Numerical methods in Fortran: Lec. IV

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/afs/ictp/public/d/dbhakuni/NM_course/NM_lec4.pdf
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Topics to be covered

- Random numbers generators, middle square method, linear congruent method, in-built FORTRAN random number generators.
- Non-uniform random number generators: Transformation method, exponential and Gaussian random number generator.
- Rejection method and Monte-Carlo methods.
- Non parametric density estimation.

Outline:

- Get the empirical cumulative distribution function.
 - How to sort a the numbers in a vector? The bubble method.
- A first approximation to the PDF: Histograms.
 - Choosing the number of bins
- More elaborated PDF: Kernel Distribution Function.

Cumulative distribution function

• Footprint of your distribution.

$$C(x) = \int_{x}^{x} f(x')dx'$$

- It is easy to obtain numerically from a given set of points:
 - 1. Given a set of M random numbers of unknown distribution $\{x_i\}$
 - 2. Sort them in ascending order obtaining $\{\chi_r\}$.
 - 3. The empirical cumulative distribution function is

$$C_{emp}(x) = \frac{r}{M}$$

Cumulative distribution function

• Footprint of your distribution.

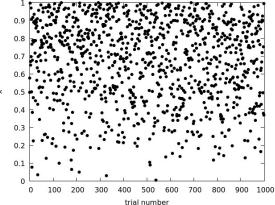
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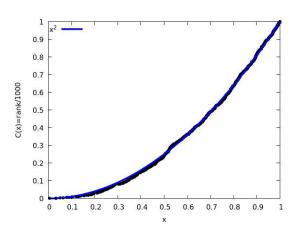
$$C_{emp}(x) = \frac{r}{M}$$

Example

• I generated 1000 points (M=1000) from the PDF f(x)=2x



Example



Sorting (inefficiently)

```
subroutine bubble (v,m)
! m is the number of elements in v
     integer :: i,newn,m,n
     real*8 :: v(m), tmp
     n=m
     do while (n>1)
         newn=0
         do i=2, n
               if (v(i-1)>v(i)) then
                  tmp=v(i)
                  v(i) = v(i-1)
                  v(i-1) = tmp
                  newn=i
               endif
         enddo
        n=newn
     enddo
end subroutine bubble
```

Probability distribution function

- Directly approximate f(x)
- •The naïve way is the Histogram:
 - -If you have a distribution of a variable x between [xmin,xmax]:
 - 1. Divide the x range into bins: $\Delta x = (x_{max} x_{min})/N_{bin}$ N_{bin} : # of bins
 - 2. Create an array for the histogram H[1:N_{hin}] (initialize to 0)
 - Each time you generate x check which "bin" it falls into.
 - 4. H[bin]=H[bin]+1
 - 4. Highing-rights, $M = \sum_{i=1}^{N_{bin}} H[i]$ $H[bin] = H[bin]/(M\Delta x)$ $x[i] = x_{min} + (i - 0.5)\Delta x$

•With the IF construction: Check that x belongs to the interval between x_{min} +(nbin-1)* Δx and x_{min} +nbin* Δx ... How would you program it in FORTRAN?

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```
DO j=1,MAXBINS 
 IF ((x>xmin+(j-1)*dx).AND.(x<=xmin+j*dx)) H(j)=H(j)+1 
 ENDDO
```

•With the IF construction: Check that x belongs to the interval between x_{min} +(nbin-1)* Δx and x_{min} +nbin* Δx ... How would you program it in FORTRAN?

•Is there a wiser manner?

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```
DO j=1,MAXBINS  \label{eq:continuous}  \text{IF } ((x>xmin+(j-1)*dx).AND.(x<=xmin+j*dx)) \text{ } \text{H}(j)=\text{H}(j)+1 \\ \text{ENDDO}
```

•Is there a wiser manner?

```
j=FLOOR((x-xmin)/dx)+1
H(j)=H(j)+1
```

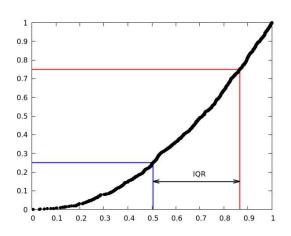
Freedman-Diaconis rule

$$\Delta x \approx IQR \frac{2}{M^{\frac{1}{3}}}$$

• IQR= Interquartile range. Easy to compute: $IQR = C_{emn}^{-1}(0.75) - C_{emn}^{-1}(0.25)$

