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1 Coupling grid nanoindentation and surface chemical analysis to infer

2 the mechanical properties of shale mineral phases

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13	
14	Highlights:
15	• A protocol was developed for coupling analysis of grid nanoindentation and local
16	mineralogy.
17	• Quartz and illite phases can be subdivided between intact particles and nano-porous

- 18 aggregates.
- 19 Self-consistent modeling provides a basis for upscaling properties from intact particles
- 20 to mineral aggregates.
- 21
- 22

23 Abstract

24 Shale is a low-permeability, multi-phase, and multi-scale composite material with intrinsic 25 heterogeneity in micro-texture and mineralogical composition. In this paper, relationships 26 between the microscale texture and constituent phases of shale were investigated using the 27 SEM (scanning electron microscopy)/BSE (backscatter electron imaging)-EDS (energy-28 dispersive x-ray spectroscopy) methods, and grid nanoindentation experiments. The 29 mechanical properties of the constituent phases of the carbonate-rich Longmaxi shale 30 samples were extracted based on the spatial distribution of mineral phases and the 31 indentation interaction volume. The effects of particle size on the interpreted mechanical 32 properties were analyzed and a characteristic length scale was determined based on a 33 probabilistic analysis. The identified characteristic length for extracting the mechanical 34 properties of constituent mineral phases from the grid nanoindentation technique is about 35 5.8 - 11.7 μ m, i.e., up to 10 times greater than that proposed in prior research. A multi-36 scale mechanical model was established with considerations of a self-consistent scheme 37 for granular morphology to link the microscopic characteristics with the multi-scale 38 mechanical properties of shales. The modeling results showed that the stiffness and the 39 strength of the homogenized nano-porous illite/quartz aggregates in Longmaxi shale can 40 be assessed from the nano-scale mechanical properties of the mineral phases and the nano-41 scale porosity. This study paves the way to accessing the mechanical properties of the 42 constituent phases of composite materials based on the properties of their building blocks 43 and provides extensive insights into their complex mechanical behavior, representing a 44 major step towards developing reliable multi-scale models for engineering applications.

45

46 Keywords: Shale; grid nanoindentation; in-situ mineral identification; mechanical
47 properties of the constituent phases; multi-scale mechanical model

49 **1 Introduction**

50 Shales are fine-grained sedimentary rocks mainly composed of highly consolidated 51 aggregates of clay minerals (phyllosilicates), silt-sized particles of other common soil 52 minerals (e.g., tectosilicates such as quartz and feldspar, and carbonates), and organic 53 matter. Shale formations play important roles in applications such as the geological 54 containment of radioactive waste (Neuzil 2013), extraction of natural gas (Du et al. 2018), 55 and sequestration of carbon dioxide (Boosari et al. 2015; Elliot and Celia 2012; Middleton et al. 2014). The engineering challenges associated with these applications require an in-56 57 depth evaluation of the mechanical and mineralogical properties of shales.

58 Shale is a porous, multi-phase, and multi-scale composite material with intrinsic 59 heterogeneity in microtexture, mineralogy, and mechanical properties, and hence, the 60 accurate measurement of its bulk mechanical properties or interpretation from its 61 constituent phases continues to be a formidable challenge for the geomechanics community. The development of nanoindentation methods, first proposed by Oliver and 62 63 Pharr (1992, 2004), makes it possible to obtain the local mechanical properties of 64 composite material at micro- and nano-scales. In that framework, the grid-indentation 65 approach can probe constituents' mechanical properties in highly heterogeneous and multi-66 phase materials, such as shales and concretes (Bobko and Ulm 2008; Constantinides et al. 67 2006; Ulm and Abousleiman 2006; Ulm et al. 2007). Furthermore, statistical deconvolution 68 through a multivariate Gaussian Mixture Model (GMM) can be applied to identify the local 69 mechanical properties of individual constituent phases (Liu et al. 2018; Liu et al. 2021; 70 Luo et al. 2021; Luo et al. 2020; Wu et al. 2020; Xu et al. 2020). However, this prior work 71 has applied statistical analyses of the mechanical data without direct reference to the local 72 mineralogy and hence, phases of different mineralogy but similar mechanical response, as 73 well as their mixtures, may manifest as a single microstructural component potentially 74 resulting in bias of the statistical deconvolution analysis. Although further efforts have

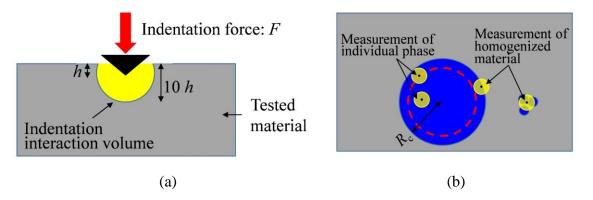
been made to reflect the mineralogy by using electron probe microanalyzer (EPMA) techniques (Abedi et al. 2016; Deirieh et al. 2012; Mashhadian et al. 2018; Veytskin et al. 2017), the direct link between the nanoindentation measurements and local mineralogy to infer mechanical properties of shale mineral phases has not been established to date.

79 Recently, Liu et al. (2022) studied the nano-scale mechanical properties of constituent 80 minerals in shales by combining nanoindentation statistical analyses and SEM-EDS-XRD 81 techniques. The residual impressions of nanoindentation grids were visually checked under 82 SEM/BSE imaging and the local mineralogy under the residual impressions was analyzed 83 using EDS measurements. In this case, the measurements should correspond to mineral 84 phases within the deformed zone, rather than those directly under the indent impression. 85 This is illustrated by considering the stress conditions beneath an indenter (Fig. 1a). 86 Elastoplastic deformations can occur within a highly stressed zone beneath the indenter tip 87 contact that is 10 times greater than the indentation depth (Durst et al. 2005; Larsson et al. 88 1996). Moreover, particle size effects and similarities in the mechanical properties of 89 certain mineral phases can result in bias in evaluating nano-scale mechanical properties.

90 In a shale sample, with a broad range of particle sizes, the process zone can either be 91 contained within a single mineral phase or be influenced by particles of distinct constituent 92 phases (and hence, represent local homogenization of the individual phases), as illustrated 93 in **Fig. 1b**. Moreover, the volume fractions of the constituent phases are closely linked to 94 the bulk mechanical properties of shale (Abedi et al. 2016; Kuila et al. 2014). However, 95 the prior studies generally infer volume fractions of minerals from mass fractions using 96 semi-quantitative XRD methods (with average densities reported independently) (Abedi et 97 al. 2016; Chen et al. 2015).

98 In light of these limitations, we have investigated the nano-scale properties of 99 individual shale phases by combining grid nanoindentation data with high-resolution 100 digital mineralogy maps. The particle size distributions of constituent mineral phases are 101 extracted with an edge detection algorithm to better estimate nano-scale mechanical 102 properties. We show how these nano-scale properties of individual shale phases can be 103 reliably determined and upscaled to estimate the macro-scale homogeneous mechanical 104 properties of the composite material.

105



106 **Fig. 1.** Effects of particle size on indentation measurements. a) Influence zone deformed in an 107 elastoplastic fashion beneath the indenter tip; b) Particles (blue) with characteristic dimension R_c 108 embedded in a matrix (grey). Indents at the edge of large particles or impinging on small inclusions will 109 measure the homogenized response of the two phases.

