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Pulse Values from Long Term Measurement

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Abstract

Long term measurement of heart rate furnishes important information of cardiac anomalies. During a period of about 24 hours pulse rate of a patient at the University Hospital Aachen has been measured. Thus, we have a lot of *pulse values* important for a medical doctor treating sick patients. To analyse these "data" the Maple Program 2015 is very useful. At first the given "data" have been interpolated by *cubic spline functions*. Then, these functions have been approximated by *non linear regression* and also by *FOURIER-Series*. The measurement started at 9 am and ended on the next day at 6 am. Thus, we have 22 pulse values within the 21 hours. For the following analyse it's most convenient to map the x-coordinates [9...30] to the abscissa [0...2*Pi] according to: $X = 2 \cdot \text{Pi} \cdot (i - 1) / 21$ for $i = 1..22$.

Pulse Values and Cubic Spline Interpolation

```
> restart: with(Statistics):
> X:=array([seq(2*Pi*(i-1)/21,i=1..22)]);
X := [ 0,  $\frac{2\pi}{21}$ ,  $\frac{4\pi}{21}$ ,  $\frac{2\pi}{7}$ ,  $\frac{8\pi}{21}$ ,  $\frac{10\pi}{21}$ ,  $\frac{4\pi}{7}$ ,  $\frac{2\pi}{3}$ ,  $\frac{16\pi}{21}$ ,  $\frac{6\pi}{7}$ ,  $\frac{20\pi}{21}$ ,  $\frac{22\pi}{21}$ ,  $\frac{8\pi}{7}$ ,  $\frac{26\pi}{21}$ ,  $\frac{4\pi}{3}$ ,  $\frac{10\pi}{7}$ ,
 $\frac{32\pi}{21}$ ,  $\frac{34\pi}{21}$ ,  $\frac{12\pi}{7}$ ,  $\frac{38\pi}{21}$ ,  $\frac{40\pi}{21}$ ,  $2\pi$  ]
> whattype(X);
symbol
> Y:=array([63,59,61,63,65,65,60,62,61,59,57,57,57,62,47,48,51,55,
60,52,44,47]);
Y := [63, 59, 61, 63, 65, 65, 60, 62, 61, 59, 57, 57, 57, 62, 47, 48, 51, 55, 60, 52, 44, 47]
> whattype(Y);
symbol
> YM:=evalf(Mean(Y),4); # mean value of Y # dimension = bpm
YM := 57.05
> DATA:=seq([X[i],Y[i]],i=1..22);
```

```
DATA := [0, 63], [2π/21, 59], [4π/21, 61], [2π/7, 63], [8π/21, 65], [10π/21, 65], [4π/7, 60],
[2π/3, 62], [16π/21, 61], [6π/7, 59], [20π/21, 57], [22π/21, 57], [8π/7, 57], [26π/21, 62], [4π/3, 47],
[10π/7, 48], [32π/21, 51], [34π/21, 55], [12π/7, 60], [38π/21, 52], [40π/21, 44], [2π, 47]
```

```
> whattype(DATA);
```

exprseq

```
> with(CurveFitting):
```

```
> Sp(x) := Spline([DATA], x, degree=3):
```

```
> alias(th=thickness, co=color, H=Heaviside):
```

```
> p[1] := plot([DATA], x=0..2*Pi, 40..80, th=3, co=black,
style=point, symbol=cross, symbolsize=50, axes=boxed,
title="Pulse-Values");
```

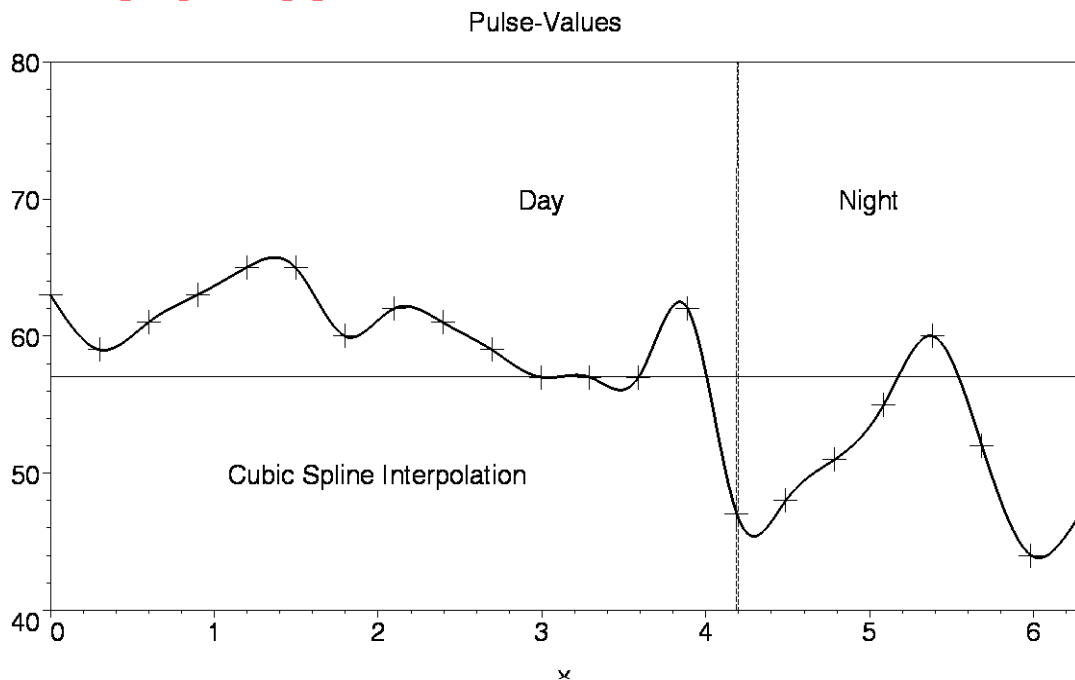
```
> p[2] := plot(Sp(x), x=0..2*Pi, th=3, co=black):
```

```
> p[3] := plots[textplot]({[3, 70, `Day`], [5, 70, `Night`],
[2, 50, `Cubic Spline Interpolation`]}):
```

```
> p[4] := plot(YM, x=0..2*Pi, linestyle=3, th=1, co=black):
```

```
> p[5] := plot(80*H(x-4.189)-40*H(x-4.199), x=0..2*Pi,
linestyle=3, th=1, co=black):
```

```
> plots[display](seq(p[k], k=1..5));
```



In this Figure 22 pulse-values (+++) are interpolated by *cubic spline*. The vertical line at $x = 4\pi/3$, i.e. 23 o'clock, divides the period $[0..2\pi]$ into two parts (Day & Night). The horizontal line at 57.05 indicates the meanvalue YM of the pulse-values. Furthermore, the measured pulse rate is important for a medical doctor to check the heart health and the fitness level. The normal pulse rates at rest are about (60 - 100) bpm for adults and (40 - 60) bpm for well-trained athletes. The values during Day should be higher than in the nighttime.

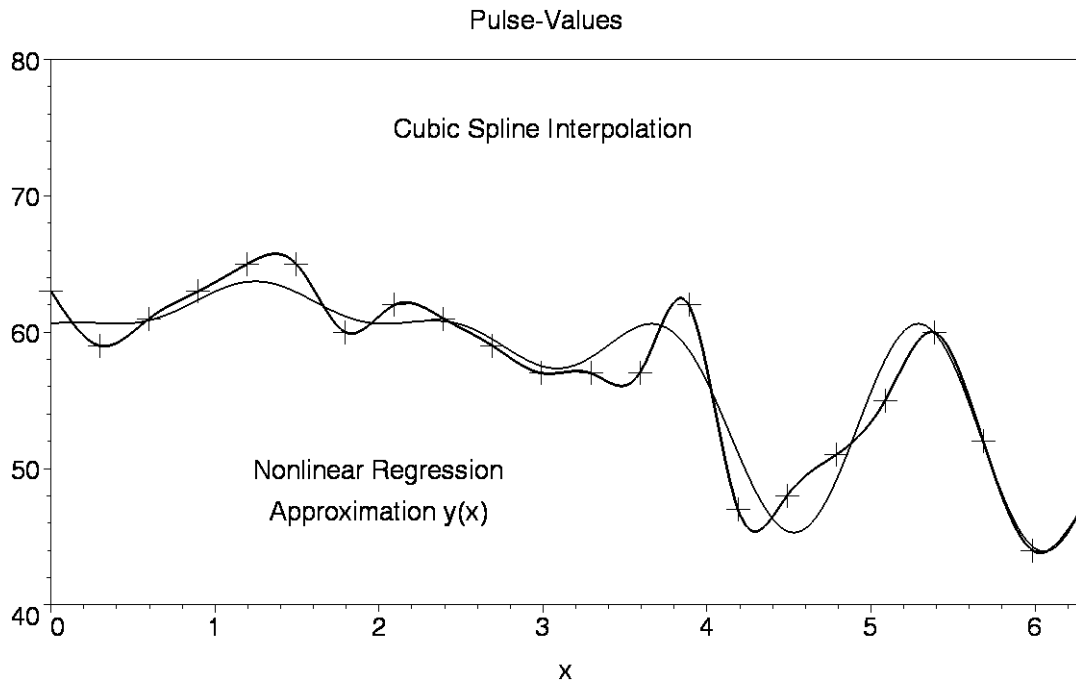
In the following the *cubic spline* is approximated by **Nonlinear Regression**:

```
> f:=(t,a,b,c,d,e,F)->a+b*t*sin(c*t+d)*cos(e*t+F)^2;

      f := (t, a, b, c, d, e, F) → a + b t sin(c t + d) cos(e t + F)2
> y:=unapply(evalf(NonlinearFit(f(t,a,b,c,d,e,F),X,Y,t)),t);
y := t → 60.6177917901184 - 3.55843771528764 t
      sin(0.628630870556338 t - 7.84446550487827)
      cos(1.93136367997716 t - 2.36238873287308)2
> y(x):=subs(t=x,y(t));
y(x) := 60.6177917901184 - 3.55843771528764 x
      sin(0.628630870556338 x - 7.84446550487826)
      cos(1.93136367997716 x - 2.36238873287308)2
> y(0):=evalf(subs(x=0,y(x)));
      y(0) := 60.6177917901184
> y(0)[data]:=63;
      y(0)data := 63
> y(2*Pi):=evalf(subs(x=2*Pi,y(x)));
      y(2 π) := 47.1050433944327
> y(2*Pi)[data]:=47;
      y(2 π)data := 47
```

