Review of manuscript **se-2020-13** submitted to Solid Earth, Linsel et al. "**High-resolution analysis of physicochemical characteristics of sandstone media at the lithofacies scale**" by Mattia Pizzati.

General remarks

Dear Authors and Editor, below you can find the review of the submitted manuscript. Revisions are made by describing the issues found line by line and also on the text file of the manuscript, in which critical points were highlighted in green color. Comments on figures, figure captions and tables are presented at the end of this file as well.

This manuscript is focused on the analysis of physical, chemical petrophysical and diagenetic properties characterizing a fluvial-lacustrine sandstone mouth bar at the sedimentological facies scale. Applied methods include both field and laboratory measurements, statistical validation and cross-checking.

Laboratory analyses and statistical tests are thoroughly described and explained, and even if the section describing methods may appear quite long compared to the rest of the manuscript, I feel it is necessary to the full comprehension of the results and discussion. This is particularly true for readers who may not have a strong statistical background and without this support they could be discouraged in going through the text.

The quality of written English is very high, sentences are clear and reading is fluent. I found some minor mistakes and added a few suggestions to further improve the structure of sentences.

Figures are schematic, clear and of good quality. I inserted a few suggestions aimed to improve their visibility and minor corrections to be done mainly on legends and captions.

Data presented fully support the thesis the authors wanted to discuss. This manuscript represents a good example of how important is the detailed study of petrophysical, diagenetic and physical properties of sandstone at the scale of the sedimentary facies. Maybe it would have been nice to have an additional paragraph inside the discussion in which the authors could have described the implications for water-oil reservoir exploitation and maybe also the up-scaling of the properties they discuss. However, this could be much beyond the original aim of the present work, but it may represent a cue for future studies.

Another point that could be of real interest is to test the differences or similarities between the properties of undeformed sandstone and fault rocks, since in the described outcrop at least two faults with few meter of displacement are present. Again consider this as a hint for next research topics.

I think this paper is worth to be accepted on Solid Earth after the completion of technicalminor revisions.

I really hope this review will be helpful to the improvement of the final version of your manuscript. It is a nice research topic and deserves to be developed in the best way.

Do not hesitate to contact me in case any questions or doubts arise from these comments: mattia.pizzati@studenti.unipr.it

Detailed comments line by line

- Line 14: Please add a hyphen between "intergranular" to "inter-granular".
- Line 19: Perhaps this is just a tiny detail, but also in textbooks I always read "P and S-waves" with capital letters. In the entire text of the manuscript it is reported with lower case. Consider if this suggestion suits you. To ease the identification, I underlined "p and s-waves" throughout all the text.
- Line 20: Here maybe "strictly" may sound better than "inevitably".
- Line 22: Consider substituting "In fact" with "Following this,".
- Line 50: Please add a hyphen in "interrelationships" to "inter-relationships".
- Line 54: Please change "are taken" with "have been sampled".
- Line 56: Perhaps the term "series" sounds a bit too generic. Is it possible to adopt the word "orogeny" or "tectonic event"? I have limited knowledge of regional geology of Germany, so evaluate if this term need to be corrected or not.
- Line 59: Check if you want to keep the lower case version "p- and s-waves" or the capitalized one "P- and S-waves".
- Line 72: When you are referring to specific hierarchical stratigraphic nomenclature (formation, group and so on) you should adopt the capital letters. Following this, please correct " *Disibodenberg formation* " with "*Disibodenberg Formation*".
- Line 73: You provided the present-day depth of the Disibodenberg Formation in the Upper Rhine Graben. Is it possible to constrain, even roughly, the maximum burial depth experienced by the studied sandstone in your field site? It could be very interesting since further below you describe the effects of mechanical and chemical compaction affecting the sandstone.
- Line 76: To avoid the repetition of the word "formation" you can substitute the second highlighted word simply with "it".
- Line 77: Here maybe "were sampled" or "were extracted" sounds better than "were taken".
- Line 81: Please change the sequence of these words from "The cubes both were..." to "The cubes were both...".
- Line 84: With the term "granularity" are you describing the angularity of grains composing the mouth-bar sandstone strata? If so I believe that "angularity" provides a clearer and more straightforward idea of what you are describing.

Lines 86-87: Since you are describing a methodology with only one bibliography reference maybe you should turn to the singular form "*This approach, however, does often...*".

Line 88: Instead of "we measured the faces of the cubes..." if you prefer adopt the form "we performed analyses on faces of the cubes...".

Line 96: Can you define the size of the elementary cell used in 3D models?

Line 97: Perhaps the title of the paragraphs could be changed to "In situ measurements" or "Field measurements".

Line 102: Check if you want to keep the lower case version "p- and s-waves" or the capitalized one "P- and S-waves".

Line 105: Check if you want to keep the lower case version "*p- and s-waves*" or the capitalized one "*P- and S-waves*".

Line 111: Since in the same sentence you used the past simple, I would recommend to keep the same style. If you agree change "do" with "did".

Line 117: Check if you want to keep the lower case version "*p- and s-waves*" or the capitalized one "*P- and S-waves*".

Line 119: Here maybe erase the second "travel" since it could be a repetition.

Line 119: Check if you want to keep the lower case version "p- and s-waves" or the capitalized one "P- and S-waves".

Line 129: Perhaps "insights on" sounds better than "insight into".

Line 130: I would rework the sentence as indicated: "can significantly impact the petrophysical properties".

Line 151: Check if here you want to use "neighborhood" or "neighbor".

Line 186: Check if you want to keep the lower case version "p- and s-waves" or the capitalized one "P- and S-waves".

Lines 191-192: Check if you want to keep the lower case version "p- and s-waves" or the capitalized one "P- and S-waves".

Line 244: Consider if "sampled" suits better than "taken".

Line 247: In the previous sentence you state that the basal part of the bed can be classified as Bauma A subdivision, which ideally implies massive medium-coarse sandstone. However, in this line it is reported the presence of "sub-horizontal layering", that may contrast with what is described above (massive should mean structureless). Can you please clarify this point? I fully understand that the Bauma subdivision is something purely ideal, and differences from the model may occur. Are these "sub-horizontal layers" similar to crude laminations, which typically occur in the upper part of the Bauma A subdivision?

Line 247: By "homogeneous" here you mean massive, without any structure or are you referring to the grain size. Sorry for being blunt here, but personally "homogeneous" fells a bit to generic.

Line 251: Here maybe instead of "transitions" you can use "decreases".

Line 252: See if this reworking suits you: "Likewise, sorting increases from poor to moderate".

Line 252: Instead of "continuously" perhaps use "throughout the entire sample volume".

Line 255: To be more explicit please add "... sub-vertically with respect to bedding".

Line 256-257: Could you please separate the percentage of feldspar and the one pertaining to the lithic fragments (mica, rock fragments)? Just to have the percentage relative to quartz, feldspar and rock fragments.

Line 258: Please define if possible the nature of "ductile, autochthonous grains".

Line 260: Here I believe you can use the in-text citation reference "Herron (1988)" instead of the form in parentheses.

Line 261: Here I would erase "produced".

Line 263: You can add "*primary*" before "*inter-granular*" since here you are referring to the original porosity of the sandstone prior to compaction.

Line 272: You can use the word "analyses" to avoid repetition of "measurements".

Line 279: Check if you want to keep the lower case version "p- and s-waves" or the capitalized one "P- and S-waves".

Line 282: Check if you want to keep the lower case version "p- and s-waves" or the capitalized one "P- and S-waves".

Line 309: Check if you want to keep the lower case version "p- and s-waves" or the capitalized one "P- and S-waves".

Line 310: Please correct the misspelled word "petrophyscial" with "petrophysical".

Line 326: Here I would erase "The" to make the sentence lighter.

Line 326: Maybe "that" is more correct than "what".

Line 368: Here you should use the plural form "*lead*" instead of the third person "*leads*", since the subject of the sentence is plural "*alteration products*".

Line 388: Please add a hyphen between "intergranular" to "inter-granular".

Line 390: See if this sentence reworking sounds: "...which might have further enforced degradation of porosity and permeability".

