

Plot Utilities

ListCollect

Levenberg Marquardt's second algorithm

$$ft(x) := \beta_2 + \frac{\beta_1 - \beta_2}{1 + \left(\frac{x}{\beta_3}\right)^{\beta_4}} - \frac{\beta_1 - \beta_2}{\left(1 + \left(\frac{x}{\beta_3}\right)^{\beta_4}\right)^2} \cdot \left(\frac{x}{\beta_3}\right)^{\beta_4} \cdot \beta_4$$

 $\varepsilon := 10^{-12}$

clear ln0 undefined

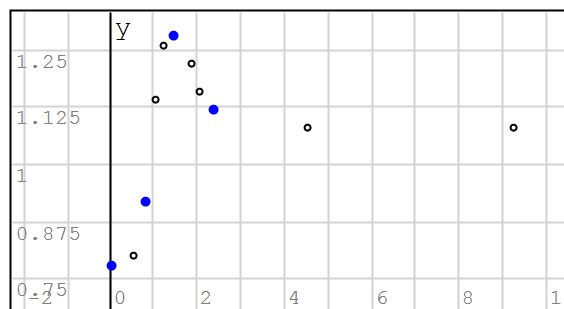
Point solve equalities

$$XY := \begin{bmatrix} \varepsilon & 0.78 \\ 0.51 & 0.8 \\ 0.82 & 0.92 \\ 1.02 & 1.14 \\ 1.23 & 1.26 \\ 1.43 & 1.28 \\ 1.85 & 1.22 \\ 2.05 & 1.16 \\ 2.36 & 1.12 \\ 4.52 & 1.08 \\ 9.22 & 1.08 \end{bmatrix}$$

```
list := [1 3 6 9]
xy := ListCollect(XY, list)
```

```
1. select the support points
2. collect for al_nleqsolve
```

```
raw := plot(XY, "o", 5, "black")
support := plot(xy, ".", 14, "blue")
```



$$xy = \begin{bmatrix} 0 & 0.78 \\ 0.82 & 0.92 \\ 1.43 & 1.28 \\ 2.36 & 1.12 \end{bmatrix}$$

4 DATA

```
x := col(xy, 1)
y := col(xy, 2)
```

t0 := time(1)

JEAN

 $\beta := [0.5 \ 1 \ 1 \ 1]^T$

Tol := 0.00000001

$$LMSA(x, y, \beta, Tol, \beta_s) = \begin{bmatrix} 0.7800 \\ 1.063 \\ 1.317 \\ 4.850 \end{bmatrix}$$

$$\begin{bmatrix} 0.78 \\ 1.06345 \\ 1.3168 \\ 4.85016 \end{bmatrix}$$

time(1) - t0 = 0.723 s

automated ... either NUMERIC/SYMBOLIC

$$\Phi(x, \alpha) := \alpha_2 + \frac{\alpha_1 - \alpha_2}{1 + \left(\frac{x}{\alpha_3}\right)^{\alpha_4}} - \frac{\alpha_1 - \alpha_2}{\left(1 + \left(\frac{x}{\alpha_3}\right)^{\alpha_4}\right)^2} \cdot \left(\frac{x}{\alpha_3}\right)^{\alpha_4} \cdot \alpha_4$$

```
f(α) :=
  ct := 1
  for j ∈ [1..rows(x)]
    Q ct := Φ(x_j, α) - y_j
    ct := ct + 1
  Q
```

$$\alpha := \beta_s = \begin{bmatrix} 0.77999 \\ 1.06345 \\ 1.31681 \\ 4.84968 \end{bmatrix}$$

Original version

$$f(\alpha) = \begin{bmatrix} \Phi(x_1, \alpha) - y_1 \\ \Phi(x_2, \alpha) - y_2 \\ \Phi(x_3, \alpha) - y_3 \\ \Phi(x_4, \alpha) - y_4 \end{bmatrix}$$

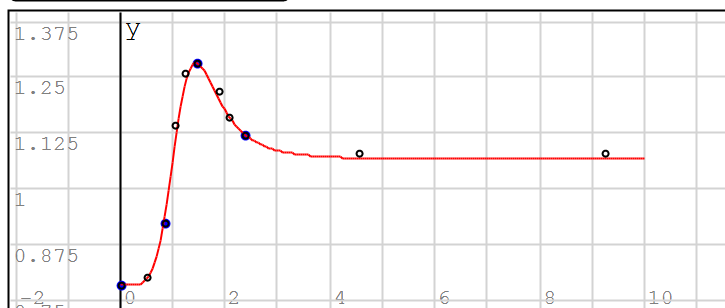
Notes...

