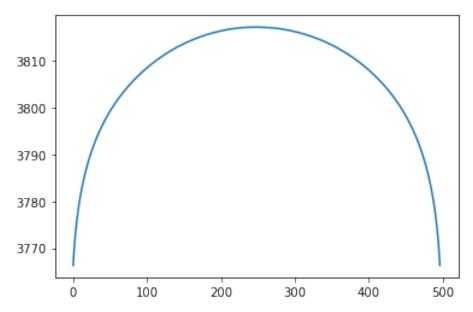
0.1 Thermal calculations on images

```
from numba import njit
    from numba.experimental import jitclass
    from numba import int64
    integer = int64
    import numpy as np
    from scipy import interpolate, special
    import os
    import tempfile
    import h5py, hickle
    import pprint
    import sys
    if 'src' not in sys.path: sys.path.append('src')
    import simulation as sim
    import wanglandau as wl
     0.1.1 Parallel Simulation
    N = 16
    M = 2**5 - 1
    Moff = 0
    I0 = Moff * np.ones(N, dtype=int)
     system_params = {
         'StatisticalImage': {
            'I0': I0,
            'I': I0.copy(),
             'M': M
10
    }
11
   # L = 16
   # system_params = {
          'IsingModel': {
              'spins': np.ones((L, L), dtype=int)
    #
5
    #
    # }
    params = {
         'system': system_params,
        'simulation': {
            'max_sweeps': 10_000_000,
            'flat_sweeps': 10_000,
            'eps': 1e-9,
            'logf0': 1,
            'flatness': 0.1
        },
         'parallel': {
            'bins': 2,
11
            'overlap': 0.25,
            'sweeps': 1_000_000
14
         'save': {
15
            'prefix': 'simulation-',
16
            'dir': 'data'
        }
18
```

19

```
params.pop('parallel', None) # Single run
wlresults = wl.run(params, log=True)
Run parameters
'M': 31}},
 'simulation': {'max_sweeps': 10000000,
              'flat_sweeps': 10000,
              'eps': 1e-09,
              'logf0': 1,
              'flatness': 0.1},
 'save': {'prefix': 'simulation-', 'dir': 'data'},
 'log': True}
Running ...
Wang-Landau START
fiter
        steps
                  max steps
1
    1120000
               97044906
2
    480000
               124608126
    800000
3
               141199505
4
    800000
               150306090
5
    1280000
               155077318
    1600000
               157519430
6
7
    2080000
               158754871
8
    3360000
               159376220
9
    3360000
               159687805
10
    4320000
               159843827
11
    5280000
               159921895
12
    10400000
               159960943
13
    19680000
               159980470
    17120000
14
               159990235
15
    28640000
               159995118
    17440000
               159997559
16
17
    98880000
               159998780
    50240000
18
               159999390
19
    35840000
               159999695
    159999848
20
               159999848
21
    36320000
               159999924
22
    88960000
               159999962
23
    63680000
               159999981
24
    38880000
               159999991
25
    30880000
               159999996
26
    125920000
               159999998
```

```
27
     90560000
                   159999999
28
     68800000
                   160000000
29
     19200000
                   160000000
30
     26720000
                   160000000
31
     57920000
                   160000000
Done: 1110559848 total MC iterations; not converged.
... done in 122 seconds.
Writing results ... done: data/simulation-v7qs5a48.h5
wlEs, S, \DeltaS = wl.join_results(wlresults['results'])
0.1.2 Results
import matplotlib.pyplot as plt
N, M = len(system_params['StatisticalImage']['I0']), system_params['StatisticalImage']['M']
for i, r in enumerate(wlresults['results']):
   plt.plot(r['Es'][:-1], r['S'] + ΔS[i])
```



Fit a spline to interpolate and optionally clean up noise, giving WL g's up to a normalization constant.

```
gspl = interpolate.splrep(w1Es, S, s=0*np.sqrt(2)) wlgs = np.exp(interpolate.splev(w1Es, gspl) - min(S))
```

