

```
In [2]: ### %%matplotlib inline
import matplotlib.pyplot as plt
import numpy as np
```

```
In [3]: def oracle(i, z, f=lambda x: x):
        if f(i) > f(z):
            return -1
        else:
            return 1
```

Amplitude Amplification with a superposed threshold

We have register 1 $|a\rangle = \sum_i a_i |i\rangle$ that will be amplified compared with the threshold value $|b\rangle = \sum_v b_v |v\rangle$, in register 2.

Amplitude amplification with a threshold transforms the amplitude a_i of state $|i\rangle$, of the computational basis,

$$a'_i = \sum_j \left(\frac{2}{N} - \delta_{ij} \right) \Theta(j, v), \text{ where } \Theta(j, v) = \begin{cases} -1 & \text{if } F(j) > F(v) \\ 1 & \text{if } F(j) \leq F(v) \end{cases}$$

Thus for an initial state $|a\rangle |b\rangle$ we get the final state

$$|final\rangle = \sum_v \sum_i \left\{ b_v \sum_j a_j \left(\frac{2}{N} - \delta_{ij} \right) \Theta(j, v) \right\} |i\rangle |v\rangle$$

Uniform in $|a\rangle$, register 1

For a uniform distribution in $|a\rangle$, $a_i = \frac{1}{\sqrt{N}}$, we get the new probability distribution

$$P(i) = \sum_v \frac{|b_v|^2}{N} \left(2 - \Theta(i, v) - 4 \frac{t_v}{N} \right)^2$$

where t_v is the number of j values that satisfy $\Theta(j, v) = -1$, or $F(j) > F(v)$.

First we study this case for uniform a_i .

```

In [4]: # Some distributions

# Step distribution
def step(N, minval, maxval, step):
    bb = np.ones(N)
    if minval>0 and maxval<N:
        bb[minval:maxval] = np.sqrt(step / (maxval - minval))
        bb[:minval] = np.sqrt((1-step) / (N-minval+maxval))
        bb[maxval:] = np.sqrt((1-step) / (N-minval+maxval))
    elif minval == 0 and maxval<N:
        bb[:maxval] = np.sqrt(step / maxval)
        bb[maxval:] = np.sqrt((1-step) / (N+maxval))
    elif maxval == N and minval>0:
        bb[minval:] = np.sqrt(step / (N - minval))
        bb[:minval] = np.sqrt((1-step) / (2*N-minval))
    else:
        bb[:] = 1/np.sqrt(N)
    return bb

# linearly increasing distribution
# nu -> sqrt(nu) !!!
def lininc(N):
    return np.array( [ nu for nu in range(N)] ) * np.sqrt( 6 / N / (
N-1) / (2*N-1))

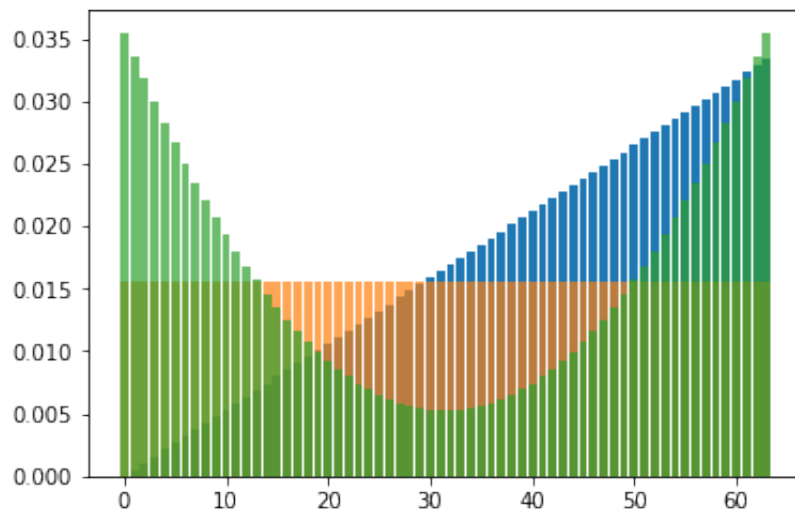
# linearly decreasing distribution
def lindec(N):
    return np.array( [ N-1-nu for nu in range(N)] ) * np.sqrt( 6 / N
/ (N-1) / (2*N-1))

# Poisson distribution
def poisson(N, lam):
    return np.sqrt(np.array( [ lam**nu / np.math.factorial(nu) for
nu in range(N)] ) * np.exp(-lam))

# Discreticed
def discreticed(N, iarr):
    bb = np.zeros(N)
    bb[iarr] = 1/np.sqrt(len(iarr))
    return bb

```

```
In [4]: N = 64  
ii = np.arange(N)  
pp = np.zeros(N)  
minval = 0  
bb = step(N, minval=minval, maxval=N, step=1)  
  
minval: 0  
Prob >49: 0.3469  
Prob <14: 0.3469  
Prob >57: 0.2108
```



```

In [5]: top = []
top_th = 50
bot = []
bot_th = 14

excel = []
excel_th = 57
good = []
good_th = 32

for mv in range(0, N):
    pp = np.zeros(N)
    bb = step(N, minval=mv, maxval=N, step=1)

    function = lambda x: x

    for i in ii:
        for nu in range(N):
            pp[i] += bb[nu]**2 / N * (2 - oracle(i, nu, f=function)
- 4 * (N-1-nu)/N )**2

    top.append(sum(pp[top_th:]))
    bot.append(sum(pp[:bot_th]))

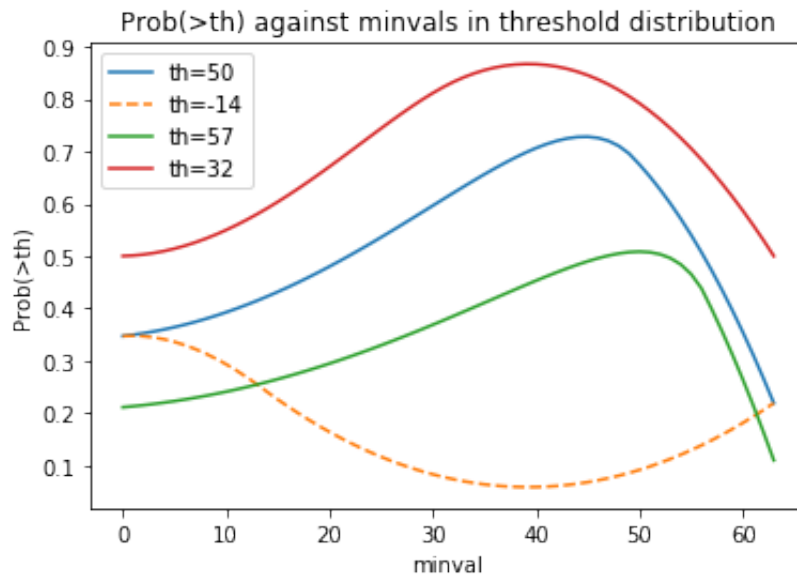
    excel.append(sum(pp[excel_th:]))
    good.append(sum(pp[good_th:]))

plt.plot(top, label='th={}'.format(top_th))
plt.plot(bot, '--', label='th=-{}'.format(bot_th))
plt.plot(excel, label='th={}'.format(excel_th))
plt.plot(good, label='th={}'.format(good_th))

plt.legend()
plt.title('Prob(>th) against minvals in threshold distribution')
plt.ylabel('Prob(>th)')
plt.xlabel('minval')

plt.show()

