

1 **Comment on “*The role of scaling laws in upscaling*”**

2 **by *B.D. Wood***

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6 In his recent article, Wood (2009) proposes a thorough and very interesting analysis of
7 the process of coarse-graining in complex subsurface hydrologic systems. No doubt this
8 article will eventually help elucidate a number of complicated aspects of upscaling.
9 Nevertheless, one of Wood’s (2009) key results is somewhat surprising. It implies that “the
10 ability to develop a coarse-grained equation for describing transport in heterogeneous
11 porous media has much less to do with the particulars of the mathematical methods used for
12 averaging than it does the assumptions (scaling laws) adopted when developing the upscaled
13 model” (p. 735). Specifically, Wood (2009) asserts that “averaging *per se* does not lead to a
14 reduction in the information content of any particular problem”. I would advocate that
15 exactly the opposite is true, namely that the averaging process eliminates way too much
16 information, and that a significant challenge in years to come will be to find ways to volume
17 average while preserving some essential information about microscale spatial complexity.

18 One of the key features of the volume averaging formula used by Wood (2009, eq. (8)),
19 which for the microscale solute concentration $c(\mathbf{y},t)$ is

20
$$\langle c \rangle_{x,t} = \frac{1}{V} \int_{y \in V_x} c(\mathbf{y},t) d\mathbf{y} \quad (1)$$

1 where $\langle c \rangle_{\mathbf{x},t}$ is the average (macroscopic) value of solute concentration at (\mathbf{x},t) , and V is the
 2 volume of the domain $V_{\mathbf{x}}$ over which the averaging is performed, is that it is designed to
 3 cause considerable information loss. Indeed, once the integral of eq. (1) is carried out, all the
 4 detailed information associated with the microscale concentration $c(\mathbf{y},t)$ effectively
 5 disappears. Only if $\langle c \rangle_{\mathbf{x},t}$ is known rigorously at every single point \mathbf{x} of the medium, and if $V_{\mathbf{x}}$
 6 is strictly invariant in space and time, can eq. (1) be deconvoluted to reclaim the original
 7 information. However, this is never possible in practice.

8 Under these conditions, since volume averaging in and of itself clearly leads to massive
 9 information loss, where did Wood's conclusion come from? It seems to stem from several
 10 assumptions made in the analysis. The first is the introduction of a particular form of
 11 "deviation concentration", defined as the difference between the microscale and average
 12 concentrations at the point where the macroscale concentration is evaluated (Wood's eq.
 13 (11)):

$$14 \quad c(x,t) = \langle c \rangle_{\mathbf{x},t} + \delta c(\mathbf{x},t) \quad (2)$$

15 Other types of deviation concentrations or local fluctuations could have been used instead,
 16 such as the "fluctuating components" of Efendiev et al. (2000, eq. (4a)), leading to different,
 17 coarse-grained equations that are not strictly continuous, but rather discretized.

18 The second assumption introduced by Wood (2009) is to consider that the microscale
 19 convection-dispersion equation

$$20 \quad \frac{\partial c}{\partial t} = \nabla \cdot (\underline{\underline{D}} \cdot \nabla c) - \nabla \cdot (\mathbf{v}c) \quad (3)$$

1 where $\underline{\underline{D}}$ is the microscale dispersion tensor and \mathbf{v} is the microscale pore water velocity, is
 2 an acceptable starting point for the derivation of macroscale transport equations, via
 3 upscaling. Several other perspectives could have been adopted, such as starting from
 4 traditional conservation equations (of mass, momentum, or energy) at the microscale,
 5 volume averaging them, using averaging theorems (e.g., Baveye and Sposito, 1984, 1985) to
 6 simplify the volume-averaged equations, and introducing constitutive relations at the
 7 macroscopic scale.

8 A third assumption in Wood's analysis is that constitutive relations at the macroscale are
 9 considered in connection not with a volume-averaged form of eq. (3), but with the so-called
 10 "deviation equation"

$$11 \quad \frac{\partial \tilde{c}}{\partial t} - \nabla \cdot (\underline{\underline{D}} \cdot \nabla \tilde{c}) + \nabla \cdot (\mathbf{v} \tilde{c}) = \nabla \cdot (\tilde{\mathbf{f}}_1 + \tilde{\mathbf{f}}_2 + \mathbf{I}) \quad (4)$$

12 where

$$13 \quad \tilde{\mathbf{f}}_1 = [\underline{\underline{D}} \cdot \nabla \langle c \rangle - \langle \underline{\underline{D}} \cdot \nabla \langle c \rangle \rangle] \quad (5)$$

$$14 \quad \tilde{\mathbf{f}}_2 = -[\mathbf{v} \langle c \rangle - \langle \mathbf{v} \langle c \rangle \rangle] \quad (6)$$

$$15 \quad \mathbf{I} = -[\langle \underline{\underline{D}} \cdot \nabla \delta \phi \rangle - \langle \mathbf{v} \delta \phi \rangle] \quad (7)$$