• It is equivalent to $IQR = x_{^3M/_4} - x_{^M/_4}$ in the sorted vector $\{x_r\}$

IQR



Freedman-Diaconis rule

$$(\Delta x)_{prox} = IQR \frac{2}{M^{\frac{1}{3}}}$$

- Compute IQR and (Δx)_{prox}
- Compute the number of bins

$$N_{bin} = floor \left(\frac{\left(x_{\text{max}} - x_{\text{min}} \right)}{\left(\Delta x \right)_{\text{meas}}} \right) + 1$$

• Compute real value of $\Delta x = (x_{max} - x_{min})/N_{bin}$

LET'S BO IT

1. Obtain the information from your data (this allows you to set the parameters for the histogram).

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- 2. Perform the counts for the histogram (this is the computational part).

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- 2. Perform the counts for the histogram (this is the computational part).
- Normalize (This allows you to compare histograms with different parameters or with other ways of computing the pdf).

Allay Ul			
random			
numbers			
1.5			
3.2			
-0.3			
0.6			
-2.3			
-1.5			
-0.7			
1.2			
2.1			
2.2			
1.7			
-0.5			

Arrayof

Array of random numbers 1.5 3.2 -0.3 0.6 -2.3 -1.5 -0.7 1.2 2.1 2.2 1.7

Obtain the information from your data:

M=12

-0.5

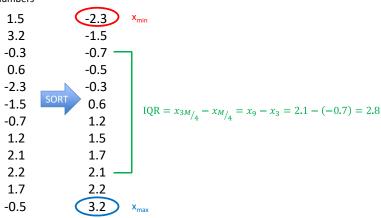
Array of random numbers	Obtain the information from your data:		
1.5	-2.3		
3.2	-1.5		
-0.3	-0.7		
0.6	-0.5		
-2.3	-0.3		
-1.5	SORT 0.6		
-0.7	1.2		
1.2	1.5		
2.1	1.7		
2.2	2.1		
1.7	2.2		
-0.5	3.2		

Array of Obtain the information from your data: random numbers -2.3 1.5 X_{min} 3.2 -1.5 -0.3 -0.7 0.6 -0.5 -2.3 -0.3 -1.5 0.6 -0.7 1.2 1.2 1.5 2.1 1.7 2.2 2.1 1.7 2.2

 $\mathbf{X}_{\mathsf{max}}$

-0.5

Array of random numbers



Array of random numbers

```
-2.3
1.5
                                 X_{min}
3.2
                    -1.5
-0.3
                    -0.7
0.6
                     -0.5
-2.3
                     -0.3
-1.5
                      0.6
                                    IQR = x_{3M/4} - x_{M/4} = x_9 - x_3 = 2.1 - (-0.7) = 2.8
-0.7
                      1.2
1.2
                      1.5
                                    (\Delta x)_{prox} = IQR \frac{2}{M^{\frac{1}{3}}} = 2.8 \frac{2}{12^{\frac{1}{3}}} = 2.446
2.1
                      1.7
2.2
                      2.1
1.7
                      2.2
-0.5
                                 X<sub>max</sub>
```

Array of random numbers

1.5
3.2

1.5
3.2
-1.5
-0.3
-0.7
0.6
-0.5
-2.3
-1.5
-0.7
1.2
1.2
1.2
1.2
1.7
2.2
2.1
$$N_{bin} = floor(\frac{(3.2 - (-2.3))}{2.446}) + 1 = floor(\frac{5.5}{2.446}) + 1 = 3$$

```
Array of
random
numbers
```

1.5

3.2 -0.3 0.6 -2.3 -1.5 -0.71.2 2.1 2.2 1.7 -0.5

M=12

Counts for the histogram:

$$N_{bin} = 3$$

$$N_{bin} =$$

$$n - S$$

$$\Delta x = \left(\frac{\left(x_{\text{max}} - x_{\text{min}}\right)}{N_{bin}}\right) = \left(\frac{\left(3.2 - (-2.3)\right)}{3}\right) = 1.833$$

$$\Delta x = 0$$

$$\Delta x = \left(\frac{1}{2}\right)$$

$$\Delta x = \int \frac{1}{2} dx$$

$$N_{bin} =$$

Array of random numbers

Counts for the histogram:

1.5

$$N_{bin} = 3$$

$$\Delta x = \left(\frac{(x_{\text{max}} - x_{\text{min}})}{N_{bin}}\right) = \left(\frac{(3.2 - (-2.3))}{3}\right) = 1.833$$

1-12

Array of random numbers

Counts for the histogram:

1.5

$$N_{bin} = 3$$

3.2

$$\Delta x = \left(\frac{\left(x_{\text{max}} - x_{\text{min}}\right)}{N_{bin}}\right) = \left(\frac{\left(3.2 - (-2.3)\right)}{3}\right) = 1.833$$

0.6

-2.3 -1.5

1.7

M=12

-2.3

j = 1 -0.467

```
Array of random numbers 1.5
3.2
-0.3
0.6
-2.3
-1.5
```

1.22.12.21.7

Counts for the histogram:

$$\Delta x = \left(\frac{\left(x_{\text{max}} - x_{\text{min}}\right)}{N_{bin}}\right) = \left(\frac{\left(3.2 - (-2.3)\right)}{3}\right) = 1.833$$

```
Array of random numbers 1.5 \\ 3.2 \\ -0.3 \\ 0.6 \\ -2.3 \\ -1.5
```

-0.7 1.2 2.1

Counts for the histogram:

$$\Delta x = \left(\frac{(x_{\text{max}} - x_{\text{min}})}{N_{bin}}\right) = \left(\frac{(3.2 - (-2.3))}{3}\right) = 1.833$$

2.2
1.7
-0.5
M=12
$$-2.3$$
 $j = 1$ -0.467 $j = 2$ 1.367 $j = 3$ 3.2

Array of Counts for the histogram: random numbers $N_{bin}=3$ 1.5 $\Delta x = \left(\frac{\left(x_{\text{max}} - x_{\text{min}}\right)}{N_{bin}}\right) = \left(\frac{\left(3.2 - (-2.3)\right)}{3}\right) = 1.833$ 3.2 -0.3 0.6 -2.3 -1.5 -0.7 1.2 Н 2.1 2.2 1.7 -0.5 M=12

-0.467

i = 1

1.367

i = 2

x

3.2

i = 3

-2.3

Array of Counts for the histogram: random numbers $N_{bin}=3$ 1.5 $\Delta x = \left(\frac{\left(x_{\text{max}} - x_{\text{min}}\right)}{N_{\text{bin}}}\right) = \left(\frac{\left(3.2 - (-2.3)\right)}{3}\right) = 1.833$ 3.2 -0.3 0.6 For all the random points: j=FLOOR((x-xmin)/dx)+1-2.3 $H(\dot{j}) = H(\dot{j}) + 1$ -1.5 -0.7 1.2 Н 2.1 2.2 1.7 -0.5

-0.467

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1.367

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x

3.2

j = 3

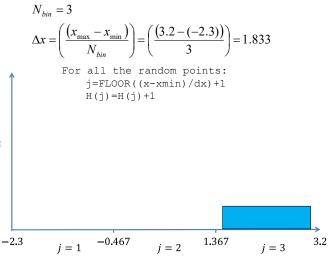
M=12

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M=12

Counts for the histogram:



x

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M=12

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i = 2

x

j = 3

i = 1

Array of Counts for the histogram: random numbers $N_{bin}=3$ 1.5 $\Delta x = \left(\frac{\left(x_{\text{max}} - x_{\text{min}}\right)}{N_{\text{bin}}}\right) = \left(\frac{\left(3.2 - (-2.3)\right)}{3}\right) = 1.833$ 3.2 -0.3 0.6 For all the random points: j=FLOOR((x-xmin)/dx)+1-2.3 $H(\dot{j}) = H(\dot{j}) + 1$ -1.5 -0.7 1.2 Н 2.1 2.2 1.7 -0.5 M=12

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-0.467

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1.367

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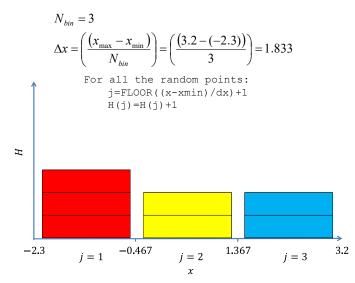
x

