1 The Wang-Landau algorithm (density of states)

We determine thermodynamic quantities from the partition function by obtaining the density of states from a simulation.

```
from numba import njit, jit_module
    import numpy as np
    import copy # for parallel systems
    import os, struct # for using `urandom`
         Utility functions.
    @njit(cache=True, inline='always')
    def bisect_right(a, x, lo=0, hi=None):
        if lo < 0:
             raise ValueError('lo must be non-negative')
        if hi is None:
            hi = len(a)
        while lo < hi:</pre>
            mid = (lo + hi) // 2
            if x < a[mid]:</pre>
                 hi = mid
10
11
             else:
                 lo = mid + 1
12
        return lo
13
    @njit(cache=True)
15
    def binindex(a, x):
        return bisect_right(a, x, hi=len(a) - 1) - 1
17
    @njit(cache=True)
    def flat(H, \varepsilon = 0.2):
        """Determines if a histogram is approximately flat to within \varepsilon of the mean height."""
        return not np.any(H < (1 - \epsilon) * np.mean(H)) and np.all(H \neq \theta)
```

1.1 Algorithm

A Wang-Landau algorithm, with quantities as logarithms and with monte-carlo steps proportional to $f^{-1/2}$ (a "Zhou-Bhat schedule").

We use energy bins encoded by numbers E_i for $i \in [0, N]$, so that there are N bins. The energies E covered by bin i satisfy $E_i \le E < E_{i+1}$. For the bounded discrete systems that we are considering, we must choose E_N to be an arbitrary number above the maximum energy.

```
Qnjit
    def wanglandau(system,
                                       # The energy bins
                      Es,
                      M = 1_00_000, # Monte carlo step scale
                      \varepsilon = 1e-10,
                                       # f tolerance
                      logf0 = 1,
                                       # Initial log f
                      flatness = 0.1, # Desired histogram flatness
                      logging = False # Log progress of f-steps
         if M \leq 0 or \epsilon \leq 1e-16 or not (0 < logf0 \leq 1) or not (0 \leq flatness < 1):
10
             raise ValueError('Invalid Wang-Landau parameter.')
11
12
         # Initial values
           Es = system.Es # Testing
14
         E0 = Es[0]
         Ef = Es[-1]
16
         N = len(Es) - 1
         logf = 2 * logf0
18
         logftol = np.log(1 + \epsilon)
         S = np.zeros(N) # Set all initial g's to 1
20
         H = np.zeros(N, dtype=np.int32)
         i = binindex(Es, system.E)
22
         converged = True
23
         if logging:
25
             mciters = 0
26
             fiter = 0
27
             fiters = int(np.ceil(np.log2(logf0) - np.log2(logftol)))
28
             print("Wang-Landau START")
29
         while logftol < logf:</pre>
31
             H[:] = 0
32
             logf /= 2
33
             iters = 0
34
             niters = int((M + 1) * np.exp(-\log f / 2))
35
             if logging:
36
37
             while not flat(H, flatness) and iters < niters:</pre>
38
                 system.propose()
                 Ev = system.Ev
40
                 j = binindex(Es, Ev)
                 if E\theta~\leq~E\nu~<~Ef and (
42
                      S[j] < S[i] or np.random.rand() < np.exp(S[i] - S[j]):
                      system.accept()
44
                      i = j
```

```
H[i] += 1
46
                 S[i] += logf
47
                 iters += 1
48
             mciters += iters
49
             if niters \leq iters:
                 converged = False
51
             if logging:
                 print("f: ", fiter, " / ", fiters, "\t(", iters, " / ", niters, ")")
53
        if logging:
55
             print("Done: ", mciters, " total MC iterations.")
56
        return Es, S, H
57
```

