# **Multiscale analysis of living benthic foraminiferal heterogeneity: Ecological advances from an intertidal mudflat (Loire estuary, France)**

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#### **Abstract :**

An unprecedented sampling effort on the Loire estuary allowed a multi scale approach to identify parameters controlling density variations of benthic foraminifera. Indeed, the distances between the samples analysed for this study vary from 1 cm to hundreds of kilometres. To catch this range of distance variations, a model called Scale Variance Analysis was build describing the participation of each scale to the total observed variance. The SVA model requires, for each scale, the stability of relative variance. A comparison with the Moran's Index and experimental variogram is proposed showing coherent conclusions with the SVA analysis. The analysis shows that in order to maximize information on foraminiferal density variation, sampling campaigns should be designed with stations distant from few meters to 1 km, with a particular focus on the hectometre scale. A range of scale too rarely investigated in the community of benthic foraminifera ecology. Next, based on two intertidal mudflat stations separated of few hundred meters, the present study shows that for Ammonia tepida, the scale dependant preponderant parameters is the Chl a concentration in the top first centimetre. Contrastingly, the indicators of food quality such as the lability index and the oxygen penetration depth do not seem to affect A. tepida densities. This high quantity, low quality diet is interpreted as an opportunistic behaviour that is indirectly confirmed by a kinetic approach. This approach compares the deep infaunal microhabitat density with the shallow infaunal microhabitat density. The identical ratio indicates quick saturation of the available resources.

#### **Highlights**

▶ A new indicator to quantify spatial heterogeneity among different scale, the Scale Variance Analysis (SVA), is proposed. ► It revels higher heterogeneity of Ammonia tepida between 1 and 1000 m, in agreement with Moran's Index and experimental variogram. ► At the hectometer scale, direct correlation of *A. tepida* density with Chl a is observed. ► Equilibrium between deep and shallow infaunal population is rapidly reached for *A. tepida* probably due to rapid saturation of environmental capabilities.

### 41 **2 Introduction**

42

43 Benthic foraminifera are ubiquitous in sediment, giving them the potential to be bio-44 indicators of ecosystem functioning in all marine environments including transitional areas 45 such as intertidal mudflats (e.g. Debenay et al., 2001; Schönfeld et al., 2012). Moreover, their 46 distribution evolves monthly to seasonally (Alve and Murray, 2001; Kitazato et al., 2000) 47 which fits with the frequency of most survey sampling and smooths the influence of 48 environmental parameters that changes more rapidly (*e. g.* semi-diurnal tides). However, a 49 comprehensive framework of their spatial heterogeneity is still lacking, while it could be an 50 important step towards a standardization of the sampling strategy required for the 51 optimization and generalization of their use as bio-indicators (Schönfeld et al., 2012). 52 Additionally, assuming that a consequence has similar heterogeneity that its cause, multiscale 53 analysis of heterogeneity would improve our understanding of the preponderant parameters 54 controlling species distribution (*e. g.* Talley 2007).

55 Since the 50's, ecological studies take advantage of some mining engineer geostatistical 56 methods to express the spatial distribution variability using synthetic indices (Legendre and 57 Fortin, 1989). For example, the Moran's Index (Moran, 1950) has been applied on benthic 58 foraminifera by Hohenegger et al. (1993) and Thibault de Chanvalon et al. (2015) to identify 59 scales of patchiness. However, in the case of our study from the Loire estuary, to compare 60 samples acquired at different spatial resolutions, and with a different sampling size, scale 61 variance analysis is better-suited (Moellering and Tobler, 1972). In the Loire estuary, 62 Mojtahid et al. (2016) documented the spatial distribution patterns of living foraminifera at a 63 kilometric to decametric scale using a Van Veen grab sampler while Thibault de Chanvalon et 64 al. (2015) described foraminiferal distribution based on 1  $\text{cm}^3$  samples from the intertidal 65 mudflat "Les Brillantes". New data from 6 sites on the "Les Brillantes" mudflat distant from 66 few metres to hundreds of metre are here gathered with the Mojtahid et al. (2016) and the 67 Thibault de Chanvalon et al. (2015) datasets.

68 In the present paper, we will illustrate the role of such geostatistical tools to characterise the 69 preponderant factor controlling some species distribution following a three steps demarche 70 consisting in i) exploring, ii) identifying and iii) validating the causality relationship. In the 71 case of the *Ammonia tepida* density distribution in the Loire estuary, we will i) determine the 72 most significant scale to assess spatial foraminiferal density variation based on multiscale 73 analyses, in order to ii) identify the preponderant mechanism controlling foraminifera density 74 assuming it has a similar heterogeneity. This second step will be achieved looking at direct 75 correlation with foraminifera density. To validate this deterministic approach we will iii) test 76 the steady state between foraminifera density and the controlling factor based on vertical 77 distribution analysis. Hence, with depth, different microhabitat characterised by different 78 feeding time, exchange their population due to mobility or bioturbation. We will demonstrate 79 that equilibrium between the density of these microhabitats indicate steady state regime at all 80 depths.

## 82 **3 Materials and methods**

83 **3.1 Study area** 

84 "Les Brillantes" mudflat is located in the inner part of the Loire River estuary (Figure 1), 85 the outlet of a 117,045  $\text{km}^2$ -drainage basin composed of both sedimentary and granitic rocks. 86 The mean discharge of the Loire River is 900  $m^3s^{-1}$ , varying from 120  $m^3s^{-1}$  in summer to over  $5000 \text{ m}^3\text{s}^{-1}$  during winter flood, and leading to a high seasonal variability of water salinity at 88 "Les Brillantes". The Loire estuary is macrotidal and hyper-synchronous (Le Floch, 1961) 89 with a tidal range from 2 to 7 m, producing large intertidal areas ("Les Brillantes" is 1350 ha) 90 and important sediment resuspension. Therefore, sediment grain size characteristics at "Les 91 Brillantes" is quite homogeneous with silty-clay unimodal deposits with a median size of 92 between 10 and 20  $\mu$ m (Coynel et al., 2016). The 12 hours tidal cycling produces large daily 93 changes of salinity and structuration of water column, especially when it chimes with low 94 flow (that mostly varies seasonally, see Thibault de Chanvalon et al., 2016) and high tidal 95 intensity (whose main cycle lasts 2 weeks). Depending on the period of the year, the moment 96 of the lunar cycle and the time of the tidal cycle, salinity at the sampling point can vary from 97 35 to 0 and was about 20 in May 2013 and 14 in September 2012.