```



```
In [6]: # How does the position of the maximum vary?
sup = []
isup = []

for thres in range(0, N):
    top = []
    for mv in range(0, N):
        pp = np.zeros(N)
        bb = step(N, minval=mv, maxval=N, step=1)

        function = lambda x: x

        for i in ii:
            for nu in range(N):
                pp[i] += bb[nu]**2 / N * (2 - oracle(i, nu, f=func
ion) - 4 * (N-1-nu)/N )**2

        top.append(sum(pp[thres:]))

    sup.append(max(top))
    for i, s in enumerate(top):
        if s == sup[-1]:
            isup.append(i)
            break
```

```
In [7]: plt.plot(isup)
plt.title('The $minval$ with the maximum probability for with thres
hold $z$')
plt.ylabel('best $minval$')
plt.xlabel('$z$')
plt.show()

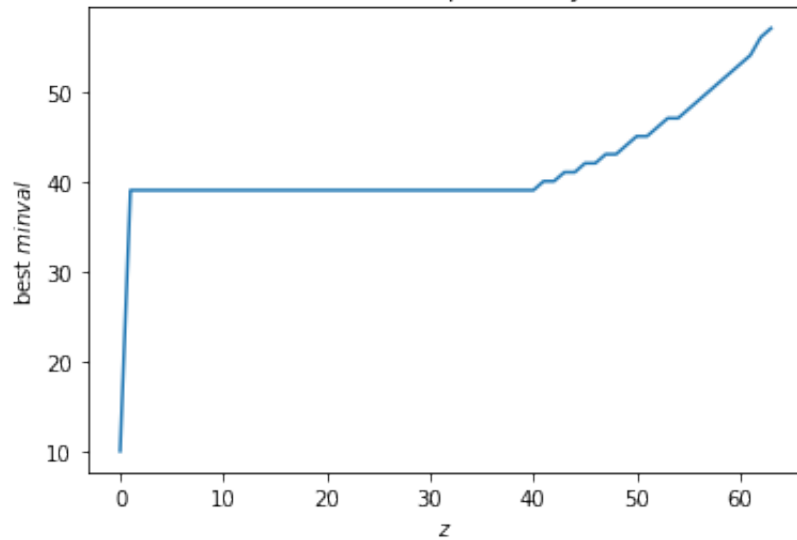
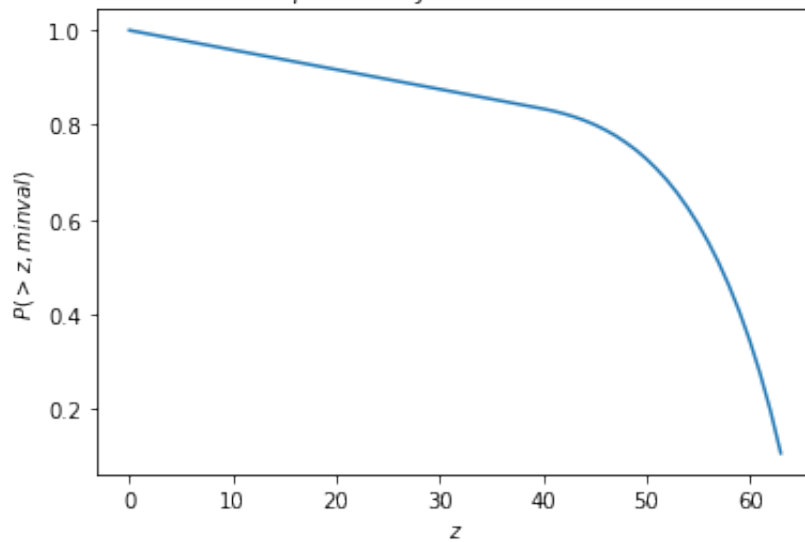
plt.plot(sup)
plt.title('The maximum $probability$ for an index with threshold $z
$')
plt.ylabel('$P(>z, minval)$')
plt.xlabel('$z$')
plt.show()

plt.plot(top, label='th={}'.format(top_th))
plt.plot(bot, '--', label='th=-{}'.format(bot_th))
plt.plot(excel, label='th={}'.format(excel_th))
plt.plot(good, label='th={}'.format(good_th))

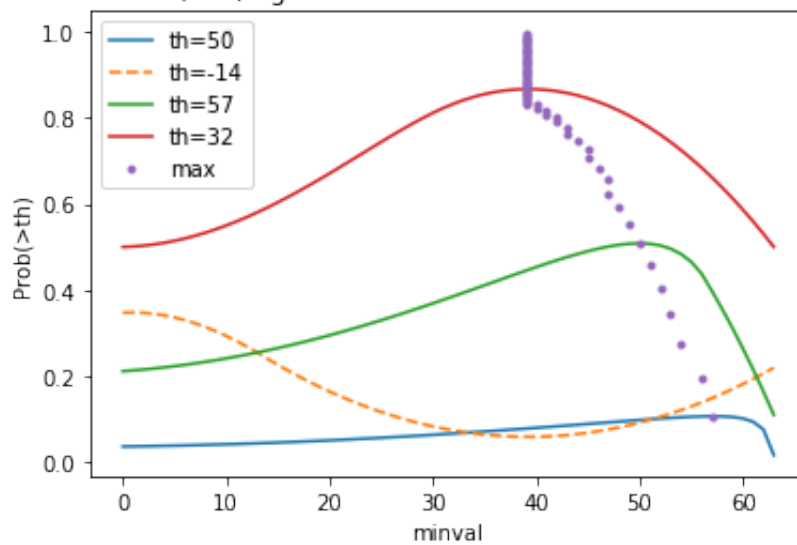
plt.plot(isup, sup, '.', label='max')

plt.legend()
plt.title('Prob(>th) against minvals in threshold distribution')
plt.ylabel('Prob(>th)')
plt.xlabel('minval')

plt.show()
```

The *minval* with the maximum probability for with threshold z The maximum *probability* for an index with threshold z 

Prob(>th) against minvals in threshold distribution



NON-uniform in $|a\rangle$, register 1

For a non-uniform distribution in $|a\rangle$, we get the new probability distribution

$$P(i) = \sum_v \left| b_v \sum_j a_j \left(\frac{2}{N} - \delta_{ij} \right) \Theta(j, v) \right|^2$$

```
In [6]: def prob(i, a, b, function=lambda x: x):
# assumes the given amplitudes are real
N = len(a)

p = 0
for nu in range(N):

    pnu = 0
    for j in range(N):
        pnu += a[j] * oracle(j, nu, function)
    pnu *= 2/N
    pnu -= a[i] * oracle(i, nu, function)
    pnu *= b[nu]

    p += pnu**2
return p
```

```

In [9]: N = 64
ii = np.arange(N)
pp = np.zeros(N)
mina, maxa = 0, N
minb, maxb = 45, 55
aa = step(N, minval=mina, maxval=maxa, step=1)
bb = discreticed(N, [50])#step(N, minval=minb, maxval=maxb, step=1)

function = lambda x: x

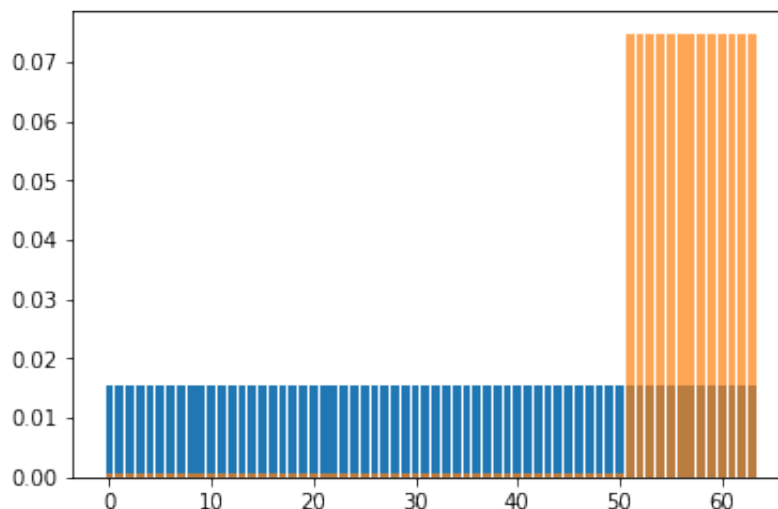
for i in ii:
    pp[i] += prob(i, aa, bb, function=function)

print(' {:10s} | {:10s} | {:10s} | {:10s} | {:10s}'.format('a range', 'b range', 'bad (<14)', 'good (>49)', 'excel (>57)'))
print(' {:10s} | {:10s} | {:10s} | {:10s} | {:10s}'.format('-'*10, '-'*10, '-'*10, '-'*10, '-'*10))
print(' [{:3}, {:3}] | [{:3}, {:3}] | {:10f} | {:10f} | {:10f}'.format(mina, maxa, minb, maxb, sum(pp[:14]), sum(pp[50:]), sum(pp[57:]))))

plt.bar(ii, aa*aa)
#plt.bar(ii, bb*bb, alpha=0.7)
plt.bar(ii, pp, alpha=0.7)
plt.show()