The result of the *Nonlinear Regression* with its *Approximation* $y(x)$ is represented in the next Figure:

```
> alias(th=thickness,co=color):
> p[1]:=plot([DATA],x=0..2*Pi,40..80,th=3,co=black,
      style=point,symbol=cross,symbolsize=50,axes=boxed,
      title="Pulse-Values"):
> p[2]:=plot(Sp(x),x=0..2*Pi,th=3,co=black):
> p[3]:=plot(y(x),x=0..2*Pi,th=2,co=black):
> p[4]:=plots[textplot]([ [3,75,`Cubic Spline Interpolation`],
      [2,50,`Nonlinear Regression`], [2,47,`Approximation y(x)`] }):
> plots[display](seq(p[k],k=1..4));
```



The quality of the approximation $y(x)$ with respect to the cubic spline or to the given pulse-values can be checked by the *error norms* $L[2]$ or $l[2]$, respectively.

```
> L[2]:=
sqrt((1/(2*Pi))*Int((SPLINE-APPROX)^2,x=0..2*Pi))=
evalf(sqrt((1/(2*Pi))*int((Sp(x)-y(x))^2,x=0..2*Pi)))/YM;
```

$$L_2 := \frac{1}{2} \sqrt{2} \sqrt{\frac{1}{\pi} \int_0^{2\pi} (\text{SPLINE} - \text{APPROX})^2 dx} = 0.02934655276$$

```
> with(linalg):
> for i from 1 to 22 do
  v[i]:=evalf(subs(x=DATA[i][1],y(x))-DATA[i][2]) od:
> V:=vector([seq(v[i],i=1..22)]);
```

```
V := [-2.38220820988157, 1.66475872791110, -0.131970019738494, -0.627688982867983,
-1.30877883007925, -2.06856503490148, 1.15867243788686, -1.35971455517728,
-0.187139060881871, 0.473448472020975, 0.518937263714959, 1.15202553060854,
3.45736736351700, -3.30261707123917, 4.16910917516865, -2.57974794155758,
-1.89518693126377, 2.97275415919216, -0.0109951552355960, -0.117458093442199,
0.301276130214685, 0.105043394432700]
```

```
> l[2]:=
(1/sqrt(number_of_points))*Norm(V,2)=
evalf((1/sqrt(22))*norm(V,2))/YM;
```

$$l_2 := \frac{\text{Norm}(V, 2)}{\sqrt{\text{number_of_points}}} = 0.0332870367049969$$

The above *error norms* show that the approximation $y(x)$ is a suitable nonlinear model function fit to given data.

Note: The above Approximation $\mathbf{y}(\mathbf{x})$ has been calculated using *floating-point arithmetic default* in order to find *high precise results*. Furthermore, we have tested the calculation, for instance, with *digits = four* and *five* and found **not acceptable** results!

>

Approximation of the Splinefunction by *FOURIER-Series*

> **restart:**

> **with(Statistics):**

> **X:=array([seq(2*Pi*(i-1)/21,i=1..22)]);**

$X := \left[0, \frac{2\pi}{21}, \frac{4\pi}{21}, \frac{2\pi}{7}, \frac{8\pi}{21}, \frac{10\pi}{21}, \frac{4\pi}{7}, \frac{2\pi}{3}, \frac{16\pi}{21}, \frac{6\pi}{7}, \frac{20\pi}{21}, \frac{22\pi}{21}, \frac{8\pi}{7}, \frac{26\pi}{21}, \frac{4\pi}{3}, \frac{10\pi}{7}, \frac{32\pi}{21}, \frac{34\pi}{21}, \frac{12\pi}{7}, \frac{38\pi}{21}, \frac{40\pi}{21}, 2\pi \right]$

> **whattype(X);**

symbol

> **Y:=array([63,59,61,63,65,65,60,62,61,59,57,57,57,62,47,48,51,55,60,52,44,47]);**

$Y := [63, 59, 61, 63, 65, 65, 60, 62, 61, 59, 57, 57, 57, 62, 47, 48, 51, 55, 60, 52, 44, 47]$

> **whattype(Y);**

symbol

> **YM:=evalf(Mean(Y),4); # mean value of Y # dimension = bpm**

$YM := 57.05$

> **DATA:=seq([X[i],Y[i]],i=1..22);**

$DATA := [0, 63], \left[\frac{2\pi}{21}, 59 \right], \left[\frac{4\pi}{21}, 61 \right], \left[\frac{2\pi}{7}, 63 \right], \left[\frac{8\pi}{21}, 65 \right], \left[\frac{10\pi}{21}, 65 \right], \left[\frac{4\pi}{7}, 60 \right], \left[\frac{2\pi}{3}, 62 \right], \left[\frac{16\pi}{21}, 61 \right], \left[\frac{6\pi}{7}, 59 \right], \left[\frac{20\pi}{21}, 57 \right], \left[\frac{22\pi}{21}, 57 \right], \left[\frac{8\pi}{7}, 57 \right], \left[\frac{26\pi}{21}, 62 \right], \left[\frac{4\pi}{3}, 47 \right], \left[\frac{10\pi}{7}, 48 \right], \left[\frac{32\pi}{21}, 51 \right], \left[\frac{34\pi}{21}, 55 \right], \left[\frac{12\pi}{7}, 60 \right], \left[\frac{38\pi}{21}, 52 \right], \left[\frac{40\pi}{21}, 44 \right], [2\pi, 47]$

> **with(CurveFitting):**

> **# The cubic splinefunction fit to data is given by:**

> **Sp(x):=Spline([DATA],x,degree=3):**

> **# This function is not printed because of its length.**

However, printing is possible, if we insert a semicolon (;) instead of the doppel point (:) at the of the command for Sp(x).

> **FOURIER_series(x):=**

a[0]/2+sum(a[k]*cos(k*x)+b[k]*sin(k*x),k=1..infinity);

$$\text{FOURIER_series}(x) := \frac{1}{2}a_0 + \left(\sum_{k=1}^{\infty} (a_k \cos(kx) + b_k \sin(kx)) \right)$$

> **a[k]:=(1/Pi)*Int(f(x)*cos(k*x),x=0..2*Pi);**

$$a_k := \frac{1}{\pi} \int_0^{2\pi} f(x) \cos(kx) dx$$

> `a[0]:=simplify(subs(k=0,a[k]));`

$$a_0 := \frac{1}{\pi} \int_0^{2\pi} f(x) dx$$

> `b[k]:=(1/Pi)*Int(f(x)*sin(k*x),x=0..2*Pi);`

$$b_k := \frac{1}{\pi} \int_0^{2\pi} f(x) \sin(kx) dx$$

> `with(CurveFitting): F(x):=Sp(x):`

> `A[0]:=evalf(subs(f(x)=F(x),a[0]));`

$$A_0 := 114.1960062$$

> `A[k]:=simplify(value(subs(f(x)=F(x),a[k]))):`

> `A[k]:=subs({sin(k*Pi)=0.,(cos(k*Pi))^2=1.},%):`

> `A[k]:=evalf(%):`

> `for i in [seq(k,k=1..5)] do A[i]:=subs(k=i,A[k]) od:`

> `for i in [seq(k,k=1..15)] do P[i]:=subs(k=i,A[k]) od:`

> `for i in [seq(k,k=1..100)] do R[i]:=subs(k=i,A[k]) od:`

> `B[k]:=simplify(value(subs(f(x)=F(x),b[k]))):`

> `B[k]:=subs({sin(k*Pi)=0.,(cos(k*Pi))^2=1.},%):`

> `for i in [seq(k,k=1..5)] do B[i]:=subs(k=i,B[k]) od:`

> `for i in [seq(k,k=1..15)] do Q[i]:=subs(k=i,B[k]) od:`

> `for i in [seq(k,k=1..100)] do S[i]:=subs(k=i,B[k]) od:`

> `# The coefficients A[i]...S[i] are not printed because of their length. However, it can be done by inserting semicolons (;) instead of doppelpoints (:)` at the ends of the commands. The FOURIER-Series with `k = 5, 15, 100` are expressed by `y(x), z(x), T(x)`, respectively, as follows:

>

> `y(x):=evalf(A[0]/2+sum(A[k]*cos(k*x)+B[k]*sin(k*x),k=1..5));`

`y(x) := 57.09800310 - 1.037475982 cos(x) + 5.750747994 sin(x) - 1.043625943 cos(2. x)`
`+ 1.017438123 sin(2. x) - 2.581563326 cos(3. x) - 0.3011718296 sin(3. x)`
`- 2.144147642 cos(4. x) + 2.555986247 sin(4. x) + 1.715299509 cos(5. x)`
`+ 1.882993266 sin(5. x)`

