fit (ppm wt. H ₂ O)	Solubility used in fit (ppm wt. H ₂ O)	logD // [100]
1D	Slice	1	fast	0	1.8*	-10.2
		8	slow	0.4*	3*	-14.0
3D-NPI	3D-WB	1	fast	0	4	-10.4
		20	slow	4	10	-12.3
3D-WB	3D-WB	1	fast	0	3	-10.2
		20	slow	3	10	-14.0
3D-WB	3D-WB	1	fast	0	3	-10.4
		20	slow	3	10	-12.3
3D-WB	3D-WB	1 and 20	fast	0	3	-10.2
		1 and 20	slow	0	7	-12.2±0.3
3D-WB	3D-WB	1 and 20	single	0	10	-11.3±0.1

*Kohlstedt and Mackwell 1998 use a different calibration scheme than Demouchy and Mackwell 2006 and report their concentrations in H / 10⁶ Si, wh

logD // [010]	logD // [001]	r2 for fits to both 1 hr and 20 hr 3D-WB measurements of Demouchy and Mackwell 2006	Reference
-11.7	-11.2	N/A	Kohlstedt and Mackwell (1998) (KM98)
-14.2	-12.9		
-11.7	-12.0	0.89	Demouchy and Mackwell (2006) (DM06)
-13.3	-12.0		
-11.7	-11.2	0.61	This study: 3D-WB applied using KM98
-14.2	-12.9		
-11.7	-12.0	0.87	This study: 3D-WB applied using DM06
-13.3	-12.0		
-11.7	-11.2	0.83	This study: Simultaneous fit assuming KM98
-13.2±0.5	-12.3±0.3		fast mechanism
-13.6±0.6	-12.1±0.2	0.79	This study: Single mechanism

which we convert to ppm wt. H₂O by dividing by 16.3

	Position // [100]* (mm)	Hydrogen // [100]* (ppm wt. H2O)	Position // [010] (mm)	Hydrogen // [010] (ppm wt. H2O)	Position // [001] (mm)
3D-WB FTIR measurements	-1.68	8.8	-0.75	9.0	-0.83
	-1.58	15.2	-0.65	20.7	-0.73
	-1.43	21.2	-0.50	27.5	-0.58
	-1.33	24.5	-0.40	26.8	-0.48
	-1.23	25.1	-0.30	28.9	-0.38
	-1.13	24.8	-0.20	31.8	-0.28
	-1.03	25.5	-0.10	31.1	-0.18
	-0.83	25.2	0.00	33.3	-0.08
	-0.63	26.4	0.10	31.9	0.02
	-0.33	26.2	0.20	31.4	0.12
	0.08	28.9	0.30	31.4	0.22
	0.38	28.2	0.40	29.9	0.32
	0.68	26.0	0.50	29.9	0.42
	0.78	27.0	0.60	30.5	0.52
	0.88	27.8	0.65	27.0	0.62
	0.98	28.3	0.75	20.0	0.72
	Slice FTIR measurements	N/A	N/A	-0.75	1.0
			-0.66	14.5	-0.73
			-0.51	26.6	-0.63
			-0.41	25.6	-0.53
			-0.21	32.2	-0.43
			-0.01	28.2	-0.33
			0.19	33.1	-0.23
		0.49	35.1	-0.13	
		0.59	28.9	-0.03	
		0.69	16.1	0.07	

			0.79	10.3	0.17
					0.27
					0.37
					0.47
					0.63
					0.73
					0.83
Slice SIMS measurements	N/A	N/A	-0.75	3.8	-0.83
			-0.66	25.0	-0.73
			-0.51	31.8	-0.63
			-0.41	33.7	-0.53
			-0.31	33.4	-0.43
			-0.21	34.7	-0.33
			-0.11	35.5	-0.23
			0.09	34.1	-0.03
			0.19	44.9	0.07
			0.29	34.5	0.17
			0.39	34.5	0.27
			0.49	34.2	0.37
			0.59	29.6	0.47
			0.69	7.8	0.63
			0.79	3.4	0.73
					0.83

Hydrogen
// [001] (ppm wt. H2O)

15.8
20.9
24.9
25.9
27.8
27.7
28.8
28.7
29.0
28.8
29.1
27.9
27.4
25.8
24.5
18.1

6.6
12.5
20.0
27.7
32.8
33.0
34.0
33.9
28.2
29.0

28.8
29.4
31.0
30.2
30.2
24.8
13.6
5.6
12.9
31.3
35.1
34.2
29.7
32.3
29.7
25.9
25.3
25.4
24.9
25.5
30.9
25.6
6.6