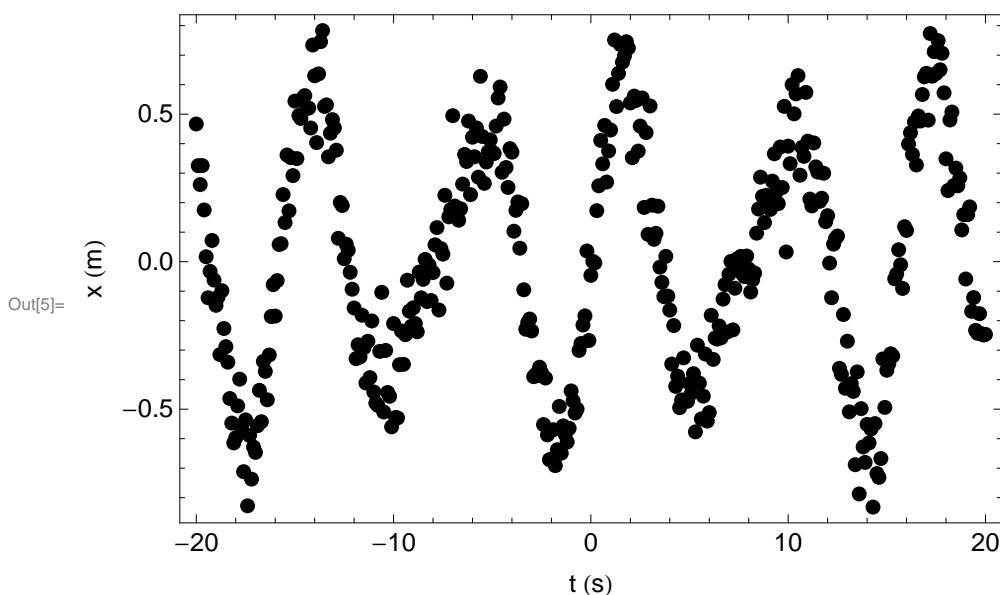


# Nelinearna regresija

```
In[1]:= SetOptions[ListPlot, ImageSize → 6 × 72, Frame → True,
  GridLinesStyle → Directive[Gray, Dashed],
  Method → {"GridLinesInFront" → True},
  PlotStyle → Directive[Thickness[Medium], PointSize[Large], Black],
  TicksStyle → Thickness[Medium], AxesStyle → Thickness[Medium],
  LabelStyle → Directive[FontFamily → "Helvetica", FontSize → 12], Axes → False];
SetOptions[Plot, ImageSize → 6 × 72, Frame → True,
  GridLinesStyle → Directive[Gray, Dashed],
  Method → {"GridLinesInFront" → True},
  PlotStyle → Directive[Thickness[Large], PointSize[Large], Red],
  TicksStyle → Thickness[Medium], AxesStyle → Thickness[Medium],
  LabelStyle → Directive[FontFamily → "Helvetica", FontSize → 12], Axes → False];
SetOptions[ContourPlot, ImageSize → 4 × 72, Frame → True,
  GridLinesStyle → Directive[Gray, Dashed], Method → {"GridLinesInFront" → True},
  TicksStyle → Thickness[Medium], AxesStyle → Thickness[Medium],
  LabelStyle → Directive[FontFamily → "Helvetica", FontSize → 12], Axes → False];

In[4]:= data =
  Table[{t, 0.5 Sin[0.8 t] + 0.2 Sin[1.2 t] + Random[NormalDistribution[0, 0.1]]},
    {t, -20, 20, 0.1}];

In[5]:= pdata = ListPlot[data, PlotRange → All, FrameLabel → {"t (s)", "x (m)"}]
```