> `y(0):=simplify(subs(x=0,y(x)));`

$$y(0) := 52.00648972$$

> `y(0)[data]:=63;`

$$y(0)_{data} := 63$$

> `y(2*Pi):=simplify(subs(x=2*Pi,y(x)));`

$$y(2\pi) := 52.00648974$$

> `y(2*Pi)[data]:=47;`

$$y(2\pi)_{data} := 47$$

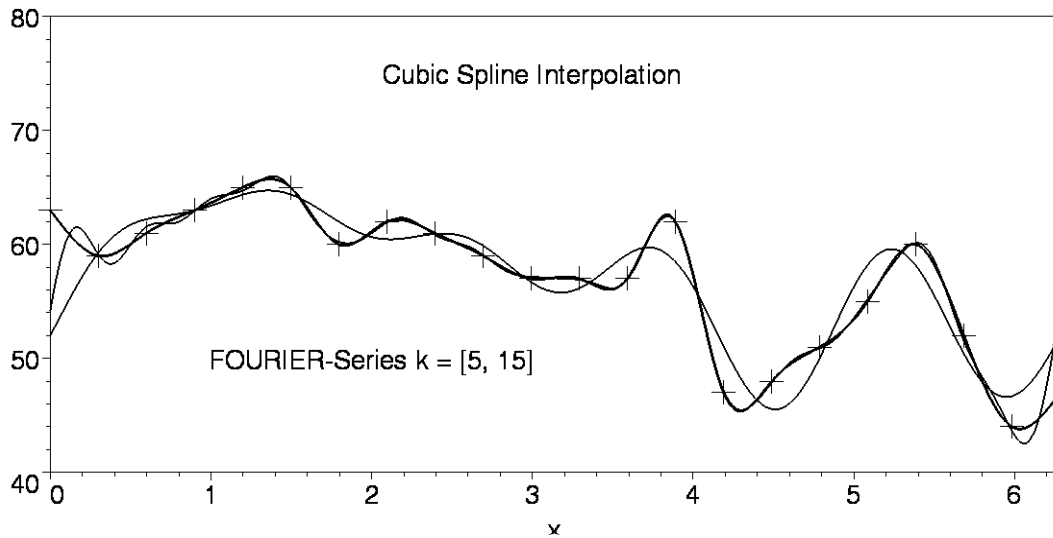
>

```

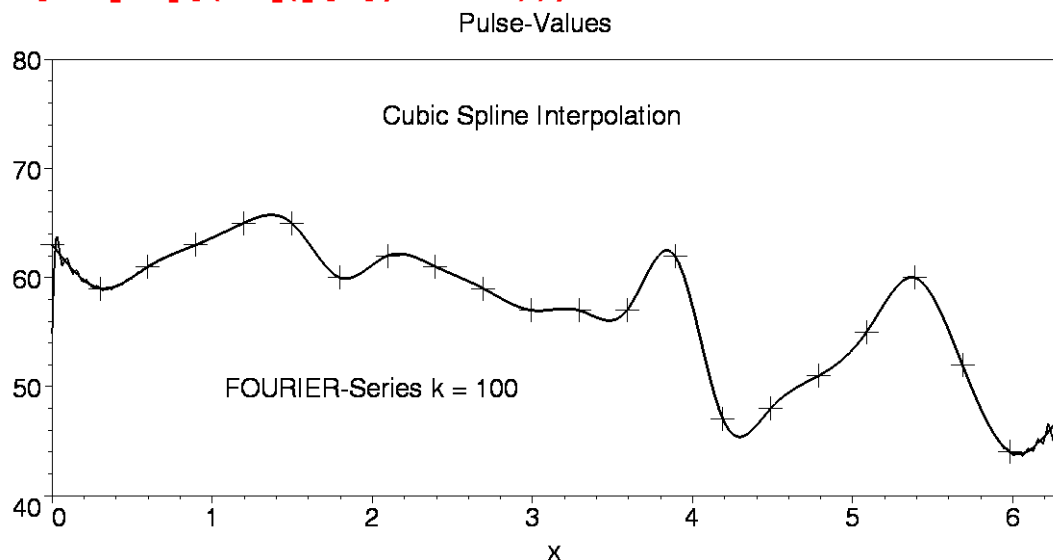
> z(x):=evalf(A[0]/2+sum(P[k]*cos(k*x)+Q[k]*sin(k*x),k=1..15));
z(x) := -1.0374754 cos(x) + 5.750747994 sin(x) + 57.09800310 + 0.04588158316 cos(15. x)
+ 0.05707791714 cos(14. x) + 0.2450638226 cos(13. x) - 0.01220520060 cos(12. x)
+ 0.1780472153 cos(11. x) + 0.2354260321 cos(10. x) - 0.1540426313 cos(9. x)
+ 1.400542814 cos(8. x) + 0.1770108345 cos(7. x) + 0.0385764726 cos(6. x)
+ 0.3026995316 sin(14. x) + 0.3339772552 sin(15. x) + 0.5473660021 sin(12. x)
+ 0.5061307261 sin(13. x) + 1.104154831 sin(10. x) + 0.05669948876 sin(11. x)
- 0.1608339671 sin(8. x) + 0.1772926149 sin(9. x) + 1.065751796 sin(6. x)
+ 1.704896570 sin(7. x) + 1.715299510 cos(5. x) + 1.882993268 sin(5. x)
- 2.144147645 cos(4. x) + 2.555986247 sin(4. x) - 1.04362594 cos(2. x)
+ 1.017438123 sin(2. x) - 2.581563325 cos(3. x) - 0.3011718296 sin(3. x)
> z(0):=simplify(subs(x=0,z(x)));
z(0) := 54.21786916
> z(0)[data]:=63;
z(0)data := 63
> z(2*Pi):=simplify(subs(x=2*Pi,z(x)));
z(2 π) := 54.21786923
> z(2*Pi)[data]:=47;
z(2 π)data := 47
>
> T(x):=evalf(A[0]/2+sum(R[k]*cos(k*x)+S[k]*sin(k*x),k=1..100)):
> # This equation is not printed because of its length with more
than 100 terms. However, printing is possible, if we insert a
semicolon (;) instead of the doppelpoint (:) at the end of the
command for T(x).
> T(0):=simplify(subs(x=0,T(x)));
T(0) := 54.88027315
> T(0)[data]:=63;
T(0)data := 63
> T(2*Pi):=simplify(subs(x=2*Pi,T(x)));
T(2 π) := 54.88027358
> T(2*Pi)[data]:=47;
T(2 π)data := 47
> # The following Figure illustrates the approximations of the
splinefunction by FOURIER-Series with k = [5, 15]:
> alias(th=thickness,co=color):
> p[1]:=plot(Sp(x),x=0..2*Pi,40..80,axes=boxed,th=3,co=black):
> p[2]:=plot([DATA],x=0..2*Pi,th=3,co=black,
style=point,symbol=cross,symbolsize=50,title="Pulse-Values"):
> p[3]:=plot({y(x),z(x)},x=0..2*Pi,th=2,co=black):
> p[4]:=plots[textplot]({[3,75,`Cubic Spline Interpolation`],
[2,50,`FOURIER-Series k = [5, 15]`]}):

```

```
> plots[display](seq(p[k],k=1..4));
Pulse-Values
```



```
> # The next Figure shows the FOURIER-Series with k = 100:
> alias(th=thickness,co=color):
> p[1]:=plot(Sp(x),x=0..2*Pi,40..80,axes=boxed,th=3,co=black):
> p[2]:=plot([DATA],x=0..2*Pi,th=3,co=black,
  style=point,symbol=cross,symbolsize=50,title="Pulse-Values"):
> p[3]:=plot(T(x),x=0..2*Pi,th=2,co=black):
> p[4]:=plots[textplot]({[3,75,`Cubic Spline Interpolation`],
  [2,50,`FOURIER-Series k = 100`]}):
> plots[display](seq(p[k],k=1..4));
```



In this Figure we cannot observe differences between the splinefunction and the FOURIER-Series. Only immediately at the beginning $x = 0$ and at the end $x = 2*\text{Pi}$:

```
> T(0):=simplify(subs(x=0,T(x)));
T(0.021):=simplify(subs(x=0.021,T(x)));
T(0) := 54.88027315
T(0.021) := 63.06636791
```


> T(0)[data]:=63;

$$T(0)_{data} := 63$$

> T(2*Pi):=simplify(subs(x=2*Pi,T(x)));

T(1.9943*Pi):=simplify(subs(x=1.9943*Pi,T(x)));

$$T(2\pi) := 54.88027358$$

$$T(6.265278230) := 47.03280626$$

> T(2*Pi)[data]:=47;

$$T(2\pi)_{data} := 47$$

Because of the above anomaly at $x = 0$ and $x = 2\pi$, let us discuss the following *error norms* $L[2]$ and $l[2]$ between $x = 2\pi/21 = 0.299$ and $x = 40\pi/21 = 5.984$ instead $x = [0..2\pi]$.

Thus, we consider only 20 pulse-values instead of the complete number of 22.

The *error norms* $L[2]$ ($k = 5, 15, 100$), of the approximations $y(x)$, $z(x)$, and $T(x)$ with respect to the cubicfunction can thus be expressed as:

> L[2][k=5]:=

sqrt((1/5.685)*Int((SP(xi)-y(xi))^2,xi=0.299..5.984))=
sqrt((1/5.685)*int((Sp(x)-y(x))^2,x=0.299..5.984))/YM;

$$L_{2_{k=5}} := 0.4194061220 \sqrt{\int_{0.299}^{5.984} (SP(\xi) - y(\xi))^2 d\xi} = 0.02973261534$$

> L[2][k=15]:=

sqrt((1/5.685)*Int((SP(xi)-z(xi))^2,xi=0.299..5.984))=
sqrt((1/5.685)*int((Sp(x)-z(x))^2,x=0.299..5.984))/YM;

$$L_{2_{k=15}} := 0.4194061220 \sqrt{\int_{0.299}^{5.984} (SP(\xi) - z(\xi))^2 d\xi} = 0.004180071437$$

> L[2][k=100]:=

sqrt((1/5.685)*Int((SP(xi)-T(xi))^2,xi=0.299..5.984))=
sqrt((1/5.685)*int((Sp(x)-T(x))^2,x=0.299..5.984))/YM;

$$L_{2_{k=100}} := 0.4194061220 \sqrt{\int_{0.299}^{5.984} (SP(\xi) - T(\xi))^2 d\xi} = 0.0006802844784$$

