

Stability, Emergence and Part-Whole-Reduction

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1 Introduction

Often we can describe the macroscopic behaviour of systems without knowing much about the nature of the constituents of the systems let alone the states the constituents are in. Thus, we can describe the behaviour of real or ideal gases without knowing the exact velocities or places of the constituents. It suffices to know certain macroscopic quantities in order to determine other macroscopic quantities. Furthermore, the macroscopic regularities are often quite simple. Macroscopic quantities are often determined by only a few other macroscopic quantities. This fact is quite remarkable, as Jerry Fodor noted:

Damn near everything we know about the world suggests that unimaginably complicated to-ings and fro-ings of bits and pieces at the extreme micro-level manage somehow to converge on stable macro-level properties. [...] [T]he ‘somehow’, really is entirely mysterious [...] why there should be (how there could be) macro level regularities at all in a world where, by common consent, macro level stabilities have to supervene on a buzzing, blooming confusion of micro level interactions. (Fodor 1997, 161)

The puzzle is that on the one hand macroscopic behaviour supervenes on the behaviour of the constituents, i.e. there is no change in the macroscopic behaviour without some change at the micro-level. On the other hand not every change in the states of the constituents leads to a change on the macro-level. To some extent the macro-behaviour is independent of what is going on at the micro-level. The question we will address in this paper is whether there is an explanation for the fact that as Fodor put it the micro-level “converges on stable macro-level properties”.

Various metaphors have been used to describe this peculiar relation of the micro-level to the macro-level. The macro-behaviour has been said to be “autonomous” (Fodor 1997), it has been characterized as “insensitive to microscopics” or due to “higher organizing principles in nature” (Laughlin & Pines 2000, 261) and above all macroscopic behaviour is said to be “emergent”, where emergence implies novelty, unexplainability and irreducibility. There is no consensus about how best to define the concept of emergence — partly due to the fact that there is no uncontroversial set of paradigmatic examples and partly due to the fact the concept of emergence plays a role in a variety of philosophical and scientific contexts. (For a discussion see Humphreys & Bedau 2008, Introduction)

As a consequence of these conceptual difficulties we will ignore the notion of emergence for the time being and will instead stick with the notion of the *stability* of macro-phenomena. One obvious advantage of this terminological choice consists in the fact that stability as opposed to emergence allows for degrees. Even though there is no precise concept of emergence it seems uncontroversial that whether or not some behaviour falls under this concept should be an all or nothing affair. The very word “stability” by contrast allows for behaviour to be more or less stable.

Even though the concept of stability has been used in various different senses within the philosophy of science recently, there is a clear conceptual core: Stability is a property we attribute to entities (things, properties, behaviour, sets, laws etc) if the entity in question does not change even though other entities, that are specified, do change. This definition captures the notions of stability that have been recently been discussed. Woodward and Mitchell call laws or causal generalizations “stable” relative to background conditions if they continue to hold, even though the background conditions change (Mitchell 2003, 140; Woodward 2007, 77). Lange calls a set of statements stable if their truth-value would remain the same under certain counterfactual changes (Lange 2009, 29). What we are interested in in this paper is the stability of the behaviour of macro-systems vis-à-vis changes with respect to the micro-structure of the system. So the changes we consider concern not external or background conditions but rather system-internal features.

An explanation for the stability of macro-phenomena is not only an interesting project in itself but also of wider significance for other areas of philosophy. It is apparent that Fodor’s seminal papers on the autonomy of the special sciences is largely motivated by the observation that macro-behaviour is stable under some changes on the micro-level. This in turn motivated the position of ‘non-reductive’ physicalism in the philosophy of mind. Even though what we will finally present is not a non-reductive but rather a reductive account of stability, this account nevertheless vindicates what we take to be two of the core-intuitions that motivate non-reductive physicalism: (1) There are macro-level properties that are distinct from micro-level properties. (2) Macro-level properties are dependent on micro-level properties. (See Rudder-Baker 2009, 110 for a similar characterisation of non-reductive physicalism that does not refer explicitly to the failure of reduction). Even though we will not discuss this point in any detail we would like to suggest that it is the stability of macro-behaviour that makes distinct macro-properties possible.

Fodor’s question, as to how it is possible that extremely complex behaviour at an atomistic level could (for many macroscopic systems we know of) give rise to stable macroscopic behaviour has in recent years been transformed into a wider and more ambitious question, viz. how it is possible that even microscopically very different systems manage to exhibit the same type of stable macroscopic behaviour. In a still wider context it has been discussed by Batterman, most notably in his paper “Multi-realizability and Universality” (2000) as well as (Batterman 2002). His central claim is that the notion of universality used to describe the surprising degree of insensitivity of critical phenomena in the vicinity of continuous phase transition points can also be used to explain how special science properties can be realized by quite heterogeneous systems. Indeed, the renormalization group (RNG) explanation of the universality of critical phenomena was advocated as a paradigm that could account for the multi-realizability of macro-behaviour in general. While we agree that RNG provides an explanation of a certain kind of stability of macro-behaviour, we wish to point out that the case of critical phenomena is very special, in that it is restricted to phenomena that obtain in the vicinity of 2nd order phase transitions, but not elsewhere. Batterman himself is aware of this problem (See Batterman 2011). A similar remark pertains to Morrison’s recent paper on emergence (Morrison 2012). Her account of emergence and stability in terms of symmetry breaking pertains only to a restricted class of those systems that exhibit stable behaviour.

For a more general explanation of the stability of macro-phenomena of individual systems, but also across classes of similar systems, other principles must therefore be invoked. In this paper we propose to rationalise the stability of macro-behaviour by pointing out that observations of macro-behaviour are usually observations on anthropomorphic scales, meaning that they are results of coarse-grained observations in both space and time. That is, they involve averages, both over individual behaviours of huge numbers of atomistic components constituting the system under study, and averages of their behaviour over time scales, which are very large compared to characteristic time-scales associated with motion at the micro-level. In fact we shall point out that large numbers of constituent atomistic components, beyond their role in observations on anthropomorphic scales, are also a prerequisite for the very existence and stability of ordered states of matter, such as crystals or magnetic materials.

Just to give an impression of orders of magnitude involved in descriptions of macroscopic amounts of matter, consider a cubic-millimetre of a gas at ambient temperature and pressure. It already contains approximately $2.7 \cdot 10^{16}$ gas molecules, and the distance between constituents particles is so small that the typical time between two successive scattering events with other molecules would for each of the molecules be of the order of 10^{-10} seconds, entailing that equilibration times in such systems are very short.

Indeed, the traditional way of taking the time-averaging aspect associated with observations of macro-behaviour into account has been to consider results of such observations to be properly captured by values characteristic of thermodynamic equilibrium, or — taking a microscopic point of view — by equilibrium statistical mechanics. It is this microscopic point of view, which holds that a probabilistic description of macroscopic systems using methods of Boltzmann-Gibbs statistical mechanics is essentially correct that is going to form a cornerstone of our reasoning. Indeed, within a description of macroscopic systems in terms of equilibrium statistical mechanics it is essential that systems consist of a vast number of constituents in order to exhibit stable, non-fluctuating macroscopic properties and to react in predictable ways to external forces and fields. In order for stable ordered states of matter such as crystals, magnetic materials, or super-conductors to exist, numbers must (in a sense we will specify below) in fact be so large as to be “indistinguishable from infinity”.

Renormalization group ideas will still feature prominently in our reasoning, though somewhat unusually in this context with an emphasis on the description of behaviour away from criticality.

We will begin by illustrating our main points with a small simulation study of a magnetic system; the simulation clearly shows that an increase of the system size leads to reduced fluctuations in macroscopic properties, and thus exhibits a clear trend towards increasing stability of macroscopic magnetic order (section 2). We then go on to describe the mathematical foundation of the observed regularities in the form of limit theorems of mathematical statistics for independent variables, using a line of reasoning originally due to Jona-Lasinio (1975), which relates limit theorems with key features of large-scale descriptions of these systems (section 3). Generalizing to coarse-grained descriptions of systems of interacting particle systems, we are lead to consider RNG ideas of the form used in statistical mechanics to analyse critical phenomena. However, for the purpose of the present discussion we shall be mainly interested in conclusions the RNG approach allows to draw about system behaviour away from criticality (section 4). We will briefly discuss the issue how the finite size of actual systems affects our argument (section 5). We will furthermore discuss to what extent an explanation of stability is a reductive explanation (section 6) and will finally draw attention to some interesting conclusions about the very project of modelling condensed matter systems (section 7).