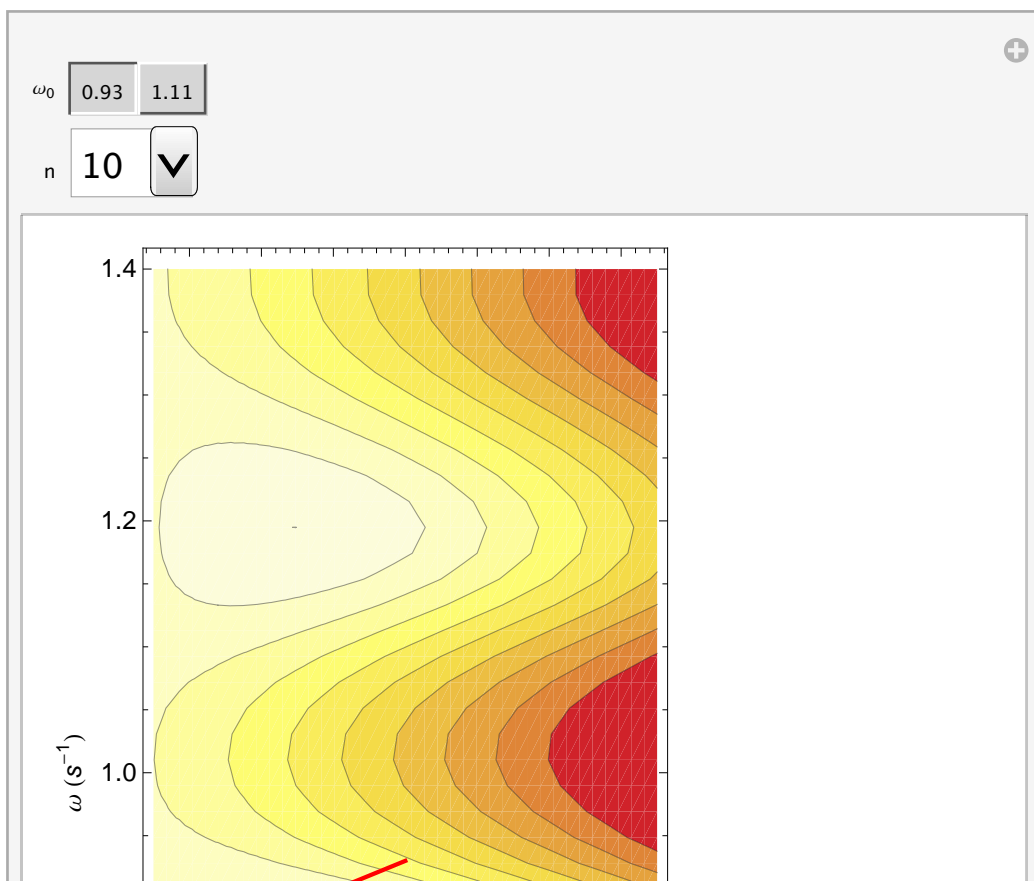


```
In[6]:= f1[a_, ω_, t_] := a Sin[ω t]
chi21[a_, ω_] :=
  Sum[(data[[i, 2]] - f1[a, ω, data[[i, 1]])^2, {i, Length[data]}) / Length[data]
contour1 = ContourPlot[Log[chi21[a, ω]], {a, 0.05, 0.75},
  {ω, 0.6, 1.4}, Contours → 20, MaxRecursion → 0,
  PlotPoints → 40, ColorFunction → ColorData["TemperatureMap"],
  AspectRatio → 2, FrameLabel → {"x0 (m)", "ω (s-1)"}];
```