110

111 **2 Materials and Methods**

112 **2.1 Samples and surface preparation**

113 The samples used in this research are from the Silurian Longmaxi Formation in 114 Sichuan Basin, China. This shale is composed of abundant brittle minerals, mainly 115 feldspar, mica, quartz, and carbonate (average 56.3%) and displays a relatively high 116 porosity ranging from 1.2 - 10.8% (Dai et al. 2014). The shale samples do not show any 117 visible bedding planes. Small cuboidal specimens with dimensions ($2cm \times 2cm \times 1cm$; 118 Length × Width × Thickness) were trimmed from large shale block samples using an 119 IsoMet 1000 precision saw (Buehler Inc., IL, USA). A total of seven trimmed shale 120 specimens (M1 - M7) were used to enable comparisons across specimens from different locations and target test patches. 121

122 Two surface preparation methods were investigated in this study. The first method 123 consists of dry-mechanical polishing (using SiC paper with grit size > 0.25mm) and 124 ultrasonication in n-decane solution; while the second involves Argon milling using a 125 cross-section polisher (JEOL Inc., MA, USA) with a maximum accelerating voltage of 6 126 kV (to minimize heat damage). Appendix A summarizes the specification of the procedures 127 of Argon milling.

128 The surface roughness of the polished specimens was measured using an Atomic 129 Force Microscope (AFM) (Oxford Instruments Asylum Research Inc., CA, USA) with a 130 wave mode or 'tapping' scan (Simpson et al. 1999). The specimens were characterized by 131 a roughness $R_q = 15$ nm and 50 nm for mechanical polishing and Argon milling, respectively. Following prior studies (Bobji and Biswas 1999; Donnelly et al. 2006), which 132 133 involve indentation depth h such that $h > 3R_{q}$, experiments in this study were designed with 134 a target penetration depth of 100 nm to 400 nm. Fig. A2 shows that the polishing method 135 had minimal impact on the engineering properties computed from grid indentation 136 experiments; hence, the data reported in this paper are all based on specimens prepared by 137 the more efficient mechanical polishing method.

138

139

2.2 Grid Nanoindentation Tests

140 Indentation measurements were performed on seven mechanically polished 141 specimens (M1 - M7) of carbonate-rich Longmaxi shale using a NanoTest indenter (Anton Paar, Graz, Austria) equipped with a Berkovich tip (half-angle 70.32° and curvature radius 142 143 of approximately 30 nm). The setup allows applying loads ranging between 0.1 and 500 mN with a resolution of 0.04 μ N. Indentation tests were performed on a square grid ($n^2 =$ 144 21×21) with 10 µm separation (total patch area, $A = 0.04 \text{ mm}^2$) to avoid interactions 145

between adjacent indents (Ulm and Abousleiman 2006). Each test followed a standard loading function to a prescribed maximum force of 4.0 mN, resulting in a maximum indentation depth varying between 100 nm and 400 nm according to the local mineralogy and particle texture. The load was held constant for 2 mins before unloading (with |dF/dt|= 15 mN/min in both loading and unloading). From each indentation test, two quantities were computed. First, the hardness, *H*, is defined as follows:

$$H = \frac{F_{\text{max}}}{A_{\text{c}}} \tag{1}$$

where F_{max} is the maximum indentation load, and A_c is the projected contact area. The contact area function is well-defined for a perfectly sharp Berkovich tip. In practice, however, the Berkovich indenter exhibits some bluntness, which can significantly affect the interpretation of results, especially for shallow indentations. Thus, in this study, A_c is determined by indentation on a material whose mechanical properties are known (e.g., fused silica, which has an indentation modulus M = 72 GPa). A_c is a function of the contact depth, h_c :

160
$$A_{c} = C_{1}h_{c}^{2} + C_{2}h_{c} + C_{3}h_{c}^{1/2} + C_{4}h_{c}^{1/4} + \dots$$
(2)

- 161
- 162

163 $h_{\rm c} / h_{\rm max} = 1 - \varepsilon F_{\rm max} / (Sh_{\rm max})$ (3)

where C_1 is usually determined by the area-to-depth constant of the perfectly sharp indenter (e.g., for a Berkovich indenter, $C_1 = 24.58$) and $\{C_i\}_{i>1}$ are the factors that attribute to the bluntness of the indenter. ε is a constant related to the geometry of the indenter, and *S* is the contact stiffness, which can be calculated by evaluating the derivative at the maximum load:

169
$$S = \frac{\delta F}{\delta h}\Big|_{h_{\text{max}}}$$
(4)

170 Second, Young's modulus, *E*, of the indented bulk material can be obtained from the 171 unload force-displacement curve:

172
$$E = (1 - v^2) / (\frac{1}{M} - \frac{1 - v_i^2}{E_i})$$
(5)

where *v* and v_i are the Poisson's ratio of the test sample and the diamond indenter (0.07), respectively; *M* is the reduced or indentation modulus, which coincides with the plane stress modulus for rigid indentation of an isotropic material; E_i is Young's Modulus of the diamond indenter (1140 GPa).

Poisson's ratio of shales has been measured and reported in the literature. Typical ranges v = 0.1 - 0.3 and 0.15 - 0.35 have been reported in prior studies for Longmaxi shale (Chen et al. 2018; Yang et al. 2016) and Barnett shale (Guo et al. 2013), respectively. Here, we assume v = 0.2 for interpreting E across all samples. Prior work has demonstrated that the uncertainty on v has limited influence on the estimation of E (Oliver and Pharr 1992; Zhang et al. 2009).

183

184 **2.3 Method to Interpret Local Mineralogy**

The test procedure, illustrated in Fig. 2, involves the selection of the area tested by 185 186 grid nanoindentation using SEM imaging, followed by the insertion of electron backscatter 187 diffraction (EBSD) analyzers to generate the BSE image and EDS map with a resolution 188 of 800×1024 pixels (0.24 mm \times 0.3 mm), each pixel corresponding to 300 nm \times 300 nm. 189 The quantitative element information stored in each pixel was then extracted through a 190 Matlab code. After acquiring the distribution of each containing element, the unique 191 elements of each mineral phase are used for mineralogical identification. For instance, 192 silicon, aluminum, and potassium are used to identify illite. If only high-intensity silicon 193 and no other unique elements (i.e., magnesium, iron, sulfur, aluminum, and potassium) are 194 detected in a pixel, this pixel will be identified as the quartz phase. However, if all high-195 intensity silicon, aluminum, and potassium are detected in a pixel, this pixel will be 196 identified as an illite phase. Carbon and oxygen are not used as unique elements to analyze the mineral phases due to their wide existence in most of the constituent phases in shale. Based on the elemental chemical composition in each pixel, the in-situ mineralogy can be uniquely identified pixel by pixel. Finally, a 'digital mineralogy map' was obtained that determines the edges of particles, the area fraction, and the spatial distribution of each component phase. The results of this approach are illustrated in **Fig. 3**. They closely match the microstructure observed in the BSE image of the same region, with six pure mineral phases (illite, quartz, calcite, dolomite, pyrite, and organics) and up to six mixed phases.

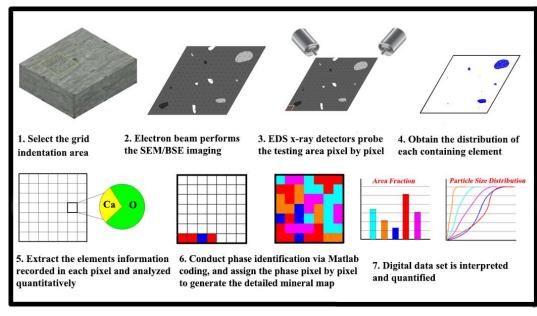
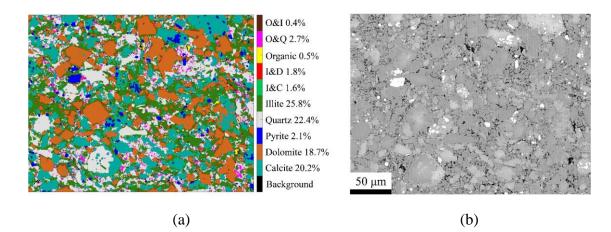


Fig. 2. Procedure for in-situ mineralogical identification.