3.2

i = 3

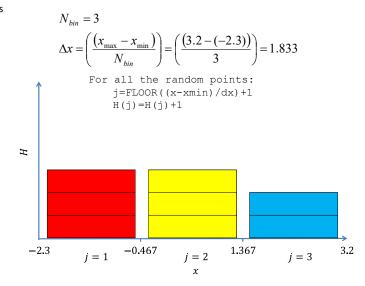
> 2.1 2.2 1.7 -0.5

M=12



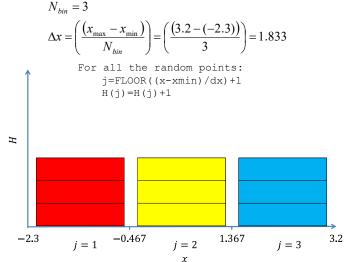
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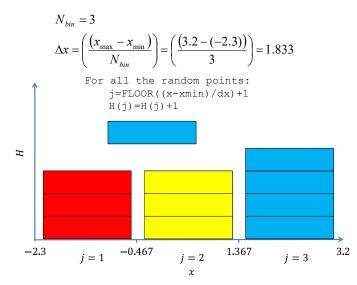
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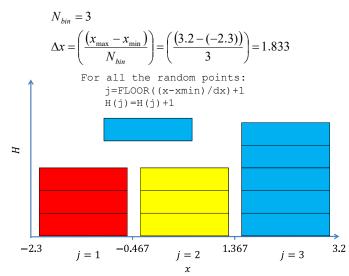
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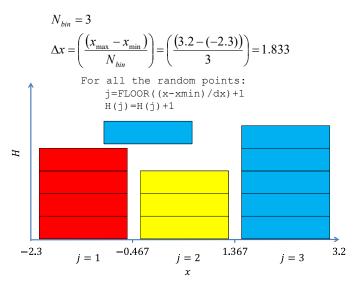
> 2.1 2.2 1.7 -0.5

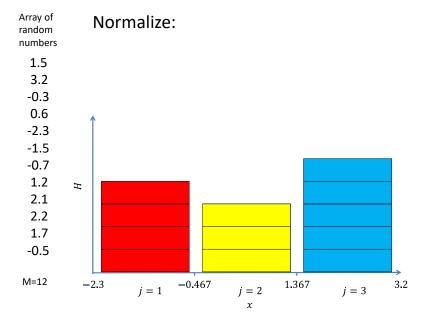
M=12



> 2.1 2.2 1.7

M=12





Array of Normalize: random numbers $M = \sum_{i=1}^{N_{bin}} H[i]$ 1.5 3.2 $H[bin] = H[bin]/(M\Delta x)$ -0.3 $x[i] = x_{\min} + (i - 0.5)\Delta x$ 0.6 -2.3 -1.5 -0.7 1.2 Н 2.1 2.2 1.7 -0.5 M=12

-0.467

j = 1

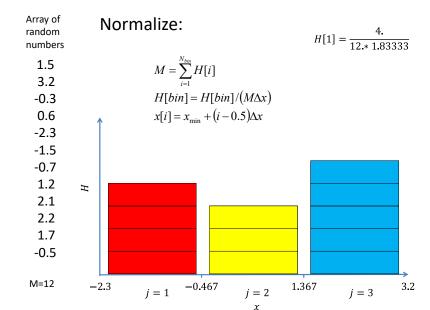
1.367

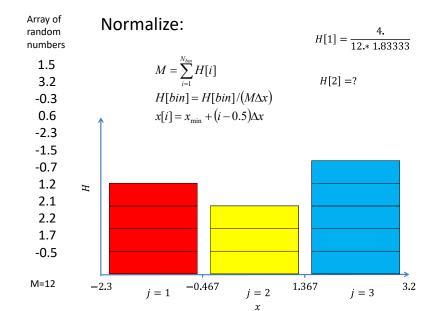
j = 2

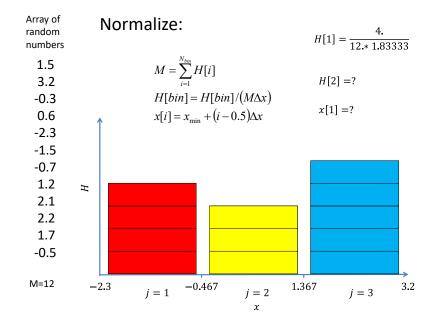
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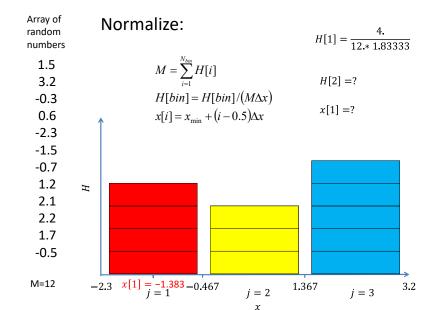
3.2

