

Energy Landscapes: Frequently Asked Questions

Thermodynamics

The **microcanonical** and **canonical** partition functions can be written as a **superposition** of contributions from the catchment basins of all the minima [1]:

$$\Omega(E) = \sum_i \Omega_i(E) \quad \text{and} \quad Z(T) = \sum_i Z_i(T),$$

where $\Omega_i(E)$ and $Z_i(T)$ are the **microcanonical** density of states and **canonical** partition function for local minimum i , respectively.

The **potential** energy surface is independent of mass, temperature and coordinate systems; $\Omega(E)$ and $Z(T)$ provide connections to the thermodynamic potentials in the **microcanonical** and **canonical** ensembles, namely **entropy** and **free energy**.

The canonical equilibrium **occupation probability** of minimum i is $p_i^{\text{eq}}(T) = Z_i(T)/Z(T)$, and the corresponding free energy is $F_i(T) = -kT \ln Z_i(T)$. The free energy of transition state \dagger is $F^\dagger(T) = -kT \ln Z^\dagger(T)$, where $Z^\dagger(T)$ does not include the mode corresponding to the **negative** Hessian eigenvalue.

In the canonical ensemble the equilibrium **occupation probability** and **free energy** of a group of minima, J are

$$p_J^{\text{eq}}(T) = \sum_{j \in J} p_j^{\text{eq}}(T) \quad \text{and} \quad F_J(T) = -kT \ln \sum_{j \in J} Z_j(T),$$

and the free energy of the **transition states** connecting J and L is then

$$F_{LJ}^\dagger(T) = -kT \ln \sum_{(lj)^\dagger} Z_{lj}^\dagger(T), \quad l \in L, j \in J,$$

$$\text{with} \quad k_{LJ}^\dagger(T) = \sum_{(lj)^\dagger} \frac{p_j^{\text{eq}}(T)}{p_J^{\text{eq}}(T)} k_{lj}^\dagger(T) = \frac{kT}{h} \exp \left[-\frac{\left(F_{LJ}^\dagger(T) - F_J(T) \right)}{kT} \right].$$

Size Effects

The total (free) energy is **extensive**, and scales with system size, along with the energy range spanned by the lowest to highest minima. However, **barrier heights** and **energy differences** between minima directly connected by transition states (saddles of Hessian index one [2]) are **intensive** quantities.

The relative **equilibrium occupation probabilities** and **rate constants** [3] between these minima depend on **ratios** of intensive quantities, namely energy **differences** and kT . Hence the equilibrium concentration of a particular defect and the defect migration rate in a crystal do **not** depend on the size of the crystal [4, 5]. An **inappropriate** comparison of the **extensive** internal energy with **intensive** barriers leads to the conclusion that ‘**activated**’ processes cease to exist as the system size increases [6, 7].

Genuine **finite size** effects are certainly of interest, for example, the sharpening of **heat capacity** peaks for **first order**-type transitions [8].

Landscapes and Entropy

It can be helpful to think of the energy landscape in terms of analogies from a mountain range, following **James Clark Maxwell** in his article for *Philosophical Magazine* 'On Hills and Dales' in 1870 [9]. Transition states then correspond to **watersheds**, and steepest-descent paths to **watercourses**.

However, **low-dimensional** projections can also be **misleading** [10–17]. In particular, the system is **never** 'above' the potential energy surface: the potential energy function is defined for any configuration.

The effect of increasing the temperature is to **shift** the **occupation probabilities** to states with **higher entropy** and **higher potential energy**. The **temperature gradients** of the $p_i^{\text{eq}}(T)$ can be related to **peaks** in the heat capacity [18].

We can distinguish **two** contributions to the entropy. There is an (anharmonic) **vibrational** entropy associated with each local minimum, determined by the phase volume contained within the **basin of attraction**.

The contribution associated with the **multiplicity** and **distribution** of the minima is the **landscape entropy** [19–21], or inherent structure [22] entropy, S_{IS} . **Rotational** contributions can also be included, but are usually dominated by the **vibrational** component, which is often treated in a **harmonic** normal mode approximation [1].

The landscape contribution is **convoluted** with the local vibrational densities of states to obtain the full **configurational** entropy, S_c .

We can calculate S_{IS} by sampling the **potential energy density** of local minima [21,23]. (Note that some authors refer to S_{IS} as the configurational entropy.)

We can also consider the local densities of states associated with distinct **pathways** between **product** and **reactant** states, and define a **pathway entropy** from the multiplicity of **alternative routes** mediated by different transition states [24–26].

Funnels, Relaxation, and Self-Organisation

The number of local minima and transition states increase **exponentially** with system size, [5, 22, 27] and the number of transition states per minimum is expected to increase **linearly** [4]. This increase in connectivity **cannot** be represented properly by surfaces in three dimensions [26].

However, **efficient relaxation** to the **global minimum** can be encoded in the landscape if there is a single **funnel** structure, corresponding to a set of kinetically convergent pathways [28]. This is the **organisation** we expect for naturally occurring **proteins** that have a single function [29, 30], and for ‘**magic number**’ clusters, such as buckminsterfullerene [31, 32].

In contrast, landscapes with distinct low energy structures **separated** by high barriers are considered ‘**frustrated**’ [29, 30]. We could also describe such multifunnel landscapes as ‘rugged’, but some authors associate this term with the existence of many local minima.

The existence of more local minima does **not** necessarily lead to more complex behaviour, since funnelled landscapes also exist for **larger** systems. Hence it may be best to refer to **multifunnel** landscapes as **frustrated**.

The presence of multiple low energy morphologies provides a mechanism to encode **multifunctional** properties [33]. Such multifunnel landscapes have been characterised for certain **biomolecules** [34, 35], which may constitute **evolved** multifunctional systems. Hence we have a **design principle** for achieving targeted properties, such as **switches**, for potential technological applications.

Disconnectivity Graphs

Disconnectivity graphs [31, 36] provide a powerful way to visualise the organisation of the energy landscape, and they can be constructed directly from a database of local minima and the transition states [2] that connect them. The contrast between single and multifunnel landscapes is clearly highlighted in this construction. A single funnel landscape suggests that the system will support efficient self-organisation, in terms of relaxation to the global minimum, while a molecular switch would probably correspond to a double funnel. Multifunnel landscapes, with competing alternative morphologies separated by a high barrier, produce challenges for sampling, with broken ergodicity and rare event dynamics, associated with multiple relaxation time scales [37, 38].

In these graphs the branches terminate at the energies of local minima, and the vertical scale is the potential (or free) energy. At regular energy thresholds we identify the sets of minima (superbasins) that can interconvert amongst

themselves without exceeding the threshold. The **segregation** into these disjoint sets of minima at each threshold is obtained from analysis of the database of minima and transition states, which constitutes a kinetic transition network [15, 17, 39, 40]. The superbasins of minima (nodes) **split** into smaller sets as the threshold decreases, and ultimately terminate at the energy of individual minima on the vertical axis.

Global thermodynamic and kinetic properties are calculated from the database of minima and transition states using standard tools of **statistical mechanics** and **unimolecular rate theory**, based on the energetics of these stationary points and their densities of states. The disconnectivity graphs provide a complementary **visualisation** of how the landscape is **organised**. Here the nodes are **ordered** on the horizontal axis according to how they **split** from the parent node in the level above. The children of each parent node can be **ordered** so that the superbasins containing the **lowest** energy minima are in

the **middle** of each set; alternatively the superbasins with the most minima can be positioned centrally. The relative **spacing** on the horizontal axis is chosen according to the number of minima in each superbasin to allow room for further branching at lower energy thresholds.

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