207



208 Fig. 3. a) 'Digital mineralogy map' generated by the in-situ mineralogical identification technique and

b) BSE image of the same region of interest (ROI). The labels 'C', 'D', 'Q', 'I', and 'O' represent calcite,
dolomite, quartz, illite, and organic, respectively. The particles of mineral phases, which appear with
different gray scales in the BSE image, agree with the identification results in the 'digital mineralogy
map'.

213

214 **2.4 Matching of Mineralogy and Indenter Footprint**

215 In order to assign the nano-scale mechanical properties inferred from the 216 nanoindentation grid to their corresponding mineral phases, it is essential to obtain 217 precisely the spatial information of the nanoindentation grid. However, due to the size of 218 each indentation footprint and the lack of contrast between the footprint and the tested shale 219 surface, it is impossible to distinguish individual indents while visualizing the entire 220 nanoindentation grid in the region of interest. Thus, in this study, BSE scanning was used 221 to locate the four corners of the grid (from the indentation footprints) and adjacent key 222 morphological features, as illustrated in Fig. 4a. Then, the individual indents were 223 determined assuming a regular pattern for the nanoindentation grid (rows and columns are 224 perpendicular with each other), as illustrated in Fig. 4b.

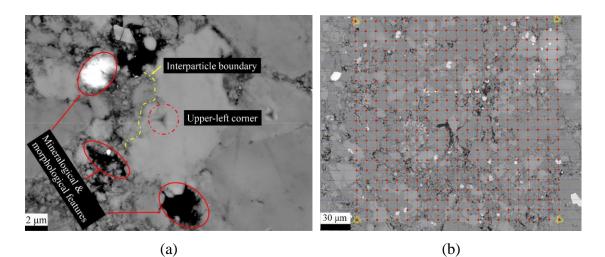


Fig. 4. Determination of nanoindentation grid coordinates. (a) Mineralogical and morphological features
 in the vicinity of the top-left corner point of the indentation grid observed by high-magnification BSE

imaging; (b) indentation grid overlaying the BSE image of the ROI. The grid is positioned assuming aperfectly squared grid based on the knowledge of the corner coordinates.

230

231 The projected area of each indent was calculated according to the measured 232 penetration depth (Eq. 2) and superimposed on the pixel-scale digital mineralogy map (Fig. 233 5). The constituent mineral phases were assigned to the pixels included in the indentation 234 interaction volume based on the spatial information of the mineral phases in the in-situ 235 'digital mineralogy map' and the coordinates of the indentation grid. The related mineral 236 phase was then classified either as a single constituent phase (> 90% of the pixels display 237 the same mineralogy) or a mixed phase based on the pixel ratio, which defines the number 238 of pixels displaying the same mineralogy in the indentation interaction volume, $N_{\rm m}$, to the 239 total number of the pixels included in the interaction volume, $N_{\rm t}$, i.e., $N_{\rm m}/N_{\rm t}$. We adopt a 240 conservative rule of thumb, if the proportion of other minor identified constituents is lower 241 than 10%, the measured phase is considered a pure mineral phase (e.g., calcite phase). The 242 measured phase is classified as a mixture phase if two or more phases, each representing 243 more than 10%, are present.

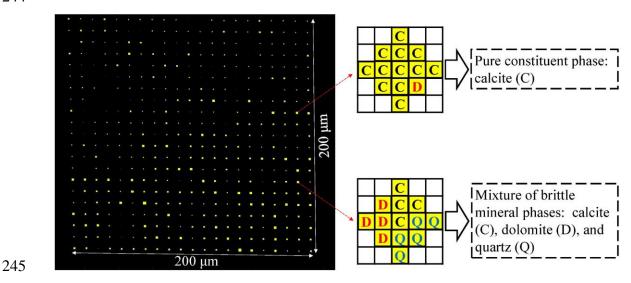




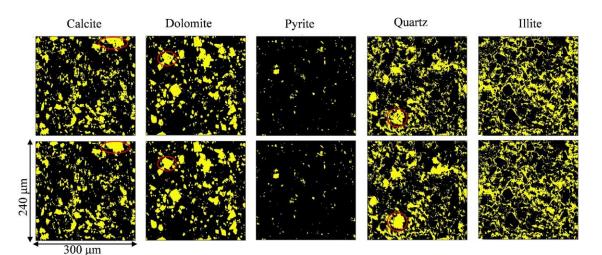
Fig. 5. Mechanical and mineralogical information matching.

247

248 **2.5 Particle Size Distribution of mineral phases**

249 Particle size distributions of shale mineral phases are quantified by processing the 250 digital mineralogy map (Fig. 3). An edge detection algorithm, which has been widely used 251 for quantitative analysis of particle and pore size in a porous medium such as soil and rock 252 (Jiu et al. 2018; Jiu et al. 2020; Wu and Yu 2012; Zhou et al. 2018), is applied using a 253 customized Matlab code. Small holes inside the particles of mineral phases (see, e.g., red 254 circles in **Fig. 6**) are filled first. Then the boundary of each particle is identified by the 255 connection algorithm, which labels connected components in 2D binary image before 256 counting and recording the number of pixels occupied by each particle. Finally, based on 257 the relationship between the actual length and pixel size, the true area of each particle is 258 calculated, which allows us to compute the cumulative distribution of particle area for each 259 mineral phase. It should be noted that particles or pores with a size smaller than the pixel 260 size $(300 \text{ nm} \times 300 \text{ nm})$ cannot be detected in this analysis.





- Fig. 6. Procedures for particle size calculation. The first row shows the distribution map of each mineral
 phase identified by the in-situ mineralogical identification. Small holes inside each isolated particle (see
 red circled) are filled before counting and recording the number of pixels occupied by each particle (the
 second row).
- 267

268 **3 Results**

3.1 Mineralogy and Particle Size Distribution

270 Fig. 7 shows two high-resolution digital maps of the mineralogy of Longmaxi shale 271 samples (specimens M3 and M4). These results highlight the complex microstructure of 272 shale, which is characterized principally by five mineral phases, i.e., quartz, calcite, 273 dolomite, illite, and pyrite. These phases constitute about 90% of the surface area, among 274 which pyrite and organics display much smaller fractions. Since each pixel in the digital 275 mineralogy map corresponds to an area of 300 nm \times 300 nm, up to 6 mixed phases (with a 276 total area fraction of about 5.5%) were detected at the junctions between different mineral 277 phases. These mixed phases were further classified into three categories: 1) brittle mixtures 278 (any combination of calcite, C, dolomite, D, quartz, Q and pyrite, P); 2) ductile mixtures 279 (combination of organic, O, and illite, I); and 3) composite mixtures (combinations of 280 minerals from both brittle and ductile groups). Specifically, the C&Q and D&Q phases 281 were classified as brittle mixtures, whereas the I&O phase was classified as a ductile 282 mixture. Finally, the I&C, I&D, and I&Q phases are classified as composite mixtures. 283 **Table 1** summarizes the area fractions of the pure and mixed phases (found in 8 regions of 284 interest) for the seven specimens of the carbonate-rich Longmaxi shale. The measured area 285 fraction of each mineral phase exhibited a relatively small standard deviation (of about 0.5 286 ~ 5%) over the ROI (240 μ m × 300 μ m).