```

a range	b range	bad (<14)	good (>49)	excel (>57)
-----	-----	-----	-----	-----
[0, 64)	[45, 55)	0.007690	0.972534	0.523376



```

In [10]: N = 64
ii = np.arange(N)

print('| {:10s} | {:10s} | {:10s} | {:10s} | {:10s}|'.format('a range', 'b range', 'bad (<14)', 'good (>49)', 'excel (>57)'))

arange_arr = [[32, N], [50, N], [32, N], [32, N], [0, 32], [32, N], [45, 55]]
brange_arr = [[50, N], [32, N], [32, N], [0, 32], [32, N], [0, 50], [45, 55]]
for arange, brange in zip(arange_arr, brange_arr):
    pp = np.zeros(N)
    mina, maxa = arange
    minb, maxb = brange
    aa = step(N, minval=mina, maxval=maxa, step=1)
    bb = step(N, minval=minb, maxval=maxb, step=1)

    function = lambda x: x

    for i in ii:
        pp[i] += prob(i, aa, bb, function=function)

    print('| {:10s} | {:10s} | {:10s} | {:10s} | {:10s} |'.format('-'*10, '-'*10, '-'*10, '-'*10, '-'*10))
    print('| [{:3}, {:3}) | [{:3}, {:3}) | {:10f} | {:10f} | {:10f}|'.format(mina, maxa, minb, maxb, sum(pp[:14]), sum(pp[50:]), sum(pp[57:])))

```

a range	b range	bad (<14)	good (>49)	excel (>57)
[32, 64)	[50, 64)	0.182007	0.455444	0.350769
[50, 64)	[32, 64)	0.135864	0.514771	0.257385
[32, 64)	[32, 64)	0.146118	0.311890	0.209778
[32, 64)	[0, 32)	0.437500	0.000000	0.000000
[0, 32)	[32, 64)	0.000000	0.437500	0.218750
[32, 64)	[0, 50)	0.322554	0.072085	0.036042
[45, 55)	[45, 55)	0.046484	0.440234	0.023242

```
In [11]: N = 64

z = 50 # Has to be higher than 31 (N/2 - 1) to work
function = lambda x: x

ii = np.arange(N)
pp = np.zeros(N)

aa = step(N, minval=0, maxval=N, step=1)

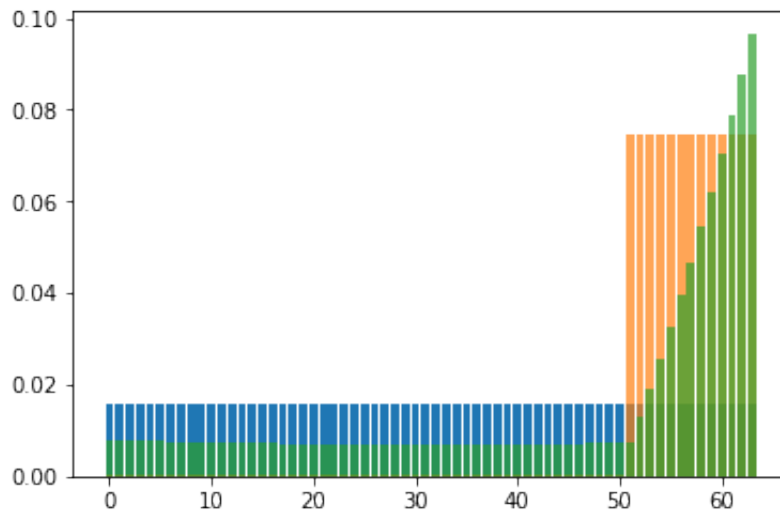
bb = np.zeros(N)
for nu in range(N):
    for mu in range(N):
        bb[nu] += oracle(mu, z, f=function)
    bb[nu] *= 2/N
    bb[nu] -= oracle(nu, z, f=function)
    bb[nu] /= np.sqrt(N)

for i in ii:
    pp[i] += prob(i, aa, bb, function=function)

print('Threshold (if any): ', z)
print('| {:10s} | {:10s} | {:10s}'.format('bad (<14)', 'good (>49)',
    , 'excel (>57)'))
print('| {:10s} | {:10s} | {:10s}'.format('-'*10, '-'*10, '-'*10))
print('| {:10f} | {:10f} | {:10f}'.format( sum(pp[:14]), sum(pp[50:
]), sum(pp[57:])))

plt.bar(ii, aa*aa)
plt.bar(ii, bb*bb, alpha=0.7)
plt.bar(ii, pp, alpha=0.7)
plt.show()
```

```
Threshold (if any): 50
| bad (<14) | good (>49) | excel (>57)
| ----- | ----- | -----
| 0.106184 | 0.641371 | 0.496820
```



Testing different fitness functions

There is a range where applying amplitude amplification amplifies the amplitude of a desired set, but depending on the state used and the number of solutions the result may be very different.

```

In [5]: def test_function(N, aa, funcname, function, z):
        ii = np.arange(N)
        bb = np.zeros(N)
        for nu in range(N):
            for mu in range(N):
                bb[nu] += oracle(mu, z, f=function)
            bb[nu] *= 2/N
            bb[nu] -= oracle(nu, z, f=function)
            bb[nu] /= np.sqrt(N)

        pp = np.zeros(N)
        for i in ii:
            pp[i] += prob(i, aa, bb, function=function)

        print('| {:12s} | {:12s} | {:12s} | {:12s} | {:12s}|'.format('f
unction', 'z & tz/N', 'bad (f<%20)', 'good (f>%80)', 'excel (f>%90)
'))
        print('| {:12s} | {:12s} | {:12s} | {:12s} | {:12s} |'.format('
-'*12, '-'*12, '-'*12, '-'*12, '-'*12))
        fp = [(function(i), pp[i]) for i in ii]
        fp.sort()
        bad = sum(fp[i][1] for i in range(N//5))
        good = sum(fp[i][1] for i in range(N-N//5, N))
        excel = sum(fp[i][1] for i in range(N-N//10, N))
        tz = sum(1 for i in range(N) if function(i)>function(z))
        print('| {:>12s} | {:3} & {:5.4f} | {:12f} | {:12f} | {:12f} |'
        .format(funcname, z, tz/N, bad, good, excel))

        plt.bar(ii, aa*aa, label='prob_reg1_initial')
        plt.bar(ii, bb*bb, alpha=0.7, label='prob_reg2_threshold')
        plt.bar(ii, pp, alpha=0.7, label='prob_reg1_final')
        plt.legend()
        plt.show()

        plt.title('The function')
        plt.bar(ii, function(ii))
        plt.show()

        plt.plot(function(ii), pp, '.')
        plt.ylabel('probability')
        plt.xlabel('fitness')
        plt.show()