98 Two stations on the unvegetated slikke were chosen to study the spatial variability at the 99 metre scale in "Les Brillantes" mudflat (Figure 1): Site 1 is located 20 m offshore from a 1 m-100 high-eroded cliff while Site 2 is 500 m offshore. The main difference between the two stations 101 lies in the longer emersion time for site 1, the closest to the shore. According to Benyoucef 102 (2014), Site 1 is characterized by a denser microbiofilm (*i.e.* microphytobenthos composed of 103 diatoms).

104

#### 105 **3.2 Sampling strategies**

106 In this study, the estimation of metric heterogeneity for micro and meiofaunal (*i.e.* 107 foraminifera) composition is based on three replicate interface cores (triplicates cores) from 108 the same site, distant from each other by few meters. For all the other measured parameters 109 (oxygen profiles, macrofauna), a dedicated core was sampled from each site. A similar 110 vertical sampling resolution was used for all analyses *i. e.* cores with inner diameter of 8.2 111 cm, were sliced every 2 mm until 2 cm and every 5 mm until 5 cm with the exception of 112 macrofauna of which the abundance was determined for the full core depth. To minimise the 113 temporal variability, the foraminifera samples were acquired in September 2012 and May 114 2013, simultaneously to those from Mojtahid et al. (2016, sampled in September 2012) and 115 Thibault de Chanvalon et al., 2016.

#### 116 **3.3 Biological compartment**

117 The core triplicate dedicated to foraminifera was sliced few hours after recovery and 118 incubated overnight in Cell-Tracker™ Green (Invitrogen Detection Technologies) / 119 dimethylsulfoxide (DMSO) mixture (final concentration of 1 µmol  $L^{-1}$ ) then preserved in 10% 120 formaldehide / 3.8% borate mixture. This method was chosen for its accuracy at 121 discriminating living from dead foraminifera since it reacts with enzymes to produce a 122 fluorescent compound (see details in Bernhard et al., 2006). Only the larger (>150 µm) 123 fraction including adult specimens was conserved for identification (see detailed procedure in 124 Langlet et al., 2014). Then, only foraminifera fluorescing continuously and brightly under an 125 epifluorescent binocular (Olympus SZX12 with a fluorescent light source Olympus URFL-T) 126 were picked out, counted and determined. Note that, the species that we refer to as *Ammonia*  127 *tepida* in the following text corresponds to the phylotype T6 according to the recent 128 classification of *Ammonia* sp. (Richirt et al., 2019). This is a common *Ammonia* phylotype in 129 the European intertidal mudflats (Bird et al., 2020).

130 Foraminiferal oxygen uptake (FOU) is calculated with the equation (1), with  $R_i(T_{13})$  being 131 the respiration rate of the species i from laboratory measurement at 13 °C in pmol  $O_2$  ind<sup>-1</sup> d<sup>-1</sup> 132 from Geslin et al., 2011, the exponential being the Arrhenius temperature correction (with  $T_A$ ) 133 a constant in  ${}^{\circ}$ C determined by Bradshaw, 1961) and  $d_{i}^{\circ}$  the measured areal density of living 134 foraminifera in the oxic layer in ind  $m<sup>2</sup>$ .

FOU = 
$$
\sum_{i}
$$
 R<sub>i</sub>(T<sub>13</sub>) exp( $\frac{T_A}{T_{13}} - \frac{T_A}{T_{obs}}$ ) d<sub>i</sub><sup>0</sup> (1)

135

136 We calculated the average living depth  $ALD<sub>x</sub>$ , initially proposed by Jorissen et al. (1995) to 137 describe quantitatively the microhabitats distribution, following equation (2) :

$$
ALD_x = \frac{\sum_i n_i D_i}{\sum_i n_i} \tag{2}
$$

138 With  $n_i$  the number of specimens in interval i,  $D_i$  the midpoint of sample interval and x the 139 lower boundary of the deepest sample.

140 Another core triplicate was dedicated to microphytobenthos (MPB) and frozen *in situ* by 141 liquid nitrogen. Pigments extraction used a cold mixture (4°C) of 90% methanol/0.2M 142 ammonium acetate and 10% ethyl acetate (90/10 vol/vol) and measurement performed by 143 HPLC (see Méléder et al., 2005 for details). To assess organic matter quality, we used the 144 lability index,  $LI = Chl$  *a*/(Chl *a* + Pheo *a*), with Pheo *a* corresponding to the total amount of 145 phaeophorbides a and pheophytins a, respectively due to grazing and microbial activity (*e.g.* 146 Bianchi and Findlay, 1991; Cartaxana et al., 2003).

147 Finally, each core of the triplicate dedicated to the macro-invertebrates was homogenized 148 over its full depth (35cm) sieved at 1 mm and preserved in 4% formaldehyde before species 149 identification and counting in the >1 mm fraction.

#### 150 **3.4 Oxygen fluxes**

151 Dissolved oxygen vertical profiles were measured in a separate core in the dark, within few 152 hours after sampling using a Clark-type microelectrode with a 50 μm thick tip (OX50, 153 Unisense, Denmark) connected to a multimeter (Unisense) in a temperature controlled bath. 154 Twelve and 10 oxygen profiles were measured in September 2012 and 24 and 4 profiles in 155 May 2013 at stations S1 and S2 respectively. Diffusive  $O_2$  uptake (DOU) was estimated with 156 the PROFILE software by fitting the measured oxygen concentration with concentration from 157 diffusion-reaction models (see details in Berg et al., 1998).

#### 158 **3.5 Statistical analysis**

### 159 3.5.1 **Moran's Index**

160 Patchiness effect was explored using spatial correlograms built using the Moran's Index (I), 161 computed with R (package "spdep" following Bivand and Wong, 2018 and Fortin and Dale, 162 2005, Equation (3)). This index calculates the similarity of pair values for one neighbourhood 163 compared to the global mean of the dataset, a neighbourhood being defined by a weighted 164 (w<sub>d</sub>) function of the distance (l<sub>ij</sub>) between the pair values  $(x_i, x_j)$ . Here, we used a weighted 165 function sensitive only to the scale of the distance, *i. e.*:

$$
I(d) = \frac{\sum_{i}^{n} \sum_{j \neq i}^{n} w_d(l_{ij})(x_i - \bar{x})(x_j - \bar{x})}{\sum_{i}^{n} (x_i - \bar{x})^2} \times \frac{n_d}{\sum_{i,j}^{n} w_d(l_{ij})}
$$
(3)

$$
\mathbf{W}
$$

$$
W_{d}(l_{ij}) = \begin{cases} 1, & 10^{d} < l_{ij} < 10^{d+1} \\ 0, & \text{otherwise} \end{cases}
$$
 (4)

ith

166 and d, the scale of interest, n, the number of samples and  $n_d$  the number of samples forming 167 at least one pair. Significance of values is estimated based on Monte-Carlo analyses provided 168 in the "spdep" package (function moran.mc, done with 9999 simulations). This function 169 compares the I value obtained from the original dataset with a distribution produced by many

170 simulated I values. First, these simulated I values were obtained by random distribution of all 171 density values. Second, to take into account that in some case, very few sample formed pairs, 172 the simulated I values were obtained by exchanging randomly 10% of the samples forming 173 pairs with random samples from the dataset.