Line 391: Maybe you can change "analogies" simply with "ones".

Line 409: Please add a hyphen between "intergranular" to "inter-granular".

Line 454: Check if this reference is correct, at the end appears an abbreviation "edn." which duplicates the info given before. Maybe it just depends on the reference organizer you adopted.

Line 484: Please put "scotland" in capital letters "Scotland".

Line 534: Check if this reference is correct, at the end appears an abbreviation "edn." which duplicates the info given before. Maybe it just depends on the reference organizer you adopted.

Line 563: Check if this reference is correct, at the end appears an abbreviation "edn." which duplicates the info given before. Maybe it just depends on the reference organizer you adopted.

Comments on figures and figure captions

Fig. 1

In Fig.1a is reported "scissor faults" are you referring to a conjugate system of extensional faults with opposing dip direction? Can they be defined simply "conjugate faults"?

In Fig.1a there is a white rectangle in the background of the label "scissor faults" which partially overlaps with the outcrop image. Please shift upward the white rectangle to avoid intersection.

In the legend in Fig.1e you indicate "current ripples", but as you wrote for "plant fragment and intraclast" you should use the singular form instead of the plural.

In the caption you state "scissor faults", see if you prefer to use the term "conjugate faults" if you feel it could be more correct.

In the caption is reported "*rip-up clasts*", here you can add the nature of the eroded clasts "*pelitic rip-up clasts*".

In the caption please correct "course sand" with "coarse sand".

Fig. 2

In the figure caption please correct "dashes lines" with "dashed lines".

Fig. 3

In the figure caption check if you want to keep the lower case version "*p-* and *s-waves*" or the capitalized one "*P-* and *S-waves*".

In the figure caption you write "*x-ray*", while in the rest of the manuscript is written with capital letters "*X-ray*". Consider revising this point to keep the same style throughout all the text.

Fig. 4

In the caption please erase one of the highlighted "the" the avoid repetition.

Fig. 6

In the caption please evaluate to change "major parts" with "the vast majority of...".

In the caption please add a hyphen between "intergranular" to "inter-granular".

Fig. 7

In the figure caption check if you want to keep the lower case version "*p- and s-waves*" or the capitalized one "*P- and S-waves*".

Fig. 8

In the figure caption check if you want to keep the lower case version "*p- and s-waves*" or the capitalized one "*P- and S-waves*".

Fig. 9

To ease the distinction of the different lines you can try to enlarge a bit both the graphs, or you may assign different colors for every face of the cube.

Table 1

In the table caption check if you want to keep the lower case version "*p- and s-waves*" or the capitalized one "*P- and S-waves*".

High-Resolution Analysis of the Physicochemical Characteristics of Sandstone Media at the Lithofacies Scale

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Abstract. The prediction of physicochemical rock properties in subsurface models regularly suffers from uncertainty observed at the sub-meter scale. Although at this scale – which is commonly termed the lithofacies scale – the physicochemical variability plays a critical role for various types of subsurface utilization, its dependence on syn- and post-depositional processes is still subject to investigation.

The impact of syn- and post-depositional geological processes, including depositional dynamics, diagenetical compaction and chemical mass transfer, onto the spatial distribution of physicochemical properties in siliciclastic media at the lithofacies scale is investigated in this study. We propose a new workflow using two cubic rock samples where eight representative geochemical, thermophysical, elastic and hydraulic properties are measured on the cubes' faces and on samples taken from the inside. The scalar fields of the properties are then constructed by means of spatial interpolation. The rock cubes represent the structurally most homogeneous and most heterogeneous lithofacies types observed in a Permian lacustrine delta formation that deposited in an intermontane basin. The spatio-temporal controlling factors are identified by exploratory data analysis and geostatistical modeling in combination with thin section and environmental scanning electron microscopy analyses.

Sedimentary structures are well preserved in the spatial patterns of the negatively correlated permeability and mass fraction of Fe_2O_3 . The Fe-rich mud fraction, which builds large amounts of the intergranular rock matrix and of the pseudomatrix, has a degrading effect onto the hydraulic properties. This relationship is underlined by a zonal anisotropy that is connected to the observed stratification. Feldspar alteration produced secondary pore space that is filled with authigenic products including illite, kaolinite and opaque phases. The local enrichment of clay minerals implies a non-pervasive alteration process that is expressed by network-like spatial patterns of the positively correlated mass fractions of Al_2O_3 and K_2O . Those patterns are spatially decoupled from primary sedimentary structures. The elastic properties, namely p- and q-wave velocity, indicate a weak anisotropy that is not inevitably oriented perpendicularly to the sedimentary structures.

The multifarious patterns observed in this study emphasize the importance of high-resolution sampling in order to properly model the variability present in a lithofacies-scale system. In fact, the physicochemical variability observed at the lithofacies scale might nearly cover the global variability in a formation. Hence, if the local variability is not considered in full-field projects – where the sampling density is usually low – statistical correlations and thus conclusions about causal relationships among physicochemical properties might be feigned inadvertently.

1 Introduction

The utilization of the subsurface in disciplines such as petroleum reservoir engineering, geothermal heat extraction, mining, carbon-capture and storage or nuclear waste disposal requires highly accurate spatial predictions of relevant physical or geochemical properties in order to assess the economic feasibility of a target region (Landa and Strebelle, 2002; Heap et al., 2017; Kushnir et al., 2018; Rodrigo-Ilarri et al., 2017). Although most of these types of utilization take place at full-field scales, geological variability present at the sub-meter scale may play an important role during the development process. The scale we are speaking of is commonly termed the lithofacies scale (Miall, 1985). Geological heterogeneities at the lithofacies scale might constitute undesirable features in the subsurface such as flow-barriers in reservoirs (Landa and Strebelle, 2002; Ringrose et al., 1993), pathways in nuclear waste disposal sites or geochemical anomalies in mining areas (Wang and Zuo, 2018). Hence, the controlling factors of sub-meter variability should be understood and at least roughly quantified before starting the development in the subsurface region.

In sedimentary bodies, the spatial distribution of the properties is mainly controlled by depositional and diagenetical processes (McKinley et al., 2011, 2013). The spatial characteristics of physicochemical properties in sedimentary rock media are complex due to strongly intersecting and interacting processes during sediment transport, deposition and diagenesis (McKinley et al., 2011). Multiple studies aimed to quantify the variability at the lithofacies scale, most of which concentrated on reservoir properties such as permeability and porosity in 2-D space (McKinley et al., 2011; Hornung et al., 2019). A 2-D analysis suits well for identifying non-visible patterns related to micro-bedding structures at multiple scales even in very homogeneous sandstones (McKinley et al., 2004). That perspective, however, involves simplifications of the physicochemical variability in 3-D space since specific rock properties such as permeability are dependent on the Cartesian direction. Also, consideration of geostatistical parameters such as variographic direction, range, sill and nugget revealed differences in 3-D compared to 2-D space (Landa and Strebelle, 2002; Hurst and Rosvoll, 1991).

With a proper knowledge of the statistical and causal relationships among physicochemical rock properties at different scales, prognostic property models can be significantly enhanced by the integration of small-scale uncertainty into upscaling or conditional simulation algorithms (Lake and Srinivasan, 2004; Verly, 1993). Especially, since multivariate geostatistics can account for interrelationships between rock properties, those relationships can be used as trends or drifts in geostatistical predictions in order to optimize their accuracy in space and time (Hudson and Wackernagel, 1994).

In order to quantify the spatial variability and the multidimensional relationships among physicochemical properties at the 3-D lithofacies scale, the quasi-continuous scalar fields of two rock cubes are modeled by means of spatial interpolation, which is constrained by laboratory measurements. The rock cubes have volumes of 0.0156 m³ and 0.008 m³ and are taken from a Permian lacustrine-deltaic sandstone formation that deposited in the intermontane Saar-Nahe basin during the Cisuralian series. The lithological characteristics of the sandstones are analyzed and both isotropic and anisotropic properties, including bulk rock geochemistry, thermophysical, hydraulic and elastic rock properties, are measured on the cubes' faces. In addition, the intrinsic gas permeability under infinite pressure gradient, the effective porosity, the elemental composition, the thermal

conductivity, the thermal diffusivity together with the p- and s-wave velocity are measured on 108 rock cylinders taken from the inside of the cubes representative for each Cartesian direction in order to account for anisotropic patterns.