```

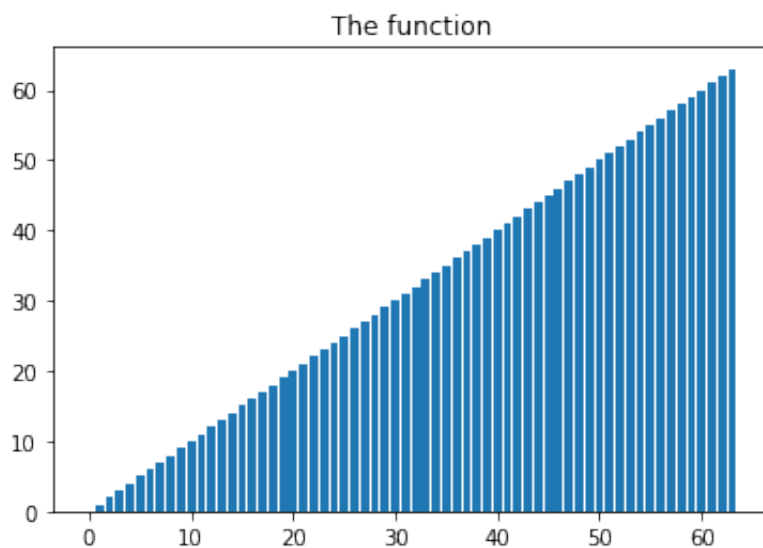
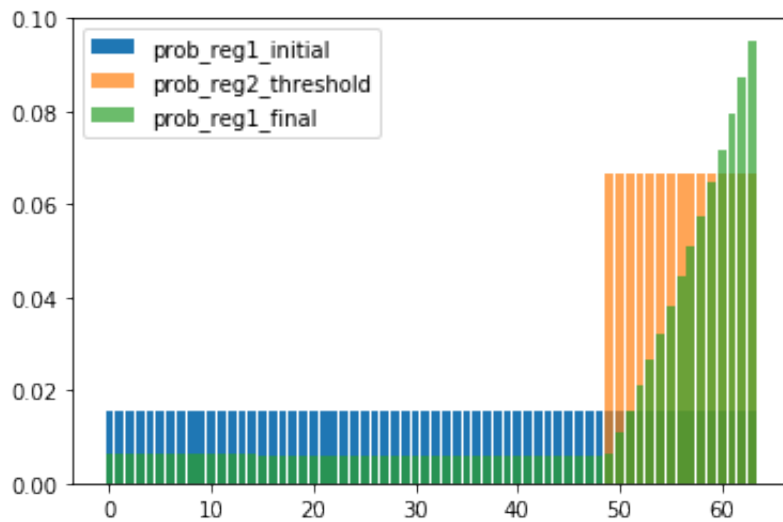
```

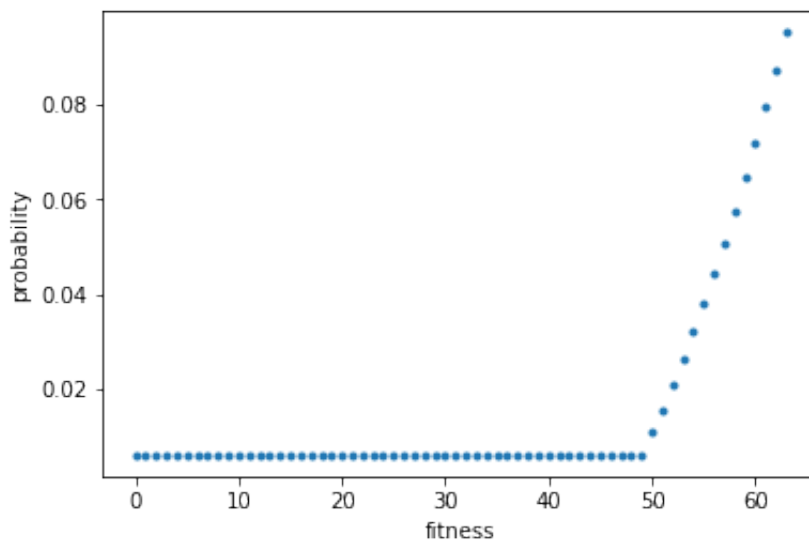
In [13]: N = 64
        aa = step(N, minval=0, maxval=N, step=1)
        test_function(N, aa, 'x', lambda x: x, 48)

```

function	z & tz/N	bad (f<%20)	good (f>%80)	exce
1 (f>%90)				
-----	-----	-----	-----	-----

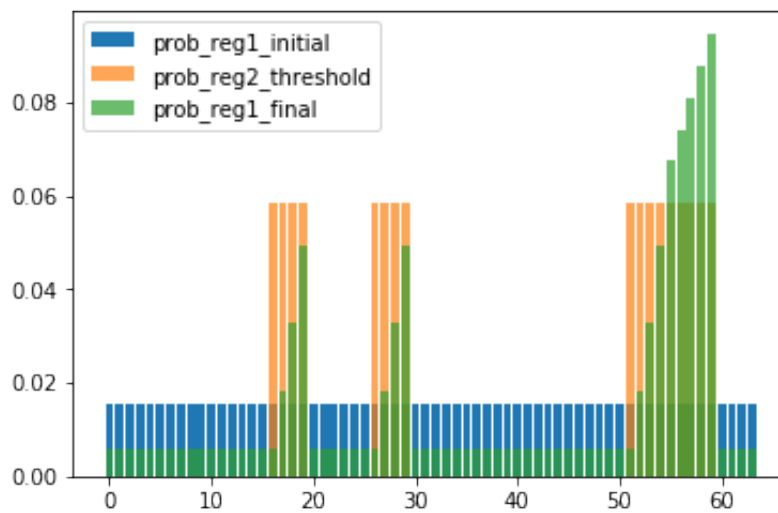
	x	48 & 0.2344	0.073941	0.667966
0.455493				

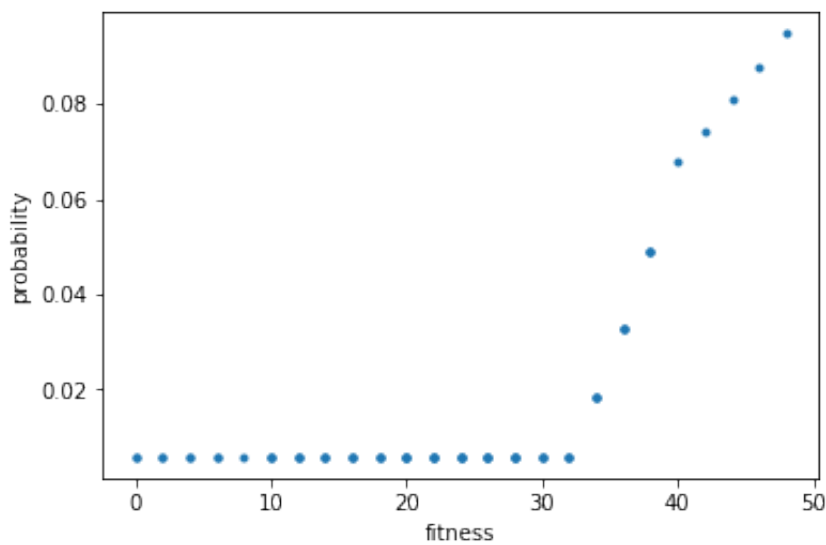
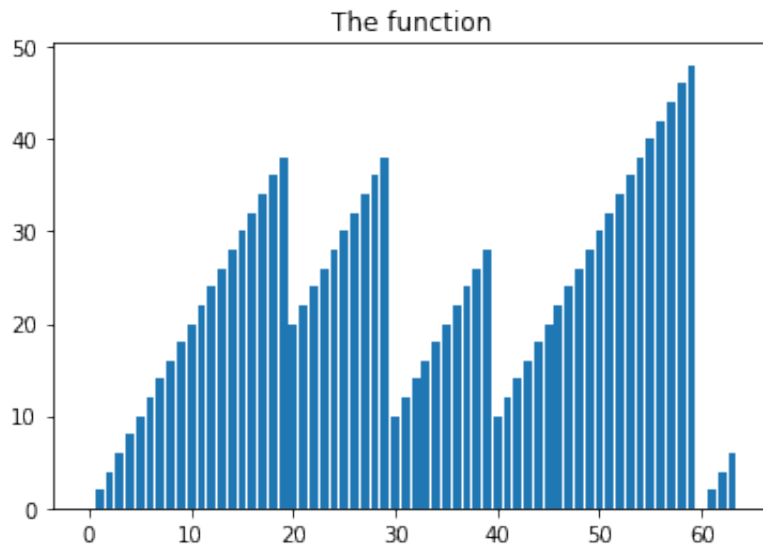




```
In [14]: N = 64
aa = step(N, minval=0, maxval=N, step=1)
test_function(N, aa, 'slopes', lambda x: x % 20 + x % 30, 50)
```

function	z & tz/N	bad (f<%20)	good (f>%80)	exce
1 (f>%90)				
slopes	50 & 0.2656	0.071228	0.669150	0.454405

