

TFY4235/FYS8904
Solution problemset 7 Spring 2015

Problem 1.

The programs that generate the sequence of random numbers are

Listing 1: r16807.f

```

1      program r16807
2      rinv=0.5/(2.**31-1)
3      ibm=4711333
4      do i=1,1000
5      ibm=ibm*16807
6      enddo
7      rold=0.
8      rnew=0.
9      do i=1,10000
10     100  rold=rnew
11         ibm=ibm*16807
12         rnew=float(ibm)*rinv+0.5
13         if(rold.lt.0.001.and.rnew.lt.0.001) then
14         write(*,*) rold,rnew
15         else
16         goto 100
17         endif
18         enddo
19         end

```

and

Listing 2: rinnb.f

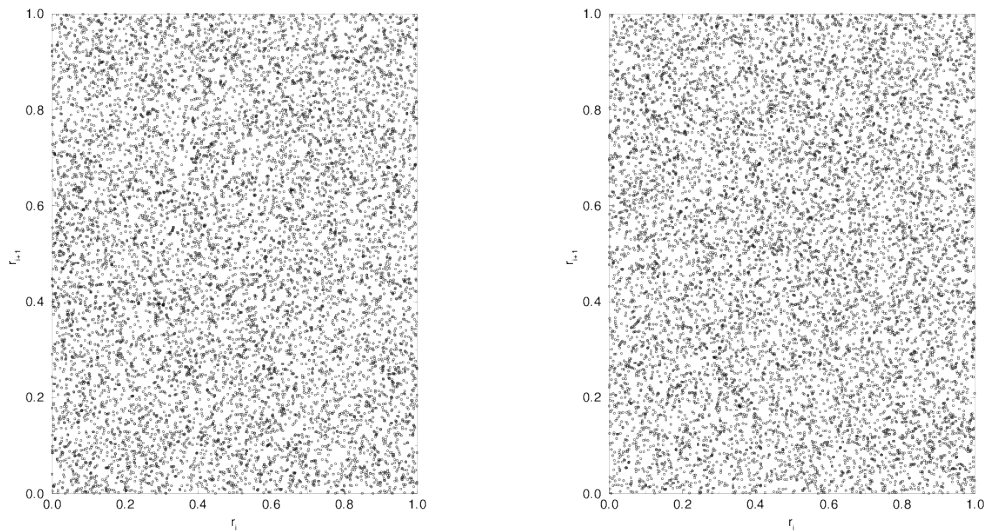
```

1      program rinnb
2      rold=0.
3      rnew=0.
4      do i=1,10000
5     100  rold=rnew
6         rnew=r_addran()
7         if(rold.lt.0.001.and.rnew.lt.0.001) then
8         write(*,*) rold,rnew
9         else
10        goto 100
11        endif
12        enddo
13        end

```

Equivalent implementations in C are given in the *Code* subdirectory.

Figure 1a shows 10 000 random numbers generated with the “16807” algorithm. Similar results for the built-in generator is shown in figure 1b.



(a) 16807

(b) Built-in

Figure 1: 10000 random numbers generated with either 16807 or the built-in algorithm.

Figure 2a shows result of 10 000 numbers generated with the “16807” generator but limited to the square $(0, 0)$, $(0, 0.001)$, $(0.001, 0)$, $(0.001, 0.001)$.

The is not so nice.

The result from the build-in generator is shown in figure 2b. It is absolutely ok.

Should we conclude that the “16807” generator is not useable? No, the only conclusion we can make is that it is not good for this particular use. Any generator should be tested *in the context it is to be used*.

Problem 2.

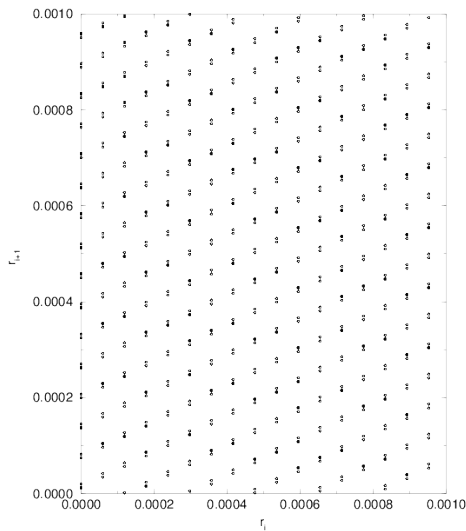
The following program uses the Box-Müller algorithm to generate random numbers following a gaussian with unit variance and zero mean. The programs groups the numbers into batches of size `nnum`= N . It identifies the largest number in each batch and then averages over these numbers to produce the average of the largest number among N numbers, $\langle x_{\max} \rangle$.