1. solves a system of [n equalities = n parameters.
2. The order of equalities is immaterial.
3. Guess wisely the best support equalities !

JEAN

$$\text{sanity} := \begin{bmatrix} \Phi(0, \alpha) \\ \Phi(1, \alpha) \\ \Phi(2, \alpha) \\ \Phi(3, \alpha) \end{bmatrix} = \begin{bmatrix} 0.78 \\ 1.066 \\ 1.172 \\ 1.083 \end{bmatrix} \quad \begin{bmatrix} 0.78 \\ 1.066 \\ 1.172 \\ 1.083 \end{bmatrix}$$

$$f(\alpha) = \begin{bmatrix} -0.000010553698 \\ 0.000007914683 \\ -0.000015697896 \\ 0.000016568730 \end{bmatrix} \quad \begin{array}{l} \text{Clear}(x) = 1 \\ u := (0 \leq x) \cdot (x \leq 10) \end{array}$$

Fit "a la carte"

$$\left\{ \begin{array}{l} \text{raw} \\ \Phi(x, \alpha) \\ u \\ \text{support} \end{array} \right.$$

Clear(x, y) = 1.000

ALL DATA

$$\left\{ \begin{array}{l} X := \text{col}(XY, 1) \\ Y := \text{col}(XY, 2) \end{array} \right.$$
Tol := 10⁻¹⁶

data := plot(XY, "o", 5, "black")

t0 := time(1)**JEAN**

$$\text{LMSA}(X, Y, \beta, \text{Tol}, \beta_s) = \begin{bmatrix} 0.7773 \\ 1.0727 \\ 1.2781 \\ 4.8008 \end{bmatrix} \quad \begin{bmatrix} 0.778997 \\ 1.072526 \\ 1.281669 \\ 4.824839 \end{bmatrix}$$

time(1) - t0 = 3.545 s

Minimize Conjugate-Gradient

$$f(x, \beta_s) := \beta_s \frac{\beta_s \frac{1 - \beta_s}{2} + \frac{\beta_s \frac{1 - \beta_s}{2}}{\beta_s^4} - \frac{\beta_s \frac{1 - \beta_s}{2}}{\left(1 + \left(\frac{x}{\beta_s}\right)^4\right)^2} \cdot \left(\frac{x}{\beta_s}\right)^4}{1 + \left(\frac{x}{\beta_s}\right)^4} \cdot \beta_s^4$$

SSD = 0.0015

corr = 1.0794

Absolute residuals

```

δ := for j ∈ [1..rows(XY)]
      Δj := eval(Yj - f(Xj, β))
      augment(X, Δ)

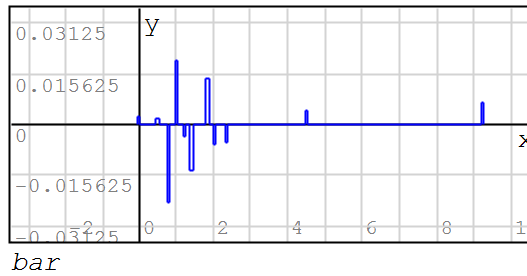
```

```

FIT := U := [0, 0.025..10]
      for i ∈ [1..rows(U)]
        vyi := eval(f(Ui, β))
        augment(U, vy)

```

```
bar := (BarSize(δ, 0.0375))
```

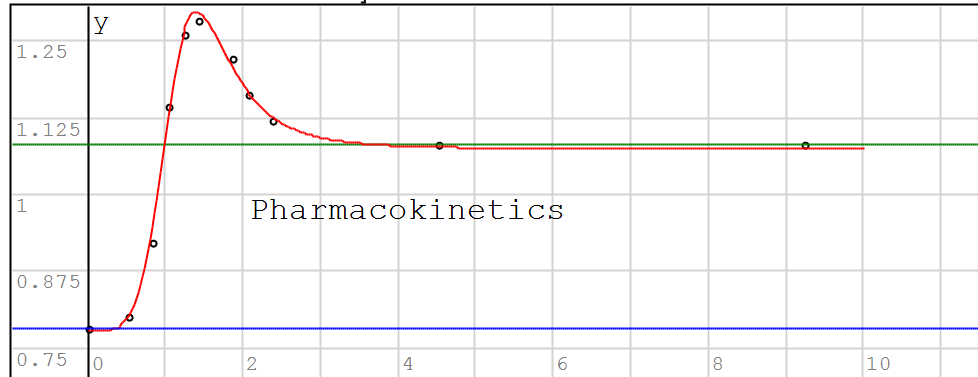


```
Title := [2 1 "Pharmacokinetics" 12 "black"]
```

```

collect := {
  FIT
  1.08
  data
  Title
}