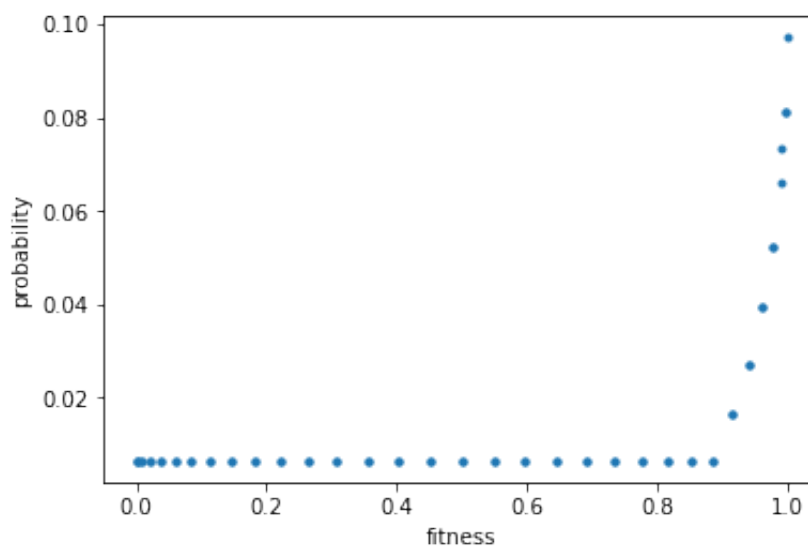
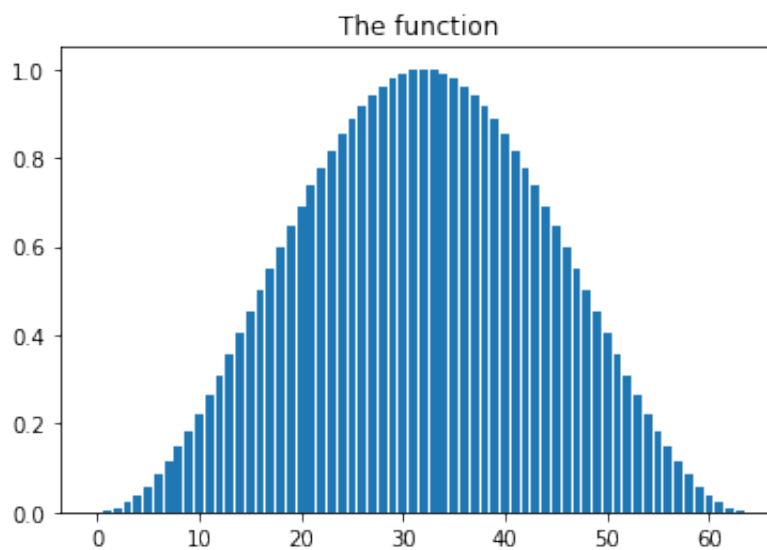
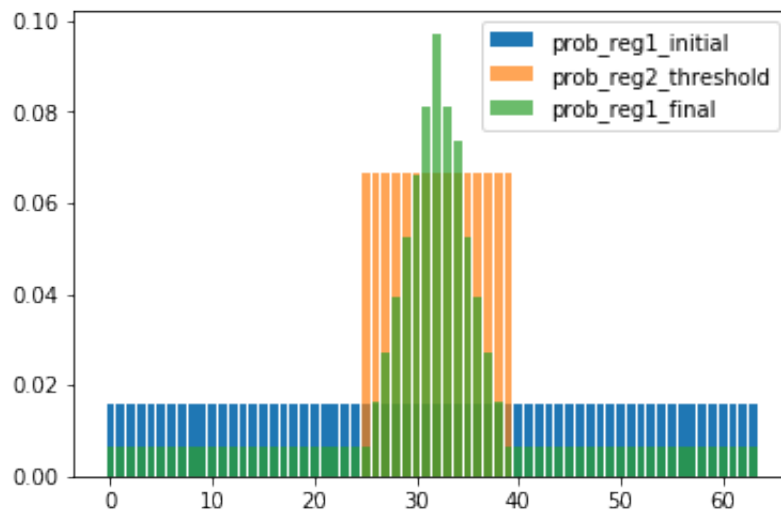




```
In [15]: N = 64
aa = step(N, minval=0, maxval=N, step=1)
test_function(N, aa, 'sin(pi*x/N)^2', lambda x: np.sin(np.pi*x/N)**
2, 40)
```

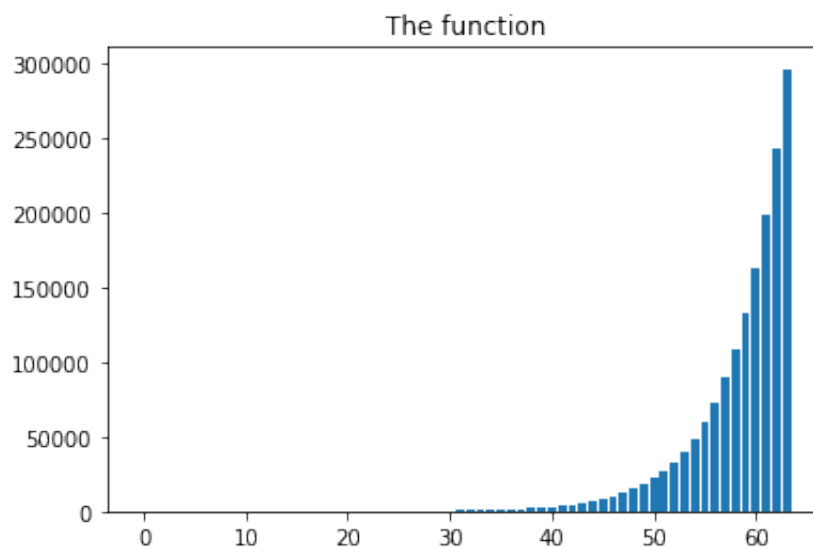
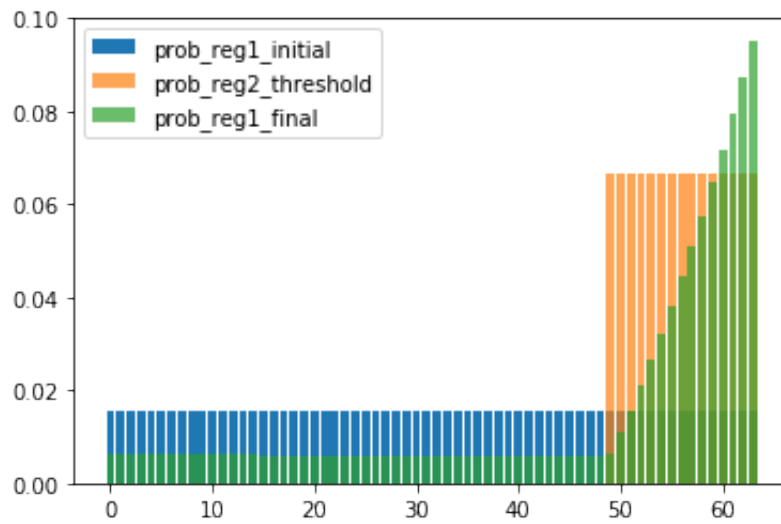
function	z & tz/N	bad (f<%20)	good (f>%80)	exce
1 (f>%90)				
-----	-----	-----	-----	----

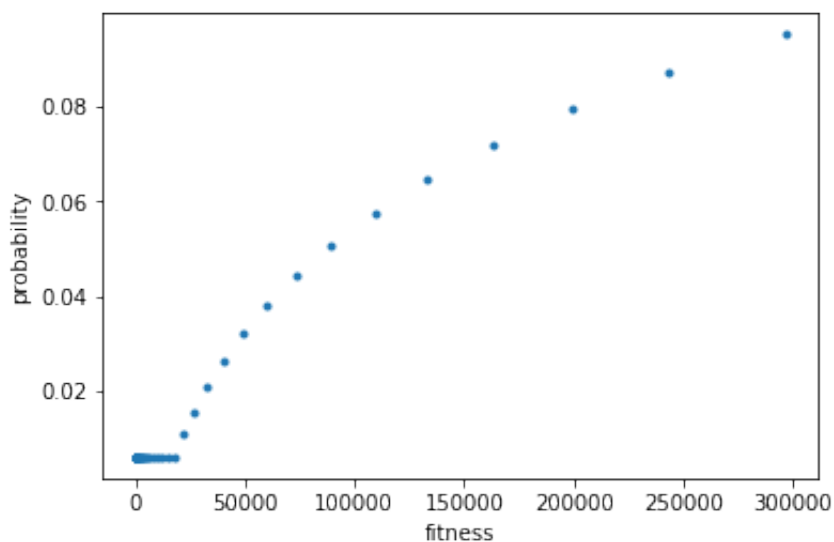
sin(pi*x/N)^2	40 & 0.2344	0.078509	0.652558	
0.451035				



```
In [16]: N = 64
aa = step(N, minval=0, maxval=N, step=1)
test_function(N, aa, 'exp(x/5)', lambda x: np.exp(x/5), 48)
```