2 Evidence from simulation: large numbers and stability

Let us then begin with a look at evidence from a simulation, suggesting that macroscopic systems — which according to our premise are adequately characterised as stochastic systems — need to contain large numbers of constituent particles to exhibit stable macroscopic properties.

This requirement, here formulated in colloquial terms, has a precise meaning in the context of a description in terms of Boltzmann-Gibbs statistical mechanics. Finite systems, are according to such a description, ergodic. They would therefore attain all possible microstates with probabilities given by their Boltzmann-Gibbs equilibrium distribution. and would therefore in general *also* exhibit fluctuating *macroscopic* properties as long as the number of constituent particles remains finite. Moreover ordered phases of matter, such as phases with non-zero magnetization or phases with crystalline order would not be absolutely stable, if system sizes were finite: for finite systems, ordered phases will always also have a finite life-time (entailing that ordered phases in finite systems are *not* stable — in a strict sense. However, life-times of ordered states of matter *can* diverge in the limit of infinite system size.

Although real world systems are clearly *finite*, the numbers of constituent particles they contain are surely unimaginably large (recall numbers for a small volume of gas mentioned above), and it is the fact that they are so very large which is responsible for the fact that fluctuations of their macroscopic properties are virtually undetectable. Moreover, in the case of systems with coexisting phases showing different forms of macroscopic order (such as crystals or magnetic systems with different possible orientations of their macroscopic magnetisation), large numbers of constituent particles are also responsible for the fact that transitions between different manifestations of macroscopic order are sufficiently rare to ensure the stability of the various different forms of order.

We are going to illustrate what we have just described using a small simulation of a magnetic model-system. The system we shall be looking at is a so-called Ising ferro-magnet on a 3D cubic lattice. Macroscopic magnetic properties in such a system appear as average over microscopic magnetic moments attached to ‘elementary magnets’ called spins, each of them capable of two opposite orientations in space. These orientations can be thought of as parallel or anti-parallel to one of the crystalline axes. Model-systems of this kind have been demonstrated to capture magnetic properties of certain anisotropic crystals extremely well.

Denoting by $s_i(t) = \pm 1$ the two possible states of the i -th spins at time t , and by $\mathbf{s}(t)$ the configuration of *all* $s_i(t)$, one finds the macroscopic magnetisation of a system consisting of N such spins to be given by the *average*

$$S_N(\mathbf{s}(t)) = \frac{1}{N} \sum_{i=1}^N s_i(t) . \quad (1)$$

In the model system considered here, a stochastic dynamics at the microscopic level is realised via a probabilistic time-evolution of the following form: Every spin experiences a ‘local field’

$$h_i(t) = \sum_{j=1}^N J_{ij} s_j(t) \quad (2)$$

generated by all other spins; the interaction constants J_{ij} denote the relative size of the contribution of $s_j(t)$ to the local field $h_i(t)$ of the i -th spin. The spins determine their new state (in random order) following a probabilistic rule of the form

$$\text{Prob}\{s_i(t + \Delta t) = \pm 1\} = \frac{1}{2} [1 \pm \tanh(\beta h_i(t))] . \quad (3)$$

The parameter β in this expression is inversely proportional to the absolute temperature T of the system; units can be chosen such that $\beta = 1/T$. The parameter β determines the degree of stochasticity of the dynamics: the larger β , (the lower the temperature), the larger the probability that $s_i(t + \Delta t)$ takes the value given by the direction (the sign) of the local field $h_i(t)$. One can easily convince oneself that positive interaction constants, $J_{ij} > 0$, encourage parallel orientation of spins, i.e. tendency to macroscopic ferromagnetic order.

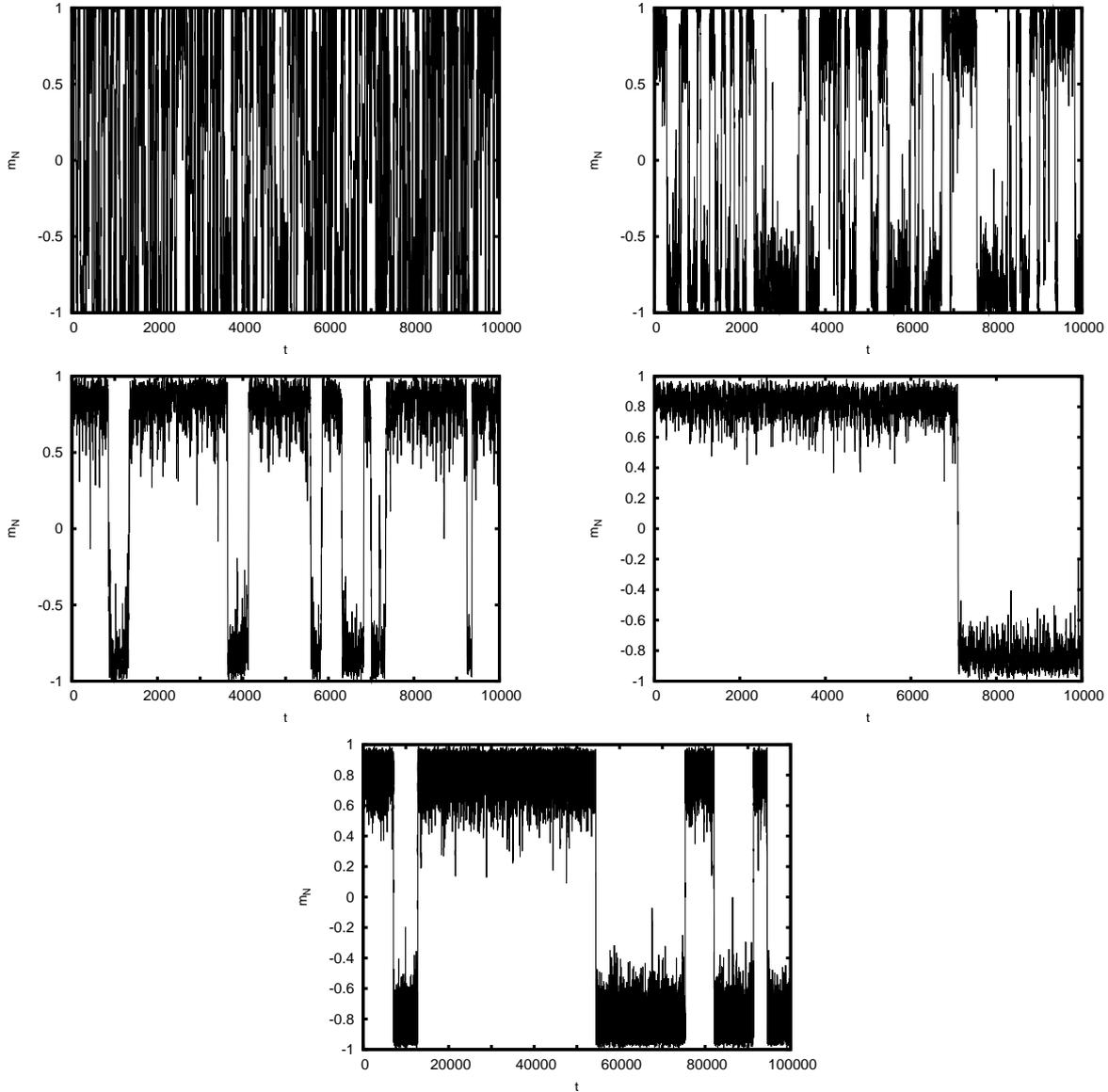


Figure 1: 1st row: magnetisation of a system of $N = 2^3$ spins (left) and $N = 3^3$ spins (right); 2nd row: magnetisation of a system of $N = 4^3$ spins (left) and $N = 5^3$ spins (right), monitored for 10000 time-steps; bottom panel: magnetisation of a system of $N = 5^3$ spins, now monitored for 10^5 time-steps. The temperature T in these simulations is chosen as $T = 3.75$, leading to an equilibrium magnetization $\simeq 0.828$ in the thermodynamic limit.

Independently of initial conditions, the stochastic dynamics (3) will approach a thermodynamic equilibrium state, described by a Gibbs-Boltzmann equilibrium distribution of

micro-states

$$P(\mathbf{s}) = \frac{1}{Z_N} \exp \left[-\beta H_N(\mathbf{s}) \right] \quad (4)$$

corresponding to the “energy function”

$$H_N(\mathbf{s}) = - \sum_{(i,j)} J_{ij} s_i s_j . \quad (5)$$

The double sum in this expression is over all possible pairs (i, j) of spins. In general, one expects the coupling strengths J_{ij} to decrease as a function of the distance between spins i and j , and that they will tend to be negligibly small (possibly zero) at distances larger than a maximum range of the magnetic interaction. In the present paper we assume that interactions are non-zero only for spins on neighbouring lattice sites.