```



```
Clear(x, β) = 1.000
```

FUNCTION

$$ft(x) := \beta_1 \cdot \beta_2 \cdot x^{\beta_2 - 1} \cdot \exp\left(-\beta_1 \cdot x^{\beta_2}\right)$$

$$x := \begin{bmatrix} 0.132 \\ 0.322 \\ 0.511 \\ 0.701 \\ 0.891 \\ 1.082 \\ 1.27 \\ 1.46 \\ 1.65 \\ 1.839 \\ 2.029 \\ 2.219 \end{bmatrix} \quad y := \begin{bmatrix} 0.1 \\ 0.258 \\ 0.543 \\ 0.506 \\ 0.606 \\ 0.622 \\ 0.569 \\ 0.453 \\ 0.438 \\ 0.316 \\ 0.29 \\ 0.195 \end{bmatrix} \quad \beta := \begin{bmatrix} 0.8 \\ 1 \end{bmatrix} \quad Tol := 0.0000000001$$

```
t0 := time(1)
```

$$LMSA(x, y, \beta, Tol, \beta_s) = \begin{bmatrix} 0.5022 \\ 2.0004 \end{bmatrix}$$

FUNCTION

$$ft(t, \beta_s) := \beta_s \beta_1 \cdot \beta_s \beta_2 \cdot t^{\beta_s \beta_2 - 1} \cdot \exp\left(-\beta_s \beta_1 \cdot t^{\beta_s \beta_2}\right)$$

```
time(1) - t0 = 1.042 s
```

```
t := x      m := rows(x)
```

Relative residuals

```

Residuals := for i ∈ [1..rows(t)]
               Δi := eval(1 - yi / ft(ti, βs))
               augment(t, Δ)

```

SSD = "Sum Square Differences"

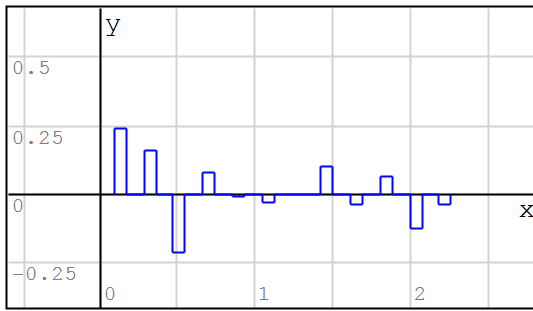
$$SSD := \sum_{i=1}^m \left((y_i - ft(t_i, \beta_s))^2 \right) = 0.0186$$

RMS difference

$$\sqrt{\frac{SSD}{m}} = 0.0394$$

$$RMSD := \sqrt{\frac{SSD}{m}}$$

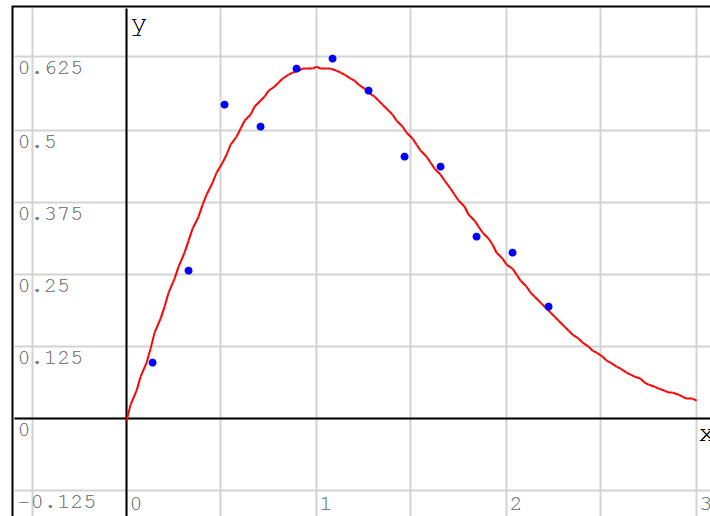
Relative residuals



```
{BarSize(Residuals, RMSD)}
```

```
data := plotG(x, y, ".", 12, "blue")
```

```
FIT := "range, populate"
      U := [0, 0.025..3]
      for i ∈ [1..rows(U)]
        vy_i := eval(ft(U_i, βs))
      augment(U, vy)
```



```
{data
  FIT}
```