>

The three values above show the *error norms* $L[2]$ of the *FOURIER-Series* $y(x)$, $z(x)$, and $T(x)$ relative to the *cubic splinefunction*. Furthermore one should discuss the *error norms* with respect to the given measured data, as follows:

> with(linalg):

> for i from 2 to 21 do

u[i]:=subs(x=DATA[i][1],y(x))-DATA[i][2] od:

> U:=evalf(vector([seq(u[i],i=2..21)]),4);

U := [0.268, 1.223, -0.025, -0.636, -0.615, 1.995, -1.547, -0.064, 0.895, -0.3071, -0.8599,
2.110, -3.518, 4.081, -2.449, -0.965, 3.229, -1.855, -1.589, 2.697]

```

> l[2][k=5]:=
  (1/sqrt(number_of_points))*Norm(U,2)=
  evalf((1/sqrt(20))*norm(U,2),4)/YM;

```

$$l_{2_{k=5}} := \frac{\text{Norm}(U, 2)}{\sqrt{\text{number_of_points}}} = 0.03381244522$$

```

> for i from 2 to 21 do
  v[i]:=subs(x=DATA[i][1],z(x))-DATA[i][2] od:
> V:=evalf(vector([seq(v[i],i=2..21)]),4);
V := [0.138, 0.557, -0.047, -0.291, 0.045, 0.236, -0.063, -0.158, 0.073, 0.1649, -0.1109, -0.123,
      0.047, 0.154, -0.170, -0.105, 0.263, 0.113, -0.479, -0.384]
> l[2][k=15]:=
  (1/sqrt(number_of_points))*Norm(V,2)=
  evalf((1/sqrt(20))*norm(V,2),4)/YM;
>

```

$$l_{2_{k=15}} := \frac{\text{Norm}(V, 2)}{\sqrt{\text{number_of_points}}} = 0.004087642419$$

```

> for i from 2 to 21 do
  w[i]:=subs(x=DATA[i][1],T(x))-DATA[i][2] od:
> W:=evalf(vector([seq(w[i],i=2..21)]),4);
W := [-0.034, 0.079, 0.041, -0.019, -0.043, 0.006, 0.032, 0.002, -0.039, -0.0201, -0.0119,
      -0.007, -0.040, -0.021, -0.020, 0.041, 0.022, 0.001, -0.056, 0.010]
> l[2][k=100]:=
  (1/sqrt(number_of_points))*Norm(W,2)=
  evalf((1/sqrt(20))*norm(W,2),4)/YM;
>

```

$$l_{2_{k=100}} := \frac{\text{Norm}(W, 2)}{\sqrt{\text{number_of_points}}} = 0.0005851007888$$

>

Note: Because of its "piecewise character" of the splinefunction the precision of the L[2] error norm cannot be arrived by *Digits*, for instance, equal to four. Therefore, we have caculeted the L[2] error norms above with the default value of ten. However, the calculation of the l[2] error norms with *Digits* = 4 is possible without loss of *precision*.

The *error norms* may sometimes be improved without changing the number of terms, if we utilize the method of *smoothing*. As an example let us apply this method to the FOURIER-Series $y(x)$ with $k = 5$ terms. Introducing the *smoothing factor* $g(k,N)$ we arrive at the *modified* Fourier-Series $G(x,n,N)$ as:

```

> g(k,N):=N*sin(Pi*k/N)/Pi/k; # smoothing factor for
FOURIER-Series

```

$$g(k, N) := \frac{N \sin\left(\frac{\pi k}{N}\right)}{\pi k}$$

```

> G(x,n,N):=
  A[zero]/2+sum(g(kappa,Nu)*(A[kappa]*cos(kappa*x))+

```

```
B[kappa]*sin(kappa*x)),kappa =1..n); # smoothing function
```

$$G(x, n, N) := \frac{1}{2} A_{zero} + \left(\sum_{k=1}^n g(k, N) (A_k \cos(kx) + B_k \sin(kx)) \right)$$

```
> # Example for y(x) with n = 5 and N = n + 1:
```

```
> g(k,6):=subs(N=6,g(k,N));
```

$$g(k, 6) := \frac{6 \sin\left(\frac{\pi k}{6}\right)}{\pi k}$$

```
> G(x,n=5,N=6):=
```

```
evalf(subs({n=5,Nu=6,kappa=k,g(kappa,Nu)=g(k,6),
A[zero]=A[0],A[kappa]=A[k],b[kappa]=B[k]},G(x,n,N)));
```

```
G(x, n = 5, N = 6) := 57.09800310 - 0.9907161038 cos(x) + 5.491559915 sin(x)
- 0.8630716053 cos(2. x) + 0.8414145698 sin(2. x) - 1.643474286 cos(3. x)
- 0.1917319392 sin(3. x) - 0.8865979059 cos(4. x) + 1.056891807 sin(4. x)
+ 0.3275980719 cos(5. x) + 0.3596252237 sin(5. x)
```

```
> G(0,n=5,N=6):=simplify(subs(x=0,G(x,n=5,N=6)));
```

```
G(0, n = 5, N = 6) := 53.04174128
```

```
> y(0):=simplify(subs(x=0,y(x)));
```

```
y(0) := 52.00648972
```

```
> y(0)[data]:=63;
```

```
y(0)data := 63
```

```
> G(2*Pi,n=5,N=6):=simplify(subs(x=2*Pi,G(x,n=5,N=6)));
```

```
G(2 π, n = 5, N = 6) := 53.04174126
```

```
> y(2*Pi):=simplify(subs(x=2*Pi,y(x)));
```

```
y(2 π) := 52.00648974
```

```
> y(2*Pi)[data]:=47;
```

```
y(2 π)data := 47
```

```
> alias(th=thickness,co=color):
```

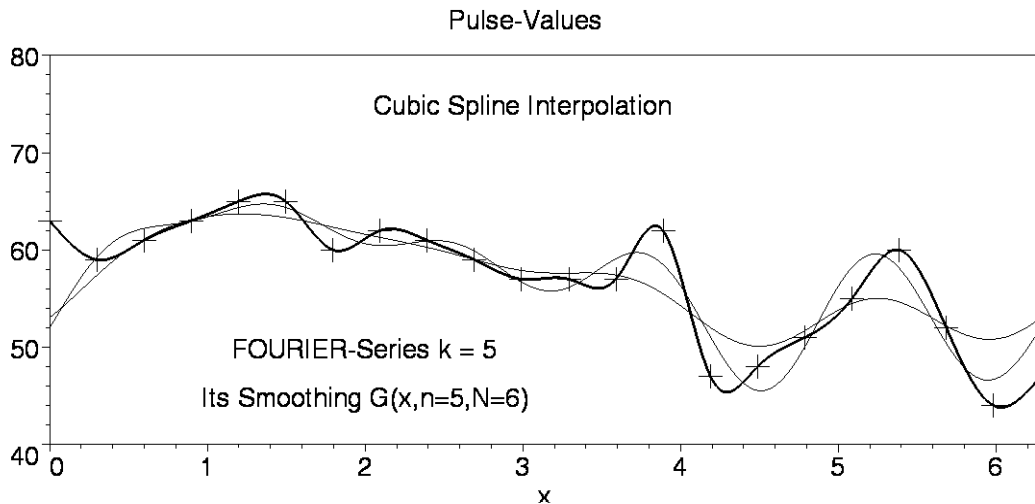
```
> p[1]:=plot(Sp(x),x=0..2*Pi,40..80,axes=boxed,th=3,co=black):
```

```
> p[2]:=plot([DATA],x=0..2*Pi,th=3,co=black,
style=point,symbol=cross,symbolsize=50):
```

```
> p[3]:=plot({y(x),G(x,n=5,N=6)},x=0..2*Pi,
th=1,co=black,title="Pulse-Values");
```

```
> p[4]:=plots[textplot]([ [3,75,`Cubic Spline Interpolation`],
[2,50,`FOURIER-Series k = 5`],[2,45,`Its Smoothing
G(x,n=5,N=6)`] ]):
```

```
> plots[display](seq(p[k],k=1..4));
```



- > # In this Figure both the FOURIER-Approximation $y(x)$ with $k = 5$ and its smoothing $G(x,n=5,N=6)$ are represented by thin lines and differ from the cubic spline interpolation (thick line) more at the beginning $x = 0$ and at the end
- > # $x = 2\pi$. Thus, we consider the data between $x = 2\pi/21 = 0.299$ and $x = 40\pi/21 = 5.984$ instead of $x = [0..2\pi]$ in order to calculate the error norms $M[2]$ and $m[2]$ for the smoothing with respect to the cubic spline and the given data, respectively.

```
> M[2][G_k=5]:=
sqrt((1/5.685)*Int((SP(xi)-G(xi))^2,xi=0.299..5.984))=
sqrt((1/5.685)*int((Sp(x)-G(x,n=5,N=6))^2,x=0.299..5.984))/YM;
```

$$M_{2_{G,k=5}} := 0.4194061220 \sqrt{\int_{0.299}^{5.984} (SP(\xi) - G(\xi))^2 d\xi} = 0.04361025500$$

```
>
> L[2][y_k=5]:=0.02973261539; # by comparison with M[2][G_k=5]
```

$$L_{2_{y,k=5}} := 0.02973261539$$

```
> with(linalg):
> for i from 2 to 21 do
  q[i]:=subs(x=DATA[i][1],G(x,n=5,N=6))-DATA[i][2] od:
> Q:=evalf(vector([seq(q[i],i=2..21)]));
```

```
Q := [-1.600718285, 0.118110312, 0.100398833, -1.296254782, -1.672640897, 2.344476781,
-0.708082836, -0.731428092, 0.008441527, 0.8974699744, 0.6244444103, 0.431036868,
-6.503707734, 5.029927996, 2.102381119, 0.703805299, -0.511270106, -5.342454724,
0.193510065, 6.828878082]
```

```
> m[2][G_k=5]:=
(1/sqrt(number_of_points))*Norm(Q,2)=
evalf((1/sqrt(20))*norm(Q,2))/YM;
```

$$m_{2_{G,k=5}} := \frac{\text{Norm}(Q, 2)}{\sqrt{\text{number_of_points}}} = 0.05003553206$$