### 174 3.5.2 **Scale Variance Analysis**

175 Scale variance analysis (SVA) decomposes the total variance of a dataset to identify the 176 contribution of each scale to the variance (Moellering and Tobler, 1972; Wu et al., 2000). The 177 SVA compares each sample to a local mean which complements the Moran's Index, in which 178 samples are compared to the global mean. This approach requires *a priori* explicit definition 179 of scales of interest and *a priori* delimitation of all regions, necessarily nested over the 180 different scales. By convention, for a dataset hierarchized over k scales of interest, the scale 1 181 is the size of the initial samples, that are gathered in local regions belonging to the scale 2. 182 Then, the mean of each local regions is treated as sample of the scale 2 and are gathered again 183 in intermediate regions belonging to the scale 3. The process is repeated until the scale k, 184 covering the extent over which the sampling has been done. On each scale, the concept of 185 "scale variances" is introduced which corresponds to the variance of samples of scale h over a 186 region of scale h+1.

187 The following details are inspired from Moellering and Tobler (1972) but using different 188 writing. For samples of the scale h, gathered in regions belonging to the scale h+1, the scale 189 variance,  $V_i^{h\to h+1}$ , is defined according to the equation (5).

$$
V_i^{h \to h+1} = \frac{1}{n^h} \sum_{j=1}^{n^h} (x_i^{h+1} - x_{i,j}^h)^2
$$
 (5)

190 With  $x_i^{h+1}$ , the mean value of all samples nested in the group i;  $x_{i,j}^h$  the different sample 191 (whose size is belonging to the scale h) value constituting the group i and  $n<sup>h</sup>$  the number of 192 sample  $x_{i,j}^h$  constituent the group i. For simplicity, we here assume that n<sup>h</sup> does not depend of 193 i, *i.e.* all groups of size belonging to the scale h+1, are constituted by the same number of 194 sample from the scale h. Then, for being representative of the importance of the variance of a 195 certain scale (h) over the whole dataset one need to look at the mean of the scale variances of 196 the scale h, according to equation (6 and 7)

$$
\overline{V^{h \to h+1}} = \frac{1}{N^{h+1}} \sum_{i=1}^{N^{h+1}} V_i^{h \to h+1}
$$
 (6)

w

 $i=1$  $N^{h+1} = \prod n^p$  $\frac{\mathbf{k}}{2}$  $p=n+1$ (7)

ith

197 Equation (7) describes that  $N^{h+1}$  is equal to the number of group whose size belonging to 198 the scale h+1. Thus, one can demonstrate (see Appendice 1) that for a dataset hierarchized 199 over k scale of interest, the variance (VAR) can be decomposed into the sum of the mean of 200 the scale variances (equation (8)):

$$
VAR = V^{1 \to k} = \sum_{h=1}^{k-1} \overline{V^{h \to h+1}}
$$
\n(8)



202 However, SVA requires a complete dataset with all values of scale 1 totally enumerated 203 (Moellering and Tobler, 1972). In our case, the data range over 8 orders of magnitude from 204 samples at the cm scale to a sampling area of hundreds of kilometres. Assuming 5 samples per 205 group, an exhaustive sampling would require  $5^7 = 78.125$  analyses. To overcome the 206 analytical limitation to produce so many analyses, a supplementary assumption is required: 207 the scale stability of relative variance. For each scale, this assumption assumes that all groups 208 i are characterized by the same relative scale variance  $(RV^{h\text{-}2h+1})$ . The  $RV^{h\text{-}2h+1}$  is defined as 209 the square of the relative standard deviation *i. e.* for any i:

$$
RV^{h \to h+1} = \frac{V_i^{h \to h+1}}{(x_i^{h+1})^2}
$$
 (9)

210 This is a heavy assumption but it allows the calculation of any mean scale variance as soon 211 as both the relative scale variance and the sum of the square of the mean of all higher scale are 212 known (see Appendice 2) according to equation (10):

For any 
$$
h \le k-2
$$
 
$$
\overline{V^{h\to h+1}} = \frac{RV^{h\to h+1}}{N^k} \prod_{i=h+1}^{k-1} (1 + RV^{i\to i+1}) \sum_{j=1}^{N^k} (x_j^k)^2
$$
 (1)

213 This relation indicates that the means of the scale variances for a certain scale can be 214 calculated as soon as the relative scale variance is known. Lastly, this relation indicates that 215 direct comparison of the relative scale variance from different scales of the same dataset is not 216 meaningful and that comparison of the mean scale variance has to be preferred.

#### 217 3.5.3 **Experimental variogram**

218 To complement Moran's Index, that compares samples to the global mean, and SVA, that 219 compares samples to a local mean, an experimental variogram was built. In this case, each 220 pair of samples is compared to the square of their difference without referring to any external 221 mean. Then, the gamma value  $(\gamma)$  is computed as the half of the mean of the values that 222 belong to a certain distance, according to equation (11).

$$
\gamma(d) = \sum_{i}^{n} \sum_{j \neq i}^{n} w_d(l_{ij}) (x_i - x_j)^2 \times \frac{1}{2 \sum_{i,j}^{n} w_d(l_{ij})}
$$
(11)

w

 $w_d(l_{ij}) = \begin{cases} 1, & 10^d < l_{ij} < 10^{d+1} \\ 0, & \text{otherwise} \end{cases}$ 0, otherwise

ith

223

#### 224 3.5.4 **Application to the Loire estuary dataset**

225 Moran's index, SVA and experimental variogramare calculated based on the average 226 density of *A. tepida* in the first centimetre depth measured in this study, in Thibault de 227 Chanvalon et al. (2015) and in the study of Mojtahid et al. (2016). This combined dataset is 228 represented in Figure 1 and is not regularly distributed in the Loire area. For example, for the 229 Moran's index calculation, each class of distance covering scales from centimetre to 230 hectometre (hundreds of meters) are represented by less than 72 sample pairs. The scale 231 variance analysis (SVA) was calculated based on regions delimited arbitrary by the black 232 lines on the Figure 1. The column "available sampling / exhaustive sampling" in Table 1 233 summarized the number of regions per scale and compared it with an exhaustive sampling as 234 theoretically requested. The scale levels 3, 4 and 5 does not contain enough samples to be 235 gathered into at least one group. Thus, the mean of the scale variance for scale 3, 4 and 5 236 could not be calculated directly and was estimated by the difference between the global 237 calculated variance and all the means of the scale variance. Moreover, all information for 238 close (< 100 m) samples belong to or are close to stations 1 and 2. Nevertheless, such 239 limitations are very frequent in foraminiferal dataset and the spatial recovery obtained with 240 this combined dataset is rare in the literature motivating the pursuit of the spatial 241 heterogeneity analysis.