⊞ — D2 ... D3

XY := D2

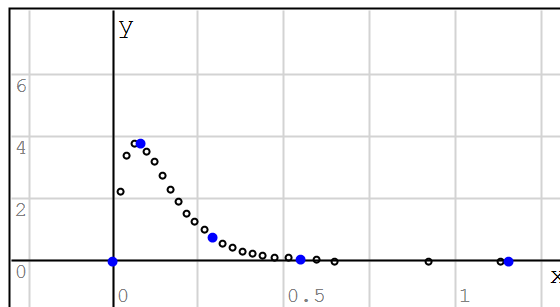
Point solve equalities

rows(D2) = 27

```
{list := [1 5 14 22 27]
  xy := ListCollect(XY, list)}
```

1. select the support points
2. collect for al_nleqsolve

```
raw := plot(XY, "o", 5, "black")
support := plot(xy, ".", 14, "blue")
```



```
{support
  raw}
```

Clear(x, β) = 1

$$ft(x) := x \cdot \left(\beta_1 + \beta_2 \cdot \exp\left(-\frac{x}{\beta_4}\right) + \beta_3 \cdot \exp\left(-\frac{x}{\beta_5}\right) \right) \quad \text{model to D2(x)}$$

$$xy = \begin{bmatrix} 0 & 0 \\ 0.079 & 3.782 \\ 0.29 & 0.77 \\ 0.545 & 0.068 \\ 1.154 & 0 \end{bmatrix}$$

$$\beta := \begin{bmatrix} 10^{-12} & 10 & -5 & 0.1 & -5 \end{bmatrix}^T$$

vector of guesses ...
from robust2crucifying !

$$\begin{cases} x := \text{col}(xy, 1) \\ y := \text{col}(xy, 2) \end{cases}$$

$$Tol := 10^{-16}$$

Jean Solution

$$t0 := \text{time}(1)$$

$$LMSA(x, y, \beta, Tol, \beta_s) =$$

$$\begin{bmatrix} 2.1283 \\ 142.4114 \\ -2.0372 \\ 0.0723 \\ -26.376 \end{bmatrix}$$

$$\begin{bmatrix} 1.1077 \\ 142.40751 \\ -1.01751 \\ 0.07235 \\ -13.58616 \end{bmatrix}$$

$$\text{time}(1) - t0 = 0 \text{ s}$$

$$\alpha := \beta_s$$

$$\Phi(x, \alpha) := x \cdot \left(\alpha_1 + \alpha_2 \cdot \exp\left(-\frac{x}{\alpha_4}\right) + \alpha_3 \cdot \exp\left(-\frac{x}{\alpha_5}\right) \right) \quad \text{model to D2(x)}$$

Implicit equalities = 0

$$f(\alpha) := \begin{bmatrix} \Phi(x_1, \alpha) - y_1 \\ \Phi(x_2, \alpha) - y_2 \\ \Phi(x_3, \alpha) - y_3 \\ \Phi(x_4, \alpha) - y_4 \\ \Phi(x_5, \alpha) - y_5 \end{bmatrix}$$

$$\alpha = \begin{bmatrix} 2.12827 \\ 142.41144 \\ -2.03717 \\ 0.07235 \\ -26.37602 \end{bmatrix}$$

Notes...Smath 7251

1. solves a system of [n equalities = n parameters.
2. The order of equalities is immaterial.
3. Guess wisely the best support equalities !