>

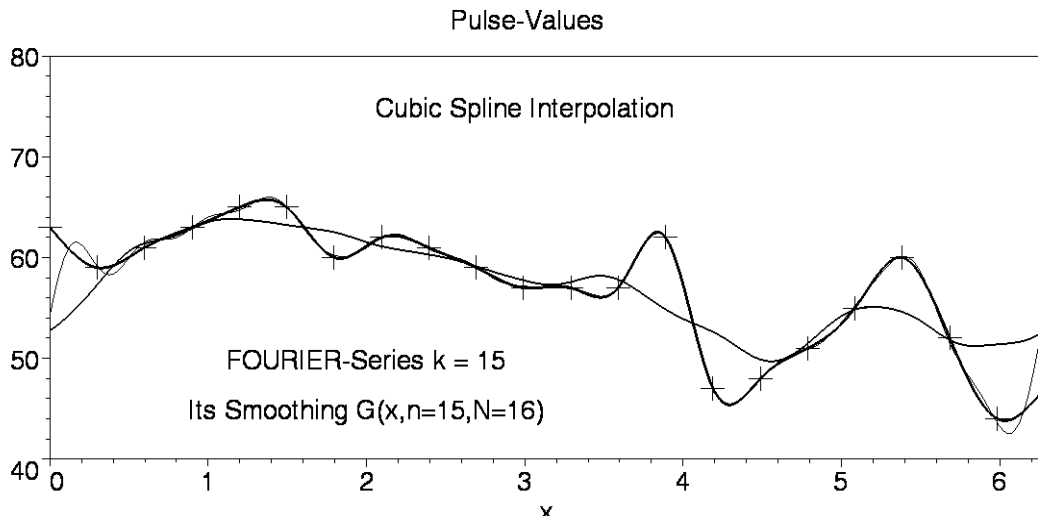
```

> l[2][y_k=5]:=0.03381244522; # by coparison with m[2][G_k=5]
                                
$$l_{y_k=5} := 0.03381244522$$

> # We see, in this example of the approximation with k = 5 terms
the error norms L[2], l[2] are less than the corresponding
values M[2], m[2]. In other examples we observe less error norms
of the smoothings than the corresponding values due to the
original approximations.
>
> # The next example is concerned with the smoothing of the
FOURIER-Series z(x) with k = 15 and N = 16:
> g(k,16):=subs(N=16,g(k,N)); # smoothing factor
                                
$$g(k, 16) := \frac{16 \sin\left(\frac{\pi k}{16}\right)}{\pi k}$$

> G(x,n=15,N=16):=
evalf(subs({n=15,Nu=16,kappa=k,g(kappa,Nu)=g(k,6),
A[zero]=A[0],A[kappa]=A[k],B[kappa]=B[k]},G(x,n,N)));
G(x, n = 15, N = 16) := -0.9907161288 cos(x) + 5.491559855 sin(x) + 57.09800310
+ 0.005841824634 cos(15. x) + 0.006743293932 cos(14. x) + 0.01800143943 cos(13. x)
- 0.01545659691 cos(11. x) - 0.03893915225 cos(10. x) + 0.03268886121 cos(9. x)
- 0.2895598960 cos(8. x) - 0.02414755710 cos(7. x) + 0.03576149974 sin(14. x)
+ 0.04252330470 sin(15. x) + 0.03717840312 sin(13. x) - 0.1826257387 sin(10. x)
- 0.004922183990 sin(11. x) + 0.03325215555 sin(8. x) - 0.03762266114 sin(9. x)
- 0.2325794711 sin(7. x) + 0.3275980719 cos(5. x) + 0.3596252233 sin(5. x)
- 0.8865979059 cos(4. x) + 1.056891806 sin(4. x) - 0.8630716019 cos(2. x)
+ 0.8414145716 sin(2. x) - 1.643474285 cos(3. x) - 0.1917319432 sin(3. x)
> alias(th=thickness,co=color):
> p[1]:=plot(Sp(x),x=0..2*Pi,40..80,axes=boxed,th=3,co=black):
> p[2]:=plot([DATA],x=0..2*Pi,th=3,co=black,
style=point,symbol=cross,symbolsize=50):
> p[3]:=plot(z(x),x=0..2*Pi,th=1,co=black,title="Pulse-Values"):
> p[4]:=plot(G(x,n=15,N=16),x=0..2*Pi,th=2,co=black):
> p[5]:=plots[textplot]({[3,75,`Cubic Spline Interpolation`],
[2,50,`FOURIER-Series k = 15`],[2,45,`Its Smoothing
G(x,n=15,N=16)`]}):
> plots[display](seq(p[k],k=1..5));

```



```
>
> # Error Norm of the smoothing with respect to the
> splinefunction:
> M[2][G_k=15]:=
sqrt((1/5.685)*Int((SP(xi)-G(xi))^2,x=0.299..5.984))=
evalf(sqrt((1/5.685)*int((Sp(x)-G(x,n=15,N=16))^2,x=0.299..5.9
84)))/YM;
```

$$M_{2_{G_k=15}} := 0.4194061220 \sqrt{\int_{0.299}^{5.984} (SP(\xi) - G(\xi))^2 dx} = 0.04649703844$$

```
> # Error Norm of the smoothing with respect to the given data:
> with(linalg):
> for i from 2 to 21 do
  r[i]:=evalf(subs(x=DATA[i][1],G(x,n=15,N=16))-DATA[i][2]) od:
> R:=vector([seq(r[i],i=2..21)]);
```

```
R := [-1.689240123, 0.4062067270, -0.1245478528, -1.195870476, -1.767450212,
      2.515720582, -0.8807586212, -0.6939730671, 0.1358380948, 0.7443021160, 0.5812004139,
      0.8126073702, -7.143995359, 5.623023666, 1.887850136, 0.4772601670, -0.154350460,
      -5.400189975, -0.1914917503, 7.379010261]
```

```
> m[2][G_k=15]:=
(1/sqrt(number_of_points))*Norm(R,2)=
evalf((1/sqrt(20))*norm(R,2))/YM;
```

$$m_{2_{G_k=15}} := \frac{\text{Norm}(R, 2)}{\sqrt{\text{number_of_points}}} = 0.05357869152$$

The next part of this worksheet is concerned with the

Dimensionless Representation

where the given pulse values have been divided by its mean value $YM = 57.05$.

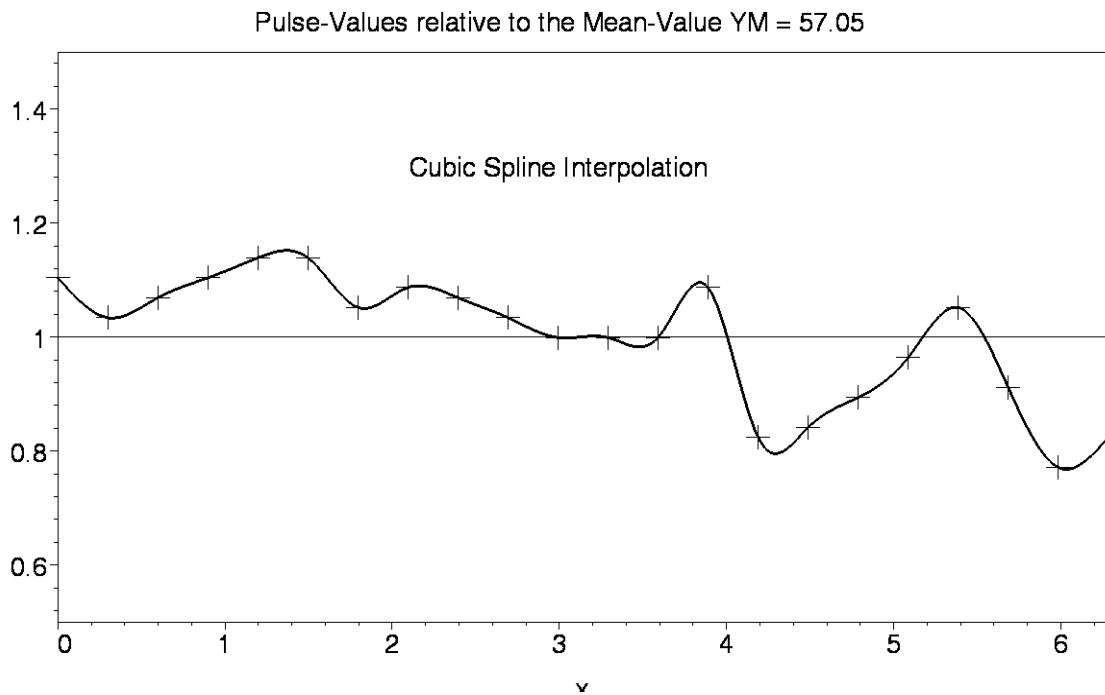
```
> restart:
> with(Statistics):
```

```

> X:=array([seq(2*Pi*(i-1)/21,i=1..22)]);
X:=
$$\left[0, \frac{2\pi}{21}, \frac{4\pi}{21}, \frac{2\pi}{7}, \frac{8\pi}{21}, \frac{10\pi}{21}, \frac{4\pi}{7}, \frac{2\pi}{3}, \frac{16\pi}{21}, \frac{6\pi}{7}, \frac{20\pi}{21}, \frac{22\pi}{21}, \frac{8\pi}{7}, \frac{26\pi}{21}, \frac{4\pi}{3}, \frac{10\pi}{7}, \frac{32\pi}{21}, \frac{34\pi}{21}, \frac{12\pi}{7}, \frac{38\pi}{21}, \frac{40\pi}{21}, 2\pi\right]$$

> whattype(X);
symbol
> Y:=array([63,59,61,63,65,65,60,62,61,59,57,57,57,62,47,48,51,55,60,52,44,47]);
Y:= [63, 59, 61, 63, 65, 65, 60, 62, 61, 59, 57, 57, 57, 62, 47, 48, 51, 55, 60, 52, 44, 47]
> whattype(Y);
symbol
> YM:=evalf(Mean(Y)); # mean value of Y # dimension = bpm
YM := 57.0454545454545
> Z:=array(evalf([seq(Y[i],i=1..22)/YM],4)); # dimensionless pulse values
Z:= [1.104, 1.034, 1.069, 1.104, 1.139, 1.139, 1.052, 1.087, 1.069, 1.034, 0.9992, 0.9992, 0.9992, 1.087, 0.8239, 0.8414, 0.8940, 0.9642, 1.052, 0.9116, 0.7713, 0.8239]
> whattype(Z);
symbol
>
> ZM:=convert(Mean(Z),rational);
ZM :=  $\frac{52376}{52381}$ 
> DATA:=seq([X[i],Z[i]],i=1..22);
DATA := [0, 1.104],  $\left[\frac{2\pi}{21}, 1.034\right]$ ,  $\left[\frac{4\pi}{21}, 1.069\right]$ ,  $\left[\frac{2\pi}{7}, 1.104\right]$ ,  $\left[\frac{8\pi}{21}, 1.139\right]$ ,  $\left[\frac{10\pi}{21}, 1.139\right]$ ,  $\left[\frac{4\pi}{7}, 1.052\right]$ ,  $\left[\frac{2\pi}{3}, 1.087\right]$ ,  $\left[\frac{16\pi}{21}, 1.069\right]$ ,  $\left[\frac{6\pi}{7}, 1.034\right]$ ,  $\left[\frac{20\pi}{21}, 0.9992\right]$ ,  $\left[\frac{22\pi}{21}, 0.9992\right]$ ,  $\left[\frac{8\pi}{7}, 0.9992\right]$ ,  $\left[\frac{26\pi}{21}, 1.087\right]$ ,  $\left[\frac{4\pi}{3}, 0.8239\right]$ ,  $\left[\frac{10\pi}{7}, 0.8414\right]$ ,  $\left[\frac{32\pi}{21}, 0.8940\right]$ ,  $\left[\frac{34\pi}{21}, 0.9642\right]$ ,  $\left[\frac{12\pi}{7}, 1.052\right]$ ,  $\left[\frac{38\pi}{21}, 0.9116\right]$ ,  $\left[\frac{40\pi}{21}, 0.7713\right]$ , [2π, 0.8239]
> with(CurveFitting):
> Sp(x):=Spline([DATA],x,degree=3):
> alias(th=thickness,co=color):
> p[1]:=plot([DATA],x=0..2*Pi,0.5..1.5,th=3,co=black,style=point,symbol=cross,symbolsize=50,axes=boxed,title="Pulse-Values relative to the Mean-Value YM = 57.05"):
> p[2]:=plot(Sp(x),x=0..2*Pi,th=3,co=black):
> p[3]:=plot(1,x=0..2*Pi,linestyle=3,th=1,co=black):
> p[4]:=plots[textplot]([3,1.3,`Cubic Spline Interpolation`]):
> plots[display](seq(p[k],k=1..4));

