# « Julia, my new friend for computing and optimization? »

- Intro to the Julia programming language, for MATLAB users
- *Date:* 14th of June 2018
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## Agenda for today [25 min]

- 1. What is Julia [3 min]
- 2. Comparison with MATLAB [3 min]
- 3. Examples of problems solved Julia [5 min]
- 4. Longer example on optimization with JuMP [10min]
- 5. Links for more information? [2 min]

## 1. What is Julia?

- Developed and popular from the last 7 years
- Open-source and free programming language (MIT license)
- Interpreted *and* compiled, very efficient
- But easy syntax, dynamic typing, inline documentation etc
- Multi-platform, imperative
- MATLAB-like syntax for linear algebra etc
- Designed and acknowledged as simple to learn and use
- Easy to run your code in parallel (multi-core & cluster)
- Used worldwide: research, data science, finance etc...

#### Ressources

- Website: JuliaLang.org for the language & Pkg.JuliaLang.org for packages
- Documentation : docs.JuliaLang.org



## Comparison with MATLAB

	Julia 😃	MATLAB 😢
Cost	Free 🤞	Hundreds of euros / year
License	Open-source	1 year user license (no longer after your PhD!)
Comes from	A non-profit foundation, and the community	MathWorks company
Scope	Mainly numeric	Numeric only
Performances	Very good performance	Faster than Python, slower than Julia

## Comparison with MATLAB

	Julia	MATLAB
Packaging	Pkg manager included.  Based on git + GitHub,  very easy to use	Toolboxes already included but \$\square\$ have to pay if you wat more!
Editor/IDE	Jupyter is recommended (Juno is also good)	Good IDE already included
Parallel computations	Very easy, low overhead cost	Possible, high overhead

## Comparison with MATLAB

	Julia	MATLAB
Usage	Generic, worldwide 💲	Research in academia and industry
Fame	Young but starts to be known	Old and known, in decline
Support?	Community <sup>1</sup> (StackOverflow, mailing lists etc).	By MathWorks
Documentation	OK and growing, inline/online	OK, inline/online

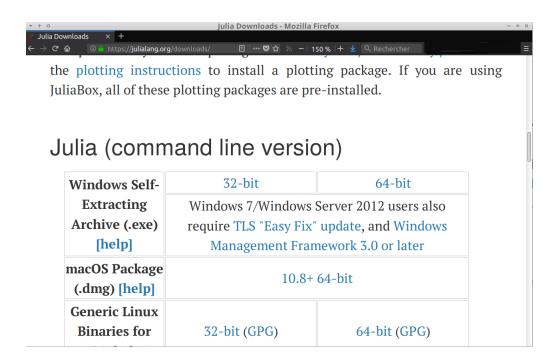
Note<sup>1</sup>: **JuliaPro** offer paid licenses, if professional support is needed.

## How to install Julia (1/2)

- You can try online for free on JuliaBox.com
- On Linux, Mac OS or Windows:
  - You can use the default installer from the website julialang.org/downloads
- Takes about 4 minutes... and it's free!

You also need Python 3 to use Jupyter \*\*, I suggest to use Anaconda.com/download if you don't have Python yet.

## How to install Julia (2/2)



- 1. Select the binary of your platform 📦
- 2. Run the binary 🚶!
- 3. Wait 0...
- 4. Done 🤞! Test with julia in a terminal