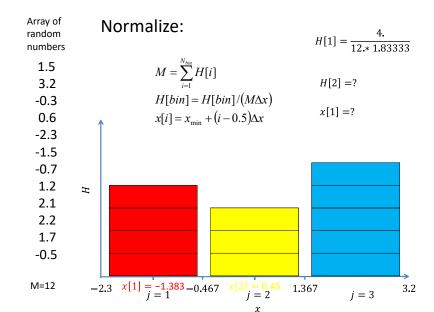
j = 3

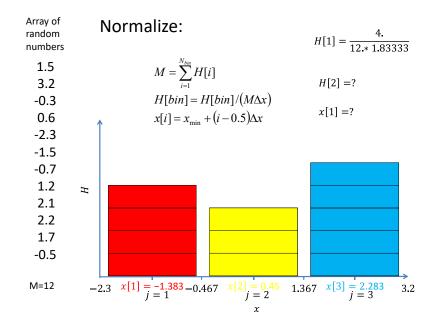




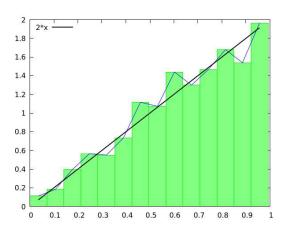


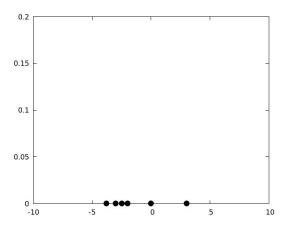


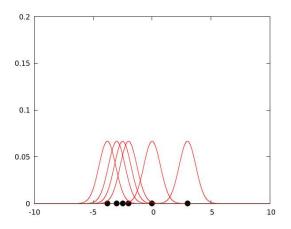


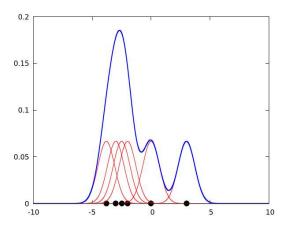


Histogram









•
$$p(x) = \frac{1}{M} \sum_{i=1}^{M} K(x, s, x_i)$$

• If the kernel is Gaussian, $K(x, s, x_i) =$

$$\mathcal{N}(x, s, x_i) = \frac{1}{s\sqrt{2\pi}} e^{-\frac{1}{2} \left(\frac{x - x_i}{s}\right)^2}$$

• *s* is the smoothing parameter, not easy to choose, as rule of thumb we will use:

choose, as rule of thumb we will use:

$$s = \frac{0.9A}{M^{1/5}}, A = \min\left(\sigma, \frac{IQR}{1.34}\right)$$

We obtain a function defined as

$$p(x) = \frac{1}{M} \sum_{i=1}^{M} K(x, s, x_i)$$

 Remember, in computational terms, a function corresponds to a table!

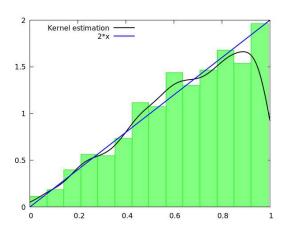
```
-1. p(-1)

-0.999 p(-0.999)

... ... ...

0.999 p(0.999)

1. p(1.)
```



Assignment-15

- 1) Generates 10^4 points (x_i) in the interval $x \in [-10, 10]$ from the dist. $(\mathcal{N}(x, s, x_i))$ is the normal dist. with mean x_i and std. dev. s) $f(x) = 15x^2\mathcal{N}(x, 0.25, -0.5) + 13\mathcal{N}(x, 0.3, -1.5) + 7\mathcal{N}(x, 1.0, 3.0)$ using the rejection method and compute -
- The empirical cumulative distribution function.
- The histogram representation using the Freedman-Diaconis rule.
- The value of the Gaussian kernel density estimation (p(x)) using the rule of thumb for the smoothing. In this case, build a table with 100000 entries for the x between -10 and 10 (Note this 10000 is not related with the size of the sample that is still 10000).

Submit your code as $surname_assignment_15.f90$ to dbhakuni@ictp.it before next Thursday.

Function assignment