0.1.3 Exact solution

We only compute to halfway since g is symmetric and the other half's large numbers cause numerical instability.

```
def reflect(a, center=True):
    if center:
        return np.hstack([a[:-1], a[-1], a[-2::-1]])
    else:
        return np.hstack([a, a[::-1]])
```

The exact density of states for uniform values. This covers the all gray and all black/white cases. Everything else (normal images) are somewhere between. The gray is a slight approximation: the ground level is not degenerate, but we say it has degeneracy 2 like all the other sites. For the numbers of sites and values we are using, this is insignificant.

```
def bw_g(E, N, M, exact=True):
    return sum((-1)**k * special.comb(N, k, exact=exact) * special.comb(E + N - 1 - k*(M + 1), E - k*(M + → 1), exact=exact)
    for k in range(int(E / M) + 1))

def exact_bw_gs(N, M):
    Es = np.arange(N*M + 1)
    gs = np.vectorize(bw_g)(np.arange(1 + N*M // 2), N, M, exact=False)
    return Es, reflect(gs, len(Es) % 2 = 1)

def gray_g(E, N, M, exact=True):
    return 2 * bw_g(E, N, M, exact=exact)
    def exact_gray_gs(N, M):
    Es = np.arange(N*M + 1)
    gs = np.vectorize(gray_g)(np.arange(1 + N*M // 2), N, M, exact=False)
    return Es, reflect(gs, len(Es) % 2 = 1)
```

Expected results for black/white and gray.

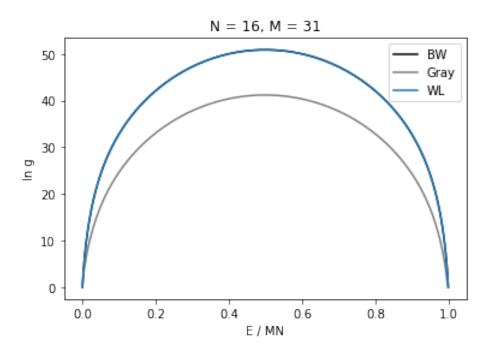
```
bw_Es, bw_gs = exact_bw_gs(N=N, M=M)
gray_Es, gray_gs = exact_gray_gs(N=N, M=-1 + (M + 1) // 2)
```

Choose what to compare to.

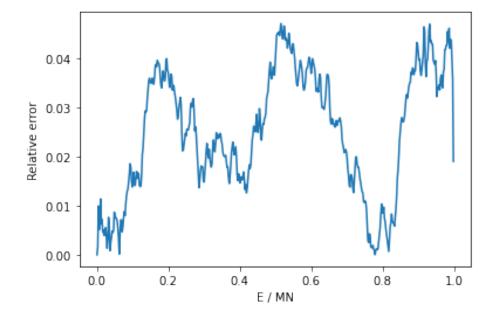
```
Es, gs = bw_Es, bw_gs
```

Presumably all of the densities of states for different images fall in the region between the all-gray and all-black/white curves.

```
plt.plot(bw_Es / len(bw_Es), np.log(bw_gs), 'black', label='BW')
plt.plot(gray_Es / len(gray_Es), np.log(gray_gs), 'gray', label='Gray')
plt.plot(wlEs / len(wlEs), S - min(S), label='WL')
plt.xlabel('E / MN')
plt.ylabel('In g')
plt.title('N = {}, M = {}'.format(N, M))
plt.legend();
```



```
plt.plot(wlEs / len(wlEs), np.abs(wlgs - bw_gs) / bw_gs)
plt.ylabel('Relative error')
# plt.plot(wlEs / len(wlEs), S - np.log(bw_gs) - min(S))
# plt.ylabel('Residuals')
plt.xlabel('E / MN');
```



print('End of job.')

End of job.