#### 242 **3.6 Model of microhabitat equilibrium**

243 The relation between deep infaunal and shallow infaunal foraminiferal faunas is modelled 244 using a dynamic 2 boxes-model (whose equations are detailed in Figure 2A) based on typical 245 assumptions drawn from ecological studies (Levin, 1976). The shallow infaunal box is 246 characterised by a population (pop<sub>sh</sub>), a first order mortality rate ( $k_{d,sh}$ ) and a reproduction rate 247 described with the Verhulst equation, that is, a first order rate  $(k_p)$  decreasing to zero as the 248 population saturates (popsat) available resources. The deep infaunal box is characterised by a 249 population (pop<sub>de</sub>) and a first order mortality rate ( $k_{d,de}$ ). The transfer between the 2 boxes 250 follows a first order rate  $(k_{\text{ech}})$ , roughly estimating biomixing. The ratio of deep over shallow 251 infaunal population,  $\alpha = pop_{de}$  / pop<sub>sh</sub>, predicted by this model after an important increase of 252 environment capability (by 500 fold in this example) is shown in Figure 2B. First the shallow 253 population increase, hence  $\alpha$  decrease. After a short delay, deep infaunal population increase 254 too, leading to an equilibrium between exponentially growing shallow and deep infaunal 255 population and  $\alpha$  reaching a plateau ( $\alpha_{exp}$ , on Figure 2B). The higher k<sub>ech</sub> is, the faster the first 256 plateau is reached. Finally, once the population reaches the limits of the environment 257 capabilities another equilibrium is observed between the two populations that is characterised 258 by a second plateau ( $\alpha_{sat}$  on Figure 2B). The higher k<sub>p</sub> is, the faster the second plateau is 259 reached, other parameters having much less influence on the rate of α changes. The α value at 260 the plateau is defined as:

$$
\alpha_{\text{sat}} = \frac{k_{\text{ech}}}{k_{\text{ech}} + k_{\text{d\_de}}}
$$
 (1)

$$
\alpha_{\rm exp} = \begin{cases} 1 - \frac{\rm b}{2} & \text{if } b \ll 2\\ \frac{\rm 1}{\rm b} & \text{if } b \gg 2 \end{cases}, \text{ with } b = \frac{\rm k_p + k_{d\_de} - k_{d\_sh}}{\rm k_{ech}}
$$

261

### 263 **4 Results**

264 **4.1 Environmental parameters** 

265 Table 2 summarizes most of the environmental parameters extracted from public survey 266 databases (banque HYDRO, SYVEL and SHOM networks, see glossary) and previous 267 publications (Benyoucef, 2014; Thibault de Chanvalon et al., 2016).

268 The two campaigns were characterized by contrasted river discharge (150 m<sup>3</sup> s<sup>-1</sup> in 269 September versus 1200 m<sup>3</sup> s<sup>-1</sup> in May), organic carbon content in the top of sediment (2.1 % 270 versus 2.8 %), salinity (22 versus 8, respectively) and temperature (17 °C versus 13 °C). 271 Contrastingly, nutrients showed mainly spatial variation with a higher concentration of 272 dissolved phosphorus at Station 1, closer to the shore, with 15.4 and 14.2  $\mu$ mol L<sup>-1</sup> for 273 September 2012 and May 2013 respectively, compared to 2.7 and 4.8  $\mu$ mol L<sup>-1</sup> for Station 2 274 (Thibault de Chanvalon et al., 2016). Finally, the oxygen penetration showed both important 275 spatial and temporal variability with a lower value in May 2013.

#### 276 **4.2 Biological parameters**

277 Table 3 indicates that the three studied biological compartments differ significantly 278 between the two stations and less clearly between the two campaigns. Station 1 shows higher 279 abundances of microphytobenthos (the average of the two campaign for chlorophyll A is 340 280 mg m<sup>-2</sup> in Station 1 versus 180 mg m<sup>-2</sup> in Station 2), macrofauna (770 ind m<sup>-2</sup> in Station 1 281 versus 290 ind m<sup>-2</sup> in Station 2 on average) and living foraminifera (78 ind  $/ 10 \text{ cm}^3$  versus 24 282 ind / 10 cm<sup>3</sup>) than Station 2. While *Ammonia tepida* (>70%) dominates foraminiferal 283 communities in both stations*,* macrofaunal assemblages switch from a dominance of the 284 polychaetes *Hediste diversicolor* (>75%) at Station 1 to a dominance of both the bivalve 285 *Scrobicularia plana* and the polychaetes *Heteromastus filiformis* at Station 2. The main 286 seasonal variation visible in both stations is for macrofauna with an increase of *H. filiformis* 287 associated to a decrease of *S. plana* in May 2013. However, in Station 1, but only there, 288 foraminifer density decreases significantly in May 2013 with, for example, a near 289 disappearance of *H. germanica* going from 36 ind  $/ 10 \text{ cm}^3$  down to 2 ind  $/ 10 \text{ cm}^3$  in the first 290 top centimetre. The lability index (LI), that is higher than 0.9 in both stations and seasons, 291 indicates important *in situ* autotrophic activity of microphytobenthos.

### 292 **4.3 Vertical distribution**

293 The densities of benthic foraminifera and Chl a concentration presented in Figure 3 show in 294 several cases an exponential decrease of densities with depth. The high density of *H.*  295 germanica in the top 2 millimetres in September 2012 at Station 1 (67 ind  $/10 \text{ cm}^3$ ) appears 296 concomitantly to particularly high densities of *A. tepida* (829 ind /10 cm<sup>3</sup>) and Chl a (201 mg  $297 \text{ m}^{-2}$ ). However, in detail, the exponential decrease associated to this high density appears more 298 progressive than for *A. tepida* and Chl a, with a minimum reached at 1.4 mm depth for *H.*  299 *germanica* versus 0.6 mm depth for the others.