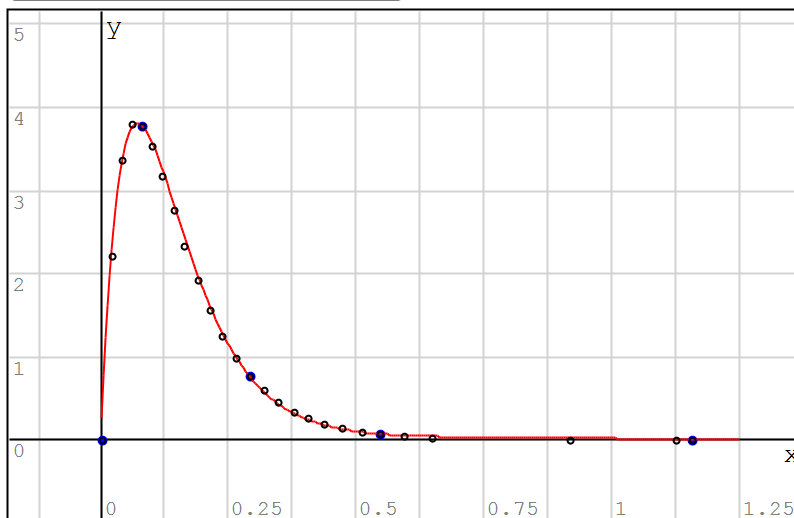
$$f(\alpha) = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

$$\begin{cases} X := \text{col}(XY, 1) \\ Y := \text{col}(XY, 2) \end{cases}$$

$$data := \text{plot}(XY, "o", 5, "black")$$

$$c(x) := \begin{cases} 1 & \text{if } (0 \leq x) \wedge (x \leq 1.25) \\ "" & \text{otherwise} \end{cases}$$

Fit "a la carte" from LMA



Support points

$$xy = \begin{bmatrix} 0 & 0 \\ 0.079 & 3.782 \\ 0.29 & 0.77 \\ 0.545 & 0.068 \\ 1.154 & 0 \end{bmatrix}$$

$$\begin{cases} data \\ \Phi(x, \alpha) \cdot c(x) \\ support \end{cases}$$

$XY := D3$

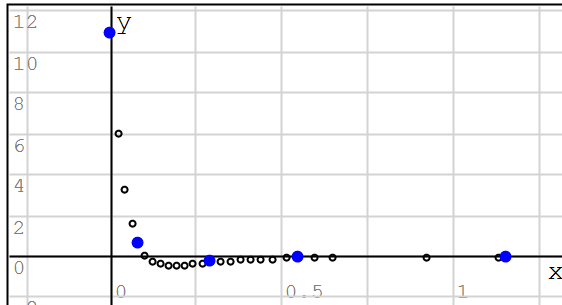
===== D3 data set =====

$\text{rows}(D3) = 27$

```
{ list := [ 1 5 14 22 27 ]
  xy := ListCollect (XY, list) }
```

1. select the support points
2. collect for LMA

```
raw := plot (XY, "o", 5, "black")
support := plot (xy, ".", 15, "blue")
```



```
{ support
  raw }
```

$\text{Clear}(x, \beta) = 1$

$$ft(x) := \beta_1 - \beta_2 \cdot \left(\exp\left(\frac{-x + \beta_3}{\beta_5}\right) - \exp\left(\left(-x + \beta_3\right) \cdot \frac{\beta_4 + \beta_5}{\beta_4 \cdot \beta_5}\right) \right)$$

model to D3(x)

$$xy = \begin{bmatrix} 0 & 10.861 \\ 0.079 & 0.706 \\ 0.29 & -0.226 \\ 0.545 & -0.031 \\ 1.154 & 0 \end{bmatrix}$$

$$\begin{cases} x := \text{col}(xy, 1) \\ y := \text{col}(xy, 2) \end{cases}$$

$$x = \begin{bmatrix} 0.000 \\ 0.07900 \\ 0.2900 \\ 0.5450 \\ 1.154 \end{bmatrix}$$

$$y = \begin{bmatrix} 10.86 \\ 0.7060 \\ -0.2260 \\ -0.03100 \\ 0.000 \end{bmatrix} \begin{bmatrix} 0 & 10.861 \\ 0.079 & 0.706 \\ 0.29 & -0.226 \\ 0.545 & -0.031 \\ 1.154 & 0 \end{bmatrix}$$

$\beta := [0 \ 1 \ 0.1 \ 0.075 \ 0.1]^T$

vector of guesses ...
from robust2crucifying !

$Tol := 10^{-16}$

$t0 := \text{time}(1)$

$LMSA(x, y, \beta, Tol, \beta_s) =$

Jean Solution

$\begin{bmatrix} 0.00071 \\ 1.0028 \\ 0.10647 \\ 0.06114 \\ 0.12774 \end{bmatrix}$

$\text{time}(1) - t0 = 0 \text{ s}$

$\beta := \beta_s$

$$\Phi(x, \beta) := \beta_1 - \beta_2 \cdot \left(\exp\left(\frac{-x + \beta_3}{\beta_5}\right) - \exp\left(\left(-x + \beta_3\right) \cdot \frac{\beta_4 + \beta_5}{\beta_4 \cdot \beta_5}\right) \right)$$

model to D3(x)

Implicit equalities = 0

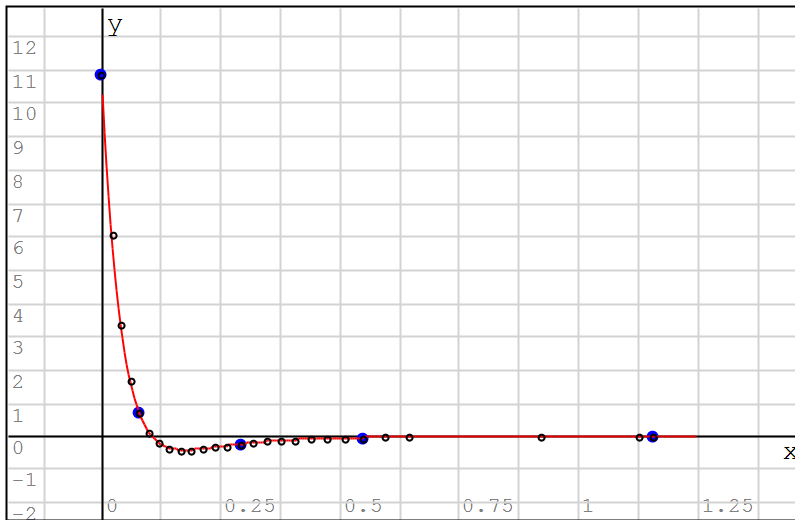
$$f(\beta) := \begin{bmatrix} \Phi(x_1, \beta) - y_1 \\ \Phi(x_2, \beta) - y_2 \\ \Phi(x_3, \beta) - y_3 \\ \Phi(x_4, \beta) - y_4 \\ \Phi(x_5, \beta) - y_5 \end{bmatrix}$$

$$\beta = \begin{bmatrix} 0.00024 \\ 1.03264 \\ 0.1063 \\ 0.06231 \\ 0.12544 \end{bmatrix}$$

$\begin{cases} [X := \text{col}(XY, 1) \ Y := \text{col}(XY, 2)] \\ \text{data} := \text{plot}(XY, "o", 5, "black") \end{cases}$

$$c(x) := \begin{cases} 1 & \text{if } (0 \leq x) \wedge (x \leq 1.25) \\ "" & \text{otherwise} \end{cases}$$

Fit "a la carte" from al_nleqsolve



```
{ data
  {  $\Phi(x, \beta) \cdot c(x)$ 
    support
```

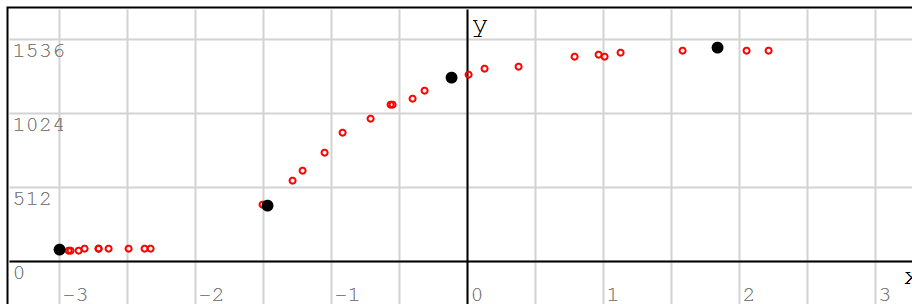
XY Thurber

Clear(x, β) = 1

```
ft(x) :=  $\beta_1 + \beta_2 \cdot \exp\left(-\exp\left(-\beta_3 \cdot (x - \beta_4)\right)\right)$  SGompertz Thurber model fnct
```

```
{ list := [ 1 13 24 34 ]
  { xy := ListCollect(XY, list)
    { 1. select the support points
      2. collect for LMA
```

```
raw := plot(XY, "o", 5, "red")
support := plot(xy, ".", 16, "black")
```



```
 $\beta := [80 \ 1400 \ 1 \ 0]^T$  vector of guesses ... from robust2crucifying !
```