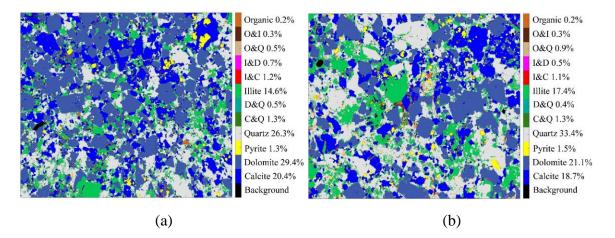


Fig. 7. Quantitative mineralogical maps (240 μ m × 300 μ m) for samples M3 (a) and M4 (b) obtained

289 by in-situ mineralogical identification.

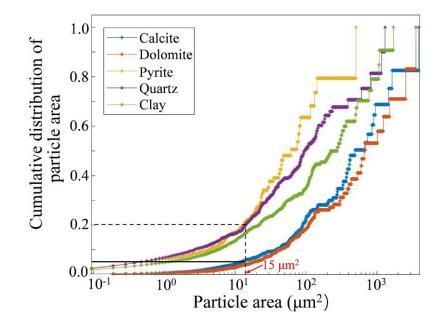
290 Table 1

Sample ID	Calcite (%)	Dolomite (%)	Pyrite (%)	Quartz (%)	Illite (%)	Organic (%)	C&Q (%)	D&Q (%)	I&C (%)	I&D (%)	O&Q (%)	O&I (%)
M1	20.2	18.7	2.1	22.4	25.8	0.5			1.6	1.8	2.7	0.4
M2	17.5	14.1	1.8	36.2	21.5	0.3	0.6		1.7	0.8	2.0	1.2
M3	20.4	29.0	1.3	26.3	14.6	0.2	0.2	0.5	1.2	0.7	0.3	0.3
M4	18.7	21.1	1.5	33.4	17.4	0.2	1.3	0.4	1.1	0.5	0.9	0.3
M5	25.8	26.1	2.1	14.6	20.5	0.06	1.4	0.6	2.1	1.5	0.6	
M6	23.5	24.5	1.8	19.4	22.6	0.3	0.8		1.2	0.7	0.2	0.5
M7	19.4	22.4	2.3	23.9	23.1	0.1		0.6	0.5	1.0	1.5	0.8
Mean	20.7±2.7	22.3±4.6	1.8±0.3	25.1±7.0	20.7±3.4	0.2±0.1	0.9±0.4	0.5 ± 0.08	1.3±0.5	1.0±0.4	1.2±0.9	0.6±0.3

291 Area fractions of mineral phases in carbonate-rich Longmaxi shale.

Note: Results were obtained from seven specimens of carbonate-rich Longmaxi shale. The labels 'C', 'D', 'Q', 'I', and 'O' represent calcite, dolomite, quartz, illite, and organic, respectively. The C&Q and D&Q phases are classified as brittle mixtures, whereas the I&O phase is classified as a ductile mixture. The I&C, I&D, and I&Q phases are classified as composite mixtures.

296 The digital mineralogy maps also illustrate the complex microstructure of the 297 Longmaxi shale, with a wide range of particle sizes for each mineral phase. Fig. 8 shows a 298 representative result of the studied shale's cumulative particle area distribution functions 299 (measured on specimen M5). This data reveals that the carbonate minerals (calcite and 300 dolomite) were associated with larger particles than quartz, illite, and pyrite. Given that the 301 maximum penetration depth for the grid indentation tests ranged from 100 to 400 nm; 302 hence, the projected area of an indentation interaction volume ranged from 5 to 15 μ m². 303 Therefore, **Fig. 8** shows that the indentation footprint can exceed up to 20% of the quartz, 304 illite, and pyrite particles but only 5% for calcite and dolomite. The high proportion of fine 305 particles ($\leq 15 \ \mu m^2$) increases the probability that a mixture of mineral phases will affect 306 the indentation measurements and corresponding engineering properties. More detailed 307 observations from the SEM/BSE images showed two distinct morphologies of quartz, illite, 308 and pyrite minerals within the Longmaxi shale: 1) larger intact particles and 2) porous 309 aggregates of fine grains with nano-scale interparticle pores. It should be noted that 310 particles or pores with a size smaller than the pixel size of the digital mineralogy map (300 311 $nm \times 300$ nm) cannot be detected in this analysis. Thus, as a result, the curve of particle 312 area distribution for illite includes both the illite particles and the nano-porous aggregates 313 of illite.



315

Fig. 8. Cumulative particle area distribution functions of calcite, dolomite, quartz, and pyrite phases measured in sample M5. The projected area of an indentation interaction volume can be up to $15 \,\mu\text{m}^2$, corresponding to ~20% for quartz, pyrite, and illite phases but only ~5% for calcite and dolomite.

Fig. 9a illustrates the co-existence of these morphologies for quartz, for which there was a significant contrast in individual particle size. These observations were consistent with results reported by Dong et al. (2019) and Xu et al. (2021), who also identified different forms of quartz in the Longmaxi shale. Similarly, the Longmaxi shale specimens included isolated larger pyrite particles and framboidal structures (**Fig. 9b**) that have been reported for other shales (e.g., Opalinus (Seiphoori et al. 2017)). In contrast, carbonate minerals were mostly found as isolated particles.

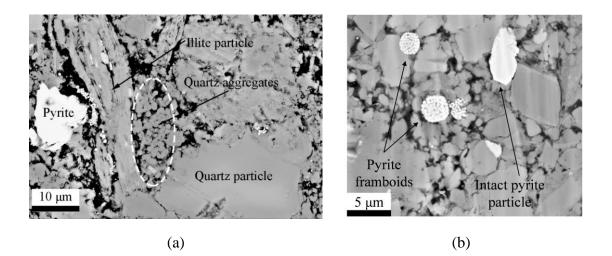


Fig. 9. Representative BSE image showing granular aggregates and isolated particles of mineral phasesin shale: (a) quartz and (b) pyrite.

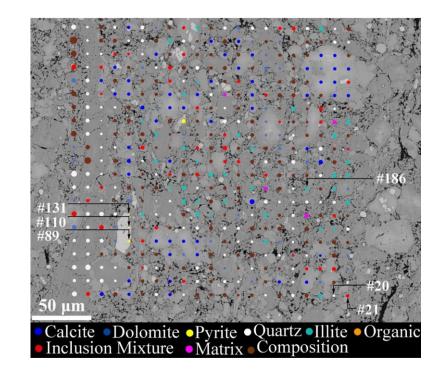
329

330 3.2 Mechanical Properties of Mineral Phases from Grid Nanoindentation

331 Fig. 10 shows the local mineral phases at the locus of each indentation superimposed 332 with the BSE image. Here, nine distinct mineral phases are identified, including six pure 333 minerals (cf. Table 1 including organics), and three mixtures (i.e., brittle, ductile, and 334 composite). Notably, although mixed phases characterize only 5.5% of the ROI surface area at the pixel scale (Fig. 7), 30 - 40% of the indentation footprints were classified as 335 336 mixtures. This is due to the greater size of the indentation interaction volume (up to $1\mu m \times$ 1 μ m) when compared to the pixel size (300 nm × 300 nm), which in turn resulted in a 337 338 higher probability of measuring a mixture. Figs 11a and 11b illustrate in more detail the 339 accuracy of the indentation-mineral map by interpreting the indentation footprints and 340 mineralogy of the randomly selected indent locations (as shown in Fig. 10). These 341 examples show that the proposed coupled analysis of grid-nanoindentation and in-situ 342 mineralogical identification provides a precise and reliable toolkit for identifying nano-343 scale mechanical mineralogical information in a complex composite material such as shale.