### 300 **4.4 Foraminifera aerobic respiration rates**

301 The respiration rates (RR) estimated for the foraminiferal population for each season and 302 station as well as the relative foraminiferal contribution to DOU are shown in Table 4. The 303 estimated respiration rates are  $3357 \pm 117$  pmol O<sub>2</sub>.ind<sup>-1</sup>.d<sup>-1</sup> and  $2154 \pm 75$  pmol O<sub>2</sub>.ind<sup>-1</sup>.d<sup>-1</sup>  $f(304)$  for *A. tepida* and  $685 \pm 134$  pmol O<sub>2</sub>.ind<sup>-1</sup>.d<sup>-1</sup> and  $439 \pm 86$  pmol O<sub>2</sub>.ind<sup>-1</sup>.d<sup>-1</sup> for *H. germanica* 305 at respectively 17°C (September) and 13°C (May). The maximal relative contribution of the 306 foraminiferal fauna to DOU was 2.3%, at station S1 in September 2012 mostly carried by the 307 dense population of *A. tepida*. For all other samplings, foraminiferal respiration rates (sum of 308 *A. tepida* and *H. germanica* respiration rates) are much lower and varies from 0.085 to 0.099 309 mmol  $O_2$  m<sup>-2</sup> d<sup>-1</sup>.

#### 310 **4.5 Statistical analysis**

311 Figure 4A shows that the Moran's index is significantly higher than zero for all scales 312 below 0.1 km, which indicates that foraminiferal densities are grouped into patches of 313 hundreds of metre's size. The negative value for Moran's index between 1 and 10 km (Figure 314 4A) indicates that most of the difference between environments occurs between 1 and 10 km 315 for *A. tepida* densities. SVA results presented in Figure 4B (black dots) show that most of the 316 variance (average scale variance > 10%) comes from the scales between 1 m and 1 km but the 317 lack of data (see Table 1) prevents us from a better accuracy. Interestingly, scale variance 318 analysis quantifies that scales between 1 cm and 1 m counting for 3.6 times less to the overall 319 variance than the scales between 1 m and 1 km. Stability of SVA, even based on our sparse 320 dataset, is illustrated by the white dots of Figure 4B. It shows that the SVA processed without 321 the particularly dense Station 1 does not modify significantly the results. The Figure 4C 322 shows the square differences of paired samples in grey dots. The variogram calculated for 323 each distance range is particularly high between 100 m and 1 km.

### 324 **5 Discussion**

### 325 **5.1 Critical scale of heterogeneity identified by multiscale analysis:**

326 In "Les Brillantes" mudflat, our data show an overall high surface foraminiferal density (up 327 to 829 ind.  $/ 10 \text{ cm}^3$ ) and very low diversity (only two different species identified) (Figure 3). 328 High surface density (over 100 ind.  $/ 10 \text{ cm}^3$ ) of foraminiferal fauna is commonly reported in 329 intertidal mudflat surfaces from estuaries (Debenay et al., 2006; Thibault de Chanvalon et al., 330 2015) or inlets (Alve and Murray, 1994; Cesbron et al., 2016; Goldstein et al., 1995) while a 331 very low diversity is more typical of macrotidal estuaries. Indeed, only the species that are the 332 most tolerant to large daily salinity variations can grow in macrotidal estuaries (Murray, 333 2006), especially on non-vegetated mud. These later are for instance known for their absence 334 of agglutinated species (Berkeley et al., 2008).

335 Debenay and Guillou (2002) demonstrated that the estuarine compatible species colonize 336 successive areas along the salinity gradient. However, they did not identify the preponderant 337 forcing among all the parameters covarying with salinity. For example, in the Loire estuary, 338 the regular dredging of the navigation channel has also been invoked to explain the extreme 339 poverty of the diversity of the foraminifera fauna with only three living species reported over 340 the whole salinity gradient (Mojtahid et al., 2016). In addition to the constrains specific to 341 estuarine environments, more common forcing such as grain size, food availability or food 342 quality would also modify foraminiferal growth opportunities and produce, *in fine*, an 343 irregular surface density distribution such as that illustrated in Figure 1. In an attempt to catch 344 such a variability, deterministic models (*e.g.* TROX model from Jorissen et al., 1995) build on 345 predefined forcing, estimate compliance between a species and an environment while they 346 hardly quantify the density variations. This issue is particularly critical in dynamic 347 environments where kinetic effects, such as new colony settling (Alve, 1999; Weinmann and

348 Goldstein, 2017) may induce changes in hydrodynamic dispersal or hysteresis associated to 349 transient environmental changes, prevail over saturation of the environment capabilities. A 350 complementary approach used in ecological survey, based on geostatistical models (*e.g.* 351 Talley, 2007) proposes at first, to synthetize spatial patterns in order to infer causality as a 352 second step. From the three different geostatistical models chosen for this study (Moran's 353 Index, SVA and experimental variogram, Figure 4) one common picture appears: most of the 354 density variation comes from the scales between the metre scale, that gathers all paired 355 samples distant from 1 to 10 m and the hectometre scale,that gathers all paired samples distant 356 from 100 to 1000 m. The Moran's Index (Figure 4A) underlines particularly the hectometre 357 scale where the Index decreases and crosses the zero line, changing from a distribution with 358 almost similar densities (Moran Index above 0) to a distribution with contrasted or random 359 densities (Moran Index value below or equal to zero, Figure 4A). The variogram plot (Figure 360 4C) confirms the preponderant role of the hectometre scale with the highest gamma value 361 calculated. Sadly, the lack of paired samples distant from 1 m to 100 m could hide unexpected 362 changes and prevents us from being more precise about the most significant scale. Identical 363 limitation is visible for the SVA that equally distributes the missing variance into the scales 364 with missing paired samples (from metre to hectometre scale). However, SVA model predicts 365 preponderant role of at least one of these scales by the difference between the overall variance 366 to the variance attributed to the other scales. At all events, in order to maximize information 367 on foraminiferal density spatial distribution, we recommend designing future sampling 368 campaigns with stations distant from 1 metre to 1 kilometre, with a particular focus on the 369 hectometre scale.

370 The geostatistical models are apparently in contradiction with the importance habitats 371 succession along salinity gradient (Debenay and Guillou, 2002) since the SVA attributes only 372 0.9 % of the total variance to the scale of salinity changes (over 100 km), the minimum 373 attributed to any scale. Qualitative analysis solves this discrepancy. For example, a qualitative 374 SVA coded with values equal to 0 when no *Ammonia tepida* is observed, to 1 when *A. tepida* 375 is a minor species (<10%) and equal to 2 when *A. tepida* is a major species (>10%) leads to 376 drastically increase the importance of the estuarine scale (black diamonds on the Figure 4B). 377 In the qualitative analysis, the scale over 10 km produces 41% of the total variance and 378 therefore in strong agreement with the importance of the salinity gradient. Taken together, 379 these results highlight the efficiency of deterministic models for qualitative predictions, 380 understood as the order of magnitude of foraminiferal population densities and their lacks to 381 quantify predictions. Geostatistical models represent promising tools to cross this gap 382 especially when performed in combination with deterministic models. For example, scale 383 analyses can by hyphened with studies of environmental processes in order to associate one 384 preponderant process to each scale of important variation. This exercise is proposed in the 385 following discussion.