Listing 3: bboxm.f

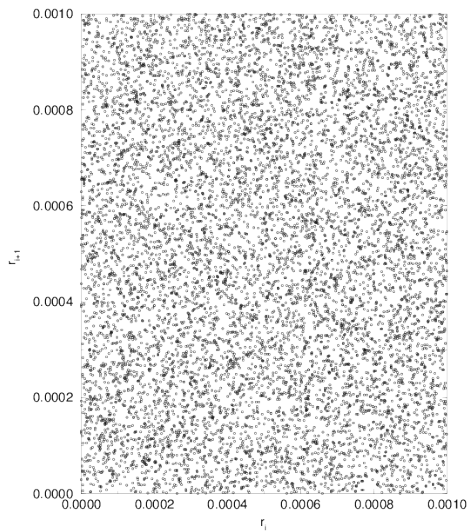
```

1      program bboxm
2      c
3      c Box-Mueller algorithm for generating gaussian random numbers
4      c
5      parameter (msamp=10000, nnum=50000)
6      c
7      ibm=1955
8      rinv=0.5/(2.**31-1.)
9      do i=1,1000

```



(a) 16807



(b) Built-in

Figure 2: 10000 random numbers generated with either 16807 or the built-in algorithm limited to the square $(0, 0)$, $(0, 0.001)$, $(0.001, 0)$, $(0.001, 0.001)$.

```

10     ibm=ibm*16807
11     enddo
12 c
13     pi2=8*atan(1.)
14 c
15     avemax=0.
16 c
17     do isamp=1,msamp
18 c
19         xmax=0.
20 c
21         do i=1,nnum/2
22             ibm=ibm*16807
23             y1=rinv*float(ibm)+0.5
24             ibm=ibm*16807
25             y2=rinv*float(ibm)+0.5
26             x1=sqrt(-2.*log(y1))*cos(pi2*y2)
27             x2=sqrt(-2.*log(y1))*sin(pi2*y2)
28             xmax=max(xmax,x1)
29             xmax=max(xmax,x2)
30         enddo
31 c
32         avemax=avemax+xmax
33 c
34     enddo

```

```

35 c
36     avemax=avemax/msamp
37     print *,nnum,avemax
38     end

```

The task was to find $\langle x_{\max} \rangle$ as a function of N . It so happens that this is a classical problem of *extreme statistics*. If the random numbers x are distributed according to a cumulative probability $P(x) = \int_{-\infty}^x p(x') dx'$, then we have that

$$P(\langle x_{\max} \rangle) = 1 - \frac{1}{N}. \quad (1)$$

The cumulative probability of a Gaussian is

$$P(x) = \frac{1}{2} \left[1 + \operatorname{erf} \left(\frac{x}{\sqrt{2}} \right) \right]. \quad (2)$$

For large x , this expression is

$$P(x \rightarrow \infty) = 1 - \frac{e^{-x^2/2\sigma^2}}{x\sqrt{2\pi/\sigma^2}} \quad (3)$$

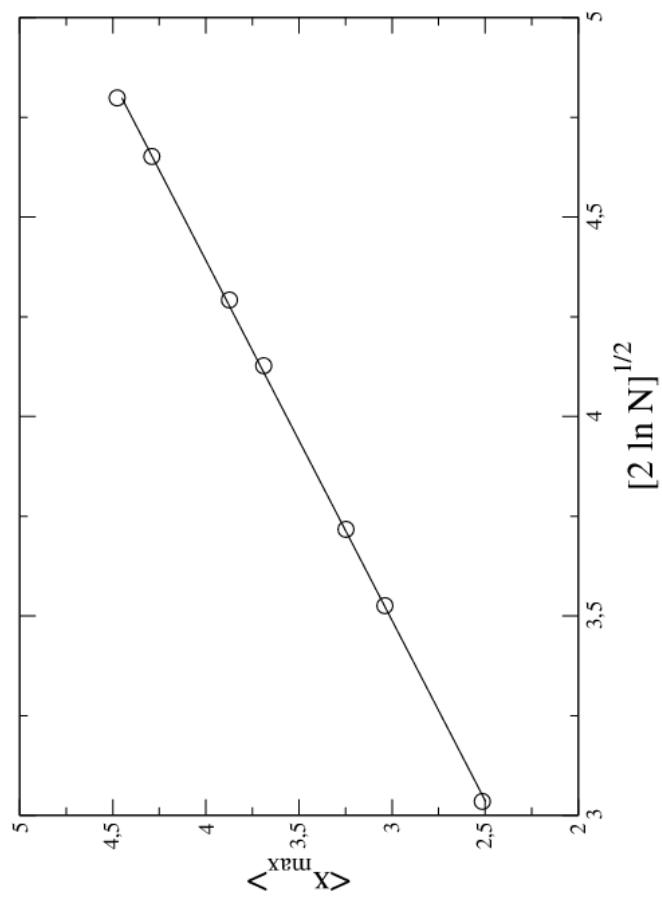
Combining Eqs. (1) and (3), we find

$$\langle x_{\max} \rangle = \sqrt{2\sigma^2 \ln N}. \quad (4)$$

We show the result of our simulation in the figure below. The straight line in the plot indicates that we find exactly the behavior expected from Eq. (4). I let N vary between 100 and 100 000, averaging each run over 10 000 samples.

I was in the lucky position here to know the functional form (4). Trying to fit $\langle x_{\max} \rangle$ vs. N without knowing this, is not so easy — but a typical task in computational physics.

The average height of people (from Wikipedia) is around 175 cm with a (guestimated) variance of around 10 cm. There are 6×10^9 people around. Plugging this into Eq. (4) gives $\langle x_{\max} \rangle = 175 \text{ cm} + 10 \text{ cm} \sqrt{2 \ln 6 \times 10^9} = 242 \text{ cm}$. This is right between the tallest and the second tallest person alive today. Hence, we may conclude that even the extremely tall follow the same Gaussian distribution as everybody else. Their height is in fact not unexpected. If they had not been that tall, somebody else would.

Figure 3: Plot of x_{max}