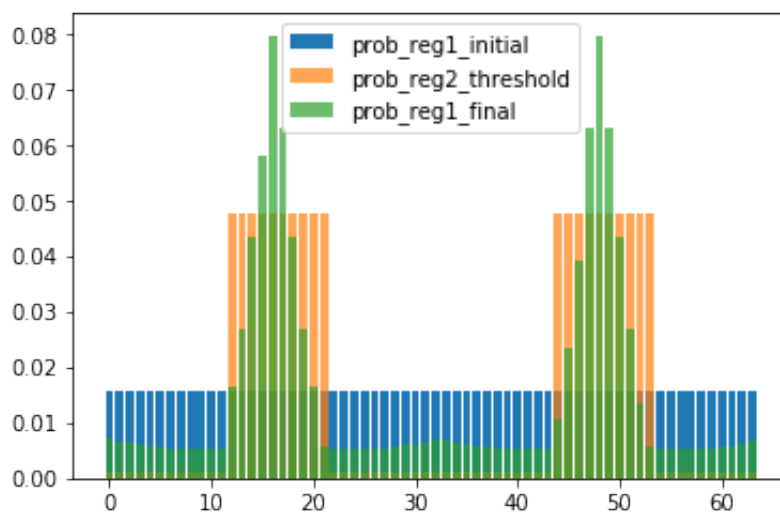
function	z & tz/N	bad (f<%20)	good (f>%80)	exce
1 (f>%90)				
exp(x/5)	48 & 0.2344	0.073941	0.667966	
0.455493				

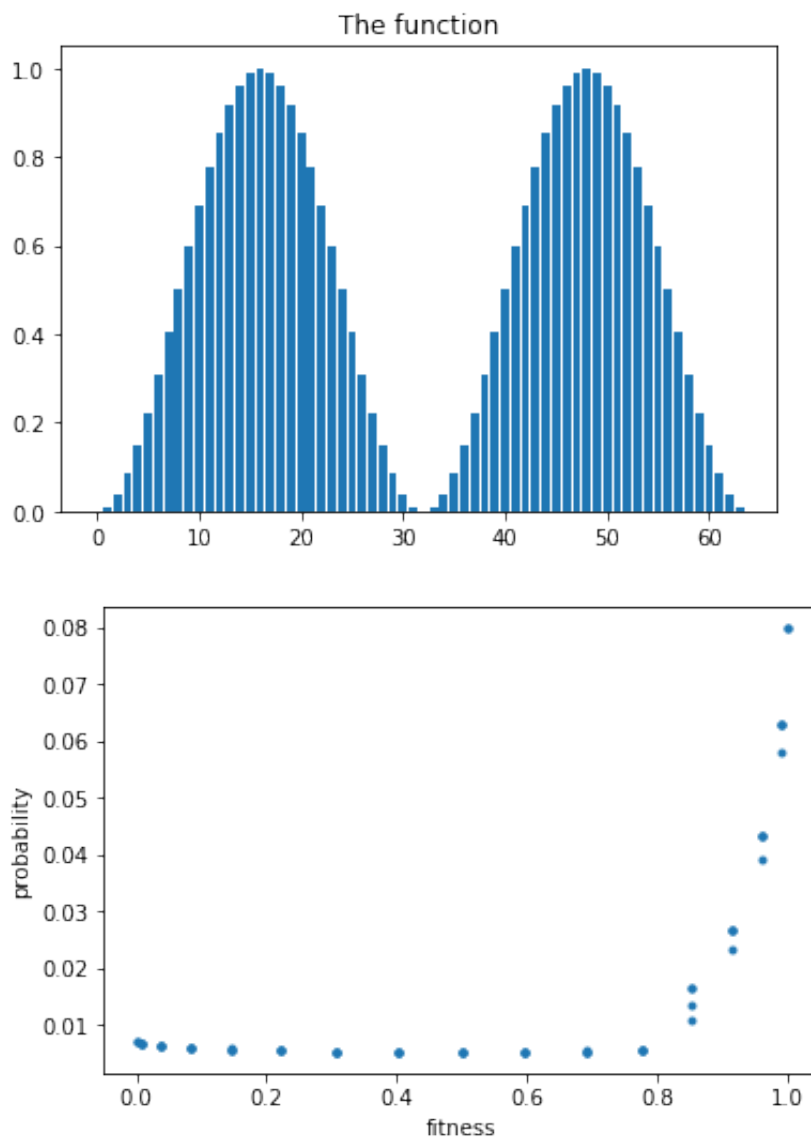




```
In [17]: N = 64
aa = step(N, minval=0, maxval=N, step=1)
test_function(N, aa, 'sin(2*pi*x/N)^2', lambda x: np.sin(2*np.pi*x/N)**2, 11)
```

function	z & tz/N	bad (f<%20)	good (f>%80)	exce
1 (f>%90)				
sin(2*pi*x/N)^2	11 & 0.3125	0.078030	0.630399	0.407164

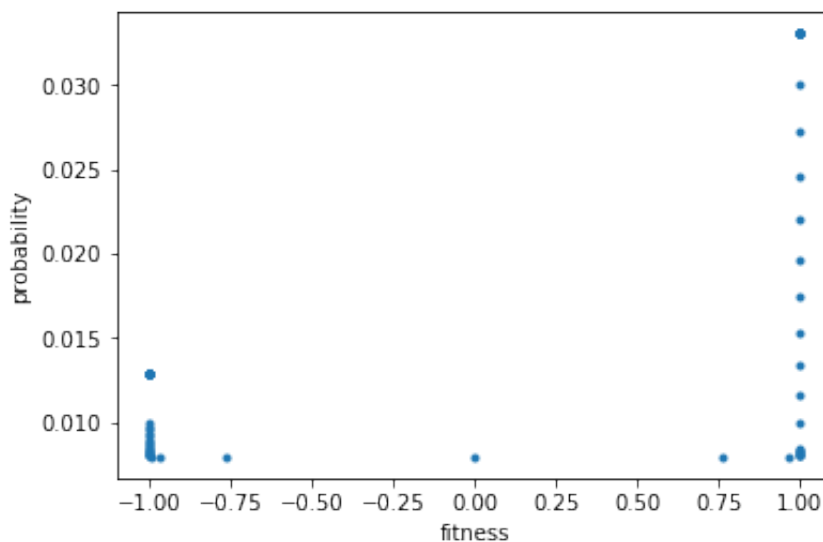
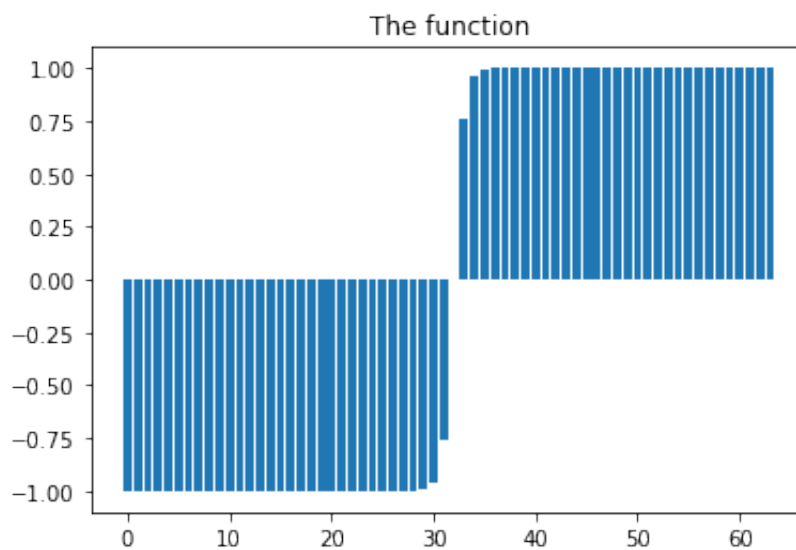
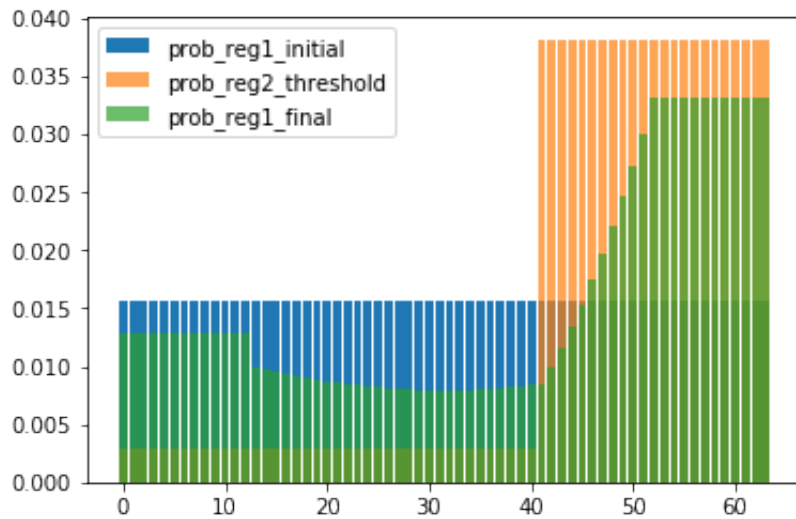




```
In [18]: N = 64
aa = step(N, minval=0, maxval=N, step=1)
test_function(N, aa, 'tanh((x-N/2))', lambda x: np.tanh((x-N/2)), 40
)
```