#### 386 **5.2 Limiting factors at the Les Brillantes mudflat scale**

387 The focus on Les Brillantes mudflat allows investigation of processes explaining density 388 variations over few hundreds of meters, a critical scale identified from the geostatistic models. 389 On one hand, our results show that the two stations present very few qualitative differences, 390 *i.e.* changes over order of magnitudes - the most significant being *H. germanica* in September 391 2012 (Table 3) with high density in Station 1 probably produced by optimal conditions for 392 development of propagules and/or reproduction during September weak riverine influence. 393 This sensitivity tends to position *H. germanica* downstream from *A. tepida* in the estuarine 394 succession as observed by Debenay et al., 2006 and Mojtahid et al., 2016 while Alve and 395 Murray, 1994, Debenay et al., 2000 and Debenay and Guillou (2002) observations state for 396 the opposite. On the second hand, quantitative differences between stations are observed on 397 every variable with 1.5 to 5 fold more abundance of microphytobenthos, meiofauna and 398 macrofauna at Station 1 (Table 3) and up to 5 fold faster respiration (Table 4). The longer 399 emersion time, hence the longer light time exposure, and the higher nutrient input, probably 400 streaming for the grazing land of the shore via a small channel (Table 2) might favour primary 401 production compared to Station 2 and consequently may support higher density of fauna.

402 The exponential relation observed between Chl a and *A. tepida* (Figure 5A) indicates a 403 possible deterministic relation between primary production and *A. tepida* at the hectometre 404 scale and makes Chl a a good limiting factor for deterministic models at this scale. However, 405 such a relation owes a lot to the opportunistic character of *A. tepida*, understood as the ability 406 for a species to saturate rapidly the capabilities of an environment. Prolonging this 407 interpretation, we can estimate that all parameters varying differently than Chl a have 408 negligible effect on *A. tepida* density. Surprisingly, the co-varying parameters LI and the 409 OPD, reputed to trace organic matter lability, evolved differently underlying the specific diet 410 regime of *A. tepida.* Indeed, this species is known in the literature for being carnivorous 411 (Dupuy et al., 2010), predating on metazoan classes (Chronopoulou et al., 2019) and thus may 412 ignore variation of primary production quality. Oppositely, the literature indicates that *H.*  413 *germanica* feeds mostly on diatoms notably to steal their chloroplast (Pillet et al., 2011; 414 Cesbron et al., 2017; Jauffrais et al., 2018; LeKieffre et al., 2018). It seems that this so-called 415 "kleptoplasty" specialisation turns into a disadvantage when facing opportunistic species in 416 low quality high quantity food environments.

#### 418 **5.3 Vertical distribution of foraminifera: Microhabitat vs Bioturbation**

#### 419 5.3.1 **Biomixing and chemotaxis forcing**

420 The fine vertical sampling resolution (Figure 3) allows a precise description of the typical 421 exponential vertical decrease of shallow infaunal microhabitat (Buzas et al., 1993). The very 422 shallow density maximum indicates a favourable environment, supposedly a reproduction 423 layer and/or propagule spawning event, due to high oxygen concentration and/or fresh organic 424 matter (Berkeley et al., 2007; de Stigter et al., 1999; Geslin et al., 2004). The progressive 425 decrease with depth is usually associated with the biomixing produced by macrofauna 426 bioturbation (*e.g*. Alve and Bernhard, 1995; Saffert and Thomas, 1998; Thibault de 427 Chanvalon et al., 2015), a predation-related strategy (De Stigter et al., 1998; Loubere, 1989) 428 or the occurrence of oxygen oases around animal burrows (Goldstein et al., 1995; Steineck 429 and Bergstein, 1979). However, the very steep decrease of *A. tepida* (minimum reached at 0.8 430 cm depth) and the systematic slight re-increase at depth (except Station 2 in May 2013, Figure 431 3) producing a shallow minimum density, corresponds to a specific pattern, likely produced 432 by the combination of biomixing and chemotaxis (BC model, Thibault de Chanvalon et al., 433 2015). In this BC model, when buried close enough to the surface, the foraminifera detect the 434 oxygenated layer and move back to the surface while, when buried deeper than their 435 pseudopod length, the foraminifera are trapped at depth in a dormancy stasis, as observed by 436 LeKieffre et al. (2017). The shallow minimum density corresponds to the chemotaxis range of 437 the foraminifera. Oases model has been discarded in these stations because of the absence of 438 correlation at the centimetre scale between deep living foraminifera and burrow traces 439 (Thibault de Chanvalon et al., 2015).

440 In September 2012, when surface densities were high enough, *H. germanica* densities 441 presented a similar pattern than *A. tepida* but with a less steep decrease since the shallow 442 minimum is reached at 1.4 cm depth for *H. germanica* versus 0.6 mm for *A. tepida* (Figure 3).

443 On the line with the BC model, we interpret this observation as a wider chemotaxis range for 444 *H. germanica*, maybe related to its pseudopod length, which feels necessary to move back to 445 the surface once buried deeper than *A. tepida*. However, this difference could also come from 446 *H. germanica* kleptoplasty, as proposed by Cesbron et al. (2017) in order to interpret similar 447 observations. In this case, *H. germanica* would be less sensitive to oxygen depletion and 448 tolerates being buried deeper before moving back to the surface.

#### 449 5.3.2 **Deep and shallow infaunal comparison**

450 Based on the BC model, specimens' behaviour depends on their position compared to the 451 shallow minimum, with upper specimens being more active and especially able to reproduce 452 and growth while lower specimens are probably in dormancy stasis. When taking into account 453 density results from the two sampling stations in September and in May, a linear positive 454 correlation between these two populations appears (Figure 5B). It is a significant result as the 455 line crosses the origin with a high  $R^2$  value (0.95) for *A. tepida*. The relation for *H. germanica* 456 is less convincing since the lower observed population induces higher uncertainties. However, 457 for *A. tepida*, this relevance is highlighted by other biological parameters measured in this 458 study such as Chl a (Figure 5B) that does not follow any linear relation.