The measurements are used to interpolate the full 3-D field of each property. Prior to interpolation, the discrete measurements are statistically analyzed for correlation and formal relationships. Interpolations are conducted using deterministic and stochastic methods including the inverse distance weighting (IDW) and simple kriging (SK) interpolation. The models are evaluated through cross-validation and the observed spatial patterns are categorized. The interpolation results providing the lowest cross-validation error are statistically analyzed again and compared with the aforementioned statistical patterns. Eventually, the geological processes, which produced the observed patterns, are interpreted and discussed with the help of qualitative thin section and environmental scanning electron microscope (ESEM) analyses.

2 Methodology

2.1 Sedimentological characterization and rock sampling

In order to cover multiple varieties of sedimentary lithofacies types, a quarry in Obersulzbach (Rhineland-Palatinate, Germany) in the Saar-Nahe basin was selected for the investigations (Fig. 1). The quarry belongs to the lacustrine-deltaic Disibodenberg formation that is assigned to the Innervariscan Rotliegend Group and comprises four lithofacies types. This formation is deeply buried (1,995 to 2,380 m b.g.s.) in the northern Upper Rhine Graben in southwestern Germany (Becker et al., 2012) and constitutes a potential target unit for hydrothermal exploitation (Aretz et al., 2015). The outcrop has been chosen in order to estimate the variability of physicochemical properties that could be expected in this formation as an uncertainty factor if the formation gets targeted in a deep geothermal project.

Two rock cubes of $0.2 \times 0.2 \times 0.2 \,\mathrm{m}$ (OSB2_c) and $0.25 \times 0.25 \times 0.25 \,\mathrm{m}$ (OSB1_c) were taken from the outcrop wall using a rock chainsaw. According to the outcrop's coordinate system, one edge of the cuboid runs east-west (x), one north-south (y) and one in altitude (z) direction. The irregular cuboids were reworked to regular cubes with a stationary rock saw. We selected two types of lithofacies (Fig. 2) – both sandstones – one representing a cross-bedded, heterogeneous, compartmentalized variety (OSB1_c) and the other one a homogeneous variety (OSB2_c). The cubes both were extracted from a distributary mouth bar building a foreset in a fluviatile-dominated lacustrine delta. OSB1_c (Fig. 2) was taken from the high-energetic basal part whereas OSB2_c was taken from the lower-energetic top. The sedimentological characteristics including grain size, sorting, granularity, sedimentary structures and mineral content were determined by visual inspection, thin section and ESEM analyses. Two different types of zonal anisotropy and spatial patterns were expected to be found with the aforementioned sampling strategy. In other studies such as McKinley et al. (2011) measurements were directly conducted in the field. These approaches, however, do often provide a drawback in accuracy and precision, especially in permeability measurements. In order to address this issue, we measured the faces of the cubes under laboratory conditions. In the next step, the cubes were cut to rock slabs, from which cylinder samples were taken. Totally, 108 rock cylinders – 79 from OSB1 c and 29 from OSB2 c –

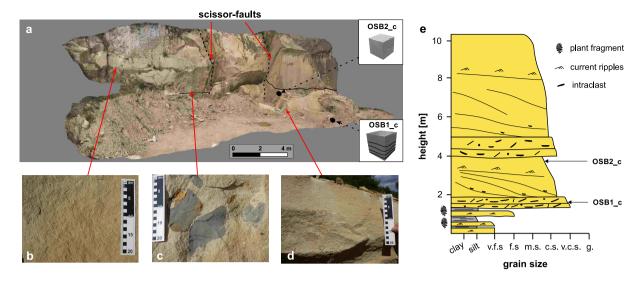


Figure 1. (a) The investigated sandstone quarry in Obersulzbach, Germany. The outcrop is compartmentalized by two scissor-faults in the central part that provide offsets of several meters. (b) Homogeneous sandstone (c) Rip-up clasts embedded in a homogeneous rock matrix (d) Ripple-cross bedded sandstone (e) Cumulative sedimentary log of the outcrop architecture. The sampling positions of OSB1_c and OSB2_c are marked. v.f.s. = very fine sand; f.s. = fine sand; m.s. = medium sand; c.s. = course sand; v.c.s. = very coarse sand; g. = granule.

were extracted from the rock cubes. It was ensured that at least five samples were produced representative for each Cartesian direction. Applying the formula for calculating a cylinder's volume V_c with

$$V_c = h \times \pi \times r^2,\tag{1}$$

where h is the height of the cylinder and r the radius, the relative volume covered by the rock cylinders in the rock cubes was calculated to be 25.4% for OSB1_c and 18.2% for OSB2_c, respectively. Eventually, target meshes are needed to interpolate the full 3-D scalar fields. Therefore, both cubes were modeled in 3-D using a regular grid consisting of 27,000 hexahedral, orthogonal cells.

2.2 Measurement campaign

First, a local metric coordinate system was defined, where each edge of the cube represents an axis in the Cartesian coordinate system in order to reference each measurement to a point in space. The sampling points were set in a raster of 9×9 points on each face for OSB1_c and 5×5 for each face of OSB2_c. All measurements were conducted in the laboratory of the Institute of Applied Geosciences in Darmstadt, Germany. After drying the rock cubes at 60° C, non-invasive measurements were conducted on each face of the cube. On the cubes' faces the p-and s-wave velocity and elemental mass fractions were determined (Fig. 3).

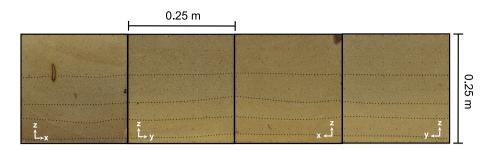


Figure 2. Lateral faces of OSB1_c displayed in the form of an open cube (from left to right: XZ Front, YZ Front, XZ Back and YZ Back). The internal bounding surfaces are indicated by the dashes lines.

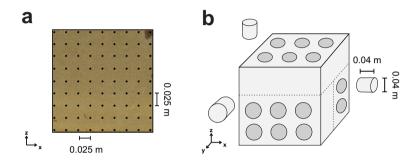


Figure 3. (a) Sampling locations for the non-invasive measurements including p- and s-wave velocity and x-ray fluorescence exemplary displayed for the face XZ Back of OSB2_c (b) Schematic of the extraction strategy for sampling the rock cylinders.

After the extraction, the rock cylinders were oven-dried at 105 °C and measured in order to determine the intrinsic gas permeability, effective porosity, p- and s-wave velocity, elemental mass fractions, thermal conductivity as well as the thermal diffusivity in unsaturated conditions. Those properties can be considered key properties of the rock matrix in porous aquifers with regard to hydrothermal systems (Agemar et al., 2014) since they constitute input variables for the governing equations for heat transfer and fluid flow in the subsurface (Carslaw and Jaeger, 1959).

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The permeability was measured with the Hassler cell permeameter, which is described in Filomena et al. (2014). This technique allows for the estimation of the intrinsic gas permeability, which is the permeability at an infinite pressure gradient. The Hassler cell permeameter was set to accept a measurement if fifteen consecutive readings do not deviate by more than 5%. The measurement error, however, can exceed that value especially in low-permeable lithologies. Effective porosity measurements were conducted using an envelope density analyzer (GeoPyc 1360). The accuracy is given by the manufacturer to be within \pm 0.55% (Micromeritics, 1998). Thermal properties under unsaturated conditions, namely the thermal conductivity and thermal diffusivity, were determined with a thermal conductivity scanner (TCS) according to the work of Popov et al. (1999). The measurement error is quantified to be \leq 3% for thermal conductivity and \leq 8% for thermal diffusivity (Popov et al., 1999). The elastic properties of p- and s-wave velocity in the rock media were measured with the sonic wave generator UKS-D (Geotron) by sending a sonic wave pulse from a pulse-providing test head (UPG-S) to a receiver (UPG-E). The wave velocity

is a function of travel length, density and travel time. The initial occurrence of the poor s-wave must be picked manually after visual inspection by the operator. Thus no measurement error can be provided since user bias cannot be assessed quantitatively. Bulk elemental analysis using the Bruker S1 TITAN handheld portable X-ray fluorescent (pXRF) analyzer was used to find correlations between the elemental composition and the petrophysical properties. The measurement device works on the basis of energy dispersive X-ray fluorescence (EDXRF) and estimates the elemental mass fractions of a sample. This device produces an ionizing X-ray beam of 1.2 cm diameter and quantifies the elemental composition based on the energy emitted by the ionized elements in the targeted area. The portable device can measure the fraction of elements with an ordinal number ≥ 12 and ≤ 235 if the threshold value defined by the measurement error for the specific element in the sample is exceeded. For this study, the device was operated in GeoChem, DualMining mode allowing for detection of the major oxides SiO₂, Al₂O₃, Fe₂O₃ and K₂O as well as a wide range of other elements. The device has been calibrated with international standards. We used the previously mentioned major oxides for analyses since those can provide insight into the iron oxide and clay mineral distribution, which can impact petrophysical properties significantly More details on the measurement devices can be found in the works of Hornung and Aigner (2002), Sass and Götz (2012), Filomena et al. (2014) and Aretz et al. (2015).

2.3 Data analysis and spatial modeling

2.3.1 Variography

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The experimental semivariogram represents the cumulative dissimilarity of a discrete set of point-pairs x with n_c as the count of point-pairs within the distance classes h of identical distance increments (Eq. 2).

$$\gamma(\mathbf{h}) = \frac{1}{2n_c} \sum_{\alpha=1}^{n_c} \left(z(x_\alpha + \mathbf{h}) - z(x_\alpha) \right)^2$$
(2)

The continuous counterpart, represented by the variogram model, is an approximation to the experimental semivariogram assuming $z(\boldsymbol{x})$ to be a stationary random field (Wackernagel, 2003). A variogram model γ_{theo} is represented by a covariance function c with the relationship $\gamma_{theo}(\boldsymbol{h}) = c(0) - c(\boldsymbol{h})$, where c is a positive definite, even function. Six covariance models are mostly used to fit the experimental semivariogram, namely the spherical, gaussian, exponential, power, kardinal sine and the linear model (Armstrong, 1998; Ringrose and Bentley, 2015). In this study, we only observe spherical relationships with nugget effect. This model is calculated as

$$c_{sph}(\boldsymbol{h}) = \begin{cases} n + b \cdot \left(1 - \frac{3|\boldsymbol{h}|}{2a} + \frac{|\boldsymbol{h}|^3}{2a^3}\right) & \text{for } 0 \le |\boldsymbol{h}| < a \\ n & \text{for } |\boldsymbol{h}| \ge a, \end{cases}$$
(3)

with the variables nugget (n), range (a) and sill (b). Semivariograms can be used to quantify the spatial or time correlation of a random property (Ringrose and Bentley, 2015; Gu et al., 2017; Rühaak et al., 2015). Further on, the differences in range and sill in dissimilar directional semivariograms can quantify the zonal and geometric anisotropy of a property (Ringrose and Bentley, 2015). The resulting covariance function is an input variable for stochastic interpolation algorithms.

2.3.2 Rock property interpolation

Spatial inter- and extrapolation can be generated with deterministic and stochastic techniques. All interpolations are based on the assumption that a point x_k with a known value $z(x_k)$ has a weight on a discrete point x_0 in space with an unknown value $z(x_0)$. The global known points, however, can be reduced to a local neighborhood of x_0 .

For deterministic interpolation the p-value inverse distance weighting (IDW) (Shepard, 1968) interpolation is used. The IDW interpolation generally calculates an unknown value $z(x_0)$ at point x_0 by weighting the distance of that point to each known value point (x_k) in space. The underlying formula for IDW is

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$$z(x_0) = \frac{\sum_{k=1}^{n} 1/d_k^p \cdot z(x_k)}{\sum_{k=1}^{n} 1/d_k^p},$$
 (4)

where d is the Euclidean distance between the point with the known value x_k and the point with the unknown value x_0 , and p is an exponent factor to bias the weights non-linearly. The p-value is mostly used for smoothing the results by controlling the distance-decay effect (Lu and Wong, 2008). IDW is a reliable and widely applied method to interpolate static rock properties in one to three-dimensional space (Rühaak, 2006).

For stochastic interpolation simple kriging (SK) is used. Kriging in general is a popular technique to interpolate geological properties in space (Goovaerts, 1997; Rühaak, 2015; Malvić et al., 2019). Through kriging, the value $z(x_0)$ at an unknown point x_0 is calculated by weighting the neighboring known values and building a linear combination of those via the formula

$$z(x_0) = \sum_{k=1}^n w_k \cdot z(x_k),\tag{5}$$

where w_k is the weight of the known point x_k with the value $z(x_k)$. To obtain the simple kriging weights, a set of n equations has to be solved. This set of equations can be written as

$$\begin{pmatrix} c(x_1 - x_1) & \cdots & c(x_1 - x_n) \\ \vdots & \ddots & \vdots \\ c(x_n - x_1) & \cdots & c(x_n - x_n) \end{pmatrix} \begin{pmatrix} w_1^{SK} \\ \vdots \\ w_n^{SK} \end{pmatrix} = \begin{pmatrix} c(x_1 - x_0) \\ \vdots \\ c(x_n - x_0) \end{pmatrix}$$

$$(6)$$

with c as covariance function and x_n as point with known value (Wackernagel, 2003). The quality of kriging interpolation is dependent on the variogram model, the defined neighborhood, the sampling density and the goodness-of-fit to the experimental values.

170 2.4 Cross-validation

Cross-validation can be used to assess the quality of a model. During cross-validation, p randomly selected samples are removed from the input data set of size n with 0 and the interpolation is performed without those samples Celisse (2014). The measures of goodness of fit being used in this study include the root-mean-square error (RMSE)

$$RMSE = \sqrt{\frac{1}{n} \sum_{k=1}^{n} (\hat{z}(x_k) - z(x_k))^2}$$
 (7)

and the mean-absolute error (MAE)

$$MAE = \frac{1}{n} \sum_{k=1}^{n} |\hat{z}(x_k) - z(x_k)|$$
 (8)

with $\hat{z}(x_k)$ as estimated value at point x_k . Those parameters allow for the quantitative assessment of an interpolation's quality. They might be prone to bias if the sampling density in the target domain is extremely scarce.

2.4.1 Anisotropy

Anisotropy describes the dependence of a physical property on a direction. Rock properties such as stiffness, permeability or thermal conductivity are anisotropic in most cases. Hence, measurements of those properties might show differing magnitudes in different directions if the medium is polar anisotropic. Anisotropy in geological media is generated by preferred orientation of mineral grains or cracks as well as by the intrinsic anisotropy of single crystals (Thomsen, 1986).