```



Nonlinear Regression:

```

> f:=(t,a,b,c,d,e,F)->a+b*t*sin(c*t+d)^2*cos(e*t+F)^2;
      f := (t, a, b, c, d, e, F) → a + b t sin(c t + d)2 cos(e t + F)2
> y:=unapply(evalf(NonlinearFit(f(t,a,b,c,d,e,F),X,Z,t)),t);
y := t → 1.09266851879094 - 0.224820920578031 t
      sin(1.04483494607003 t + 0.752812847119836)2
      cos(1.05579891650084 t + 0.730393401109659)2
> y(x):=subs(t=x,y(t));
y(x) := 1.09266851879094 - 0.224820920578031 x
      sin(1.04483494607003 x + 0.752812847119836)2
      cos(1.05579891650084 x + 0.730393401109659)2
> y(0):=evalf(subs(x=0,y(x)));
      y(0) := 1.09266851879094
> y(0)[data]:=63/YM;
      y(0)data := 1.10438247011952
> y(2*Pi):=evalf(subs(x=2*Pi,y(x)));
      y(2 π) := 0.861640200149388
> y(2*Pi)[data]:=47/YM;
      y(2 π)data := 0.823904382470120
>
> alias(th=thickness,co=color):
> p[1]:=plot([DATA],x=0..2*Pi,0.5..1.5,th=3,co=black,
      style=point,symbol=cross,symbolsize=50,axes=boxed,
      title="Pulse-Values Relative to the Mean-Value YM = 57.05");

```

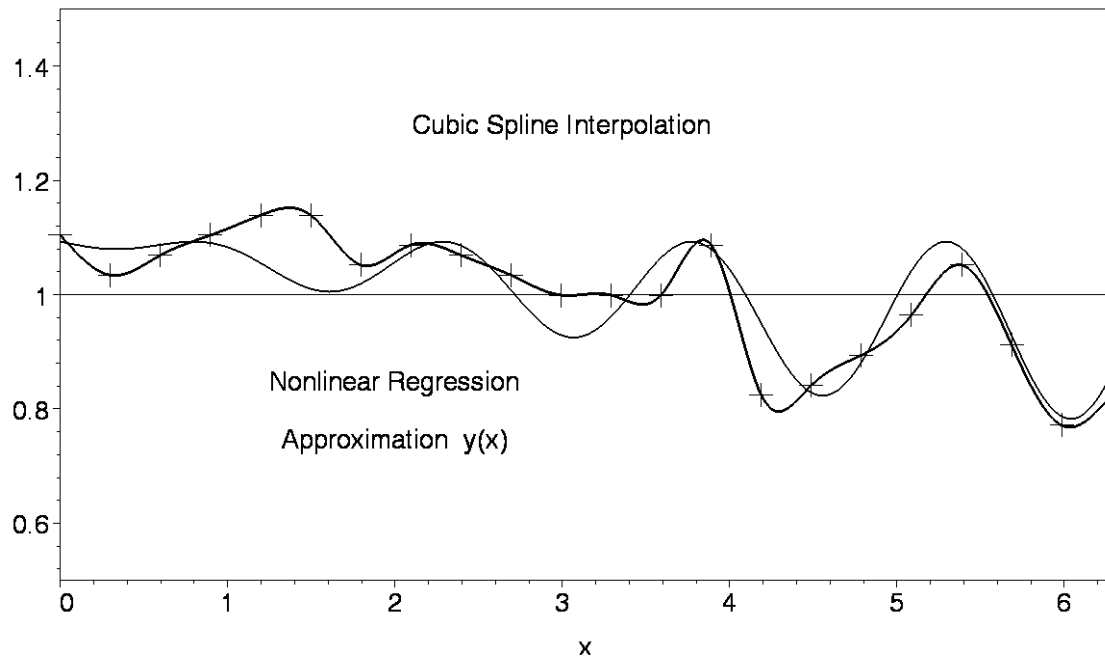


```

> p[2]:=plot(Sp(x),x=0..2*Pi,th=3,co=black):
> p[3]:=plot(y(x),x=0..2*Pi,th=2,co=black):
> p[4]:=plot(1,x=0..2*Pi,linestyle=3,th=1,co=black):
> p[5]:=plots[textplot]([3,1.3,`Cubic Spline Interpolation`],
  [2,0.85,`Nonlinear Regression`],[2,0.75,`Approximation
  y(x)`]):
> plots[display](seq(p[k],k=1..5));

```

Pulse-Values Relative to the Mean-Value YM = 57.05



error norms:

```

> L[2]:=
  sqrt((1/2/Pi)*Int((SPLINE-APPROX)^2,x=0..2*Pi))=
  evalf(sqrt((1/2/Pi)*int((Sp(x)-y(x))^2,x=0..2*Pi)));

```

$$L_2 := \frac{1}{2} \sqrt{2} \sqrt{\frac{1}{\pi} \int_0^{2\pi} (\text{SPLINE} - \text{APPROX})^2 dx} = 0.05356127607$$

```

> with(linalg):
> for i from 1 to 22 do
  v[i]:=evalf(subs(x=DATA[i][1],y(x))-DATA[i][2]) od:
> V:=vector([seq(v[i],i=1..22)]);

```

```

V := [-0.0113314812090588, 0.0457528640497927, 0.0181098702194375,
-0.0136120070822395, -0.0837358516005746, -0.128779622255690, -0.0329208595532596,
-0.0122682511941465, 0.0170208809194298, -0.0261279008333893, -0.0700035722952960,
-0.0378621710536577, 0.0655444942894189, -0.00769729809977263, 0.125626741032895,
-0.0128173799339927, -0.0124657113722093, 0.0777020071766001, 0.0295863098164297,
0.0168280871499092, 0.0157069788319250, 0.0377402001493879]

```

```

> l[2]:=
  (1/sqrt(number_of_points))*Norm(V,2)=
  evalf((1/sqrt(22))*norm(V,2));

```

$$l_2 := \frac{\text{Norm}(V, 2)}{\sqrt{\text{number_of_points}}} = 0.0540184456147690$$

>

Approximation of the Splinefunction by *FOURIER*-Series in dimensionless Representation

> **restart:**

> **with(Statistics):**

> **X:=array([seq(2*Pi*(i-1)/21,i=1..22)]);**

$X := \left[0, \frac{2\pi}{21}, \frac{4\pi}{21}, \frac{2\pi}{7}, \frac{8\pi}{21}, \frac{10\pi}{21}, \frac{4\pi}{7}, \frac{2\pi}{3}, \frac{16\pi}{21}, \frac{6\pi}{7}, \frac{20\pi}{21}, \frac{22\pi}{21}, \frac{8\pi}{7}, \frac{26\pi}{21}, \frac{4\pi}{3}, \frac{10\pi}{7}, \frac{32\pi}{21}, \frac{34\pi}{21}, \frac{12\pi}{7}, \frac{38\pi}{21}, \frac{40\pi}{21}, 2\pi \right]$