## Different tools to use Julia

• Use julia for the command line for short experiments

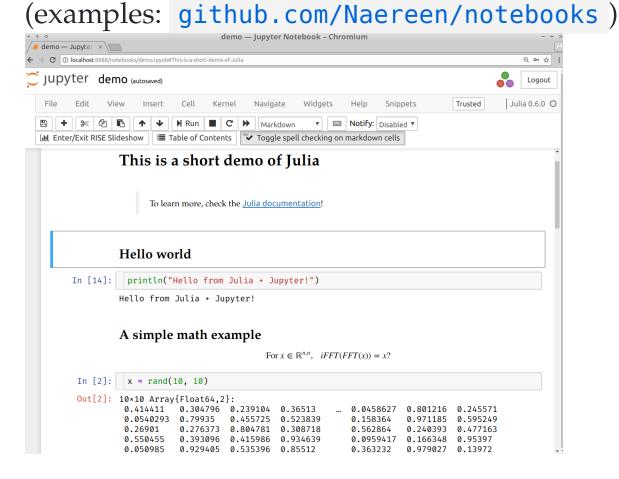
```
(lun. juin 11 -- 03:06:27)lilian@jarvis:[~]> {bashv4.4} — Konsole
$ julia
                             A fresh approach to technical computing
                             Documentation: https://docs.julialang.org
                              Type "?help" for help.
                             Version 0.6.0 (2017-06-19 13:05 UTC)
                             Official http://julialang.org/ release
                             x86_64-pc-linux-gnu
julia> println("Hello world from Julia!")
Hello world from Julia!
julia>
```

Use the *Juno* IDE to edit large projects

Demo time 💆 !

## Different tools to use Julia

• Use **Jupyter** notebooks to write or share your experiments





### We How to install modules in Julia?

• Installing is **easy**!

```
julia> Pkd.add("IJulia") # installs IJulia
```

Updating also!

```
julia> Pkg.update()
```

## How to find the module you need?

- First... ask your colleagues 😂!
- Complete list on pkg.JuliaLang.org

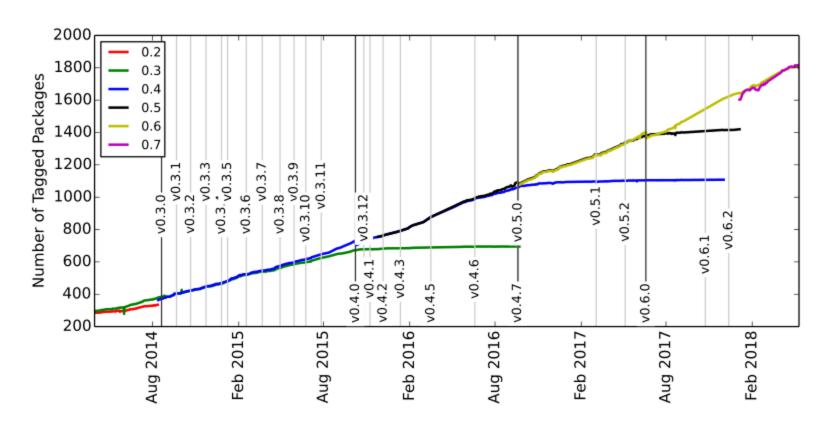


## Overview of famous Julia modules

- Plotting:
  - Winston.jl for easy plotting like MATLAB
  - PyPlot: interface to Matplotlib (Python)
- The JuliaDiffEq collection for differential equations
- The JuliaOpt collection for **optimization**
- The JuliaStats collection for **statistics**
- And many more!

Find more specific packages on GitHub.com/svaksha/Julia.jl/

## Many packages, and a quickly growing community



Julia is still in development, in version v0.6 but version 1.0 is planned soon!

## 2. Main differences in syntax between Julia and MATLAB

Ref: cheatsheets.quantecon.org

	Julia	MATLAB
File ext.	.jl	. m
Comment	# blabla	% blabla
Indexing	a[1] to a[end]	a(1) to a(end)
Slicing	a[1:100] (view)	a(1:100) (! copy)
Operations	Linear algebra by default	Linear algebra by default
Block	Use end to close all blocks	Use endif endfor etc

	Julia	MATLAB
Help	?func	help func
And	a & b	a && b
Or	a   b	a    b
Datatype	Array of any type	multi-dim doubles array
Array	[1 2; 3 4]	[1 2; 3 4]
Size	size(a)	size(a)
Nb Dim	ndims(a)	ndims(a)
Last	a[end]	a(end)

	Julia	MATLAB
Tranpose	a. '	a.'
Conj. transpose	a'	a'
<b>Matrix</b> x	a * b	a * b
<b>Element-wise</b> x	a .* b	a .* b
Element-wise /	a ./ b	a ./ b
Element-wise ^	a ^ 3	a .^ 3
Zeros	zeros(2, 3, 5)	zeros(2, 3, 5)
Ones	ones(2, 3, 5)	ones(2, 3, 5)
Identity	eye(10)	eye(10)
Range	range(0, 100, 2) or 1:2:100	1:2:100

	Julia	MATLAB
Maximum	max(a)	max(max(a)) ?
Random matrix	rand(3, 4)	rand(3, 4)
L2 Norm	norm(v)	norm(v)
Inverse	inv(a)	inv(a)
Solve syst.	a \ b	a \ b
Eigen vals	V, $D = eig(a)$	[V,D]=eig(a)
FFT/IFFT	fft(a), ifft(a)	fft(a), ifft(a)

Very close to MATLAB for linear algebra!

## 3. Scientific problems solved with Julia

Just to give examples of syntax and modules

- 1. 1D numerical integration and plot
- 2. Solving a  $2^{\mathrm{nd}}$  order Ordinary Differential Equation

## 3.1. 1D numerical integration and plot

Exercise: evaluate and plot this function on [-1, 1]:

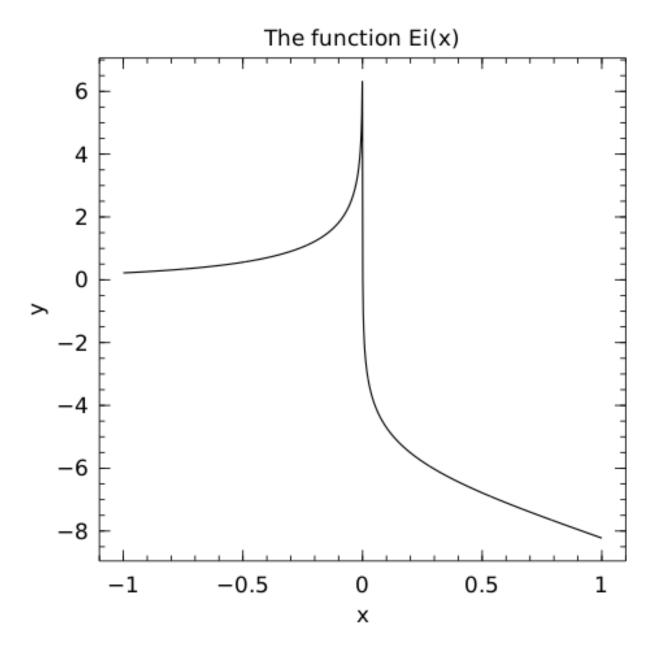
$$\mathrm{Ei}(x) := \int_{-x}^{\infty} \frac{\mathrm{e}^u}{u} \, \mathrm{d}u$$

#### How to?

Use packages and everything is easy!

- QuadGK.jl for integration
- Winston.jl for 2D plotting

```
using QuadGK
function Ei(x, minfloat=1e-3, maxfloat=100)
    f = t -> exp(-t) / t # inline function, with '- >'
    if x > 0
        return quadgk(f, -x, -minfloat)[1]
             + quadgk(f, minfloat, maxfloat)[1]
    else
        return quadgk(f, -x, maxfloat)[1]
    end
end
X = linspace(-1, 1, 1000) # 1000 points
Y = [Ei(x) for x in X]
using Winston
plot(X, Y)
title("The function Ei(x)")
xlabel("x"); ylabel("y")
savefig("figures/Ei integral.png")
```



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## 3.2. Solving a $2^{nd}$ order ODE