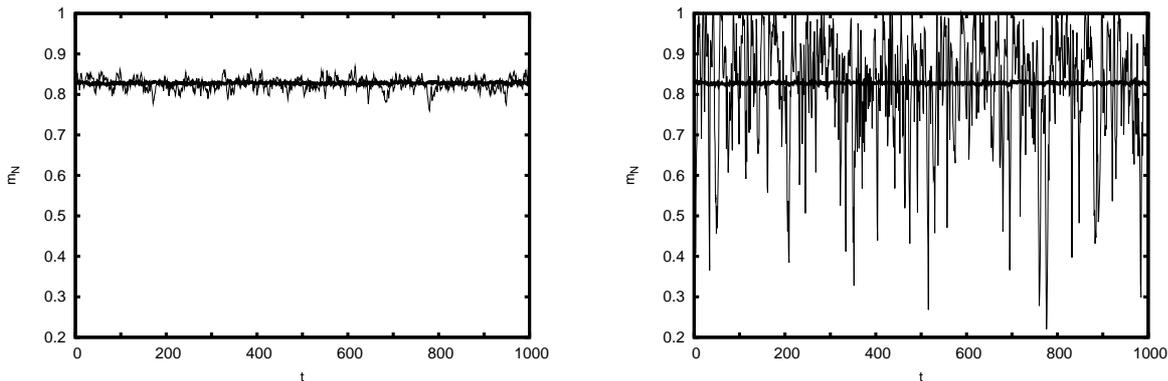


Figure 2: Magnetisation of systems containing $N = 16^3$ and $N = 64^3$ spins showing that fluctuations of the average magnetization of the system decreases with system size (left panel), and of a system containing $N=64^3$ spins, shown together with the magnetization of a smaller subsystem of this system, containing only $N_s = 3^3$ spins (right panel).

Fig. 1 shows the magnetisation (1) as a function of time for various small systems. For the purposes of this figure, the magnetisation shown is already averaged over a time unit.¹

The first panel of the figure demonstrates that a system consisting of $N = 2^3 = 8$ spins does not exhibit a stable value of its magnetisation. Increasing the number of spins $N = 3^3 = 27$ that the system appears to prefer values of its magnetisation around $S_N(\mathbf{s}(t)) \simeq \pm 0.83$. However, transitions between opposite orientations of the magnetization are still very rapid. Increasing numbers of spins further to $N = 4^3 = 64$ and $N = 5^3 = 125$, one notes that transitions between states of opposite magnetization become rarer, and that fluctuations close to the two values $S_N(\mathbf{s}(t)) \simeq \pm 0.83$ decrease with increasing system size. Only one transition is observed for the larger system during the first 10^4 time-steps. However, following the time-evolution of the magnetisation of the $N = 5^3 = 125$ system over a larger time span, which in real-time would correspond to approximately 10^{-7} seconds, as shown in the bottom panel, one still observes several transitions between the two orientations of the magnetization, on average about 10^{-8} seconds apart.

The simulation may be repeated for larger and larger system sizes; results for the magnetization (1) are shown in the left panel of Fig 2 for systems containing $N = 16^3$ and 64^3

¹The time unit in these simulations is given by the time span during which every single spin has on average once been selected for an update of its state according to (3). It is this unit of time which is comparable for systems of different sizes (Binder and Stauffer 1987, 1-36); it would correspond to a time-span of approximately 10^{-12} seconds in conventional units.

spins. The Figure shows that fluctuations of the magnetisation become smaller as the system size is increased. A second effect is that the time spans over which a stable (in the sense of non-switching) magnetisation is observed increases with increasing system size; this is shown for the smaller systems in Fig 1. Indeed, for a system containing $N = 64^3 = 262144$ spins, transitions between states of opposite magnetization are already so rare² that they are out of reach of computational resources available to us, though fluctuations of the magnetization about its average value are still discernible. Note in passing that fluctuations of average properties of a *small subsystem* do *not* decrease if the total system size is increased, and that for the system under consideration they are much larger than those of the entire system as shown in the right panel of Fig 2.

The present example system was set up in a way that states with magnetisations $S_N \simeq \pm 0.83$ would be its only (two) distinct macroscopic manifestations. The simulation shows that transitions between them are observed at irregular time intervals for small finite systems. This is the macroscopic manifestation of ergodicity. The time span over which a given macroscopic manifestation is stable is seen to increase with system size. It will diverge — ergodicity can be broken — only in the infinitely large system. However, the times over which magnetic properties are stable increase so rapidly with system size that it will be far exceeding the age of the universe for systems consisting of realistically large (though certainly finite) numbers of particles, i.e. for $N = \mathcal{O}(10^{23})$.

Only in the infinite system limit would a system exhibit a strictly constant non-fluctuating magnetisation, and only in this limit would one therefore, strictly speaking, be permitted to talk of a system with a given value of its magnetisation. Moreover, only in this limit would transitions between different macroscopic phases be entirely suppressed and only in this limit could the system therefore, strictly speaking, be regarded as macroscopically stable.

However our simulation already indicates that both properties can be effectively attained in finite (albeit large) systems. The systems just need to be large enough for fluctuations of their macroscopic properties to become sufficiently small as to be practically undetectable at the precision with which these are normally measured, and life-times of different macroscopic phases (if any) must become sufficiently large to exceed all anthropomorphic time scales by sufficiently many orders of magnitude. With respect to this latter aspect of macroscopic stability, reaching times which exceed the age of the universe could certainly be regarded as sufficient for all purposes; these are easily attained in systems of anthropomorphic dimension.

In what follows we describe how the suppression of fluctuations of macroscopic quantities in large systems can be understood in terms of statistical limit theorems. We follow a reasoning originally due to Jona-Lasinio (1975) that links these to the coarse grained descriptions and renormalization group ideas, starting in the following section with systems of independent identically distributed random variables, and generalizing in Sect 4 thereafter to systems of interacting degrees of freedom.

3 Limit Theorems and Description on Large Scales

Large numbers are according to our reasoning a prerequisite for stability of macroscopic material properties, and only in the limit of large numbers we may expect that macroscopic properties of matter are also non-fluctuating. Early formulations of equations of state of macroscopic systems which postulate deterministic functional relations, e.g. between temperature, density and pressure of a gas, or temperature, magnetisation and magnetic field in magnetic systems can therefore claim strict validity only in the infinite system limit. They

²Measured in conventional time units, such a system will exhibit a magnetisation which remains stable for times of the order of several years.

are thus seen to presuppose this limit, though in most cases, it seems, implicitly.

From a mathematical perspective, there are — as already stressed by Khinchin in his classical treatise on the mathematical foundations of statistical mechanics (Khinchin 1949) — two limit theorems of probability theory, which are of particular relevance for the phenomena just described: (i) the law of large numbers, according to which a normalized sum of N identically distributed random variables of the form (1) will in the limit $N \rightarrow \infty$ converge to the expectation $\mu = \langle s_k \rangle$ of the s_k , assuming that the latter is finite; (ii) the central limit theorem according to which the distribution of deviations from the expectation, i.e., the distribution of $S_N - \mu$ for independent identically distributed random variables will in the limit of large N converge to a Gaussian of variance σ^2/N , where $\sigma^2 = \langle s_k^2 \rangle - \langle s_k \rangle^2$ denotes the variance of the s_k .³

The central limit theorem in particular implies that fluctuations of macroscopic quantities of the form (1) will in the limit of large N typically decrease as $1/\sqrt{N}$ with system size N . This result, initially formulated for independent random variables may be extended to so-called weakly dependent variables. Considering the squared deviation $(S_N - \langle S_N \rangle)^2$, one obtains its expectation value

$$\langle (S_N - \langle S_N \rangle)^2 \rangle = \frac{1}{N^2} \sum_{k,\ell=1}^N C_{k,\ell} = \frac{1}{N^2} \sum_{k,\ell=1}^N \langle (s_k - \langle s_k \rangle)(x_\ell - \langle x_\ell \rangle) \rangle \quad (6)$$

and the desired extension would hold for all systems, for which the correlations $C_{k,\ell}$ are decreasing sufficiently rapidly with “distance” $|k - \ell|$ to ensure that the sums $\sum_{\ell=1}^{\infty} |C_{k,\ell}|$ are finite for all k .