function	z & tz/N	bad (f<%20)	good (f>%80)	exce
1 (f>%90)				
-----	-----	-----	-----	----

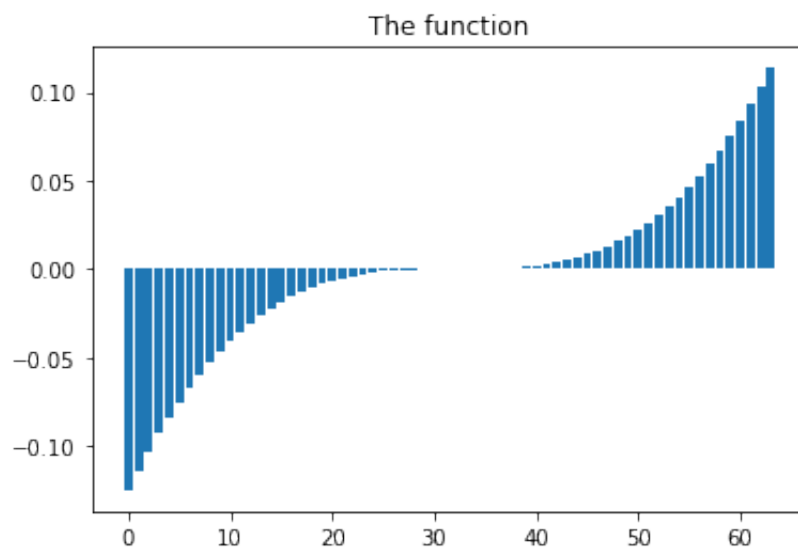
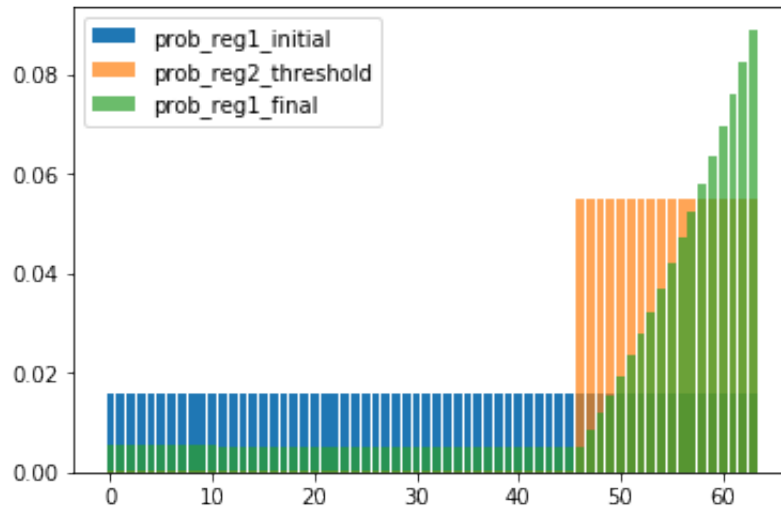
tanh((x-N/2))	40 & 0.3594	0.153872	0.396624	
0.198312				

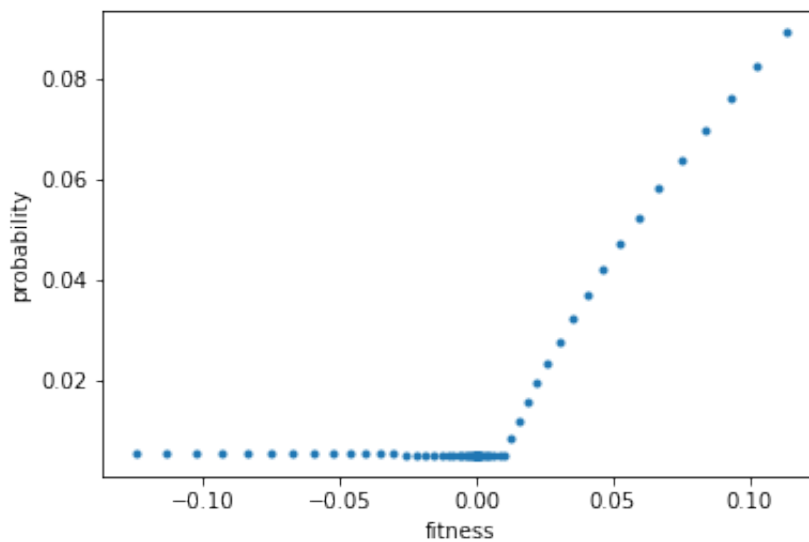


```
In [19]: N = 64
aa = step(N, minval=0, maxval=N, step=1)
test_function(N, aa, '(x/N-1/2)**3', lambda x: (x/N-1/2)**3, 45)
```

function	z & tz/N	bad (f<%20)	good (f>%80)	exce
1 (f>%90)				
-----	-----	-----	-----	----