Following, we will exemplary describe the anisotropy of elasticity and we will provide measures for anisotropy quantification under the simplifying assumption of transversal isotropy. The elastic modulus tensor can be expressed as a 4^{th} -rank tensor

$$\mathbf{C} = \begin{pmatrix} C_{11} & C_{11} - 2C_{66} & C_{13} & 0 & 0 & 0 \\ C_{11} - 2C_{66} & C_{11} & C_{13} & 0 & 0 & 0 \\ C_{13} & C_{13} & C_{33} & 0 & 0 & 0 \\ 0 & 0 & 0 & C_{44} & 0 & 0 \\ 0 & 0 & 0 & 0 & C_{44} & 0 \\ 0 & 0 & 0 & 0 & 0 & C_{66} \end{pmatrix}$$

$$(9)$$

where C_{ij} represents an elasticity modulus and the indices are related to the directional **p-** and s-wave velocity, under the assumption that z is the symmetry axis. The velocities can be calculated by

$$v_p^z = \sqrt{\frac{C_{33}}{\rho}} \tag{10}$$

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$$v_s^z = \sqrt{\frac{C_{66}}{\rho}}$$
 (11)

where v_p is the p-wave velocity and v_s is the s-wave velocity parallel to the symmetry axis and ρ is the bulk density (Yang et al., 2020). The anisotropy, here exemplary expressed for the p-wave polar anisotropy, can be quantified with the Thomsen parameters (Thomsen, 1986). For example ϵ can be expressed as

$$\epsilon = \frac{C_{11} - C_{33}}{2C_{33}}.\tag{12}$$

195 If $\epsilon \ll 1$ the material can be classified as weakly anisotropic.

2.4.2 Correlation and regression analysis

In order to quantify linear statistical relationship between two independent variables x and y, the Pearson linear product-moment correlation coefficient (R) can be used. R is expressed as

$$R = \frac{\sum_{k=1}^{n} (x_k - \overline{x})(y_k - \overline{y})}{\left(\sum_{k=1}^{n} x_k^2 - n \cdot \overline{x}^2\right)\left(\sum_{k=1}^{n} y_k^2 - n \cdot \overline{y}^2\right)},\tag{13}$$

with n representing the number of compared point pairs and \overline{x} and \overline{y} standing for the arithmetic mean of x and y.

Regression aims at finding a fitting function between samples of two or more random variables. For curvilinear regression, a function of a degree > 1 will be approximated for a discrete set of values. A second-degree polynomial function f(x) for instance would be described as

$$f(x) = b_0 + b_1 x + b_2 x^2 (14)$$

Thus, we would need to find n+1 regression coefficients, where n is the degree of f(x). In general, the regression model yields

$$f(x)_i = b_0 + b_1 x_i + b_2 x_i^2 + \dots + b_n x_i^n, \tag{15}$$

with i = 1, 2, ..., n. The regression coefficients b_i are obtained through solving a system of linear equations as

$$\begin{pmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{pmatrix} = \begin{pmatrix} 1 & x_1^1 & \cdots & x_1^m \\ 1 & x_2^1 & \cdots & x_2^m \\ \vdots & \vdots & \cdots & \vdots \\ 1 & x_n^1 & \cdots & x_n^m \end{pmatrix} \begin{pmatrix} b_0 \\ b_1 \\ \vdots \\ b_m \end{pmatrix}, \tag{16}$$

where x and y are the samples. The function approximations as produced in regression analyses are commonly evaluated by the coefficient of determination (\mathbb{R}^2), which is calculated through

$$R^2 = 1 - \frac{s_{res}}{s_{tot}} \in [0, 1], \tag{17}$$

where

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$$s_{res} = \sum_{k=1}^{n} (y_k - f(x)_k)^2 \tag{18}$$

215 is the explained sum of squares and

$$s_{tot} = \sum_{k=1}^{n} (y_i - \overline{y})^2 \tag{19}$$

is the total sum of squares.

2.4.3 Spatial modeling and statistical analyses

The spatial dependence of the discrete values is evaluated through experimental semivariograms. The semivariograms are generated for the single rock faces, where measurements are available, and for the plug measurements. The empirical semivariogram is fitted by a variogram model, which is then used for the stochastic interpolation. Interpolation analyses are performed as IDW and SK realizations (Fig. 4) that are assessed through cross-validation. The power parameter for IDW is chosen to be three since this constant provides the lowest RMSE among the realizations. The search radii for each prediction is chosen to be 0.2 m in x and y direction and 0.15 m in z direction in OSB1_c to account for the sedimentary structures. For OSB2_c the search radii are chosen to be isotropic with a length of 0.2 m. To make the methods comparable, we select the maximum number of neighboring points to be 25 representing between 5 and 95% of the measurements.

We decided to waive sequential simulation because major amounts of the cubes' volumes are covered by rock samples. Thus, we do not expect a relevant kriging variance. With this in mind, it is assumed to capture most of the total variance by the measurements themselves. The interpolation results that provide the lowest cross-validation error are used for statistical analyses in order to derive correlations and regression functions between the scalar fields. Eventually, significant correlations

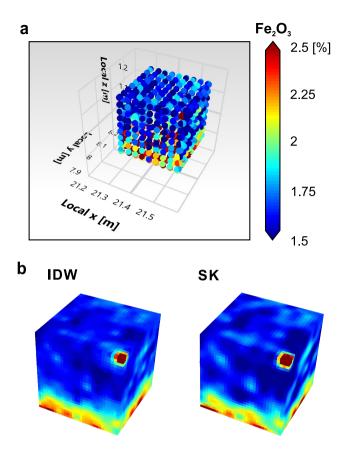


Figure 4. (a) Fe_2O_3 measurement locations on the cube faces of OSB1_c and on the rock samples extracted from the cube. The diameter of one point is 1.2 cm, which corresponds to the beam diameter of the pXRF measurement device. (b) Visual representation of the inverse distance weighting (IDW) and simple kriging (SK) realization of the 3-D scalar field of Fe_2O_3 using the the discrete points displayed in (a) as known data points.

are compared with the non-interpolated data sets. Both the spatial modeling and the statistical analyses are performed with the open-source software GeoReVi (Linsel, 2019). This software tool provides functionality for multidimensional subsurface characterization using the concept of knowledge discovery in databases, which is helpful when handling huge data sets as produced in this study.

235 3 Results

3.1 Sedimentological characteristics

The sandstones belong to a clinothem strata deposited in a fluvial-dominated lacustrine delta. More specific, the architectural element represents a distributary mouth bar, formed by rapid sandstone deposition in sheet-like bodies like exemplary described

in (Fongngern et al., 2018). The base of those bodies is typically erosive, which is why muddy rip-up clasts commonly occur above the base. Also, the beds, which deposited after the intraclast-rich basal beds, typically show trough or ripple-cross stratification with set heights of 5–15 cm. Vertical orientation of rip-up clasts can be observed in matrix-rich debrites or turbidites deposited under high-energy turbulent hyperpycnal to homopycnal flow conditions (Li et al., 2017). Those are unconformably overlying lacustrine, laminated mud strata from the prodelta environment. Accordingly, Bouma A-C layers (Bouma, 1962) with a prograding trend can be identified in the outcrop. The beds, from which the cubes were taken, correspond to Bouma A. With ongoing sedimentation, the depositional energy in a Bouma A sequence typically decreases, which leads to massive sandstones. OSB1_c was taken from a basal bed of the Bouma A sequence, characterized by a high amount of intraclasts, normal grading and sub-horizontal layering, whereas OSB2_c was taken from the topmost bed, characterized by a homogeneous structure.

The average grain size in both cubes ranges from fine to very coarse sand (200–1400 µm). While the grain size distribution in OSB2_c does not show a significant variability – mainly characterized by medium to coarse sand – a normal grading is observable in OSB1_c. Here, the grain size gradually transitions from very coarse sand at the base to medium sand at the top. Likewise, sorting transitions from poor to moderate sorting. In OSB2_c the sorting is moderate continuously. The components provide a low to medium sphericity while the grain shapes vary between sub-angular to sub-rounded. Locally, pelitic rip-up clasts occur with diameters of up to four centimeters. The rip-up clasts show a very low textural maturity and are oriented sub-vertically.

The original rigid detrital components consist of 50–60% quartz, 20–30% strongly altered feldspar as well as micas and lithic fragments. Mica grains are often bent between more rigid grains. The rock matrix accounts for approximately 10–20% and is built up by detrital grains – coated by iron oxides –, ductile, autochthonous grains and fine-grained quartz. According to the geochemical analyses, the rocks can be classified as lithic arenites to arkoses or wackes (Fig. 5) respectively, if the matrix content exceeds 15%, applying the classification of (Herron, 1988).

Thin section analysis (Fig. 6a) reveals that most of the pore space is produced secondary due to grain dissolution. The secondary pores are undeformed indicating that grain dissolution took place during structural inversion – probably during telogenesis according to the concept of Worden and Burley (2003). Most of the inter-granular volume was destroyed during mechanical compaction. ESEM analysis (Fig. 6b) confirms the presence of quartz accompanied by co-precipitated calcite, opaque phases – mainly iron oxides – and authigenic clay minerals including kaolinite and illite in the cement fraction. Thus, chemical compaction had taken place by iron oxide, quartz and clay mineral precipitation during diagenesis. Here, the earliest cement phase is represented by the opaque phases comprising a high amount of iron oxides. Following, kaolinite is formed, mainly in the secondary pore space, overgrown by illite. Often, the early cement is overgrown syntaxially by quartz. The source of SiO₂ might be internal and related to feldspar dissolution.

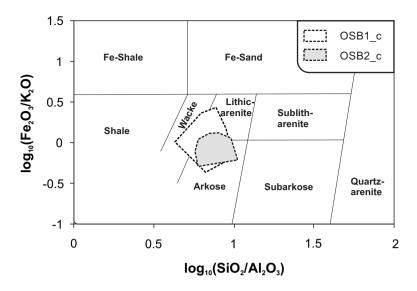


Figure 5. Petrographic classification after Herron (1988) based on the ratio of SiO_2 and Al_2O_3 and Fe_2O_3 and K_2O . The polygons show the convex hull for the measurements derived from the cubes' faces.

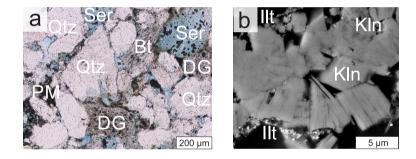


Figure 6. (a) Representative thin section taken from rock cube OSB2_c. The sandstone consists mainly of quartz (Qtz), altered feldspars with residual mineral aggregations (sericite, Ser), altered biotite (Bt) and ductile grains (DG). Feldspar dissolution lead to a high grade of secondary porosity (Molenaar et al., 2015) while major parts of the intergranular pore space are filled with primary and pseudomatrix (PM), which is rich in iron oxides. (b) Environmental scanning electrone microscope (ESEM) image of the authigenic clay minerals (mainly kaolinite (Kln) and illite (Ilt)) built in the pore space. Mineral abbreviations were taken from Whitney (2010).

270 3.2 Exploratory data analysis

In order to provide full comparability, the following section will provide an overview over the measurements derived from the rock cylinder measurements. For each property, 79 rock samples from OSB1_c and 29 from OSB2_c were investigated. An overview over the properties' ranges is provided in the Box-Whisker charts displayed in Figure 7.

The local variability of OSB1_c is significantly higher than that of OSB2_c. Intrinsic permeability of OSB1_c provides a coefficient of variation of 0.3 and a Dykstra-Parson coefficient of 0.4 while measurements from OSB2_c show values of 0.2 for

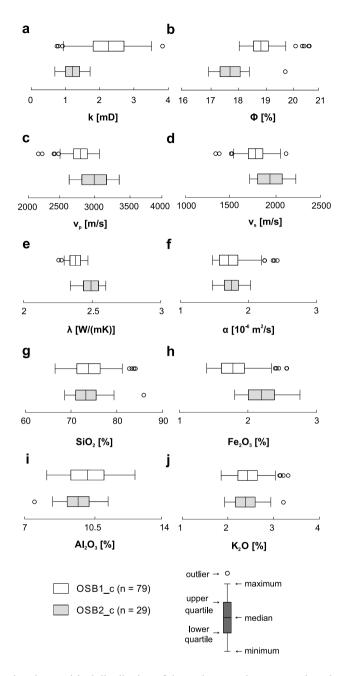


Figure 7. Box-Whisker charts showing the empirical distribution of the rock properties measured on the rock cylinders taken from the rock cubes. Outliers were detected according to Tukey's method (Tukey, 1977) where a value is tested to be in the 1.5-times inner-quartal-range of the arithmetic mean. (a) Intrinsic permeability k (b) Effective porosity ϕ (c) p-wave velocity v_p (d) s-wave velocity v_s (e) Thermal conductivity λ (f) Thermal diffusivity α and the mass fraction of (g) Silicon oxide SiO₂ (h) Iron oxide Fe₂O₃ (i) Aluminum oxide Al₂O₃ and (j) Potassium oxide K₂O.

the coefficient of variation and 0.18 for the Dykstra-Parson coefficient respectively. According to the classification provided by Corbett and Jensen (1992), the intrinsic permeability of both rock cubes can be classified as being very homogeneous.

The range of values in OSB1_c for each property is greater than the range of those in OSB2_c. OSB1_c provides lower values in p- and s-wave velocity, thermal conductivity and mass fraction of Fe₂O₃ compared to OSB2_c. Intrinsic permeability and porosity in turn are greater. The mass fraction of silicon oxide and thermal diffusivity provide similar statistical parameters in both cubes, however, the ranges are marginally larger in OSB1_c. The measurements of the elastic rock properties revealed a weak anisotropy of the p-wave attenuation especially in rock cube OSB2_c. The Thomsen parameter ϵ is 0.047 for OSB1_c and 0.096 for OSB2_c. It should be noted that OSB1_c provides visible bedding structures in contrast to OSB2_c, hence, the observed degree of anisotropy is not connectable to the bedding features in this case.

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Statistically significant linear correlations (Fig. 8), in the sense of passing a two-tailed significance test at the 0.05 level, were found between porosity and permeability, permeability and Fe_2O_3 , v_p and v_s , v_p and SiO_2 , v_p and Al_2O_3 , v_p and K_2O_3 , K_2O_3 and K_2O_3 are swell as K_2O_3 and K_2O_3 . The strongest positive linear correlation can be observed between v_p and v_s (R = 0.88), K_2O_3 and K_2O_3 (R = 0.70) and porosity and permeability (R = 0.31). The strongest negative correlation can be observed between permeability and K_2O_3 (R = -0.56). Properties not being mentioned do not provide significant statistical correlations to others.

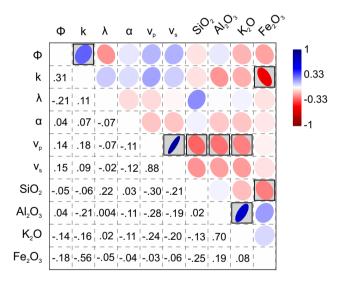


Figure 8. Matrix visualization of the Pearson correlation coefficient derived from the plug measurements. Statistically significant correlations with a p-value ≤ 0.05 are highlighted by gray boxes. The diameter of the ellipses' conjugate axes is dependent on the correlation coefficient. The smaller the length of the axis, the stronger is the correlation. The matrix is diagonal meaning that the Pearson correlation coefficient as numerical expression is located at the diagonal position relative to each ellipsis. $\Phi = \text{effective porosity}$; k = permeability; $\lambda = \text{thermal diffusivity}$; $v_p = \text{p-wave velocity}$; $v_s = \text{s-wave velocity}$.

3.3 Sub-meter scale spatial correlation

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The spatial dependence of the discrete measurements is estimated using experimental semivariograms. Therefore, the geochemical representatives SiO_2 (Fig. 2a) and Fe_2O_3 (Fig. 2b) that were measured on each of the rock faces of OSB1_c are therefore exemplary analyzed. The experimental semivariograms greatly vary from face to face in OSB1_c. The nugget effect for each experimental variogram is very low. The range of each semivariogram varies between 0.05 and 0.3 m. In the experimental semivariograms of SiO_2 , two types of patterns can be identified. The faces XY Base, XZ Back and YZ Front, which are displayed in Figure 9, all show ranges of approximately 0.08 m and a sill between 8 and $10\%^2$ until the semivariance exponentially increases when exceeding a lag distance of 0.2 m. The semivariance on the other faces runs similarly with ranges of 0.2 m and a sill of $4.7\%^2$. The semivariogram for Fe_2O_3 shows some similarities. Here, the faces XY Base, YZ Front and XZ Front show very low ranges between 0.05 and 0.15 m and sill between 0.1 and $0.15\%^2$ again with an exponential increase when exceeding a lag distance of 0.2 to 0.25 m. In contrast, the semivariance of YZ Back has the highest sill with $0.21\%^2$ and a range of 0.15 m, however, semivariance drops after exceeding a lag distance of 0.2 m. XZ Back provides the highest degree of similarity with a range of 0.3 m and a sill of $0.09\%^2$ using a spherical approximation. Both geochemical properties show a zonal anisotropy where the sill shows different magnitudes along different directions (Wackernagel, 2003; Allard et al., 2016).

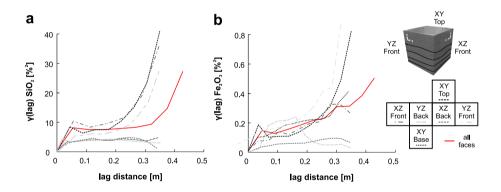


Figure 9. Empirical semivariograms of the mass fraction of SiO₂ (a) and Fe₂O₃ (b) in rock cube OSB1_c grouped by the investigated rock face.

3.4 Spatial pattern analysis

The spatial distributions of the rock properties are interpolated with Shepard's inverse distance weighting (IDW) and simple kriging (SK). Both realizations of a single scalar field provide comparable patterns. The interpolation errors are also located in similar ranges, however, IDW seems to be more sensitive to outliers resulting in much higher interpolation errors with regard to properties like p-wave velocity or mass fraction of SiO₂ (Table 1). IDW tends to underestimate the maximum and minimum values in the scalar fields. Thus, petrophyscial and geochemical contrasts are more distinctly reproduced in the stochastic approach. Also, the IDW realization shows the bull's eye effect, which is a typical artifact of IDW interpolations (Shepard, 1968). Accordingly, the simple kriging realizations are used for further analyses.

The rock properties exhibit a multitude of spatial patterns. Here, discrete, layered and homogeneous patterns, both connected and disconnected to primary sedimentary structures, could be observed in the interpolations.

Table 1. RMSE and MAE for the interpolation results of IDW and SK for OSB1_c. k = permeability; $\Phi = \text{effective porosity}$; $\lambda = \text{thermal}$ conductivity; $\alpha = \text{thermal}$ diffusivity; $v_p = \text{p-wave}$ velocity.

	RMSE IDW	RMSE SK	MAE IDW	MAE SK
k	.19	.17	.15	.14
Φ	.54	.59	.4	.42
λ	.23	.22	.18	.16
α	.14	.17	.1	.1
\mathbf{v}_p	64.19	60.95	52.21	44.74
SiO_2	4.07	3.25	3.05	2.09
Al_2O_3	.8	.83	.66	.66
K_2O	.25	.26	.19	.2
Fe_2O_3	.93	.32	.86	.21

315 3.4.1 Patterns connected to sedimentary structures

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A bedding-connected pattern is exhibited in the intrinsic permeability and Fe₂O₃ interpolation results of OSB1_c. The mass fraction of Fe₂O₃ varies between 1.25 and 5% in OSB1_c. In the histogram displayed in Figure 11 outliers were removed according to Tukey's outlier detection method (Tukey, 1977). The local histogram of OSB1_c's intrinsic permeability shows a bimodal distribution ranging from 0.7 to 3.9 mD. The application of Tukey's method revealed no statistical outliers in this scalar field.

The bedding structures in OSB1_c are well reflected by the spatial pattern of the interpolated intrinsic permeability gradually increasing from low values between 0.7 and 2 mD in the lower beds to higher values between 2 and 4 mD in the upper beds (Fig. 10).

The spatial distribution of the mass fraction of Fe_2O_3 in OSB1_c provides a reciprocal trend compared to the permeability. Here, the lowermost bed shows a significantly higher content compared to the upper beds. Both scalar fields show zonal anisotropy. The Fe_2O_3 content is an indicator for the detrital matrix, pseudomatrix and cement content what in turn would explain the reciprocal relationship with the permeability measurements. In siliciclastic systems, iron can be contained in clay minerals (up to 30wt% (Brigatti et al., 2006)), mafic components or in iron-rich oxides, hydroxides or carbonates. Local excesses in the Fe_2O_3 content exist in the spatial distribution. Those can be explained by clay-rich intraclasts observed on the rock faces. When comparing the pattern to Figure 2 at both XZ-oriented cube faces, rip-up clasts can be observed where high Fe_2O_3 mass fractions occur. Those areas provide the maximum values of the Fe_2O_3 distribution.

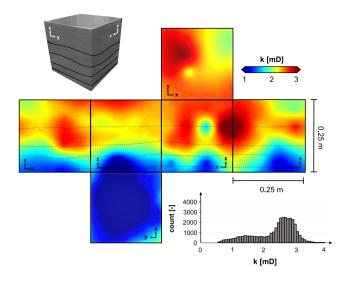


Figure 10. Spatial distribution of the intrinsic permeability modeled with a simple kriging interpolation. The histogram shows a bimodality of the distribution split up into the basal beds and uppermost beds.

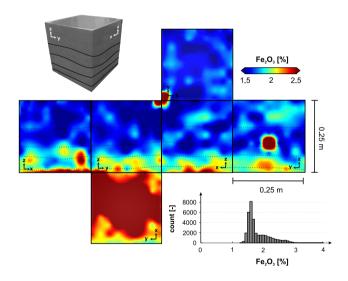


Figure 11. Spatial distribution of the mass fraction of Fe_2O_3 modeled with a simple kriging interpolation. As in the intrinsic permeability interpolation, a bimodality can be observed in the empirical histogram.

3.4.2 Patterns decoupled from sedimentary structures

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Other scalar fields are decoupled from depositional bounding surfaces. For instance, the geochemical mass fractions of K_2O (Fig. 12) and Al_2O_3 (Fig. 13) provide a significant positive correlation unconnected to visible structural boundaries. Typically, those geochemical properties are indicative for the presence of orthoclase feldspar (KAlSi₃O₈) and/or illite (KAl₃Si₃O₁₀(OH)₂)

in siliciclastic environments. The mass ratio of both components is roughly 1:3 to 1:4, which is in accordance to the illite fraction that was observed in thin section and ESEM analyses. Only minor amounts of orthoclase feldspar could be found in the thin sections. Thus, we assume that the correlation of K_2O and Al_2O_3 can traced back to the illite phases.

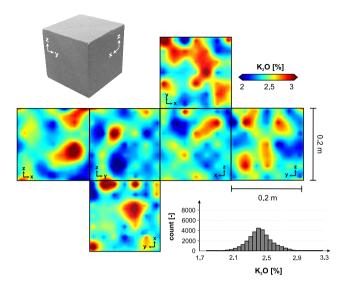


Figure 12. Spatial distribution of the mass fraction of K_2O modeled with a simple kriging interpolation. The pattern is decoupled from primary sedimentary structures and shows a network-like structure.

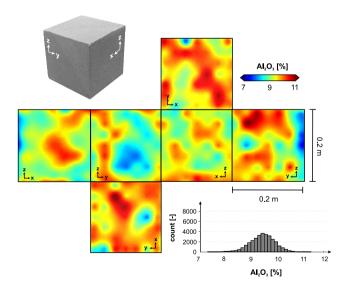


Figure 13. Spatial distribution of the mass fraction of Al_2O_3 modeled with a simple kriging interpolation. The pattern is decoupled from primary sedimentary structures and shows a network-like structure.

Higher fractions of Al₂O₃ are supposedly due to higher kaolinite (Al₂Si₂O₅(OH)₄) fractions in the clay mineral assemblages.

The patterns are diffuse showing autocorrelated areas of slightly enriched and depleted mass fractions. Enriched areas seem to be connected, building network-like patterns, while depleted areas are more isolated.

4 Discussion

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The overall aim of this study was to quantify the three-dimensional inter-dependencies of thermophysical, hydraulic, elastic and geochemical scalar fields in sandstone media at the lithofacies scale and to identify the controlling factors for the property distributions. With a high-resolution study at the lithofacies scale, statistical and spatial inter-relationships between characteristic physicochemical fields could be discovered and traced back to depositional and diagenetical processes.

4.1 Petrophysical and geochemical characteristics

Recent multi-scale modeling approaches without the usage of local constraints show that the prediction of permeability and porosity in siliciclastic systems is still challenging (Nordahl et al., 2014). Geological sampling almost never includes the entire domain that is investigated. With sampling densities of 25.4% and 18.2%, we reached a very high degree of coverage. Studies such as Hurst and Rosvoll (1991) showed that a very high sampling density is necessary to cover the entire variance of permeability at the lithofacies scale. The interpolations performed in this study reproduce the global histogram properly and also outliers are accounted for. This, in fact, implies that the sampling density was selected adequately in order to capture the total variability present in the physical and geochemical scalar fields. This condition is typically only fulfilled in sequential simulations (Robertson et al., 2006) rather than in conventional interpolations.

Although statistically significant correlations may imply a natural relationship between physicochemical properties, this relationship could also be based on random processes requiring causality to be verified. Weak correlations were found between the effective porosity and the intrinsic permeability, which are positively correlated usually (Pape et al., 1999). This relationship can be traced back to the Kozeny–Carman equation that connects the permeability with the effective pore throat radius r_{eff}^2 and a formation factor F like

$$k = r_{eff}^2/(8 \cdot F). \tag{20}$$

The formation factor is defined as the ratio of tortuosity and porosity showing that porosity and permeability provide a positive formal relationship empirically. A high amount of secondary pores, produced by feldspar dissolution, did not significantly contribute to the permeability in the investigated sandstones since those pores are often hydraulically isolated. Consequently, secondary porosity did not necessarily lead to increasing radii of the effective pore throats rather than increasing tortuosity. Also, recrystallized quartz cement – blocking a large amount of the pore throats – must be taken into account. Both effects, in turn, resulted in a degraded permeability. Additionally to the geometrical aspects previously mentioned, the alteration products in form of clay minerals occupy the pore space, which leads to larger adhesive effects that hinder the ability to transport fluids

as well. This observation is in good agreement with observations made by Molenaar et al. (2015) in Rotliegend rocks from the Donnersberg formation.

The linear correlation analysis revealed a significant negative relationship between hydraulic and geochemical properties that fits to a polynomial regression (Fig. 14). It should be considered that the geochemical measurements cover a very different measurement area – represented by a spot of 1.2 cm diameter and around 0.5 cm penetration depth compared to the hydraulic measurements performed on an entire rock cylinder of 40 mm height and diameter. Additionally, instead of using highly-precise stationary X-ray fluorescence devices for measurements, a portable, faster device was used to efficiently derive spatial trends in the objects of investigation. This technique weakens the implications for absolute values, however, the trends observed in the measurements from the portable device are in good agreement with trends observed by stationary devices. Also, the observed geochemical characteristics are in well accordance with geochemical properties of quartz-rich sandstone varieties that were investigated in Bhatia (1983) or Baiyegunhi et al. (2017).

Geochemical analyses, in contrast to petrographic ones, limit the interpretations of geological processes because mineral phases can only be assumed and not determined for certain. A high mass fraction of Fe₂O₃ may imply that the rock is rich in iron-bearing minerals like clay minerals, hematite, magnetite, goethite, lepidocrite or ferrihydrite (Costabel et al., 2018), however, a precise classification of the mineral phase is not possible. Iron oxides are more common in secondary precipitates that usually form during eo- and mesodiagenesis (Pettijohn et al., 1987). The degrading impact of iron-oxide-rich coatings on permeability and porosity in unconsolidated sand and gravel has been shown in studies like Costabel et al. (2018). The amount of detrital iron-rich phases like hematite present in the rock matrix is typically less (Walker et al., 1981; Turner et al., 1995) compared to the secondary amount. In our case, however, thin section and ESEM analyses revealed that a high degree of intergranular matrix is still preserved, especially at the base of OSB2_c where high amounts of mud and mud intraclasts are incorporated from basal erosion. The small grain size of the matrix offers a great surface area for iron-oxide-rich precipitates, which might have enforced degradation of porosity and permeability additionally. Primary matrix typically plugs the pore throats of porous media, which reduces the ability to conduct fluids compared to matrix-free analogies. However, due to progressive compaction we cannot quantify for certain how large the amount of the primary matrix is compared to the pseudomatrix produced by plastic compaction of ductile, clay-rich grains and by feldspar dissolution.

A significant correlation between K_2O and Al_2O_3 could be detected. The spatial distribution resembles a network-like structure that might be either a product of diffusive mass transport during meso- or telodiagenesis or might reflect the distribution of feldspar grains and its residues in the sandstone. During feldspar alteration, SiO_2 gets dissolved and K remains in the alteration products, which could be an implication for the meso-scale network-like structure, in which pore fluids could have had migrated. This relationship is underlined by a negative, yet non-significant correlation of K_2O with SiO_2 .

5 Conclusions

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400 Significant, non-intuitive relationships between the physical and geochemical scalar fields at the lithofacies scale have been revealed with a deductive approach of spatial field modeling and statistical data analysis.

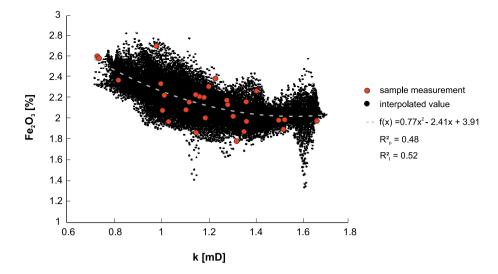


Figure 14. Regression analysis of the relationship between intrinsic permeability and mass fraction of Fe₂O₃ in the interpolated scalar fields of the rock cube OSB2_c. R_p^2 is the coefficient of determination for the plug measurements and R_i^2 is the coefficient of determination for the interpolated values.

1. As specific properties like the mass fraction of Fe₂O₃ preserve sedimentological textures well in their spatial distribution, other properties seem to be completely decoupled from depositional bounding surfaces. These scalar fields probably reflect processes that might have had taken place during diagenetical overprint of the rocks as a result of burial and exhumation. These processes produce diffuse patterns, like discussed with regard to the correlation of K₂O and Al₂O₃.

- 2. This study demonstrates that the observation of bedding structures does not necessarily indicate a stronger polar anisotropy compared to macroscopically unstructured lithologies. Here, the microscopic characteristics like the amount of secondary porosity might play a more important role in the attenuation of physical waves than the bounding surfaces.
- 3. It could be shown that hydraulic properties are dependent on the intergranular matrix and cement amount, which are in turn controlled by depositional processes and eogenetical precipitates. Those findings are not new (see Wilson and Pittman (1977) or Nordahl et al. (2014)), however, have not been evaluated in lithofacies-scale 3-D environments yet. We assume that primary matrix and ductile grain content has the most detrimental effect on rock permeability. Ductile grains were mechanically deformed during compaction leading to plugged pore throats. Feldspar dissolution has a highly productive effect on porosity but not on permeability.
- 4. We demonstrate that the strength of statistical correlation can be preserved in spatial interpolations as long as the sampling density is sufficient. If the sampling density is too low, a statistical correlation might be feigned inadvertently.

In the future work, we will examine in detail how the patterns observed in the major elements' mass fractions are connected to diagenetical processes. Therefore, it is planned to perform XRD analyses. In addition, we are planning to model the pore network within the rock cubes as a major controlling factor for diffusive mass transfer with the help of μ CT recordings.

420 Code and data availability. GeoReVi is an open-source software for Windows systems available under https://github.com/ApirsAL/GeoReVi.

The executables are available in the repository under https://github.com/ApirsAL/GeoReVi/blob/master/binaries/. The measurements are available under https://www.doi.org/10.6084/m9.figshare.11791407.v2

Sample availability. The investigated rock samples are available at the Institute of Applied Geosciences Darmstadt and can be requested under linsel@geo.tu-darmstadt.de. Also, the samples are registered in the System for Earth Sample Registration (SESAR, www.geosamples.org).

425 *Author contributions.* AL conceptualized and prepared the manuscript. AL and SW conducted the laboratory and field measurements. JH contributed to the conceptualization of the study. MH was the overall supervisor of the study.

Competing interests. The authors declare that they have no conflict of interest.

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