The relation between the above-mentioned limit theorems and the description of stochastic systems at large scales are of particular interest for our investigation, a connection that was first pointed out by Jona-Lasinio (1975).⁴ The concept of large-scale description has been particularly influential in the context of the renormalization group approach which has led to the our current and generally accepted understanding of critical phenomena.⁵

To discuss this relation, let us return to independent random variables and, generalising Eq. (1), consider sums of random variables of the form

$$S_N(\mathbf{s}) = \frac{1}{N^{1/\alpha}} \sum_{k=1}^N s_k . \quad (7)$$

The parameter α fixes the power of system size N by which the sum must be rescaled in order to achieve interesting, i.e., non-trivial results. Clearly, if the power of N appearing as the normalization constant in Eq. (7) is too large for the type of random variables considered (i.e. if α is too small), then the normalized sum (7) will almost surely vanish, $S_N \rightarrow 0$, in the large N limit. Conversely, if the power of N in Eq. (7) too small (α too large), the normalized sum (7) will almost surely diverge, $S_N \rightarrow \pm\infty$, as N becomes large. We shall in what follows restrict our attention to the two important cases $\alpha = 1$ — appropriate for sums of random variables of non-zero mean — and $\alpha = 2$ — relevant for sums of random variables of zero mean and finite variance. For these two cases, we shall recover the propositions of the law of large numbers ($\alpha = 1$) and of the central limit theorems ($\alpha = 2$) as properties of *large-scale* descriptions of (7).

³For precise formulations of conditions and proofs, see (Feller 1968).

⁴On this, see also Batterman (1998), who referred to the relation on several occasions in the context of debates on reductionism.

⁵The notion of critical phenomena refers to a set of anomalies (non-analyticities) of thermodynamic functions in the vicinity of continuous, second order phase transitions. A lucid exposition is given by Fisher (Fisher 1983).

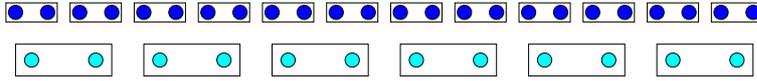


Figure 3: Repeated enlargement of scale: the present example begins with a system of 24 random variables symbolised by the dots in the first row. These are combined in pairs, as indicated by frames surrounding two neighbouring dots. Each such pair generates a renormalised variable, indicated by 12 dots of lighter shade in the second row of the figure. Two of those are combined in the next step as indicated by frames around pairs of renormalised variables, thereby starting the iteration of the renormalization procedure.

To this end we imagine the s_k to be arranged on a linear chain. The sum (7) may now be reorganised by (i) combining neighbouring pairs of the original variables and computing their averages \bar{s}_k , yielding $N' = N/2$ of such local averages, and by (ii) appropriately rescaling these averages so as to obtain renormalised random variables $s'_k = 2^\mu \bar{s}_k$, and by expressing the original sum (7) in terms of a corresponding sum of the renormalised variables, formally

$$\bar{s}_k = \frac{s_{2k-1} + s_{2k}}{2}, \quad s'_k = 2^\mu \bar{s}_k, \quad N' = N/2, \quad (8)$$

so that

$$S_N(\mathbf{s}) = 2^{(1-1/\alpha-\mu)} S_{N'}(\mathbf{s}'). \quad (9)$$

One may compare this form of “renormalization” — the combination of local averaging and rescaling — to the effect achieved by combining an initial reduction of the magnification of a microscope (to the effect that only locally averaged features can be resolved) with an ensuing change of the contrast of the image it produces.

By choosing the rescaling parameter μ such that $\mu = 1 - 1/\alpha$, one ensures that the sum (7) remains invariant under renormalization,

$$S_N(\mathbf{s}) = S_{N'}(\mathbf{s}'). \quad (10)$$

The renormalization procedure may therefore be iterated, as depicted in Fig. 3: $s_k \rightarrow s'_k \rightarrow s''_k \rightarrow \dots$, and one would obtain the corresponding identity of sums expressed in terms of repeatedly renormalised variables,

$$S_N(\mathbf{s}) = S_{N'}(\mathbf{s}') = S_{N''}(\mathbf{s}'') = \dots \quad (11)$$

The statistical properties of the renormalised variables s'_k will, in general be different from (though, of course, dependent on) those of the original variables s_k , and by the same token will the statistical properties of the doubly renormalised variables s''_k be different from those of the s'_k , and so on. However, one expects that statistical properties of variables will after sufficiently many renormalization steps, i.e., at large scale, eventually become independent of the microscopic details and of the scale considered, thereby becoming largely independent of the statistical properties of the original variables s_k , and invariant under further renormalization. This is indeed what happens under fairly general conditions.

It turns out that for sums of random variables s_k with non-zero average $\mu = \langle s_k \rangle$ the statement of the law of large numbers is recovered. To achieve asymptotic invariance under repeated renormalization, one has to choose $\alpha = 1$ in (7), in which case one finds that the repeatedly renormalised variables $s_k^{''''\dots}$ converge under repeated renormalization to the average of the s_k , which is thereby seen to coincide with the large N limit of the S_N .

If sums of random variables of zero mean (but finite variance) are considered, the adequate scaling is given by $\alpha = 2$. In this case the repeatedly renormalised variables $s_k^{''''\dots}$, and

thereby the S_N , are asymptotically normally distributed with variance σ^2 of the original variables s_k , even if these were not themselves normally distributed. The interested reader will find details of the mathematical reasoning underlying these results in an appendix at the end of this paper.

Let us not fail to mention that other stable distributions of the repeatedly renormalised variables $s_k^{''''\dots}$, thus of the S_N — the so-called Lévy α -stable distributions (Feller 1968)— may be obtained by considering sums random variables of infinite variance. Although such distributions have recently attracted some attention in the connection with the description of complex dynamical systems, such as turbulence or financial markets, they are of lesser importance for the description of thermodynamic systems in equilibrium, and we shall therefore not consider these any further in what follows.

4 Interacting systems and the renormalization group

For the purpose of describing macroscopic systems the concept of large-scale descriptions of a system, used above to elucidate the two main limit theorems of mathematical statistics, needs to be generalised to interacting, thus *correlated* or dependent random variables. Such a generalisation was formulated at the beginning of the 1970s as renormalization group approach to interacting systems.

Starting point of this approach is the Boltzmann-Gibbs equilibrium distribution of microscopic degrees of freedom taking the form (4). The idea of the renormalization group approach to condensed matter systems is perhaps best explained in terms of the normalisation constant Z_N appearing in (4), the so-called partition function. It is related to the dimensionless free energy \bar{f}_N of the system via $Z_N = e^{-N\bar{f}_N}$ and thereby to its thermodynamic functions and properties.⁶ To this end the partition function in (4) is written in the form

$$Z_N = Z_N(\mathbf{K}) = \sum_{\mathbf{s}} e^{-\bar{H}_N(\mathbf{s}; \mathbf{K})} , \quad (12)$$

in which $\bar{H}_N(\mathbf{s}; \mathbf{K})$ denotes the dimensionless energy function of the system, i.e., the conventional energy function multiplied by the inverse temperature β , while \mathbf{K} stands for the collection of all coupling constants in H_N (multiplied by β). These may include two-particle couplings as in (5), but also single-particle couplings as well as a diverse collection of many-particle couplings. Renormalization investigates, how the formal representation of the partition function changes, when it is no longer interpreted as a sum over all micro-states of the original variables, but as a sum over micro-states of renormalised variables, the latter defined as suitably rescaled local averages of the original variables in complete analogy to the case of independent random variables.

In contrast to the case of independent variables, geometric neighbourhood relations play a crucial role for interacting systems, and are determined by the physics of the problem. E.g., for degrees of freedom arranged on a d -dimensional (hyper)-cubic lattice, one could average over the b^d degrees of freedom contained in a (hyper)-cube B_k of side-length b to define locally averaged variables, as illustrated in Fig. 4 for $d = 2$ and $b = 2$, which are then rescaled by a suitable factor b^μ in complete analogy to the case of independent random variables discussed above,

$$\bar{s}_k = b^{-d} \sum_{i \in B_k} s_i , \quad s'_k = b^\mu \bar{s}_k . \quad N' = N/b^d \quad (13)$$

⁶The dimensionless free energy is just the product of the standard free energy and the inverse temperature β . At a formal level, the partition function is closely related to the characteristic function of a (set of) random variables, in terms of which we analysed the idea of large-scale descriptions for sums of independent random variables in appendix A.