(x/N-1/2)**3	45 & 0.2812	0.064875	0.677118	
0.438981				

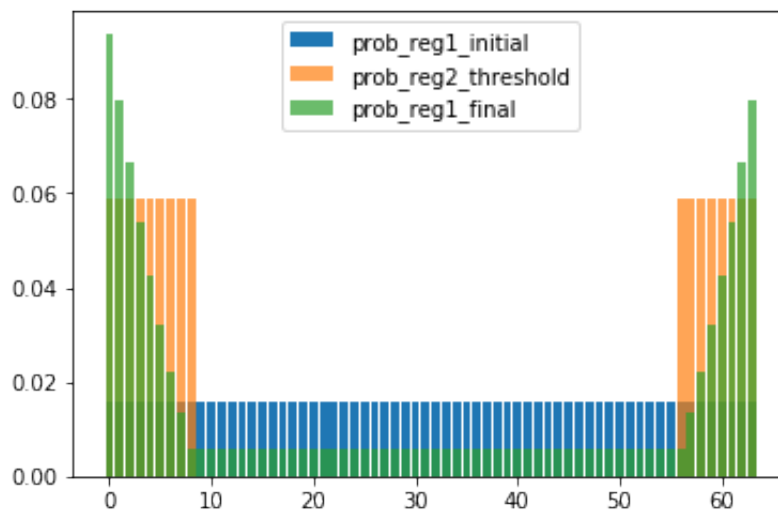


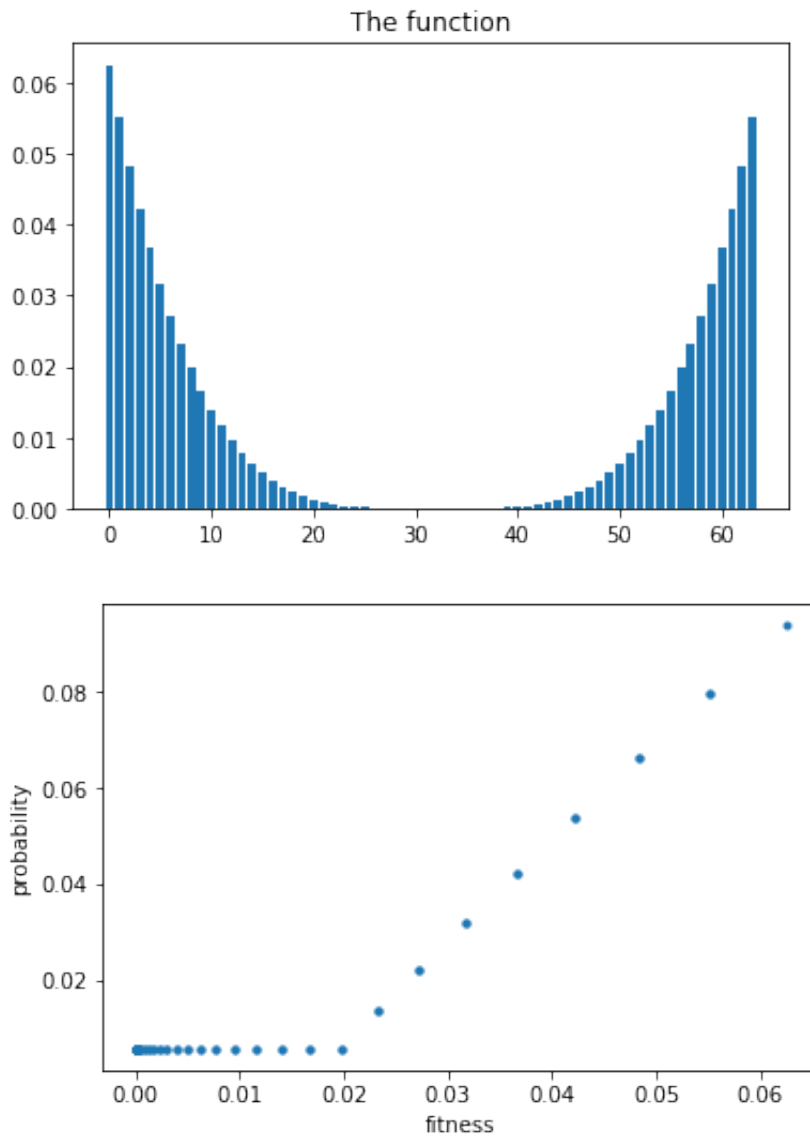


```
In [20]: N = 64
aa = step(N, minval=0, maxval=N, step=1)
test_function(N, aa, '(x/N-1/2)**4', lambda x: (x/N-1/2)**4, 55)
```

function	z & tz/N	bad (f<%20)	good (f>%80)	exce
1 (f>%90)				
-----	-----	-----	-----	----

(x/N-1/2)**4	55 & 0.2656	0.070574	0.664758	
0.439607				

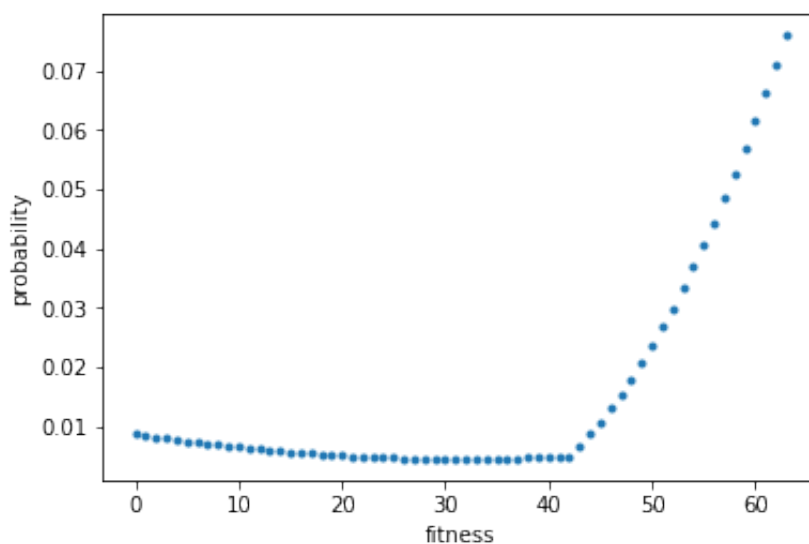
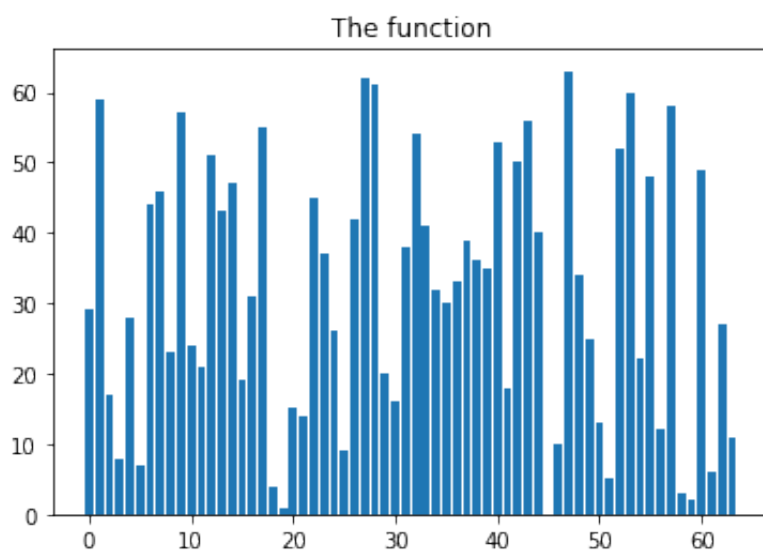
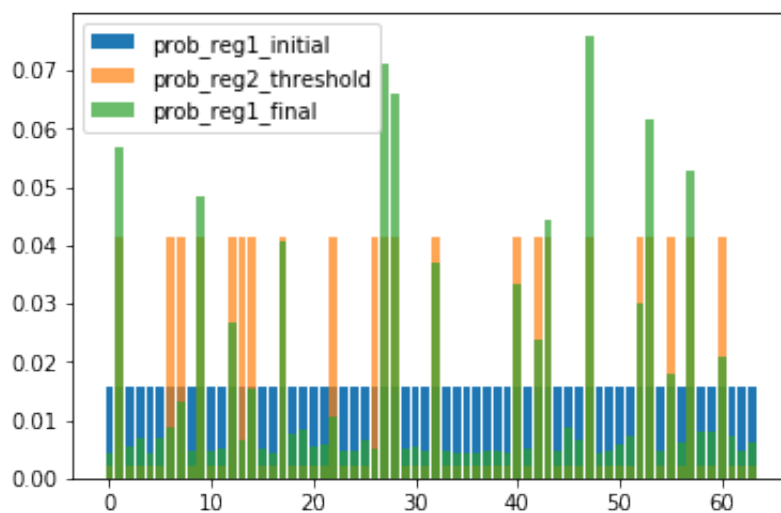




```
In [9]: N = 64
aa = step(N, minval=0, maxval=
          N, step=1)
array = np.array(range(N))
np.random.seed(12)
np.random.shuffle(array)
test_function(N, aa, 'unsorted', lambda x: array[x], 33)
```

function	z & tz/N	bad (f<%20)	good (f>%80)	exce
1 (f>%90)				
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unsorted	33 & 0.3438	0.088594	0.617464	
0.384160				



In []: