

Gamma(x) Lanczos overlord

X($\Phi, \alpha, \beta, \varepsilon$) RootSecant algo

$z := 2.5 \cdot i$

$\Gamma(z) = 0.0161290540478 - 0.0267491089459 \cdot i$

$a := 0.5$

$\Phi(x) := \text{eval}(\gamma(a; x) - \Gamma(a; x))$

$\alpha := 0.2 \quad \beta := 0.5 \quad \varepsilon := 10^{-15}$

$\text{sol} := X(\Phi; \alpha; \beta; \varepsilon)$

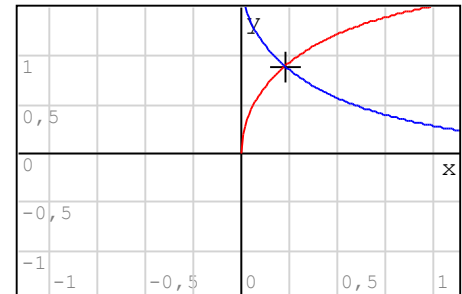
$\text{sol} = \begin{bmatrix} 1 \cdot 10^{-15} \\ 0.238739753660938 \\ 0.233859672474681 \end{bmatrix}$

Explode the nested 'sol'

$R := \text{row}(\text{sol}; \text{rows}(\text{sol}))_1$

$[R \ \Gamma(a; R)] = [0.234 \ 0.876]$

$\text{appVersion}(4) = "0.99.7822.147"$
Applications = " 8 s "



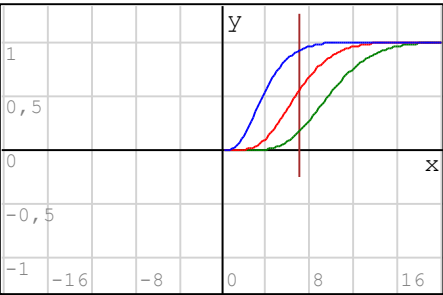
Example_1: Multistage XFR

`t0 := time(1)`

`T := 1` System TimeConstant `s := 1` Laplace UnitStep

$\lambda(x) := \begin{cases} 1 & \text{if } (0 \leq x) \wedge (x \leq 20) \\ "" & \text{otherwise} \end{cases}$ `time := 7`

$$f(n; t) := \frac{\Gamma(n+1) - \Gamma\left(n+1; \frac{t}{T}\right)}{s \cdot \Gamma(n+1)}$$
 Laplace XFR multistage system
'n' stages, 'T' time constant



$$\begin{bmatrix} f(3; time) \\ f(6; time) \\ f(9; time) \end{bmatrix} = \begin{bmatrix} 0.92 \\ 0.55 \\ 0.17 \end{bmatrix}$$

```

XY :=
0 127.42
0.1 123.469
0.2 118.008
0.3 111.187
0.4 102.811
0.5 93.554
0.6 84.168
0.7 74.818
0.8 66.373
0.9 58.587
1 51.678
1.1 45.739
1.2 40.675
1.3 36.404
1.4 32.74
1.5 29.581
1.6 26.848
1.7 24.472
1.8 22.395
1.9 20.573
2 18.97
2.1 17.555
2.2 16.3
2.3 15.184
2.4 14.188
2.5 13.295
2.6 12.492
2.7 11.768
2.8 11.12
2.9 10.547
3 10.028

```

Example_2: Fitting Mathsoft data set.

```

plot(data; char; size; clr) := for k ∈ [1..rows(data)]
    [ r3_k := char r4_k := size r5_k := clr
    augment(data; r3; r4; r5)

```

```
pts := plot(XY; "o"; 5; "black")
```

```
[yo := 10.028 a := 143 b := 1.77 η := 2.5]
```

$\varepsilon := 10^{-15}$

```
f(x) := yo + a · Γ(b; x)η
```

```

Ω := { u := [ε; 0.05..3]
      for i ∈ [1..rows(u)]
        vy_i := eval(f(u_i))
      augment(u; vy)
}

```

```

X := [ε; 0.05..3]
for i ∈ [1..rows(X)]
  Y_i := eval(f(X_i))
S(x) := cinterp(X; Y; x)

```

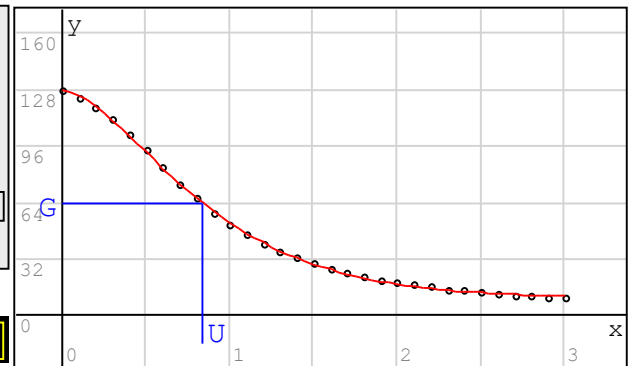
```
G := 64 ... 'G' ▼ ordinate solve abscissa
```

```
U := solve(S(x) - G = 0; x; ε; 3) = 0.836
```

```

plot := { Ω
        pts
        [-0.175 G + 7 "G" 12 "blue"]
        [U 0 "U" 12 "blue"]
}

```



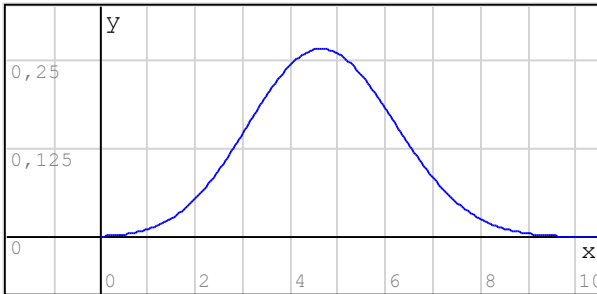
$[G \ U] = [64 \ 0.836]$

Example_3: Hypergeometric statistical function.**Hypergeom ...** $\Gamma(x) := \text{Gamma}(x)$ no need for this $\Gamma(x) := \text{Lanczos}(x)$ this neither

$$\text{hypergeom}(N; n; p; x) := \frac{\Gamma(N+1) \cdot \Gamma(n+1) \cdot \Gamma(N+n-p+1) \cdot \Gamma(p+1)}{\Gamma(x+1) \cdot \Gamma(-x+n+1) \cdot \Gamma(N+n+1) \cdot \Gamma(x+N-p+1) \cdot \Gamma(-x+p+1)}$$

 $N := 39 \quad n := 12 \quad p := 20$
 $\text{hypergeom}(x) := \text{hypergeom}(N; n; p; x)$

Continuous PDF Hypergeometric



```

if 0 ≤ x
  hypergeom(x)
else
  "outside lower bound"

```

In probability theory and statistics, the hypergeometric distribution is a discrete probability distribution that describes the probability of 'k' successes in 'n' draws, without replacement, from a finite population of size 'N' that contains exactly 'K' successes, wherein each draw is either a success or a failure. In contrast, the binomial distribution describes the probability of 'k' successes in 'n' draws with replacement.

 $\text{time}(1) - t0 = 8 \text{ s}$
RemToDo ▲ code recursive