

```

import sys
import numpy as np

def load_parameter(fnm):
    k=[]
    q0=[]
    for rl in open(fnm,'r'):
        srl=rl[:-1].split()
        k.append(float(srl[3]))
        q0.append(float(srl[2]))

    return (np.array(k, dtype=np.float64), np.array(q0, dtype=np.float64))

```

```
def eval_ei(beta, k, q0, q_n):
```

```

nstate = len(k)
if nstate!=len(q0):
    print 'ERROR: data k and q0 do not match.'
    sys.exit()

n_q = len(q_n)
_e = np.zeros((nstate, n_q), dtype=np.float64)

for i in range(nstate):
    wi_q = (q_n - q0[i])*(q_n - q0[i])*0.5*k[i]
    _e[i,:] = np.exp((-beta)*wi_q)

return _e

```

```
def histogram(qdata, nbin, start_pos, binwidth):
```

```

Nr = np.zeros(nbin, dtype=np.float64)
q_n = np.zeros(nbin, dtype=np.float64)
lb = start_pos - 0.5*binwidth

```

```

for si in qdata:
    for di in si:

```

```

        __v = di - lb
        Nr[int(__v/binwidth)]+=1.0

```

```
return (Nr, np.array([start_pos + n*dq for n in range(nbin)], dtype=np.float64))
```

```

def determine_histogram_size(_min_q, _max_q, dq):
    # divide region [_min_q, max_q] into bins withwidth of dq
    # the center of bins should be multiple of dq

```

```

if _min_q<0.0:
    _a = -_min_q
    _b = int(_a/dq)*dq
    if _a > _b + 0.5*dq:
        _b+= dq # all the bins are defined as follows [q-0.5*q, q+0.5*dq) and q = N*dq
    _b=_b
else:
    _a = _min_q
    _b = (int(_a/dq)+1)*dq
    if _a< _b - 0.5*dq:
        _b-=dq
    _b=dq

return (_b, int((_max_q-_b+0.5*dq)/dq)+1)

```

#load the force constants and restraint position

## Main Code start Here

```
k, q0 = load_parameter('us.sh') # k, q0 ndarray
```

# key parameters

```
beta = 1/0.58
```

```
dq = 0.05
```

```
#load data points
```

#the data for each state are stored in a txt file in which each line list

# q and unbiased energy in each row at a given time

```
nstate = len(k)
```

```
print 'There are',nstate,'thermodynamic states'
```

```
print 'state parameters'
```

```
print 'k:'
```

```
print k
```

```
print 'q0:'
```

```
print q0
```

```
qdata =[]
```

```
edata =[]
```

```
overall_max=-10000.0
```

```
overall_min=10000.0
```

```
for i in range(nstate):
```

```
    _q, _e = np.loadtxt(fname='test%d.d.log'%i, dtype=np.float64, unpack=True, usecols=(1,2))
```

```
    qdata.append(_q)
```

```
    edata.append(_e)
```

```
    _max_q = max(_q)
```

```
    _min_q = min(_q)
```

```
    if _min_q< overall_min: overall_min = _min_q
```

```
    if _max_q> overall_max: overall_max = _max_q
```

#data count for each simulation

```
n_r = np.array([len(i) for i in qdata], dtype=np.int32)
```

```
print 'state frame count:'
```

$\sum_i e^{-\beta W_i(q_m)} \cdot \frac{\sum_j H_j(q_m)}{\sum_k e^{-\beta W_k(q_m)}} \cdot \frac{1}{Z_{R'}}$

If doesn't matter where data come from as long as  $q \in [q_m, q_m + dq]$

$Z_{R'} = \sum_k e^{-\beta W_k(q_m)}$

(load data

in col 1, col 2  
into two np array  
 $-q, -e$

$M_k \leftrightarrow n_r$

```

print n_r
# special for histogram
print 'min data', overall_min
print 'max data', overall_max
start_pos, nbin = determine_histogram_size(overall_min, overall_max, dq)
print 'starting position of the first bin', start_pos
print 'number of bins', nbin
# start_pos should be multiple times of dq, nbin should be sufficient to cover all data
# 1d array Nr(q) = count frames in which q in bin [qi-dq/2, qi+dq/2]
# q_n center positions of each bin
Nr, q_n = histogram(qdata, nbin, start_pos, dq)
print 'Histogram:'
print Nr
print 'q of bins:'
print q_n
# initial guess all Z=1
Z = np.ones(nstate, dtype=np.float64)
Z_old = np.ones(nstate, dtype=np.float64)
# eval e_i = {exp(-beta w_i(q1)), exp(-beta w_i(q2)), ...}
# e[i,j] = exp(-beta w_i(qj))
# w_i(qj) = 0.5*k_i*(qj - qj0)^2
_e = eval_ei(beta, k, q0, q_n)
print 'e_i:'
print _e
NR = np.zeros(nbin, dtype=np.float64)
### NOW START OPT of Z###
step=0
while True:
    Z_1 = np.array([1./i for i in Z], dtype=np.float64)
    W = n_r*Z_1
    NR[:] = Nr[:]
    for j in range(nbin):
        T = (W*_e[:,j]).sum()
        NR[j] /= T
    for i in range(nstate):
        Z[i] = (_e[i,:]*NR).sum()
    Z_def = np.absolute((Z - Z_old)*Z_1)
    Z_old[:] = Z[:]
    step+=1
    if step%100==0: print step, max(Z_def)

```

define bin info  
adaptively:

$$Nr(q) = \sum_j H_j(q)$$

$q_n$

$$Z = \sum_i e^{-\beta w_i(q_n)}$$

Need to construct

$$Q_{ij} = e^{-\beta w_i(q_j)}$$

i state, position of jth bin

$$M_i = \frac{1}{Z_i}$$

$$M_k = \frac{1}{Z_k}$$

$$NR[j] = \frac{\sum_i H_i(q_j)}{\sum_k e^{-\beta w_k(q_j)} \frac{M_k}{Z_k}}$$

$$Z_i = \sum_m e^{-\beta w_i(q_m)} \cdot \frac{P_o(q_j) \Delta q}{P_o(q_m) \Delta q}$$

```

if max(Z_def)<1e-5: break

print 'optimized Z:'
for i in Z: print i

# calculate any quantity
# e.g. p_0(q)
# from the last round of opt, NR correspond to Z0*p(qj)*dq
Z_0 = NR.sum()      By constant Z_0 = 1
print 'Prob distr:'
for j in range(nbin):
    print q_n[j], NR[j]/Z_0/dq

```

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