459 The model of microhabitat equilibrium (§3.6 and Figure 2) details explicitly, despite 460 evident oversimplifications, how the shallow and deep infaunal population interaction can be 461 describe by their ratio, so called α, which depends of the intrinsic species dynamics 462 (propagule spawning / reproduction and mortality rates), biomixing rate and the delay since 463 the last change of available resources. After a certain point, function of biomixing and 464 reproduction rate, the model shows that the time does not influence the  $\alpha$  value anymore, slow 465 growing species being in exponential growth while faster growing species, such as 466 opportunists, having already saturated the environment capabilities. The constancy of the  $\alpha$ 467 value found for *A. tepida* ( $\alpha = 1.1$ , Figure 5B) despite a 5 fold change of density indicates that 468 during each campaign, depth repartition of foraminifera population has reached an 469 equilibrium since the last change of available resources, more likely a saturation equilibrium. 470 Hence biomixing is fast relatively to foraminifera resource changes. This is not the case for 471 Chl a probably because its main resource (available light) changes too fast compared to 472 biomixing events while foraminifera populations average short term variations of food 473 availability. Moreover, equation (11) indicates that an  $\alpha$  about 1 indicates high biomixing rate 474 compared to mortality in anoxia, a first step to estimate biomixing rate using foraminifera 475 vertical distribution. Taken together, analysis of vertical distribution confirms the steady state 476 reached between surface resources and *A. tepida* density at surface and at depth and the 477 importance of Chl a concentration at Les Brillantes. While the bioturbation intensity was 478 expected to be a supplementary depth cause due to mortality increase at depth, this effect was 479 not found to be significant since the highest density of foraminifera (including *H. germanica*) 480 matches with the highest density of macrofauna (figure 5A).

## 481 **6 Conclusion**

482 Because of the several extreme conditions characterizing intertidal mudflat habitats, 483 amongst which we can cite the risk of burial in anoxic sediments and the large daily salinity 484 variation, the two species observed in "Les Brillantes" mudflat developed contrasting skills. 485 *H. germanica* suffers from freshwater conditions during river flood periods but seems to get 486 longer-range chemotaxis to face anoxia while *A. tepida* appears to be much less sensitive to 487 freshwater inputs and favours dormancy as a strategy to overcome burial into anoxic depths. 488 These differences could come from their different feeding strategies,*H. germanica* having a 489 more specific diet while *A. tepida,* feeding from different sources, is emancipated from 490 primary producer dependency and shows an opportunistic behaviour.

491 Geostatistic models confirm the effectiveness of average salinity to describe qualitatively 492 the habitats distribution. However, they indicate that the density of foraminifera in these 493 habitats are controlled by other parameters, such as Chl a, that varies over distance from 1 m 494 and 1 km. These distances are often underrepresented in publications looking at foraminiferal 495 heterogeneity and require supplementary investigations to state about their importance. Thus, 496 we recommend to future models to fit geostatistical and deterministic approaches, for 497 example, by associating a particular preponderant mechanism to each scale characterized by 498 high heterogeneity.

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## 508 **8 Glossary**

509 SYVEL (Surveillance system of the Loire estuary) network maintains 6 high frequency 510 stations between Nantes and Paimboeuf for physicochemical parameters of subsurface waters 511 (temperature, salinity, dissolved oxygen concentration, and turbidity). Founded by the region 512 Pays de la Loire.

513 SHOM is a french national military service for marine and coastal geographic information.

# 514 Banque HYDRO is a national public gathering hub of river flows, alimented mainly by 515 numerous national services.

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# 676 **10Tables**

### 677 **10.1 Table 1: Scale Variance Analysis**



678 \*Number of samples of order h required to constitute a sample of order h+1, † calculated after exclusion of S1 from the dataset

### 680 **10.2 Table 2 : Geochemical parameters**



- $*$ Most of the profiles show bioturbation  $\frac{1}{2}$  measured at Mont-Jean sur Loire (banque
- 682 HYDRO)<sup>2</sup>GIP Loire <sup>3</sup>SHOM, <sup>4</sup>from Benyoucef (2014) <sup>5</sup>from Thibault de Chanvalon et al.
- 683 (2016)

684

# 685 **10.3 Table 3: Biological parameters**



### 686 **10.4 Table 4: Respiration rate calculation**



# **11Figures**

**11.1 Figure 1: Bathymetry of the Loire estuary with surface density of** *A. tepida* **from Mojtahid et al. (2016) sampled in September 2012, Thibault de Chanvalon et al. (2015) sampled in May 2013 and this study (both May 2013 and September 2012). Black lines indicate regions used in the scale variance analysis for the scale 7 (A1 to A7) and the scale 6 (B1 to B7). Bottom right insert focus on the "Les Brillantes" mudflat (Map produced on R using leaflet package, bathymetry from the SHOM).** 



# 694 **11.2 Figure 2: A. Deterministic model to explain these relations based on biomixing**  695 **and chemotaxis forcing.**

696 vprod, vdeath, vsh->de and vde->sh correspond to the rate of reproduction, the rate of death, the 697 rate of exchange from shallow to deep microhabitat and the rate of exchange from deep to 698 shallow microhabitat respectively.  $k_p$ ,  $k_d$ <sub>sh</sub>,  $k_{d-de}$  and  $k_{ech}$  are the associated parameters while 699 popsh and popde are the population of shallow and deep microhabitat. B. Example of a 700 representative result from the model following a 500 fold increase of environment capability , 701 *i. e.* 500 fold pop<sub>sat</sub> increase.



702

# **11.3 Figure 3: Vertical distribution in the sediment column of living benthic foraminifera and Chlorophyll a.**



# 709 **11.4 Figure 4: Geostatistical model processes with the dataset from Figure 1 (black**  710 **dots).**

711 Error bars in Figure A corresponds to twice the standard deviation of the distribution 712 obtained when the samples are randomly distributed (number of simulation is 104; star 713 attribution is based on p-value;  $\leq 0.01***$ ;  $\leq 0.05**$ ;  $\leq 0.1*$ ). In Figure B open circle 714 represents the scale variance analysis (SVA) after exclusion of S1 from the dataset (open 715 circle) and using qualitatively transformed dataset (black diamonds, see text for details). In 716 the experimental variogram (Figure C), grey dots correspond to the square of the difference of 717 each possible pairs, plotted against their distance. The black dots correspond to the mean and 718 the error bars to a third of the standard deviation for each scale of distance between samples.



**11.5 Figure 5: A. Evolution of biological parameters in Les Brillantes according to Chl a (data from Table 2). B. Relation between deep infaunal and shallow infaunal population of** *A. tepida* **and** *H. germanica* **in Les Brillantes.** 



## 726 **12 Appendices**

# 727 **12.1 Appendix 1: Demonstration that the variance of a dataset hierarchized over k**  728 **scale is equal to the sum of the mean scale variance of each scale.**

729 Based on the definition of the variance and on the formalism previously described, 730 samples' value are the  $x_i^1$  (they belong to the scale 1) and the global mean is  $x^k$ , we have :

$$
VAR = V^{1 \to k} = \frac{1}{N^1} \sum_{i=1}^{N^1} (x_i^1 - x^k)^2
$$

$$
V^{1 \to k} = \frac{1}{N^1} \left( \sum_{i=1}^{N^1} (x_i^1)^2 - N^1 (x^k)^2 \right)
$$

The samples can be gathered in  $n^{k-1}$  group whose extension belong to the scale k-1. There is 732 N<sup>1</sup> / n<sup>k-1</sup> samples per group and the  $x_j^{k-1}$  are the means of each groups and the  $x_i^1$  are now 733 written as  $x_{i,j}^1$ , with the subscript j indicates to which groups the sample belong.

$$
V^{1\to k} = \frac{1}{N^1} \left( \sum_{i=1}^{N^1} (x_{i,j}^1)^2 - \sum_{j=1}^{n^{k-1}} \frac{N^1}{n^{k-1}} (x_j^{k-1})^2 + \sum_{j=1}^{n^{k-1}} \frac{N^1}{n^{k-1}} (x_j^{k-1})^2 - N^1 (x^k)^2 \right)
$$
  
\n
$$
V^{1\to k} = \frac{1}{N^1} \sum_{j=1}^{n^{k-1}} \left( \sum_{i=1}^{n^{k-1}} (x_{i,j}^1)^2 - \frac{N^1}{n^{k-1}} (x_j^{k-1})^2 \right) + \frac{1}{n^{k-1}} \left( \sum_{j=1}^{n^{k-1}} (x_j^{k-1})^2 - n^{k-1} (x^k)^2 \right)
$$
  
\n
$$
V^{1\to k} = \frac{1}{N^1} \sum_{j=1}^{n^{k-1}} \frac{N^1}{n^{k-1}} V_j^{1\to k-1} + V^{k-1\to k}
$$
  
\n
$$
V^{1\to k} = \frac{1}{n^{k-1}} \sum_{j=1}^{n^{k-1}} V_j^{1\to k-1} + V^{k-1\to k}
$$

The property expressed by A3 is true also for any j and p,  $V_j^{1\to k-p}$ , especially

$$
V_j^{1 \to k-1} = \frac{1}{n^{k-2}} \sum_{l=1}^{n^{k-2}} V_{j,l}^{1 \to k-2} + V_j^{k-2 \to k-1}
$$

735 Once you inject A4 into A3 you get,

$$
V^{1\to k} = \frac{1}{n^{k-1}} \frac{1}{n^{k-2}} \sum_{j=1}^{n^{k-1}} \sum_{l=1}^{n^{k-2}} V_{j,l}^{1\to k-2} + \frac{1}{n^{k-1}} \sum_{j=1}^{n^{k-1}} V_j^{k-2\to k-1} + V^{k-1\to k}
$$

736 Then using the property expressed by A3 for p =2 you get

$$
V^{1\to k} = \frac{1}{n^{k-1}} \frac{1}{n^{k-2}} \sum_{j=1}^{n^{k-1}} \sum_{l=1}^{n^{k-2}} \sum_{m=1}^{n^{k-3}} V_{j,l,m}^{1\to k-3}
$$
  
+ 
$$
\frac{1}{n^{k-1}} \frac{1}{n^{k-2}} \sum_{j=1}^{n^{k-1}} \sum_{l=1}^{n^{k-2}} V_{j,l}^{k-3\to k-2} + \frac{1}{n^{k-1}} \sum_{j=1}^{n^{k-1}} V_j^{k-2\to k-1} + V^{k-1\to k}
$$

737 Repeating this for all p until  $p=k-2$ 

$$
V^{1 \to k} = \frac{1}{N^2} \sum_{j=1}^{n^{k-1}} V_j^{1 \to 2} + \dots + \frac{1}{N^{k-2}} \sum_{j=1}^{N^{k-2}} V_j^{k-3 \to k-2} + \frac{1}{N^{k-1}} \sum_{j=1}^{N^{k-1}} V_j^{k-2 \to k-1} + V^{k-1 \to k}
$$
 A71  
 
$$
VAR = V^{1 \to k} = \sum_{h=1}^{k-1} \overline{V^{h \to h+1}}
$$

738

### 739 **12.2 Appendix 2: Demonstration of the expression of the mean of the scale variance**

### 740 **as a function of the relative scale variance**

741 Based on the relative scale variance definition, for any i:

$$
RV^{h \to h+1} = \frac{1}{(x^{h+1})^2} \left(\frac{1}{n^h} \sum_{i=1}^{n^h} (x_i^h)^2 - (x^{h+1})^2\right)
$$
 B1

742 Once reorganized, we get:

$$
\sum_{i=1}^{n^h} (x_i^h)^2 = (1 + RV^{h \to h+1}) n^h (x^{h+1})^2
$$

743 By sum over all the j group from the scale h+1,

$$
\sum_{j=1}^{N^{h+1}} \sum_{i=1}^{n^h} (x_i^h)^2 = \sum_{j=1}^{N^{h+1}} (1 + RV^{h \to h+1}) n^h (x^{h+1})^2
$$
  

$$
\sum_{i=1}^{N^h} (x_i^h)^2 = (1 + RV^{h \to h+1}) n^h \sum_{j=1}^{N^{h+1}} (x_j^{h+1})^2
$$
 B2

- 744 The equation B2 express the sum of the square at the scale h, as a function of the sum of the
- 745 square at the scale h+1. Injecting p times the equation B2 into itself, we get:

$$
\sum_{i=1}^{N^h} (x_i^h)^2 = \prod_{i=h}^{h+p-1} (1 + RV^{i \to i+1}) n^i \sum_{j=1}^{N^{h+p}} (x_j^{h+p})^2
$$
 B3

<sup>*i*=1</sup><sup>*i*=h</sup> *j*=1<br>
746 On another hand, taking the definition of the mean scale variance, we have:

$$
\overline{V^{h \to h+1}} = \frac{1}{N^{h+1}} \sum_{i=1}^{N^{h+1}} (x_i^{h+1})^2 R V^{h \to h+1}
$$

747 Injecting B3 into B4, with p=k-1-h,

$$
\overline{V^{h\to h+1}} = \frac{RV^{h\to h+1}}{N^{h+1}} \prod_{i=h+1}^{k-1} (1 + RV^{i\to i+1}) n^i \sum_{j=1}^{N^k} (x_j^k)^2
$$

$$
\overline{V^{h\to h+1}} = \frac{RV^{h\to h+1}}{N^k} \prod_{i=h+1}^{k-1} (1 + RV^{i\to i+1}) \sum_{j=1}^{N^k} (x_j^k)^2
$$

748

749