16 The deviation expression in eq. (4) is obtained by replacing c by eq. (2) in eq. (3), volume
 17 averaging the resulting equation and subtracting eq. (3). As ingenious as this manipulation is,
 18 again alternative approaches could have been considered, leading to different intermediate
 19 transport equations, not necessarily involving variables that cannot be directly measured, as
 20 eq. (4) does. These different equations would require different heuristic constitutive

1 relationships to end up with the same macroscale equation, eventually. Nevertheless, like
2 Wood's (2009) perspective, the various possible approaches along those lines would all be
3 equally defensible. As always in continuum mechanics, and particularly in the continuum
4 mechanics of transport processes in polyphasic systems, the ultimate test of a particular
5 heuristic construct is whether it is able to derive empirically-obtained transport expressions.

6 To the extent that it involves the local deviation \tilde{c} at every point in the medium, eq. (4)
7 formally encompasses the same information as eq. (3). However, clearly, this is in spite of
8 the massive information loss caused by volume averaging, and results directly from the
9 additional assumptions made by Wood (2009), which in effect re-introduce local microscale
10 information in the picture. So, the statement that "averaging *per se* does not lead to a
11 reduction in the information content of any particular problem" does not represent what
12 really happens.

13 This discussion about the loss of information associated with the integration in eq. (1) has
14 more than theoretical interest. In a number of different contexts, including the ecology of
15 soil microorganisms (Balser et al., 2006) or the metagenomic analysis of soils (Baveye, 2009),
16 it has significant practical implications. Fundamentally, the link with practice is due to the
17 fact that eq. (1) mimics the measurement process in soils (Baveye and Sposito, 1984, 1985).
18 A neutron moisture meter, a gamma-ray densitometer, or a time-domain reflectometer
19 probe, located at a specific point \mathbf{x} in a soil at a time t , are all affected by the position of
20 myriads of atoms present around \mathbf{x} , but they manage to condense the information associated
21 with the spatial coordinates of these atoms into a single number, either the volumetric water
22 content or the bulk density of the soil at (\mathbf{x}, t) . The phenomenal loss of information that

1 results characterizes every measurement made in soils at this stage, even in the case of the
2 minute (micron-sized) voxels delineated by computed tomography (e.g., Baveye et al., 2002).

3 In recent years, it has become clear that the volume averaging (or convolution product
4 with instrumental response functions) carried out by measuring instruments, loses
5 information that, in some situations, is crucially needed to describe soil processes,
6 particularly (but not exclusively) those involving microorganisms. A vivid example of such a
7 situation is the microscopic spatial distribution of Cu in vineyard soils in Burgundy
8 (Jacobson et al., 2007). In these soils, bulk Cu concentrations reach levels of hundreds of
9 parts per million that should strongly inhibit microbial growth and metabolism. Yet,
10 microorganisms thrive. The key to this apparently contradictory observation, revealed by
11 synchrotron-based X-ray fluorescence spectroscopy, is that Cu distribution is extremely
12 heterogeneous at the micron scale, with “hotspots” in the vicinity of which microorganisms
13 are likely to have a hard time colonizing the soil, and with relatively large portions of the soil
14 that are devoid of Cu. This heterogeneity cannot be captured by classical bulk (macroscopic)
15 Cu concentrations. At a minimum, beyond a simple volume average, some measure of the
16 spatial dispersion of Cu and of microorganisms is needed, as well as a parameter quantifying
17 the spatial disconnect between them.

18 Garnier et al. (2009) introduced such a parameter in their model of straw biodegradation
19 in soils, and showed that with this parameter, which they termed a “contact factor”, a
20 satisfactory fit of their model to experimental data could be obtained. At this stage, the
21 model of Garnier et al. (2009) is empirical and still has to be established on a firm theoretical
22 basis. It might be possible to tweak Wood’s (2009) upscaling scheme slightly, i.e., introduce
23 different constitutive relations, to produce a macroscale convection-dispersion equation

1 containing a disconnect parameter *à la* Garnier et al. (2009). Such an approach, despite its
2 heuristic character, would be very enlightening, in that it would allow us to determine what
3 information, lost during the classical volume averaging process, is required to describe things
4 correctly. Reintroducing all the microscale information, in this context, would definitely be
5 an overkill, but it is not clear what minimum set of data is absolutely needed and in what
6 form.

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