 $Tol := 10^{-16}$

```
xy =  $\begin{bmatrix} -2.981 & 84.248 \\ -1.46 & 390.724 \\ -0.109 & 1260 \\ 1.841 & 1470 \end{bmatrix}$   $\begin{cases} x := \text{col}(xy, 1) \\ y := \text{col}(xy, 2) \end{cases}$ 
```

 $t0 := \text{time}(1)$

```
LMSA(x, y,  $\beta$ , Tol,  $\beta_s$ ) =  $\begin{bmatrix} 84.24796852 \\ 1396.05301219 \\ 1.61215217 \\ -1.20180484 \end{bmatrix}$   $\begin{bmatrix} 84.25 \\ 1396.05 \\ 1.61 \\ -1.2 \end{bmatrix}$ 
```

 $\text{time}(1) - t0 = 1 \text{ s}$

$$f(\beta) = \begin{bmatrix} 5.588 \cdot 10^{-12} \\ 6.377 \cdot 10^{-13} \\ 5.117 \cdot 10^{-11} \\ -1.627 \cdot 10^{-10} \\ 1.112 \cdot 10^{-10} \end{bmatrix} \quad \text{Jean} \quad \begin{bmatrix} 2.932 \cdot 10^{-6} \\ 2.063 \cdot 10^{-6} \\ 0.0001973 \\ -0.0006440 \\ 0.0004336 \end{bmatrix}$$

SGompertz...

is extremely **reflexive**, declared unfittable in Smath [L-M, CG]. Possible attempt OriginLab, Mathcad 11. As offered, and fitted by al_nleqsolve, at this point, you may want to round/refactor the coefficients. The fillet region is a bit sparse !

$$\Phi(x, m) := m_1 + m_2 \cdot \exp\left(-\exp\left(-m_3 \cdot (x - m_4)\right)\right)$$

SGompertz Thurber model fnct

Implicit equalities = 0

$$f(\mu) := \begin{bmatrix} \Phi(x_1, \mu) - y_1 \\ \Phi(x_2, \mu) - y_2 \\ \Phi(x_3, \mu) - y_3 \\ \Phi(x_4, \mu) - y_4 \end{bmatrix}$$

$$\mu := \beta s = \begin{bmatrix} 84.25 \\ 1396.05 \\ 1.61 \\ -1.2 \end{bmatrix}$$

Notes...Smath 7251

1. solves a system of [n equalities = n parameters].
2. The order of equalities is immaterial.
3. Guess wisely the best support equalities !

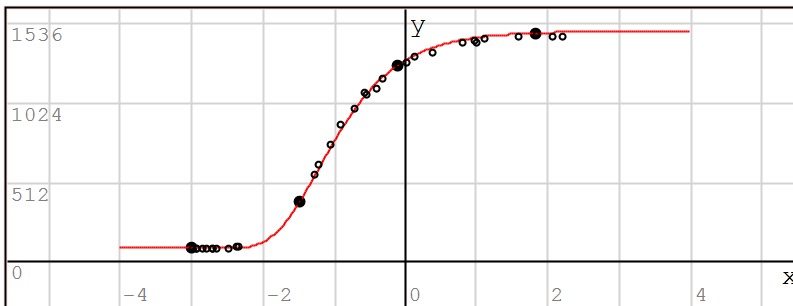
$$f(\mu) = \begin{bmatrix} 0.000000000051 \\ 0 \\ -0.000000000001 \\ -0.000000000031 \end{bmatrix} \quad \begin{bmatrix} 0.000000000000 \\ -0.000000000003 \\ 0.000000000000 \\ 0.000000000002 \end{bmatrix}$$

$$\begin{cases} X := \text{col}(XY, 1) & Y := \text{col}(XY, 2) \\ data := \text{plot}(XY, "o", 5, "black") \end{cases}$$

$$c(x) := \begin{cases} 1 & \text{if } (-4 \leq x) \wedge (x \leq 4) \\ "" & \text{otherwise} \end{cases}$$

$$\mu := \begin{bmatrix} 84.25 \\ 1396 \\ 1.61 \\ -1.2 \end{bmatrix}$$

$$j := 1$$



Support points

$$xy = \begin{bmatrix} -2.981 & 84.248 \\ -1.46 & 390.724 \\ -0.109 & 1260 \\ 1.841 & 1470 \end{bmatrix}$$

$$\begin{cases} data \\ \Phi(x, \mu_{1j}) \cdot c(x) \\ support \end{cases}$$