> **whattype(X);**

symbol

> **Y:=array([63,59,61,63,65,65,60,62,61,59,57,57,57,62,47,48,51,55,60,52,44,47]);**

$Y := [63, 59, 61, 63, 65, 65, 60, 62, 61, 59, 57, 57, 57, 62, 47, 48, 51, 55, 60, 52, 44, 47]$

> **whattype(Y);**

symbol

> **YM:=evalf(Mean(Y),4); # mean value of Y # dimesion = bpm**

$YM := 57.05$

> **Z:=array(evalf([seq(Y[i],i=1..22)/YM],4)); # dimensionless pulse values**

$Z := [1.104, 1.034, 1.069, 1.104, 1.139, 1.139, 1.052, 1.087, 1.069, 1.034, 0.9992, 0.9992, 0.9992, 1.087, 0.8239, 0.8414, 0.8940, 0.9642, 1.052, 0.9116, 0.7713, 0.8239]$

> **whattype(Z);**

symbol

> **DATA:=seq([X[i],Z[i]],i=1..22);**

$DATA := [0, 1.104], \left[\frac{2\pi}{21}, 1.034 \right], \left[\frac{4\pi}{21}, 1.069 \right], \left[\frac{2\pi}{7}, 1.104 \right], \left[\frac{8\pi}{21}, 1.139 \right], \left[\frac{10\pi}{21}, 1.139 \right], \left[\frac{4\pi}{7}, 1.052 \right], \left[\frac{2\pi}{3}, 1.087 \right], \left[\frac{16\pi}{21}, 1.069 \right], \left[\frac{6\pi}{7}, 1.034 \right], \left[\frac{20\pi}{21}, 0.9992 \right], \left[\frac{22\pi}{21}, 0.9992 \right], \left[\frac{8\pi}{7}, 0.9992 \right], \left[\frac{26\pi}{21}, 1.087 \right], \left[\frac{4\pi}{3}, 0.8239 \right], \left[\frac{10\pi}{7}, 0.8414 \right], \left[\frac{32\pi}{21}, 0.8940 \right], \left[\frac{34\pi}{21}, 0.9642 \right], \left[\frac{12\pi}{7}, 1.052 \right], \left[\frac{38\pi}{21}, 0.9116 \right], \left[\frac{40\pi}{21}, 0.7713 \right], [2\pi, 0.8239]$

> **with(CurveFitting):**

> **# The cubic splinefunction fit to the data is given by:**

> **Sp(x):=Spline([DATA],x,degree=3):**

> **# This function is not printed because of its length. However,**

```

printing is possible, if we insert a semicolon (;) instead of
the doppel point (:) at the end of the command for Sp(x).
> FOURIER_Series(x):=
a[0]/2+sum(a[k]*cos(k*x)+b[k]*sin(k*x),k=1..infinity);
      FOURIER_Series(x) :=  $\frac{1}{2}a_0 + \left( \sum_{k=1}^{\infty} (a_k \cos(kx) + b_k \sin(kx)) \right)$ 
> a[k]:=(1/Pi)*Int(f(x)*cos(k*x),x=0..2*Pi);
       $a_k := \frac{1}{\pi} \int_0^{2\pi} f(x) \cos(kx) dx$ 
> a[0]:=simplify(subs(k=0,a[k]));
       $a_0 := \frac{1}{\pi} \int_0^{2\pi} f(x) dx$ 
> b[k]:=(1/Pi)*Int(f(x)*sin(k*x),x=0..2*Pi);
       $b_k := \frac{1}{\pi} \int_0^{2\pi} f(x) \sin(kx) dx$ 
> with(CurveFitting):
> F(x):=Sp(x):
> A[0]:=evalf(subs(f(x)=F(x),a[0]));
       $A_0 := 2.001661937$ 
> A[k]:=simplify(value(subs(f(x)=F(x),a[k]))):
> A[k]:=subs({sin(k*Pi)=0.,(cos(k*Pi))^2=1.},%):
> A[k]:=evalf(%):
> for i in [seq(k,k=1..5)] do A[i]:=subs(k=i,A[k]) od:
> for i in [seq(k,k=1..15)] do P[i]:=subs(k=i,A[k]) od:
> for i in [seq(k,k=1..100)] do R[i]:=subs(k=i,A[k]) od:
> B[k]:=simplify(value(subs(f(x)=F(x),b[k]))):
> B[k]:=subs({sin(k*Pi)=0.,(cos(k*Pi))^2=1.},%):
> for i in [seq(k,k=1..5)] do B[i]:=subs(k=i,B[k]) od:
> for i in [seq(k,k=1..15)] do Q[i]:=subs(k=i,B[k]) od:
> for i in [seq(k,k=1..100)] do S[i]:=subs(k=i,B[k]) od:
> # The coefficients A[i]...S[i] are not printed because of their
length. However,
> # it can be done by inserting semicolons (;) instead of
doppelpoints (:) at the # ends of the commands. The
FOURIER-Series with k = 5, 15, 100 are expressed by # y(x),
z(x), T(x), respectively, as follows:
>
> y(x):=evalf(A[0]/2+sum(A[k]*cos(k*x)+B[k]*sin(k*x),k=1..5));
y(x) := 1.000830968 - 0.01825313978 cos(x) + 0.1006481742 sin(x)
      - 0.01830805872 cos(2. x) + 0.01773632949 sin(2. x) - 0.04520463741 cos(3. x)
      - 0.005376925601 sin(3. x) - 0.03759917455 cos(4. x) + 0.04490080094 sin(4. x)
      + 0.02997855942 cos(5. x) + 0.03300989107 sin(5. x)

```

```

> y(0):=simplify(subs(x=0,y(x)));
                                y(0) := 0.9114445169
> y(0)[data]:=63/YM;
                                y(0)data := 1.104294479
> y(2*Pi):=simplify(subs(x=2*Pi,y(x)));
                                y(2 π) := 0.9114445173
> y(2*Pi)[data]:=47/YM;
                                y(2 π)data := 0.8238387381
> z(x):=evalf(A[0]/2+sum(P[k]*cos(k*x)+Q[k]*sin(k*x),k=1..15));
z(x) := -0.018253144 cos(x) + 0.1006481786 sin(x) + 0.004293500446 cos(13. x)
+ 0.00959191560 sin(12. x) - 0.0002209624171 cos(12. x) + 0.003122387705 cos(11. x)
+ 0.00098482511 sin(11. x) + 0.01934109920 sin(10. x) + 0.004129050911 cos(10. x)
- 0.002720989955 cos(9. x) + 0.00307326010 sin(9. x) + 0.02453893885 cos(8. x)
- 0.00283646628 sin(8. x) + 0.02993027422 sin(7. x) + 0.003127297193 cos(7. x)
+ 0.0007299447424 cos(6. x) + 0.01860477310 sin(6. x) + 0.0008050868405 cos(15. x)
+ 0.00584797498 sin(15. x) + 0.001001552415 cos(14. x) + 0.00529394883 sin(14. x)
+ 0.00886349836 sin(13. x) + 0.02997855942 cos(5. x) + 0.03300989099 sin(5. x)
- 0.03759917454 cos(4. x) + 0.0449008008 sin(4. x) - 0.04520463744 cos(3. x)
- 0.0053769260 sin(3. x) - 0.01830805854 cos(2. x) + 0.0177363298 sin(2. x)
+ 1.000830968
> z(0):=simplify(subs(x=0,z(x)));
                                z(0) := 0.9502503196
> z(0)[data]:=63/YM;
                                z(0)data := 1.104294479
> z(2*Pi):=simplify(subs(x=2*Pi,z(x)));
                                z(2 π) := 0.9502503209
> z(2*Pi)[data]:=47/YM;
                                z(2 π)data := 0.8238387381
>
> T(x):=evalf(A[0]/2+sum(R[k]*cos(k*x)+S[k]*sin(k*x),k=1..100)):
> # This equation has not be printed because of its length with
more than 100 terms.However, printing is possible, if we insert
a semicolon (;) instead of the doppelpoint (:) at the end of the
command for T(x).
> T(0):=simplify(subs(x=0,T(x)));
                                T(0) := 0.9618525872
> T(0)[data]:=63/YM;
                                T(0)data := 1.104294479
> T(2*Pi):=simplify(subs(x=2*Pi,T(x)));
                                T(2 π) := 0.9618525945
> T(2*Pi)[data]:=47/YM;

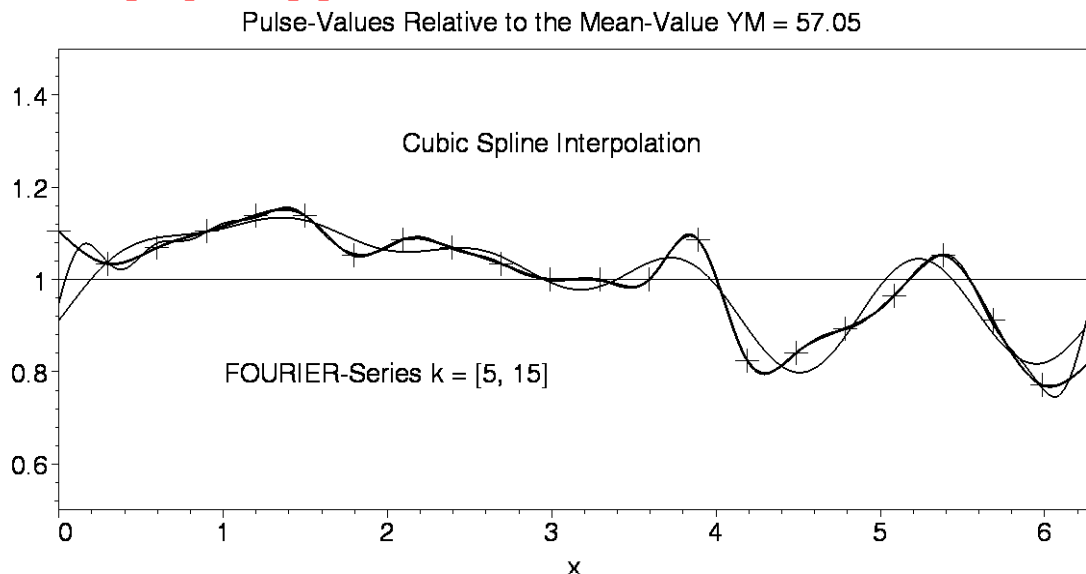
```