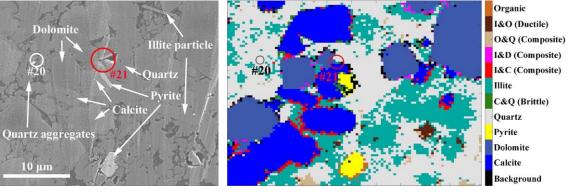
All the indentation data from grid indentation experiments on specimens M3 - M7 (in total $441 \times 5 = 2205$ indentations) were integrated. These data provided sufficient information for statistical analyses of each phase's mechanical properties and area
fractions. Area fractions of the five pure mineral phases obtained from the statistical
analysis were consistent with those measured from the in-situ mineralogical identification
(Table 1).

350



351

Fig. 10. Indentation-mineral map (sample M2) showing local mineral phases at the locus of each
 indentation superimposed with the BSE image. Indentation points are randomly selected to verify the
 accuracy of the measured indentation-mineral map.



(a)

(b)

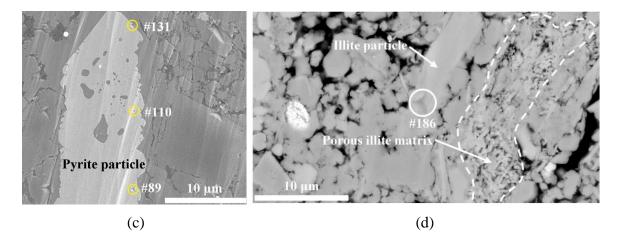
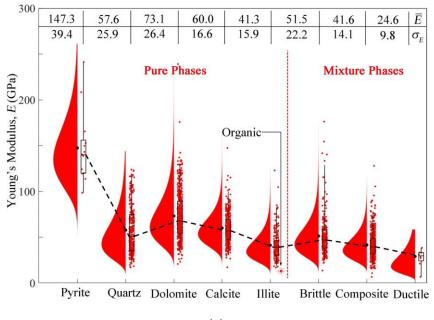
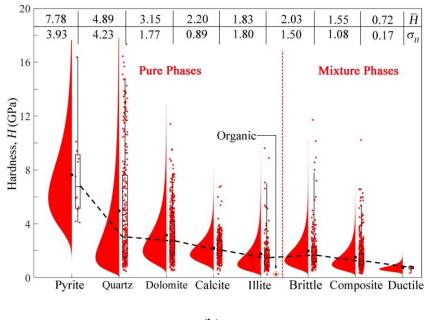


Fig. 11. Verification of the indentation-mineral map: a) Indentation footprints on a quartz phase (indent #20) and the brittle mineral mixtures (indent #21). b) Local mineralogy of indents #20 and #21. c) Indentation footprints on a single pyrite particle (indents #89, #110, #131), and d) indentation footprint on an illite particle (indent #186). These observations result from high-magnification SEM imaging and BSE imaging, respectively. The residual footprints of the selected points are circled in the SEM and BSE images.



(a)



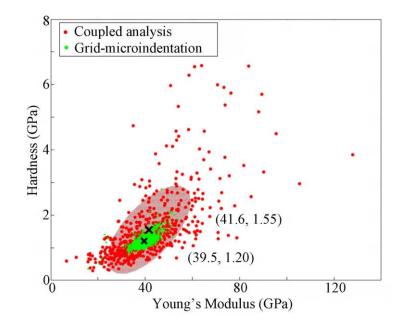
(b)

Fig. 12. Mechanical properties of the mineral phases from grid nanoindentation of M1-M7 specimens: a) Young's modulus and b) hardness. The data are presented using a raincloud format that combines conventional boxplots with median, interquartile range (IQR) box, and outlier whiskers (upper quartile $+ 1.5 \times IQR$ and lower quartile $- 1.5 \times IQR$); together with interpretations assuming a single lognormal distribution function for each phase (expectation and standard deviation properties are reported in the inset in each figure). The black dots at the bottom of each distribution show the positions of the expectation (\overline{E} and \overline{H}).

368

369 Figs. 12a, 12b show raincloud plots (Allen et al. 2019) of Young's modulus, E, and 370 hardness, H, computed from grid nanoindentation on the Longmaxi shale for five pure 371 mineral phases and three mixed phases. These figures combine boxplots of individual 372 indents with an interpretation assuming that these properties can be described by a single 373 lognormal distribution for each classified phase. The data was relatively sparse for the 374 organic phase, and the properties of the organics were reported using a mean value (E =375 8.2GPa and H = 0.95 GPa), which is consistent with results reported in the literature by 376 Eliyahu et al. (2015), Li et al. (2018), and Shukla et al. (2013). In addition, the boxplot data 377 for quartz and illite show significant skewness (particularly for H, Fig. 12b), which 378 indicates that the structures of the mechanical data of quartz and pyrite phases are not 379 monophasic. The skewness confirms prior observations that particle morphology (i.e., 380 large intact particles vs. nano-porous aggregates of quartz and illite particles) affects the 381 results. The statistical mean of Young's modulus of clay phase shown in Fig. 12b (i.e., 382 41.3 GPa) is greater than the results from some other prior studies (Bobko and Ulm 2008; 383 Liu et al. 2018). This is because the indentation data of the clay phase obtained from the 384 coupling analysis incorporate the measurement of both the intact clay particles and the 385 nano-porous clay aggregates. However, in prior studies, the indentation measurement 386 mainly focused on the porous clay composite, which has a much lower modulus and 387 hardness than those of the intact illite particles. A comparison between Young's modulus 388 and hardness of the five main mineral phases measured in this study and those reported in 389 prior publications are summarized in Table 2.

390 The data of the composite mixture phase obtained from the above coupled analysis can 391 be compared to measurements performed at a mesoscopic scale, beyond which the 392 mechanical response is homogeneous. Fig. 13 compares Young's modulus and hardness 393 values of the composite mixture phase with microindentation data (penetration depths 8 -394 10 µm) reported for the very same specimens of Longmaxi shale (Du et al. 2021). While 395 there is a much larger scatter in properties nanoindentation determined at the nanoscale 396 (reflecting the different phase mixtures), the two data sets show very similar mean values 397 of the two-dimensional Gaussian distribution. This result indicates that the homogenized 398 mechanical properties of complex composite materials such as shale can be assessed by its 399 nano/microscale representative unit composed of both the brittle and ductile mineral 400 phases.



402

403 **Fig. 13.** Comparison between the mechanical properties (*E* and *H*) of the composite phase extracted 404 from the coupled analysis of nanoindentation (h = 100 - 400 nm; red dots) and the homogenous 405 mechanical properties measured by grid microindentation with characteristic indentation depth h = 8-406 10 µm (green dots) (Du et al. 2021). Values in the brackets show the statistical mean of Young's modulus 407 and hardness. The shaded ellipsoid is drawn based on the statistical confidence level of 95% and the 408 covariance matrix, which determines the length of the major and minor axes, and the orientation of the 409 axes, respectively.