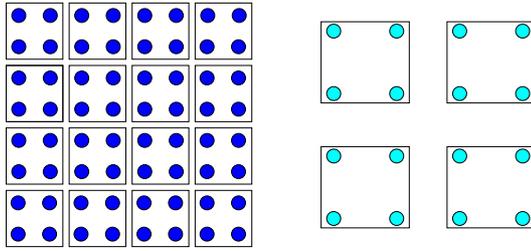


Figure 4: Iterated coarsening of scale

The partition sum on the coarser scale is then evaluated by first summing over all microstates of the renormalised variables \mathbf{s}' and for each of them over all configurations \mathbf{s} compatible with the given \mathbf{s}' , formally

$$Z_N(\mathbf{K}) = \sum_{\mathbf{s}'} \left[\sum_{\mathbf{s}} P(\mathbf{s}', \mathbf{s}) e^{-\bar{H}_N(\mathbf{s}; \mathbf{K})} \right] \equiv \sum_{\mathbf{s}'} e^{-\bar{H}_{N'}(\mathbf{s}'; \mathbf{K}')} = Z_{N'}(\mathbf{K}') \quad (14)$$

where $P(\mathbf{s}', \mathbf{s}) \geq 0$ is a projection operator constructed in such a way that $P(\mathbf{s}', \mathbf{s}) = 0$, if \mathbf{s} is incompatible with \mathbf{s}' , whereas $P(\mathbf{s}', \mathbf{s}) > 0$, if \mathbf{s} is compatible with \mathbf{s}' , and normalised such that $\sum_{\mathbf{s}'} P(\mathbf{s}', \mathbf{s}) = 1 \forall \mathbf{s}$. The result is interpreted as the partition function corresponding to a system of $N' = b^{-d}N$ renormalised variables, corresponding to a dimensionless energy function $\bar{H}_{N'}$ of the same format as the original one, albeit with renormalised coupling constants $\mathbf{K} \rightarrow \mathbf{K}'$, as expressed in (14). The distance between neighbouring renormalised degrees of freedom is larger by a factor b than that of the original variables. Through an ensuing rescaling of all lengths $\ell \rightarrow \ell/b$ one restores the original distance between the degrees of freedom, and completes the renormalization group transformation as a mapping between *systems of the same format*.

As in the previously discussed case of independent random variables, the renormalization group transformation may be iterated and thus creates not only a sequence of repeatedly renormalised variables, but also a corresponding sequence of repeatedly renormalised couplings

$$\mathbf{K} \rightarrow \mathbf{K}' \rightarrow \mathbf{K}'' \rightarrow \mathbf{K}''' \rightarrow \dots \quad (15)$$

As indicated in Fig. 5, this sequence may be visualised as a renormalization group ‘flow’ in the space of couplings.

The renormalization transformation entails a transformation of (dimensionless) free energies $\bar{f}_N(\mathbf{K}) = -N^{-1} \ln Z_N$ of the form

$$\bar{f}_N(\mathbf{K}) = b^{-d} \bar{f}_{N'}(\mathbf{K}') \quad (16)$$

For the present discussion, however, the corresponding transformation of the so-called correlation length ξ which describes the distance over which the degrees of freedom in the system are statistically correlated, is of even greater interest. As a consequence of the rescaling of all lengths, $\ell \rightarrow \ell/b$, in the final step of the renormalization transformation, one obtains

$$\xi_N(\mathbf{K}) = b \xi_{N'}(\mathbf{K}') \quad (17)$$

Repeated renormalization amounts to a description of the system on larger and larger length scales. The expectation that such a description would on a sufficiently large scale eventually become independent of the scale actually chosen would correspond to the finding that the renormalization group flow would typically approach a fix-point: $\mathbf{K} \rightarrow \mathbf{K}' \rightarrow \mathbf{K}'' \rightarrow$

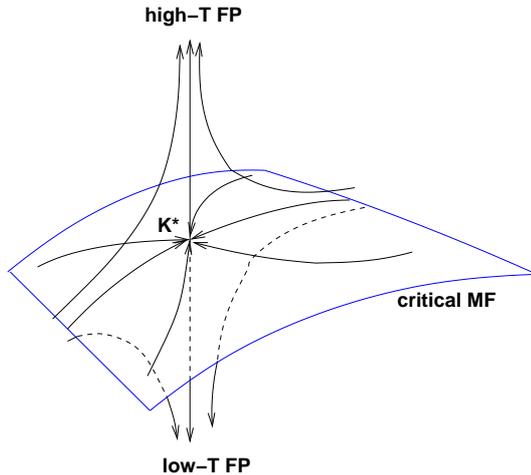


Figure 5: Renormalization group flow in the space of couplings

$\dots \rightarrow \mathbf{K}^*$. As in the case of the renormalization of sums of independent random variables exponent μ in the rescaling operation in Eq. (13) must be judiciously chosen to allow approach to a fixed point describing non-trivial large scale behaviour.

The existence of fixed points is of particular significance in the limit of infinitely large system size $N = N' = \infty$, as in this limit Eq. (17), $\xi_\infty(\mathbf{K}^*) = b\xi_\infty(\mathbf{K}^*)$ will for $b \neq 1$ *only* allow for the two possibilities

$$\xi_\infty(\mathbf{K}^*) = 0 \quad \text{or} \quad \xi_\infty(\mathbf{K}^*) = \infty . \quad (18)$$

The first either corresponds to a so-called high-temperature fixed point, or to a low-temperature fixed point. The second possibility with infinite correlation length corresponds to a so-called critical point describing a continuous, or second order phase transition. In order to realise the second possibility, the initial couplings \mathbf{K} of the system must be adjusted in such a way that they come to lie precisely *on* the critical manifold in the space of parameters, defined as the basin of attraction of the fixed point \mathbf{K}^* for the renormalization group flow. If the system's initial couplings \mathbf{K} are not exactly *on* the critical manifold, but close to it, repeated renormalization will result in a flow that visits the vicinity of the fixed point \mathbf{K}^* , but is eventually driven away from it (see Fig. 5). Within an analysis of the RG transformation that is linearized in the vicinity of \mathbf{K}^* this feature can be exploited to analyse critical behaviour of systems in the vicinity of their respective critical points and quantify it in terms of critical exponents, and scaling relations satisfied by them (Fisher 1975, Fisher 1983).

The distance from the critical manifold is parameterized by the so-called “relevant couplings”.⁷ Experience shows that relevant couplings typically form a low-dimensional manifold within the high-dimensional space of system parameters \mathbf{K} . For conventional continuous phase transitions it is normally two-dimensional — parametrized by the deviation of temperature and pressure from their critical values in the case of gasses, by corresponding deviations of temperature and magnetic field in magnetic systems, and so forth.⁸ The fact that all systems in the vicinity of a given critical manifold are controlled by the same fixed point does

⁷These correspond to a subset of the couplings encoded in \mathbf{K} , whose distance from the critical manifold is increased under renormalization. The critical manifold itself is parameterized by the so-called irrelevant couplings; their distance from the critical point \mathbf{K}^* is decreased under successive renormalizations.

⁸Note that in practical RG analyses of the liquid gas critical point, chemical potential is usually used instead of pressure. Also, proper independent coordinates of the manifold of relevant couplings are not necessarily the physical parameters themselves; they could be, and often are, constructed from suitable combinations thereof. For a detailed discussion, see e.g. Lavis & Bell (1998).

in itself have the remarkable consequence that there exist large classes of microscopically very diverse systems, the so-called universality classes, which exhibit essentially the same behaviour in the vicinity their respective critical points (Fisher 1975, Fisher 1983).

For the purpose of the present discussion, however, the phenomenology close to *off-critical* high and low-temperature fixed points is of even greater importance. Indeed, all non-critical systems will eventually be driven towards one of these under repeated renormalization, implying that degrees of freedom are virtually *uncorrelated* on large scales, and that the description of non-critical systems within the framework of the two limit theorems for independent variables discussed earlier is therefore *entirely adequate*, despite the correlations over small distances created by interactions between the original microscopic degrees of freedom.

5 The thermodynamic limit of infinite system size

We are ready for a first summary: only in the thermodynamic limit of infinite system size $N \rightarrow \infty$ will macroscopic systems exhibit non-fluctuating thermodynamic properties; only in this limit can we expect that deterministic equations of state exist which describe relations between different thermodynamic properties as well as the manner in which these depend on external parameters such as pressure, temperature or electromagnetic fields. Moreover, only in this limit will systems have stable macroscopic properties in the sense that transitions between different macroscopic phases of matter (in case they exist) will not be observed in finite time. Indeed stability in this sense is a consequence of the absence of fluctuations, as (large) fluctuations would be required to induce such macroscopic transformations. We have seen that these properties can be understood in terms of coarse-grained descriptions, and the statistical limit theorems for independent or weakly dependent random variable describing the behaviour averages and the statistics of fluctuations in the large system limit, and we have seen how RNG analyses applied to *off-critical* systems can provide a rationalization for the applicability of these limit theorems.