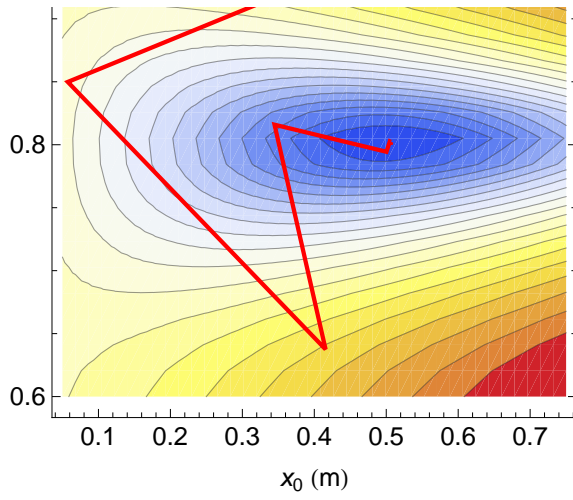
```

In[19]:= Manipulate[
  Module[{fit, eval, p1, p2},
    eval = {};
    fit = NonlinearModelFit[data, f1[a,  $\omega$ , t], {{a, 0.4}, { $\omega$ ,  $\omega_0$ }},
      t, EvaluationMonitor  $\rightarrow$  AppendTo[eval, {a,  $\omega$ , chi21[a,  $\omega$ ]}]];
    p1 = Show[contour1, Graphics[{Red, Thickness[Large],
      Line[eval[[1 ;; n, 1 ;; 2]]]}]];
    p2 = Show[pdata, Plot[f1[eval[[n, 1]], eval[[n, 2]], t], {t, -20, 20}]];
    Row[{
      p1,
      " ",
      Column[{
        StyleForm[ToString[TraditionalForm[ $x_0 \sin[\omega t]$ ]],
          FontSize  $\rightarrow$  20, FontFamily  $\rightarrow$  "Helvetica", FontColor  $\rightarrow$  Red],
        p2,
        StyleForm[
          ToString[TraditionalForm[ $\chi^2/N$ ]] <> " = " <> ToString[eval[[n, 3]]],
          FontSize  $\rightarrow$  20, FontFamily  $\rightarrow$  "Helvetica", FontColor  $\rightarrow$  Red],
        StyleForm["Korelacijska matrika v minimumu:", FontColor  $\rightarrow$  Red],
        StyleForm[TableForm[fit["CorrelationMatrix"],
          TableHeadings  $\rightarrow$  {{ $x_0$ , " $\omega$ "}, { $x_0$ , " $\omega$ "}}], FontColor  $\rightarrow$  Red]
        }, Center, Spacings  $\rightarrow$  2]
      ]}
  ],
  { $\omega_0$ , {0.93, 1.11}}, {n, Range[1, 10]}
]

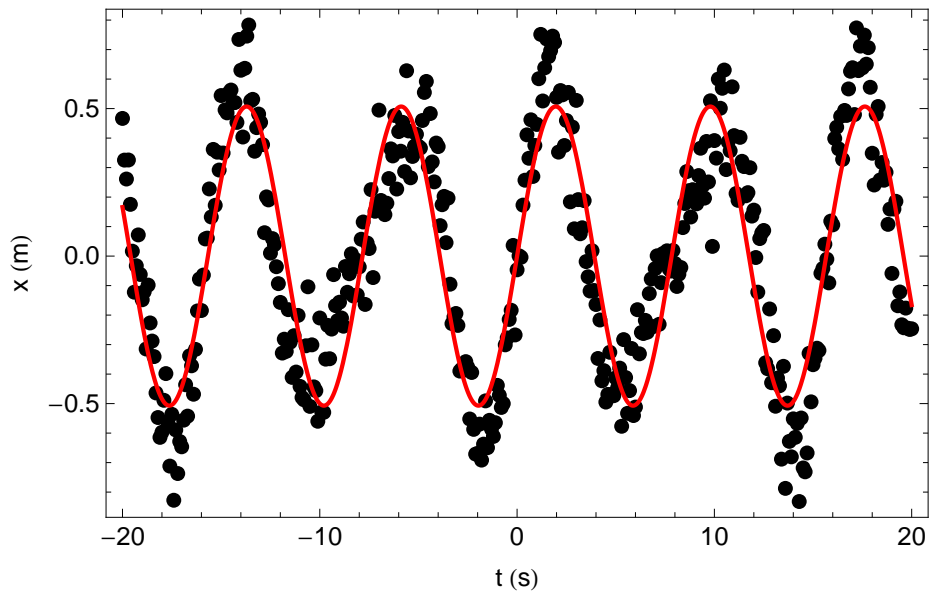
```



Out[19]=



$$x_0 \sin(t \omega)$$



$$\frac{\chi^2}{N} = 0.0290324$$

Korelacijska matrika v minimumu:

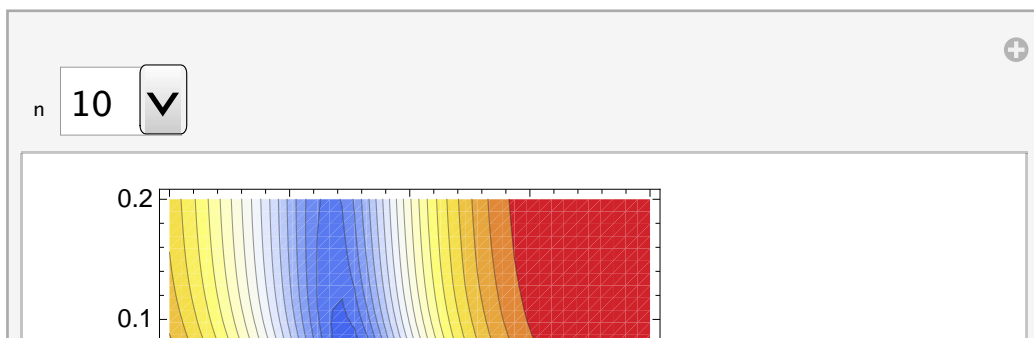
	$x_0$	$\omega$
$x_0$	1.	0.03758
$\omega$	0.03758	1.

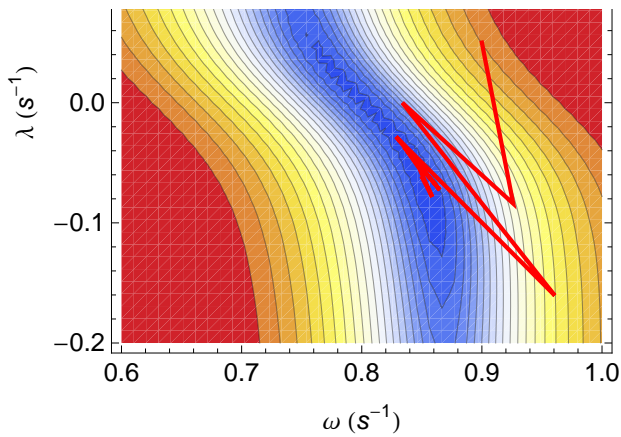
```

In[10]:=  $\omega_0 = 0.8;$ 
         t1 = 40;
         t2 = 60;
         f2[ $\omega$ _,  $\lambda$ _, t_] := 0.5 Sin[ $\omega$  t + Tanh[ $\lambda$  t]]
         chi22[ $\omega$ _,  $\lambda$ _] :=
           Sum[(data[[i, 2]] - f2[ $\omega$ ,  $\lambda$ , data[[i, 1]])^2, {i, Length[data]}] / Length[data]
         contour2 = ContourPlot[Log[chi22[ $\omega$ ,  $\lambda$ ]], { $\omega$ , 0.6, 1},
           { $\lambda$ , -0.2, 0.2}, Contours -> 20, MaxRecursion -> 0,
           PlotPoints -> 40, ColorFunction -> ColorData["TemperatureMap"],
           AspectRatio -> 1, FrameLabel -> {" $\omega$  (s-1)", " $\lambda$  (s-1)"}];

In[16]:= Manipulate[
  Module[{fit, fit1, eval, p1, p2},
    eval = {};
    fit = NonlinearModelFit[data, f2[ $\omega$ ,  $\lambda$ , t], {{ $\omega$ , 0.9}, { $\lambda$ , 0.05}},
      t, EvaluationMonitor -> AppendTo[eval, { $\omega$ ,  $\lambda$ , chi22[ $\omega$ ,  $\lambda$ ]}];
    fit1 = NonlinearModelFit[data, f2[ $\omega$ , 0, t], {{ $\omega$ , 0.9}}, t];
    p1 = Show[contour2,
      Graphics[{Red, Thickness[Large], Line[eval[[1 ;; n, 1 ;; 2]]]}];
    p2 = Show[pdata, Plot[fit1[t], {t, -20, 20},
      PlotStyle -> Directive[Blue, Thickness[Large]]],
      Plot[f2[eval[[n, 1]], eval[[n, 2]], t], {t, -20, 20}];
    Row[{
      p1,
      " ",
      Column[{
        StyleForm[ToString[TraditionalForm[x0 Sin[ $\omega$  t]]],
          FontSize -> 20, FontFamily -> "Helvetica", FontColor -> Blue],
        StyleForm[ToString[TraditionalForm[x0 Sin[ $\omega$  t + Tanh[ $\lambda$  t]]],
          FontSize -> 20, FontFamily -> "Helvetica", FontColor -> Red],
        p2,
        StyleForm[
          ToString[TraditionalForm[ $\chi^2/N$ ]] <> " = " <> ToString[eval[[n, 3]]],
          FontSize -> 20, FontFamily -> "Helvetica", FontColor -> Red],
        StyleForm["Korelacijska matrika v minimumu:", FontColor -> Red],
        StyleForm[TableForm[fit["CorrelationMatrix"],
          TableHeadings -> {{" $\omega$ ", " $\lambda$ "}, {" $\omega$ ", " $\lambda$ "}}], FontColor -> Red]
      ], Center, Spacings -> 2]
    ]],
  ],
  {n, Range[1, 10]}
]