410

411 **3.3 Effects of Particle Size on Mechanical Properties**

412 Determining the 'characteristic' lengths of the constituent mineral phases is essential 413 for extracting their mechanical properties from the grid nanoindentation technique. By 414 assuming the particles for each phase are spherical with a diameter, D, and that the 415 indentation interaction volume is five times greater than the indentation depth h, Ulm and 416 Abousleiman (2006) proposed that a ratio h/D < 1/10, will allow a reliable interpretation 417 of mechanical properties for the constituent phases. However, this assumption does not 418 consider indentation footprints closer to the edge of a particle, where a mixture of phases 419 influences the mechanical response. Furthermore, the probability of indenting at the center 420 of a spherical particle is negligible, and hence, the characteristic length is generally421 underestimated.

422 In this study, the critical characteristic lengths of the constituent mineral phases are 423 determined based on a probabilistic analysis. Considering a nanoindentation grid randomly 424 performed on a ROI, the mechanical properties of a given constituent mineral phase require 425 an adequate surface area for the indentation footprint within a single particle. The footprint 426 should be located at a minimum distance of 3h from the particle boundary to avoid the 427 influence of adjacent phases or pores. As a result, compared to the maximum indentation 428 depth, the particle area should be large enough to ensure that the probability of the 429 indentation loci within the applicable area, η , obeys the conservative rule of thumb, i.e., η 430 = 90%. Such a rule makes it possible to extract the mechanical properties of a specific 431 constituent phase by statistical analysis of all the particles of that phase. Therefore, the 432 critical characteristic length for measuring the mechanical properties of the constituent 433 phases can be determined as follows:

434

$$\pi (R_{\rm c} - 3h)^2 / \pi R_{\rm c}^2 = 0.9 \tag{6}$$

where R_c is the particle radius. In this study, maximum indentation depths range from 100 to 200 nm for the pure mineral phases (i.e., calcite, dolomite, quartz, pyrite, and illite). Therefore, the characteristic length and area of the pure mineral phases are about 5.8 - 11.7 μ m and 100 - 430 μ m², respectively, i.e., up to 10 times greater than that proposed in prior research.

The skewed distributions of mechanical properties for quartz and illite phases in Fig. 12 suggest that the pure mineral phases must be partitioned to reflect two or more particle structures. We have investigated particle size effects for quartz and illite phases based on detailed observations of their particle morphology. For example, Figs. 14a and 14b compare nanoindentation values of E and H for nano-porous illite aggregate with those of a much larger intact particle. The intact illite phase has a much higher modulus and hardness than those of the nano-porous phase. 447

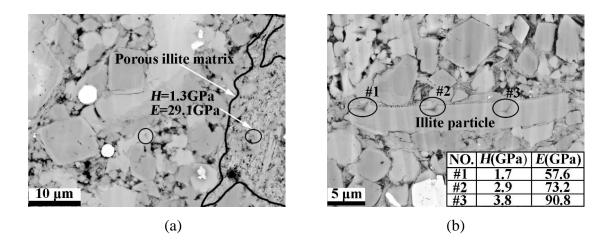


Fig. 14. BSE images of illite phase in Longmaxi shale samples. a) Nano-porous illite matrix with much
lower hardness and Young's modulus than b) Intact illite particle. The residual indentation footprints
are circled in the images.

451

The Bayesian Information Criterion (BIC) and Multidimensional Gaussian Mixture
Model (GMM) (Abedi et al. 2016; Deirieh et al. 2012) were used to identify and classify
morphological components for illite and quartz:

455

$$BIC = -2\log\hat{L}(\theta) + n\log(N) \tag{7}$$

$$j = 1, n; D\left(X_{i}; \mu_{j}^{X}, s_{j}^{X}\right) = \frac{1}{s_{j}^{X}\sqrt{2\pi}} \int_{-\infty}^{X_{i}} \exp\left(\frac{-(u - \mu_{j}^{X})^{2}}{2(s_{j}^{X})^{2}}\right) du$$
(8)

457 where $\log \hat{L}(\theta)$ denotes the value of the maximized loglikelihood objective function, *n* is 458 the number of components, and *N* is the number of data points, X_i is the *i*th mechanical 459 data, μ is the statistic mean, and *s* is the covariance matrix, which is a square matrix giving 460 the covariance between each pair of elements of a given vector.

BIC analyses are performed 500 times for each preset component number n (from 2 to 10) and each type of covariance matrix (diagonal or full; shared or unshared) to avoid a random error. The covariance matrix determines the shape of the confidence ellipsoid drawn over each cluster. In this study, we investigated whether the covariance matrices are diagonal or full, and whether all components have the same covariance matrix (shared or unshared). **Figs. 15a** and **15b** summarize the BIC analysis for quartz and illite phases, 467 respectively, indicating that a two-component mixture (n = 2) with a full-unshared 468 covariance matrix provides the best representation. The modulus and hardness of these 469 intact particles and nano-porous aggregates are then extracted by GMM clustering, as 470 shown in **Figs. 15c** and **15d**. The shaded ellipsoids are generated based on the statistical 471 confidence level of 95% and the covariance matrix, which determines the length of the 472 major and minor axes, and the orientation of the axes, respectively.

473

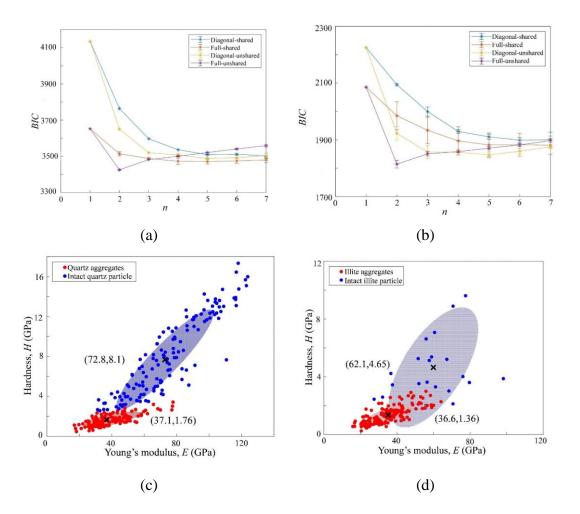


Fig. 15. Refined mechanical properties obtained by BIC and GMM analysis for quartz and illite. a) and b) Optimal components (n=2) based on BIC analysis for quartz and illite, respectively; c) and d) GMM analyses used to establish statistical mean values of Young's modulus and hardness for the intact and nano-porous aggregate components of quartz and illite, respectively.

479 To further address the effect of particle size on the mechanical properties, the 480 measured Young's modulus and hardness of a pure mineral phase were linked to the 481 particle sizes obtained from the edge detection algorithm. In all five pure mineral phases 482 in the Longmaxi shale (i.e., excluding organics), Young's modulus and hardness initially 483 increased with the particle area and then were relatively constant after the particle area 484 exceeded a specific threshold value. Young's modulus and hardness of intact mineral 485 particles were thus obtained (Table 2). Compared to the results from the coupled analysis 486 only (Fig. 12), with these considerations of particle size effects, the mechanical properties 487 of the constituent phases were further refined and show a better agreement with the values 488 for the same mineral phases reported in the literature, as shown in **Table 2**.