Real systems are, of course, always finite. They are, however, typically composed of huge numbers of atomic or molecular constituents, numbers so large in fact that they are for the purposes or determining macroscopic physical properties “indistinguishable from infinity”: fluctuations of macroscopic properties decrease with increasing system size, to an extent of becoming virtually undetectable in sufficiently large (yet finite) systems. Macroscopic stability (in the sense of absence of transitions between different macroscopic phases on time-scales which exceed the age of the universe) in large finite systems ensues. The stability we observe, i.e. the phenomenon to be explained, is compatible both with the idealization of infinite system size and with the finite, but very large, size of real systems. The fact that strict stability requires the infinite limit poses no problem because we are in a region where the difference between the finite size model and the infinite size model cannot be observed.

The role of the thermodynamic limit and the issue of stability of macroscopic system properties are more subtle, and more interesting, in the case of phase transitions and critical phenomena. To discuss them, it will be helpful to start with a distinction.

What we have discussed so far is the stability of macroscopic physical properties vis-à-vis incessant changes of the system’s dynamical state on the micro-level. Let us call this “actual stability” and contrast it with “counterfactual stability”. Counterfactual stability is the stability of the macro-behaviour with respect to non-actual counterfactual changes a system’s composition at the micro-level might undergo: e.g., in a ferro-magnetic system one might add next-nearest neighbour interactions to a system originally having only nearest neighbour interactions, and scale down the strength of the original interaction in such a manner that would leave the macroscopic magnetization of a system invariant.

The notion of counterfactual stability applies in particular to the phenomenon of universality of critical phenomena. Critical phenomena comprise a set of anomalies (*algebraic singularities*) of thermodynamic functions that are observed at second order phase transitions in a large variety of systems. Critical phenomena, and specifically the critical exponents introduced to characterize the non-analyticities quantitatively have various remarkable properties. For instance, critical exponents exhibit a remarkable degree of universality. Large classes of systems are characterized by identical sets of critical exponents, despite the fact that interactions at the microscopic level may be vastly different. Within the RNG approach described in the previous section this is understood as a global property of a renormalization group flow: all systems with Hamiltonians described by couplings \mathbf{K} in the vicinity of a given critical manifold will be attracted by the same RNG fixed point \mathbf{K}^* , and therefore exhibit identical critical exponents.

The case of critical behaviour is thus a special and particularly impressive case of (counterfactual) stability. Note, however, that even though the notion of universality provides a notion of stability (see Batterman 2002, 57ff), the range it applies to is fairly restricted and does not cover all the cases Fodor had in mind when he was referring to the stability of macro-level. In particular, universality of critical phenomena as uncovered by the RNG approach *only* refers to asymptotic critical exponents⁹, describing critical singularities only in the *immediate vicinity* of critical points. The thermodynamic properties of a system *not* exactly *at* its critical point will, however, be influenced by the presence of irrelevant couplings, and thus show properties which are system-specific, and *not* universal within universality classes. We will discuss some further notes of caution in Sect 6 below.

The case of critical phenomena requires special discussion also with respect to the infinite system limit. We have seen that the thermodynamic limit is a prerequisite for systems to exhibit non-fluctuating macroscopic physical properties, but that this type of behaviour is well approximated (to the extent of being indistinguishable from it for all practical purposes) in finite sufficiently large systems. Thermodynamically, phase transitions and critical phenomena are associated with non-analyticities in a system's thermodynamic functions. Given certain uncontroversial assumptions, such non-analyticities cannot occur in finite systems (cf. Menon and Callendar, 2013, 194) in the canonical or grand-canonical ensembles of Statistical Mechanics. For phase transitions to occur, and systems to exhibit critical phenomena it thus appears that an “ineliminable appeal to the thermodynamic limit and to the singularities that emerge in that limit” (Batterman 2011, 1038) is required.

Is this an appeal to idealizations that differs from other cases? Within a RNG analysis of large finite systems, the inverse L^{-1} of the system's linear dimension L is a *relevant* variable in the RNG sense due to the final rescaling $\ell \rightarrow \ell/b$ of all lengths (thus $L^{-1} \rightarrow bL^{-1} > L^{-1}$) in the final RG step. The condition for this to be the only modification of the RNG transformation is that the system must be sufficiently large for all renormalized couplings to be embeddable in the system (for a review, see Barber 1983). A finite value of L then implies that a finite system can never be exactly *on* a critical manifold, and thus never be exactly at a critical point. If carried through, this finite-size modification of the RNG transformation gives rise to so-called finite-size-scaling theory (FSS) which describes in quantitative detail the way in which critical singularities are rounded due to finite size effects. Thus while proper singularities of thermodynamic functions disappear in systems of finite size, deviations from the behaviour described in terms of the corresponding infinite-system singular functions will become noticeable only in ever smaller parameter regions around “ideal” critical points, as system sizes grow, and will in sufficiently large systems eventually be indistinguishable from it using experiments of conceivably attainable precision. In this sense, the infinite system idealization of singular behaviour of thermodynamic functions in the vicinity of critical points

⁹This includes asymptotic scaling functions and even asymptotic corrections to scaling (Wegner 1976)

is an idealization, which is controllable in large systems in a well-defined sense, which does not appear to be fundamentally different from that of non-fluctuating thermodynamic functions and absolute stability discussed earlier. (For a discussion of these points see also Butterfield 2011, section 7 as well as Menon and Callendar 2013, section 3.2)

6 Stability, Universality and Part-Whole-Explanation

When the macro behaviour of physical systems is stable, many details of the exact micro-state are irrelevant. This is particularly impressive in the case of universality.

Margret Morrison argues that if the microphysical details are irrelevant, the phenomenon in question is not reducible to its microphysical constituents and that the supervenience relation is “inapplicable in explaining the part-whole aspects” of such phenomena (Morrison 2012, 156). Morrison classifies universal behaviour at phase transitions as “emergent”, characterizing emergence as follows:

... what is truly significant about emergent phenomena is that we cannot appeal to microstructures in explaining or predicting these phenomena, even though they are constituted by them. (Morrison 2012, 143).

Even though we agree with Morrison that universal phenomena are in an interesting sense ontologically independent of the underlying microstructure we reject her claim that this violates the “reductionist picture” (Morrison 2012, 142). We will argue that ontological independence in the sense of actual or counterfactual stability is compatible with the “reductionist picture” in the sense she uses this expression.

What Morrison means by the “reductionist picture” can be captured by the following two claims:

- (i) *supervenience*: the constituents’ properties and states *determine* the behaviour of the compound system,
- (ii) *explanation*: we can *explain* the macro-phenomena in terms of the properties and states of the constituents.

Morrison rejects both of these claims. The main idea seems to be that (i.) fails and therefore (ii.) is impossible.

Why would one suppose that supervenience fails in the case of stable macro-behaviour and universal behaviour in particular? In Morrison’s paper we find two arguments. The first has to do with the thermodynamic limit. The stable macro-behaviour to be explained presupposes an infinite number of constituents. Real systems are finite. The behaviour to be explained is not determined by the behaviour of the finite number of constituents. We have already dealt with this issue in the previous section and have indicated why we are not convinced by this line of argument. We thus move to her second argument:

If we suppose that micro properties could determine macro properties in cases of emergence, then we have no explanation of how universal phenomena are even possible. Because the latter originate from vastly different micro properties, there is no obvious ontological or explanatory link between the micro and macro levels. (Morrison 2012, 162).

We fail to see why it should be impossible that vastly different micro-properties determine the same micro-property. To illustrate: the integer 24 may be obtained as a sum of smaller

integers in many different ways ($24 = 13 + 11 = 10 + 14 = 9 + 8 + 7$ etc). However, this is not a valid argument for the claim that the summands fail to determine the sum. Similarly, the fact that a multiplicity of micro-states gives rise to the same macro-state is no objection to the claim that the micro-state determines the macro-state.

In fact, the simulation in section 2 and the explanations in sections 3 and 4 aim at explaining how this is possible. The simulation in section 2 has *illustrated* that, by simply increasing the sample size, fluctuations decrease and macroscopic modifications become more and more rare, i.e. the macro-state becomes more and more stable despite of changes in the micro-states. In section 3 we discussed an *explanation* for this phenomenon for the case of non-interacting constituents/variables. It is in virtue of the central limit theorem that fluctuations of macroscopic observables decrease with the number N of constituents as $\frac{1}{\sqrt{N}}$ for large system sizes, and that stable behaviour exists for large systems. In section 4 we moved to the case of interacting variables/constituents. Again, what is provided is an *explanation* for why so many features of the micro-system are irrelevant. RNG explains how systems that are characterized by very different Hamilton operators nevertheless give rise to very similar macro-behaviour. This only works because, given a particular Hamiltonian, i.e. the micro-properties, the macro-behaviour is fixed and thus determined. These explanations tell us *why* and *in which sense* certain features of the constituents or their interactions become irrelevant: note that irrelevance in the RNG sense acquires a technical meaning, which coincides with the plain-English meaning of the word only as far as the determination of asymptotic critical exponents (in infinitely large systems) is concerned.