```

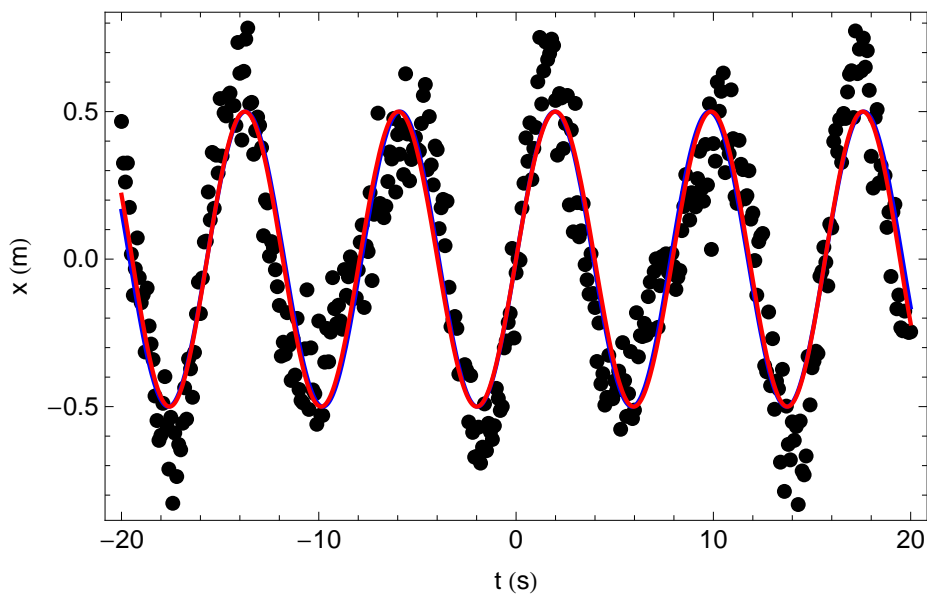




$$x_0 \sin(t \omega)$$

$$x_0 \sin(\tanh(\lambda t) + t \omega)$$

Out[16]=



$$\frac{\chi^2}{N} = 0.0279961$$

Korelacijska matrika v minimumu:

	$\omega$	$\lambda$
$\omega$	1.	-0.914781
$\lambda$	-0.914781	1.

**1. naloga:** Teorija kemijske kinetike napove za sigmoidno krivuljo iz podatkov *Adrenalin.dat* naslednjo odvisnost  $F / F_{\max} = c / (a + c)$ , kjer pomeni  $a$  koncentracijo s polovičnim maksimalnim učinkom. S pomočjo nelinearne regresije določi koeficienta  $F_{\max}$  in  $a$ . Rezultate primerjaj z rezultati naloge 7.2.

**2. naloga:** V datoteki *RefractiveIndex.txt* so zbrane meritve lomnega količnika nekega materiala v odvisnosti od valovne dolžine pri različnih temperaturah. Odvisnost lomnega količnika od valovne dolžine opisuje Sellmeierjeva formula

$n(\lambda)^2 - 1 = \sum_{i=1}^N \frac{S_i}{1 - \left(\frac{\lambda_i}{\lambda}\right)^2}$ , kjer so  $\lambda_i$  valovne dolžine elektronskih prehodov v opazovanem materialu,  $S_i$  pa pripadajoče uteži. Z nelinearno regresijo določi  $\lambda_i$  in  $S_i$  (a) ob predpostavki, da meritev lahko opišemo z enim samim elektronskim preходом ( $N=1$ ) in (b) ob predpostavki, da lahko meritev opišemo z dvema takima prehodoma ( $N=2$ ). Nariši temperaturno odvisnost valovnih dolžin  $\lambda_i$  in uteži  $S_i$ .