489 **Table 2**

Mineral Phase	-	analysis . 12)	Refinement by BIC & GMM (Fig. 15)		Par	for Intact ticles hold size)	Published Results			
	E (GPa)	H (GPa)	E (GPa)	H (GPa)	E (GPa)	H (GPa)	E (GPa)	H (GPa)	Source	
Calcite	60.0	2.20	64.3	2.4	61.2	2.3	73-83	2.1-2.4	(1)	
Dolomite	57.6	3.15	85.9	4.9	93.1	4.7	53-82	3.5-7.5	(2)	
Quartz	73.1	4.89	72.8	8.10	69.3	6.9	82-96	13.3-14.8	(1)	
Pyrite	147.3	7.78			226.6	13.6	264-330	16-27	(3)-(6)	
Illite	41.3	1.83	62.1	4.65	71.1	3.9	62		(7)	

490 Summary of Young's modulus and hardness of the 5 main mineral phases in Longmaxi shale samples obtained in this study and prior publications

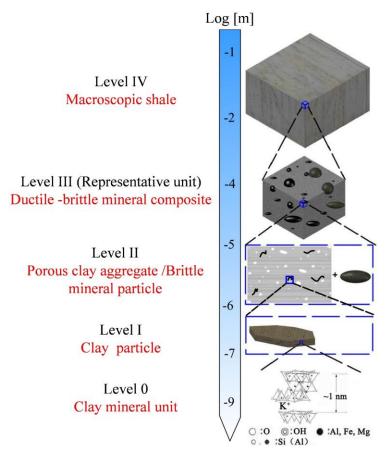
491 Note: (1) Broz et al. (2006); (2) Viktorov et al. (2014); (3) Li et al. (2022); (4) Kumar et al. (2012); (5) Aguilar-Santillan (2008); (6) Loberg et al.

492 (1985); (7) Xian et al. (2019)

494 **4 Discussion**

Shales are composite materials with high heterogeneity and complex multi-phase
characteristics. Fig. 16 presents a conceptual 5-level (multi-scale) representation of shale
extending from the nanoscale clay unit to the centimeter-scale macroscopic shale.

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Fig. 16. Multi-scale model of shale structure.

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Level 0 corresponds to the elementary unit of clay minerals at the nanometer scale (0.7 nm and 1.0 nm in thickness for the 1:1 and 2:1 phyllosilicate clay minerals, respectively (Zhang et al. 2010)). Level I corresponds to the aggregated clay, which typically comprises assemblies of particles in face-face configurations: illite aggregates 506 generally are up to 10 nm thick with surface dimensions ~500 - 1500 nm (Nadeau 1985). 507 Level II is the scale of the ductile porous clay/organic aggregate and the brittle mineral 508 particles. Level III corresponds to the representative unit of the macroscopic shale at the 509 submillimeter scale, including both the ductile and brittle minerals. Thus, this scale is 510 considered the elementary building block of shale. Level IV is the scale of macroscopic 511 shale, which is the object of conventional, macroscopically-homogeneous laboratory 512 studies. Precise prediction and characterization of the multi-scale mechanical properties of 513 shale are of great scientific importance and engineering significance.

514 Upscaling of Level II to III has been widely carried out in prior studies using self-515 consistent and Mori-Tanaka approaches (Liu et al. 2018; Liu et al. 2022; Luo et al. 2021). 516 In this study, a multi-scale model for upscaling mineral particles (Level I) to their nano-517 porous aggregates (Level II) was analyzed for the first time based on a self-consistent 518 scheme for granular morphology, accounting for the nanoscale porosity of shale materials 519 (Seiphoori et al. 2017). Young's moduli of Longmaxi shale samples at Levels I and II are 520 obtained by the coupled analysis described in this study. These experimental data can 521 provide both the input parameters and the experimental validation for the multi-scale 522 mechanical model.

523 In the case of linear, isotropic, and elastic behavior of the mineral phases, a self-524 consistent model (Dormieux et al. 2006; Zaoui 2002) was established for homogenizing 525 the stiffness of a mixture with granular morphology:

526
$$K_{\text{hom}} = \sum_{i} \varphi_{i} K_{i} (1 + \alpha_{0} (\frac{K_{i}}{K_{\text{hom}}} - 1))^{-1} \left[\sum_{i} \varphi_{i} (1 + \alpha_{0} (\frac{K_{i}}{K_{\text{hom}}} - 1))^{-1} \right]^{-1}$$
(9)

527
$$G_{\text{hom}} = \sum_{i} \varphi_{i} G_{i} (1 + \beta_{0} (\frac{G_{i}}{G_{\text{hom}}} - 1))^{-1} \left[\sum_{i} \varphi_{i} (1 + \beta_{0} (\frac{G_{i}}{G_{\text{hom}}} - 1))^{-1} \right]^{-1}$$
(10)

528 where K_{hom} and G_{hom} are the homogenized bulk modulus and shear modulus, 529 respectively; K_i , G_i and φ_i are the bulk modulus, shear modulus, and volume fraction of 530 the *i*-th phase, respectively. Finally, α_0 and β_0 are defined by $\alpha_0 = \frac{3K_{\text{hom}}}{3K_{\text{hom}} + 4G_{\text{hom}}}$, and

531
$$\beta_0 = \frac{6(K_{\text{hom}} + 2G_{\text{hom}})}{5(3K_{\text{hom}} + 4G_{\text{hom}})}.$$

532 The upscaling of strength properties requires the use of non-linear homogenization 533 theory (Bobko et al. 2011):

534

$$H_{\text{hom}} = h_s \times \Pi_H(\eta, \mu) \tag{11}$$

where Π_H is a dimensionless function to be developed, H_{hom} and h_s are the homogenized hardness and the particle hardness, respectively, and η and μ are the solid packing density and friction coefficient, respectively. Based on data analysis, Gathier (2008) obtained the scaling relations between the particle hardness, h_s , and the homogenized hardness, H_{hom} , for a porous composite exhibiting self-consistent morphology (after Bobko et al. (2011)):

541
$$\Pi_{H}(\eta,\mu) = \Pi_{1}(\eta) + \mu(1-\eta)\Pi_{2}(\eta,\mu)$$
(12)

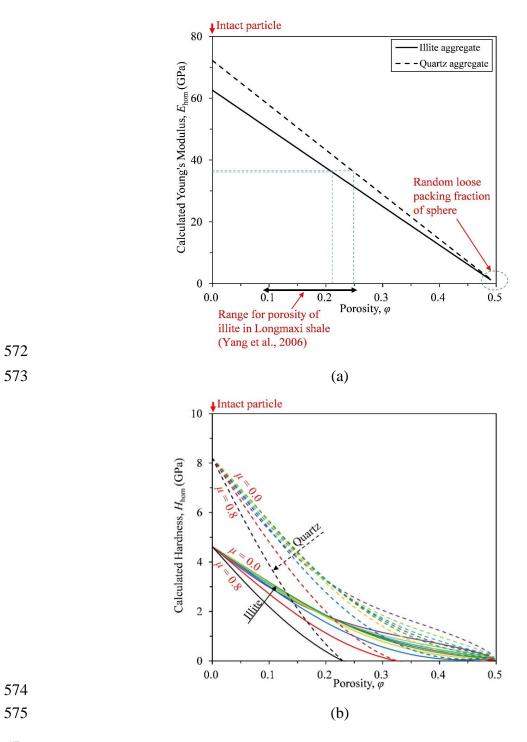
542
$$\Pi_{\rm I}(\eta) = \frac{\sqrt{2(2\eta-1)} - (2\eta-1)}{\sqrt{2} - 1} \left[1 - 5.3678(1-\eta) + 12.1933(1-\eta)^2 - 10.3071(1-\eta)^3\right]$$
(13)

543
$$\Pi_{\rm II}(\eta,\mu) = \frac{2\eta - 1}{2} [6.7374 - 39.5893(1 - \eta) + 34.3216(1 - \eta)\mu - 21.2053\mu^3]$$
(14)

The quartz and illite phases of Longmaxi shale include intact particles and nanoporous aggregates (level II; **Fig. 15**) that can be considered as a granular material and upscaled using self-consistent assumptions. Young's modulus and hardness of the intact particles are determined by the GMM fitting of the grid nanoindentation data. However, the packing density/porosity of the nano-porous aggregates ($\eta = 1-\varphi$) is not measured directly in this study.