The fact that certain features of the constituents are irrelevant in the technical RNG sense does therefore not imply that the properties and states of the constituents fail to influence the macro-behaviour. Rather, it is only a small number of features of these that does the work for asymptotic critical exponents. Interestingly, these same features are also responsible for driving an off-critical system away from critical RNG fixed points and towards one of the fixed-points at which the (coarse-grained) system appears as a collection of *uncorrelated* degrees of freedom, for which the limit theorems for uncorrelated or weakly correlated random variables provide an appropriate description.

Whenever a system is not exactly at its critical point, there will always be a residual effect of the so-called irrelevant variables on thermodynamic behaviour. A *finite* system, in particular, is never exactly at a critical point, as $1/L$ (with L denoting the linear dimension of the system) is always a relevant variable in the renormalization group sense (increasing under renormalization/coarse graining); this leads to rounding of critical singularities, which can be quantitatively analysed (see section 5). As a consequence, Morrisons argument, if valid, would not apply to finite systems (nor to systems which are off-critical for other reasons), and therefore the so-called irrelevant variables will contribute to the determination of the macro-behaviour of all finite systems.

More positively, the reductionist might characterize what is going on in the special case of critical phenomena as follows: With respect to universal behaviour RNG has a twofold explanatory role. These two roles concern two sets of questions that ought to be distinguished.

The first set of questions concerns the behaviour of single critical systems: Why does the compressibility of water diverge according to some fixed function of temperature etc.? With regard to questions like this RNG is simply a coarse-graining procedure that allows us to calculate approximately correct results. The explanation of the single system's critical exponents starts with the description of the microstructure. In this context RNG is effectively used as a tool for micro-reductive explanation of the system's behaviour. RNG is merely an effective tool to evaluate thermodynamic functions of interacting systems (albeit in the majority of cases only approximately), where exact evaluations are infeasible. In some sense, RNG can be regarded as a successor theory of mean field theory, which was unable to produce

even approximately correct critical exponents.

There is, however, a second set of questions. Micro-reductive explanations may appear to be unable to answer these. These questions concern universality and scaling (i.e. the fact that the critical exponents obey certain system-independent relations). Why are there universality classes at all, i.e. why is it that systems with extremely different microstructure such as alloys ferro-magnets and liquids obey exactly the same laws near their critical points, i.e. why is it that the values of the critical exponents of the members of such classes coincide? Furthermore, why is it the case that all critical systems — disregarding their universality class — obey the scaling relations?

RNG appears to answer the above questions non-reductively. The essential ingredients of such an explanation are not properties of the constituents of single critical systems, one might argue, but rather properties of the renormalization-group-flow - topological features of the space of coupling constants. (see Morrison 2012, 161/2). The renormalization-group-transformation induces a mapping between Hamiltonians or the coupling-constants of the Hamiltonians in question. The iteration of such a mapping defines a flow in the space of coupling constants as represented in figure 5. Fixed points come along with basins of attraction. All Hamiltonians/physical systems within such a basin flow towards the same critical point, i.e. their behaviour is identical. It is — basically — the existence of these basins of attraction that explain why physical systems with widely diverging microstructure behave identically near the critical point.

The explanation of universality and the scaling relation do not appear to appeal to the microscopic constitution of individual systems that show this behaviour. Instead it appeals to topological features of the space of coupling constants/Hamiltonians.

But the reductionist is able to defend her view: Why should micro-explanations be unable to explain that a variety of systems behave identical near the critical point? There is a feature on the micro-level that all of these systems have in common. And it is in virtue of this common feature that RNG works:

The starting point in the renormalization group approach is to realize that the most important fluctuations at the critical point have no characteristic length. Instead the important fluctuations have *all* wavelengths ranging from the atomic spacing up to the correlation length; close to the critical point the correlation length is much larger than the atomic spacing. (Wilson 1976)

The absence of a characteristic length (the divergence of the correlation length) *at* the critical point implies invariance of a system under coarse graining, or scale transformations. An RNG transformation which describes the effect of coarse graining in terms of a transformation $\mathbf{K} \rightarrow \mathbf{K}'$ of the systems' set of couplings will therefore identify a critical point with a fixed point \mathbf{K}^* of that transformation. The reductionist will thus think of topological features of the transformation such as basins of attraction (i.e. universality classes) as an unavoidable consequence of this type of analysis.

This is finally the point to recall that there is a second way in which fluctuations of thermodynamic properties of a system can fail to exhibit a characteristic length: this is the situation where where the correlation length *vanishes* and the system is statistically fully homogeneous (and therefore also scale invariant). This possibility is realized at the high- or low-temperature fixed points of the RNG transformations (as discussed in Sect 4), which will come with *their own* basins of attraction and associated notions of insensitivity to detail, as embodied in statistical limit theorems.

To sum up: What we claim to have shown is that the reductionist picture according to which the constituents' properties and states *determine* the behaviour of the compound system, and the macro-phenomena can be explained in terms of the properties and states

of the constituents is neither undermined by stable phenomena in general nor by universal phenomena in particular.

Let us add, however, that the reductionist picture as outlined above does not imply that an explanation of the macro-behaviour of a system always has to appeal to the micro-level. The fact that it is possible to give such an explanation does not imply that it is the best explanation let alone the only available explanation. In fact, once we know that many details are irrelevant, we have a perfectly good reason to focus on those factors that are not irrelevant. This is the topic of the following section.

7 Post Facto Justification of Modelling

Let us close with the following observation: A remarkable feature of our analysis of stability of macroscopic system properties based on a renormalization group approach is the fact that it provides us with a justification for certain kinds of idealisation. Models that are used to describe critical (or other collective) behaviour of given physical systems are often grossly simplified, and it is a non-trivial problem to understand why such models can nevertheless be as successful, as they apparently are. RNG illuminates this point as follows. It teaches us that systems, which are described by Hamiltonians that differ only with respect to irrelevant couplings (systems within the basin of attraction of the RNG fixed point relevant for the large scale phenomenon under study) will under renormalization be attracted to the same fixed point, and will therefore exhibit the same type of collective behaviour. The presence of irrelevant couplings can in this sense be regarded as belonging to the inessential details which do not significantly affect the problem under study, such as critical behaviour of a given many-particle system, or macroscopic off-critical properties. In a quantitatively well-defined sense, such irrelevant couplings can therefore be neglected when analysing collective phenomena in such systems — critical or not. A description of a system which neglects or abstracts from these details constitutes what would properly be described as an idealized description. Within the RNG setting we therefore have a well-defined and even quantifiable notion of the sense in which an idealized description of a complex system will nevertheless capture the essence of its collective behaviour.

It is a remarkable fact that conventional second order phase transitions are characterized by just two relevant couplings in the RNG sense; this tells us that all but two operators within an infinite set characterizing the universe of possible Hamiltonians for a given system are irrelevant, and can therefore be neglected when analysing asymptotic critical behaviour associated with second order phase transitions. Moreover it is the presence of these same two relevant couplings which is responsible for driving the RNG flow to off-critical fixed points at which systems are statistically fully homogeneous and their macroscopic properties thus describable in terms of statistical limit theorems for uncorrelated or weakly correlated random variables.

It is due to this remarkable fact that simplified models do at all allow us to capture the essence of collective behaviour. In this sense RNG provides a justification for the idea of modelling per se, and it gives us a glimpse on the reasons why “simple” models of condensed matter could at all be successful. Note finally that it is essential that systems are large to allow levels of description which are sufficiently coarse-grained compared to atomistic scales. Stability, and as we have just discussed, simplicity only arises at this sufficiently coarse scale. It has been pointed out (Kühn 2008) that in this sense large numbers appear to be an essential prerequisite for facilitating successful theory formation for condensed matter systems.

A Renormalization and characteristic functions

The renormalization group transformation for the case of sums of independent random variables is best investigated in terms of their *characteristic functions*.