$$T(2\pi)_{data} := 0.8238387381$$

```

> # The following Figure illustrates the approximations of the
  splinefunctions by FOURIER-Series with k = [5, 15]:
> alias(th=thickness,co=color):
> p[1]:=plot(Sp(x),x=0..2*Pi,0.5..1.5,axes=boxed,th=3,co=black):
> p[2]:=plot([DATA],x=0..2*Pi,th=3,co=black,
  style=point,symbol=cross,symbolsize=50,
  title="Pulse-Values Relative to the Mean-Value YM = 57.05"):
> p[3]:=plot({y(x),z(x)},x=0..2*Pi,th=2,co=black):
> p[4]:=plot(1,x=0..2*Pi,linestyle=3,th=1,co=black):
> p[5]:=plots[textplot]({[3,1.3,`Cubic Spline
  Interpolation`],[2,0.8,`FOURIER-Series k = [5, 15]`]}):
> plots[display](seq(p[k],k=1..5));

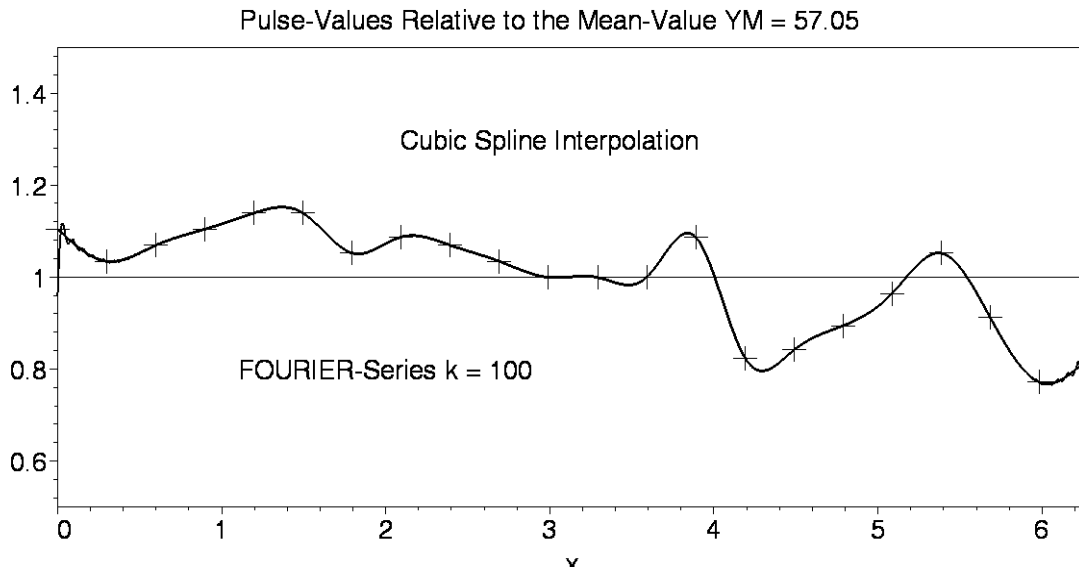
```



```

> # The next Figure represents the approximation of the
  splinefunction by a FOURIER-Series with k = 100:
> alias(th=thickness,co=color):
> p[1]:=plot(Sp(x),x=0..2*Pi,0.5..1.5,axes=boxed,th=3,co=black):
> p[2]:=plot([DATA],x=0..2*Pi,th=3,co=black,style=point,symbol=cro
  ss, symbolsize=50,title="Pulse-Values Relative to the Mean-Value
  YM = 57.05"):
> p[3]:=plot(T(x),x=0..2*Pi,th=2,co=black):
> p[4]:=plot(1,x=0..2*Pi,linestyle=3,th=1,co=black):
> p[5]:=plots[textplot]({[3,1.3,`Cubic Spline
  Interpolation`],[2,0.8,`FOURIER-Series k = 100`]}):
> plots[display](seq(p[k],k=1..5));

```



In this Figure we cannot observe differences between the splinefunction and the FOURIER-Series. Only immediately at the beginning $x = 0$ and at the end $x = 2\pi$ there are some differences:

```

> T(0):=simplify(subs(x=0,T(x)));
  T(0.0207):=simplify(subs(x=0.0207,T(x)));
                    T(0) := 0.9618525872
                    T(0.0207) := 1.104157090
> T(0)[data]:=63/YM;
                    T(0)data := 1.104294479
> T(2*Pi):=simplify(subs(x=2*Pi,T(x)));
  T(1.9942*Pi):=simplify(subs(x=1.9942*Pi,T(x)));
                    T(2 π) := 0.9618525945
                    T(6.264964071) := 0.8228604027
> T(2*Pi)[data]:=47/YM;
                    T(2 π)data := 0.8238387381

```

The *error norms* $L[2](k=5,15,100)$, of the approximations $y(x)$, $z(x)$, and $T(x)$ with respect to the cubic splinefunction can be expressed as:

```

> L[2][k=5]:=
  sqrt((1/5.685)*Int((SP(xi)-y(xi))^2,xi=0.299..5.884))=
  sqrt((1/5.685)*int((Sp(x)-y(x))^2,x=0.299..5.984));
                    L2k=5 := 0.4194061220  $\sqrt{\int_{0.299}^{5.884} (SP(\xi) - y(\xi))^2 d\xi} = 0.02973952528$ 
> L[2][k=15]:=
  sqrt((1/5.685)*Int((SP(xi)-z(xi))^2,xi=0.299..5.984))=
  sqrt((1/5.685)*int((Sp(x)-z(x))^2,x=0.299..5.984));

```

$$L_{2_{k=15}} := 0.4194061220 \sqrt{\int_{0.299}^{5.984} (\text{SP}(\xi) - z(\xi))^2 d\xi} = 0.004174537492$$

```
> L[2][k=100]:=
sqrt((1/5.685)*Int((SP(xi)-T(xi))^2,xi=0.299..5.984))=
sqrt((1/5.685)*int((Sp(x)-T(x))^2,x=0.299..5.984));
```

$$L_{2_{k=100}} := 0.4194061220 \sqrt{\int_{0.299}^{5.984} (\text{SP}(\xi) - T(\xi))^2 d\xi} = 0.0006794214249$$

>

The three values above show the $L[2]$ error norms of the FOURIER-Series $y(x)$, $z(x)$, and $T(x)$ relative to the *cubic splinefunction*. Furthermore, one should discuss the *error norms* with respect to the given measured data, as follows:

```
> with(linalg):
> for i from 2 to 21 do
  u[i]:=subs(x=DATA[i][1],y(x))-DATA[i][2] od:
> U:=evalf(vector([seq(u[i],i=2..21)]),4);
U := [0.00476, 0.02145, -0.00058, -0.01125, -0.0108, 0.03477, -0.02718, -0.00102, 0.01562,
      -0.00552, -0.01503, 0.03711, -0.06172, 0.07168, -0.04284, -0.01674, 0.05665, -0.03253,
      -0.02779, 0.04708]
```

```
> l[2][k=5]:=
(1/sqrt(number_of_points))*Norm(U,2)=
evalf((1/sqrt(20))*norm(U,2),4);
```

$$l_{2_{k=5}} := \frac{\text{Norm}(U, 2)}{\sqrt{\text{number_of_points}}} = 0.03383$$

```
> for i from 2 to 21 do
  v[i]:=subs(x=DATA[i][1],z(x))-DATA[i][2] od:
> V:=evalf(vector([seq(v[i],i=2..21)]),4);
V := [0.00241, 0.00974, -0.00089, -0.00523, 0.00074, 0.00406, -0.00120, -0.00279, 0.00126,
      0.00282, -0.00197, -0.00211, 0.00088, 0.00266, -0.00300, -0.00180, 0.00469, 0.00199,
      -0.00840, -0.00680]
```

```
> l[2][k=15]:=
(1/sqrt(number_of_points))*Norm(V,2)=
evalf((1/sqrt(20))*norm(V,2),4);
```

$$l_{2_{k=15}} := \frac{\text{Norm}(V, 2)}{\sqrt{\text{number_of_points}}} = 0.004099$$

```
> for i from 2 to 21 do
  w[i]:=subs(x=DATA[i][1],T(x))-DATA[i][2] od:
> W:=evalf(vector([seq(w[i],i=2..21)]),4);
W := [-0.00060, 0.00135, 0.00065, -0.00051, -0.0007, 0.00005, 0.00068, 0.00003, -0.00071,
      -0.00039, -0.00025, -0.00012, -0.00062, -0.00034, -0.00014, 0.0007, 0.00049, 0.00009,
      -0.00100, 0.00030]
```

```
> l[2][k=100]:=
```

```
(1/sqrt(number_of_points))*Norm(W,2)=  
evalf((1/sqrt(20))*norm(W,2),4);
```

$$l_{2_{k=100}} := \frac{\text{Norm}(W, 2)}{\sqrt{\text{number_of_points}}} = 0.0005874$$

Résumé: The above *error norms* - $L[2]$ and $l[2]$ - show that the FOURIER-Series $y(x)$, $z(x)$, and $T(x)$ are *best approximations* fit to the given measured data. Furthermore, the *Nonlinear Regression* furnish *nonlinear model functions* suitable to fit the given data, too.

>