Fig. 17a shows the relationships of the homogenized stiffness [Eqs. (9), and (10)] as a function of the stiffness of intact particles (i.e., at $\varphi = 0$) and the porosity of the nanoaggregates for illite and quartz phases. These results show that the self-consistent formulation explains the difference in mean values of *E* for intact and nano-porous aggregates with $\varphi = 0.21$, 0.25 for illite and quartz, respectively. The estimate of nano555 porosity for illite is also within the range reported from nitrogen adsorption measurements 556 on Longmaxi shale by Yang et al. (2016). The increased porosity results in lower 557 homogenized stiffness and hardness for both the illite and quartz aggregates. As the 558 porosity equals zero, the composite material becomes an intact particle. Thus, the 559 calculated stiffness and hardness [E, H] reflect the properties of the intact illite and quartz 560 particles, i.e., [62.1 GPa/4.65 GPa] and [72.8 GPa/8.10 GPa], respectively.

561 Fig. 17b shows the homogenized hardness for the nano-porous quartz and illite aggregates [based on Eqs. (9) - (11)] as functions of intact particle hardness, nano-porosity, 562 and selected values of the friction coefficient μ . There have been extensive studies of the 563 564 inter-particle friction for quartz (e.g., summarized in Terzaghi et al., 1996), suggesting an 565 average value $\mu = 0.4$, and thus an estimated homogenized hardness, H = 2.4 GPa. This 566 value is higher than the mean (1.8 GPa) reported in Fig. 15a but is within the range 567 measured for nano-porous quartz in Longmaxi shale. Similarly, the interparticle friction of 568 illite has been reported in the range $\mu = 0.40 \pm 0.04$, corresponding to a homogenized 569 hardness, $H = (1.7 \pm 0.02)$ GPa from the self-consistent analyses (Fig. 17b), which is within 570 the range identified by GMM analyses of nano-porous illite (Fig. 15b).



576 Fig. 17. Self-consistent homogenized a) Young's modulus, *E*; and b) hardness, *H*, for nano-porous 577 quartz and illite aggregates as functions of assumed nano-porosity, φ . *H* values are shown at selected 578 friction ratios, $\mu = 0.0 - 0.8$ (0.1 interval).

580 5 Conclusions

In this study, we developed a coupled analysis of grid nanoindentation and in-situ mineralogical identification considering the particle size effect to access the mechanical properties of the constituent mineral phases. Furthermore, multi-scale mechanical models for upscaling the elastic stiffness and strength properties were established, considering the self-consistent scheme for granular morphology to link the microscopic characteristics with the multi-scale mechanical properties of shales. The key conclusions of this study are as follows:

 By combining digital images of mineral distribution and control coordinates for grid nanoindentation, Mechanical properties (hardness and Young's modulus) for five pure mineral phases (i.e., calcite, dolomite, quartz, pyrite, and illite) and three mineral mixture phases (i.e., brittle, ductile, and composite mixtures) are extracted for specimens of carbonate-rich Longmaxi shale.

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2. Threshold particle sizes were identified for estimating intact mineral properties
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5983. The mechanical properties of shale mineral phases are influenced by particle size599and morphology, particularly for quartz and illite phases. These phases are further600partitioned (through statistical BIC and two-dimensional GMM analyses) to601distinguish between the mechanical properties of individual particles and nano-602porous aggregates.

A multi-scale model for upscaling from mineral particles (Level I) to their nano porous aggregates (Level II) is analyzed for the first time based on a self consistent scheme for granular morphology. The self-consistent modeling

606approach can explain the measured differences in elastic stiffness and hardness607properties of the mineral aggregates and their intact particles. This provides a clear608basis for upscaling properties from the microscopic characteristics (i.e., the609volume fractions and the mechanical properties of the intact mineral particles)610measured by the proposed coupling analysis.

611

612 **CRediT authorship contribution statement**

Jianting Du: Formal analysis, Investigation, Methodology, Writing – original draft.
Andrew J. Whittle: Conceptualization, Methodology, Funding acquisition, Project
administration, Supervision, Writing – review & editing. Liming Hu: Conceptualization,
Funding acquisition, Project administration, Supervision, Writing – review & editing.
Thibaut Divoux: Investigation, Project administration, Writing – review & editing. Jay
N. Meegoda: Supervision, Writing – review & editing.

619

620 Declaration of Competing Interest

621 The authors declare that they have no known competing financial interests or personal622 relationships that could have appeared to influence the work reported in this paper.

623

624 Data Availability

625 Data will be made available on request.

626

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- 634

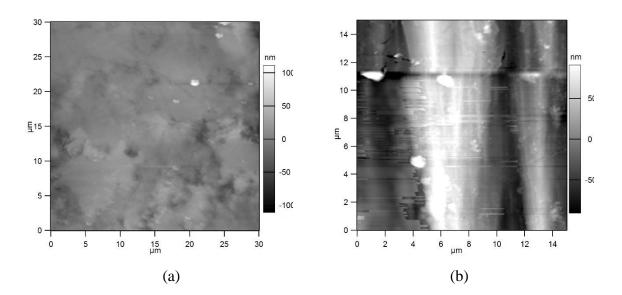
635 Appendix A. Surface preparation

636 Argon Milling

The trimmed shale samples are polished with the coarse abrasive paper into 1 cm \times 1 cm \times 0.5 cm thick blocks as required by the JEOL cross-section polisher. The maximum accelerating voltage of 6 kV is applied to minimize the heat damage, and the milling duration is set to 10 -12 h. Typically, 50 - 100 µm deep materials are removed from the surface to ensure that the milled surface is free of mechanical polishing artifacts as well as smearing due to razor blade cutting.

643 AFM roughness characterization of shale samples prepared by mechanical polishing 644 and Argon milling (**Fig. A1**) shows RMS roughness, $R_q = 15$ nm and 50 nm, respectively. 645 Trenching is a usually unwanted side effect of ion beam milling (Youngner and Haynes 646 1982), which may increase the roughness of a milled surface.

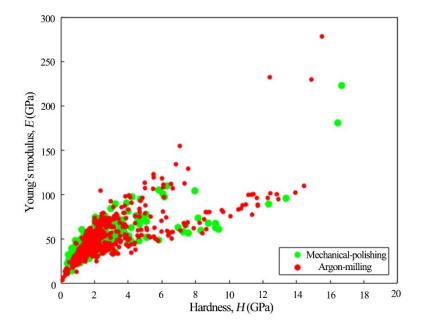
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648 Fig. A1. Representative AFM maps of interested regions prepared by mechanical polishing (a) and 649 Argon milling (b). The dimensions of AFM images are 30 μ m by 30 μ m and 15 μ m by 15 μ m, 650 respectively, and the measured surface roughness, $R_q = 16$ nm and $R_q = 50$ nm, respectively.

The surface roughness effect on nanoindentation testing from mechanical polishing and Argon milling is analyzed contrastively. As shown in **Fig. A2**, the Young modulus and hardness data of the mechanical-polishing sample and Argon-milling sample are in good accordance with each other, i.e., these two surface preparation create qualified surface conditions for nanoindentation testing. Due to its ease of use and high efficiency, mechanical polishing is employed for the surface preparation in this study.

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659

Fig. A2. Comparison of Young's modulus and hardness data between the mechanical-polishing
sample (green points) and Argon-milling sample (red points).

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