The characteristic function of a random variable X is defined as the Fourier transform of its probability density p_X ,¹⁰

$$\varphi_X(k) = \langle e^{ikX} \rangle = \int dx p_X(x) e^{ikx} . \quad (19)$$

Characteristic functions are important tools in probability. Among other things, they can be used to express moments of a random variable in compact form via differentiation,

$$\left(-i \frac{d}{dk} \right)^n \varphi_X(k) \Big|_{k=0} = \langle X^n \rangle = \int dx p_X(x) x^n . \quad (20)$$

The second important property needed here is that the characteristic function of a sum $X+Y$ of two independent random variables X and Y is given by the product of their characteristic functions. For, denoting by p_X and p_Y the probability densities corresponding to the two variables, one finds

$$\varphi_{X+Y}(k) = \int dx dy p_X(x) p_Y(y) e^{ik(x+y)} = \varphi_X(k) \varphi_Y(k) . \quad (21)$$

We investigate properties of (7),

$$S_N(\mathbf{s}) = \frac{1}{N^{1/\alpha}} \sum_{k=1}^N s_k$$

under renormalization, using its characteristic function to be denoted by ϕ_N . Let φ_1 denote the characteristic function of the original variables, φ_2 that of the renormalised variables s'_k (constructed from sums of two of the original variables), and more generally, let φ_{2^ℓ} denote the characteristic function of the ℓ -fold renormalised variables, constructed from sums involving 2^ℓ original variables. We then get

$$\begin{aligned} \phi_N(k) &:= \langle e^{ikS_N} \rangle = \varphi_1 \left(\frac{k}{N^{1/\alpha}} \right)^N = \varphi_2 \left(\frac{k}{(N/2)^{1/\alpha}} \right)^{N/2} \\ &= \varphi_4 \left(\frac{k}{(N/4)^{1/\alpha}} \right)^{N/4} = \dots = \varphi_{2^\ell} \left(\frac{k}{(N/2^\ell)^{1/\alpha}} \right)^{N/2^\ell} \end{aligned} \quad (22)$$

Assuming that multiply renormalised variables will acquire asymptotically stable statistical properties, i.e. statistical properties that remain invariant under further renormalization, the φ_{2^ℓ} would have to converge to a limiting function φ^* ,

$$\varphi_{2^\ell} \rightarrow \varphi^* \quad , \quad \ell \rightarrow \infty . \quad (23)$$

This limiting function φ^* would have to satisfy a functional *self-consistency* relation of the form

$$\varphi^*(k) = \varphi^*(2^{1/\alpha} k)^{1/2} \quad (24)$$

which follows from (22), using the invariance of S_N under repeated renormalization.

¹⁰In this appendix, we follow the mathematical convention to distinguish in notation between a random variable X and its realisation x .

The solutions of this self-consistency relation for $\alpha = 1$ and $\alpha = 2$ are seen to be given by

$$\varphi^*(k) \equiv \lim_{N \rightarrow \infty} \phi_N(k) = \exp(-ck^\alpha), \quad (25)$$

and it can be shown that these solutions are unique in the space of characteristic functions for random variables having finite moments.

One identifies the characteristic function of a non-fluctuating (i.e. constant) random variable with $c = -i\langle X \rangle = -i\mu$ for $\alpha = 1$, and the characteristic function of a Gaussian normal density with $c = \frac{1}{2}\sigma^2$ for $\alpha = 1/2$, and thereby verifies the statements of the two limit theorems.

One can also show that the convergence (23) is realised for a very broad spectrum of distributions for the microscopic variables, both for $\alpha = 1$ (the law of large numbers), and for $\alpha = 1/2$ (the central limit theorem). For $\alpha = 1$, there is a “marginal direction” in the infinite-dimensional space of possible perturbations of the invariant characteristic function (corresponding to a change of the expectation value of the random quantities being summed), which doesn’t change its distance to the invariant function $\varphi^*(k)$ under renormalization. All other perturbations are irrelevant in the sense that their distance from the invariant characteristic function will diminish under repeated renormalization. For $\alpha = 2$ there is one “relevant direction” in the space of possible perturbations, in which perturbations of the invariant characteristic function will be amplified under repeated renormalization (it corresponds to introducing a non-zero mean of the random variables being added), and a marginal direction that corresponds to changing the variance of the original variables. All other perturbations are irrelevant and will be diminished under renormalization. The interested reader will find a formal verification of these statements in the following Appendix B.

B Linear stability analysis

Statements about the stability of invariant characteristic functions under various perturbations are proved by looking at the linearisation of the renormalization group transformation in the vicinity of the invariant distribution. We shall adhere to the formulation using characteristic functions which makes the full analysis somewhat easier than the one in terms of probability densities used in (Sinai 1992).

Let R_α denote the renormalization transformation of a characteristic function for the scaling exponent α . From (22), we see that its action on a characteristic function φ is defined as

$$R_\alpha[\varphi](2^{1/\alpha}k) = \varphi(k)^2 \quad (26)$$

Assuming $\varphi = \varphi^* + h$, where h is a small perturbation of the invariant characteristic function, we have

$$R_\alpha[\varphi^* + h](2^{1/\alpha}k) = (\varphi^*(k) + h(k))^2 \simeq \varphi^*(k)^2 + 2\varphi^*(k)h(k), \quad (27)$$

where the expansion on the r.h.s. has been carried to first order in the small perturbation h . Using an expansion of the transformation R_α in the vicinity of φ^* , and denoting by $D_\alpha = D_\alpha[\varphi^*]$ the operator of the linearised transformation in the vicinity of φ^* on the l.h.s., one has $R_\alpha[\varphi^* + h] \simeq R_\alpha[\varphi^*] + D_\alpha h$ to linear order in h , thus

$$R_\alpha[\varphi^*](2^{1/\alpha}k) + D_\alpha h(2^{1/\alpha}k) \simeq \varphi^*(k)^2 + 2\varphi^*(k)h(k). \quad (28)$$

By the invariance of φ^* under R_α , we get

$$D_\alpha h(2^{1/\alpha}k) = 2\varphi^*(k)h(k) \quad (29)$$

to linear order. The stability of the invariant characteristic function is then determined by the spectrum of D_α , found by solving the eigenvalue problem

$$D_\alpha h(2^{1/\alpha}k) = 2\varphi^*(k)h(k) = \lambda h(2^{1/\alpha}k) . \quad (30)$$

In the case where $\alpha = 1$ we have $\varphi^*(k) = e^{-i\mu k}$, and the eigenvalue equation reads

$$2e^{-i\mu k}h(k) = \lambda h(2k) . \quad (31)$$

To solve it we introduce f by writing

$$h(k) = e^{-i\mu k}f(k)$$

which transforms the eigenvalue equation into

$$2f(k) = \lambda f(2k) . \quad (32)$$

Clearly this equation is solved by homogeneous functions:

$$2f_n(k) = \lambda_n f_n(2k) , \quad f_n(k) = \frac{(ikx_0)^n}{n!}, \quad \lambda_n = 2^{1-n} . \quad (33)$$

In order for $\varphi^* + h_n$ with $h_n(k) = e^{-i\mu k}f_n(k)$ to be a characteristic function (of a system with finite moments), we must have $n \geq 1$, so that $\lambda_1 = 1$ (the corresponding perturbation being marginal), and $\lambda_n < 1$ (the corresponding perturbations thus being irrelevant) for all $n > 1$. The marginal perturbation amounts to changing the mean of the random variable to $\mu + x_0$, as mentioned earlier.

In the case where $\alpha = \frac{1}{2}$ we have $\varphi^*(k) = e^{-\frac{1}{2}s^2k^2}$, so the eigenvalue equation is

$$2e^{-\frac{1}{2}s^2k^2}h(k) = \lambda h(\sqrt{2}k) . \quad (34)$$

To solve it we now define f by writing

$$h(k) = e^{-\frac{1}{2}s^2k^2}f(k)$$

which transforms the eigenvalue equation into

$$2f(k) = \lambda f(\sqrt{2}k) . \quad (35)$$

Clearly this equation, too, is solved by homogeneous functions:

$$2f_n(k) = \lambda_n f_n(\sqrt{2}k) , \quad f_n(k) = \frac{(ikx_0)^n}{n!}, \quad \lambda_n = 2^{1-\frac{n}{2}} . \quad (36)$$

Once more, for $\varphi^* + h_n$ with $h_n(k) = e^{-\frac{1}{2}s^2k^2}f_n(k)$ to be a characteristic function (of a system with finite moments), we must have $n \geq 1$, so that $\lambda_1 = 2^{\frac{1}{2}}$ (the corresponding perturbation being relevant), $\lambda_2 = 1$ (the corresponding perturbation being marginal), and $\lambda_n < 1$ for all $n > 2$ (the corresponding perturbations thus being irrelevant). In the present case, the relevant perturbation amounts to introducing a nonzero mean x_0 of the original random variables, while the marginal perturbation changes the variance to $s^2 + x_0^2$, as mentioned earlier. All other perturbations change higher order cumulants of the random variables considered and are irrelevant.

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