1.1.1 Parallel construction of the density of states

```
Onjit
    def find_bin_systems(system, Es, Ebins, N = 1_000_000, method = 'wl'):
        Find systems with energies in the bins given by `Es` by stepping `sys`.
4
            system: The initial system to search from. This is usually a ground state.
            Es: The energies of the system.
            Ebins: The energy bins to find systems for.
            N: The maximum number of steps to try.
10
            method: The string name of the search method to try.
                 'wl': Wang-Landau steps where we prefer energies we have not visited
                'increasing': Only accept increases in energy. This only works for
                    steps that are not trapped by local maxima of energy.
14
        Returns:
16
17
            A list of independent systems with energies in Ebins.
18
        Raises:
            ValueError: The method argument was invalid.
20
            RuntimeError: Bin systems could not be found after N steps.
21
        if method == 'wl':
23
            S = np.zeros(len(Es), dtype=np.int32)
        systems = [None] * (len(Ebins) - 1)
25
        n = 0
        1 = len(Ebins) - 1
27
        systems = [system] * 1
28
        empty = np.repeat(True, 1)
        i = binindex(Es, system.E)
```

```
while np.any(empty) and n < N:</pre>
31
             for s in range(1):
32
                 if empty[s] and Ebins[s] \leq system.E < Ebins[s + 1]:
33
                      systems[s] = system.copy()
34
                      empty[s] = False
36
             system.propose()
             j = binindex(Es, system.Ev)
38
             if method == 'wl':
                 if S[j] < S[i]:
40
                      i = j
                      system.accept()
42
                 S[i] += 1
43
             elif method = 'increasing':
44
                 if system.E < system.Ev:</pre>
45
                      system.accept()
             else:
47
                  raise ValueError('Invalid method argument for finding bin systems.')
48
             n += 1
49
         if N \leq n:
51
             raise RuntimeError('Could not find bin systems (hit step limit).')
52
         return systems
53
```

We can choose overlapping bins for the parallel processes to negate boundary effects.

```
def extend_bin(bins, i, k = 0.05):

if len(bins) ≤ 2: # There is only one bin

return bins

k = max(0, min(1, k))

return (bins[i] - (k*(bins[i] - bins[i-1]) if 0 < i else 0),

bins[i+1] + (k*(bins[i+2] - bins[i+1]) if i < len(bins) - 2 else 0))
```

Now we can construct our parallel systems.

```
def parallel_systems(system, Es, bins = 8, overlap = 0.1, steps = 1_000_000):
    Ebins = np.linspace(Es[0], Es[-1], bins + 1)
    systems = find_bin_systems(system, Es, Ebins, steps)
    states = [s.state() for s in systems]
    binEs = [(lambda E0, Ef: Es[(E0 ≤ Es) & (Es ≤ Ef)])(*extend_bin(Ebins, i, overlap))
    for i in range(len(Ebins) - 1)]
    return zip(states, binEs)
```

We also need a way to reset the random number generator seed in a way that is time-independent and different for each process.

```
def urandom_reseed():
    """Reseeds numpy's RNG from `urandom` and returns the seed"""
    seed = struct.unpack('I', os.urandom(4))[0]
    np.random.seed(seed)
    return seed
```

Once we have parallel results, we stitch the pieces of $\ln g(E)$ together.

```
def stitch_results(wlresults):
        E0, S0, _ = wlresults[0]
        E, S = E0, S0
        for i in range(1, len(wlresults)):
            Ev, Sv, _ = wlresults[i]
            # Assumes overlap is at end regions
            _, i0s, ivs = np.intersect1d(E0[:-1], Ev[:-1], return_indices=True)
            # Simplest: join middles of overlap regions
            l = len(i0s)
            m = 1 // 2
10
            Sv = Sv[ivs[m]] - S0[i0s[m]]
11
            # Simplest: average the overlaps to produce the final value
            E = np.hstack((E, Ev[1+1:]))
            S[-1:] = (Sv[ivs] + SO[i0s]) / 2
            S = np.hstack((S, Sv[1:]))
            E0, S0 = E\nu, S\nu
        return E, S
```