Goal: solve and plot the differential equation of a pendulum:

$$\theta''(t) + b\theta'(t) + c\sin(\theta(t)) = 0$$

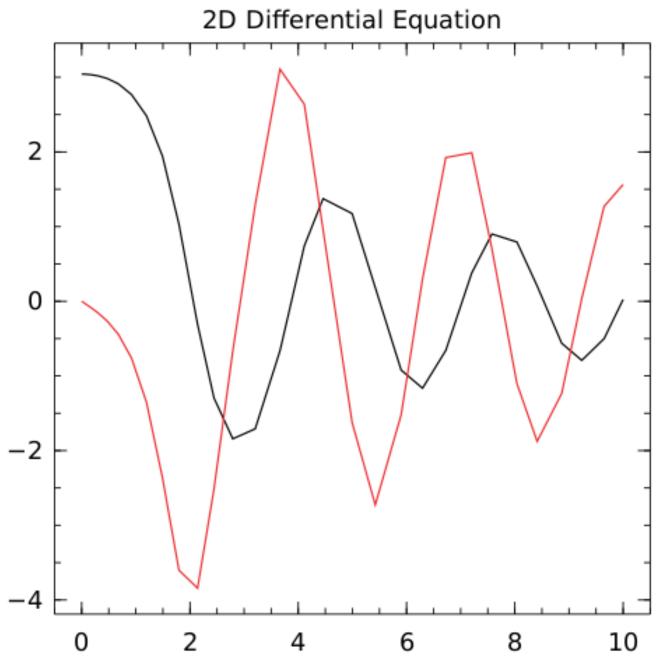
For 
$$b=1/4$$
,  $c=5$ ,  $\theta(0)=\pi-0.1$ ,  $\theta'(0)=0$ ,  $t\in[0,10]$ 

#### How to?

Use packages!

- Differential Equations.jl function for ODE integration
- Winston.jl for 2D plotting

```
using DifferentialEquations
b, c = 0.25, 5.0
# macro magic!
pend2 = @ode def Pendulum begin
  d\theta = \omega \# < -- yes, this is UTF8
  d\omega = (-b * \omega) - (c * \sin(\theta))
end
prob = ODEProblem(pend, y0, (0.0, 10.0))
sol = solve(prob) # \(\preceq solve on interval [0,10]\)
t, y = sol.t, hcat(sol.u...)'
using Winston
plot(t, y[:, 1], t, y[:, 2])
title("2D Differential Equation")
savefig("figures/Pendulum solution.png")
```



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### Conclusion (1/2)

#### Sum-up

- I hope you got a good introduction to Julia
- It's not hard to migrate from MATLAB to Julia
- Good start:

docs.JuliaLang.org/en/stable/manual/getting-started

### Conclusion (2/2)

Thanks for joining 🤎!

#### Your mission, if you accept it... 🛪

- 1. Padawan level: Train yourself a little bit on Julia
  - $\hookrightarrow$  JuliaBox.com? Or install it on your laptop!

And ead introduction in the Julia manual!

- 2. **I** *Jedi level:* Try to solve a numerical system, from your research or teaching, in Julia instead of MATLAB
- 3. *Master level:* From now on, try to use open-source & free tools for your research (Julia, Python and others)... 😽

## Examples

- 1. **Iterative computation**: signal filtering
- 2. Optimization: robust regression on RADAR data

## Iterative computation

The classical saying:

"Vectorized code often runs much faster than the corresponding code containing loops." (MATLAB doc)

does not hold for Julia, because of its **just-in-time compiler**.

Example for the smoothing of a signal  $\{u_k\}_{k\in\mathbb{N}}$ :

$$y_k = ay_{k-1} + (1-a)u_k, \quad k \in \mathbb{N}^+$$

Parameter a tunes the smoothing, between none (a=0) and strong smoothing ( $a \to 1^-$ ).

## **Optimization problem**

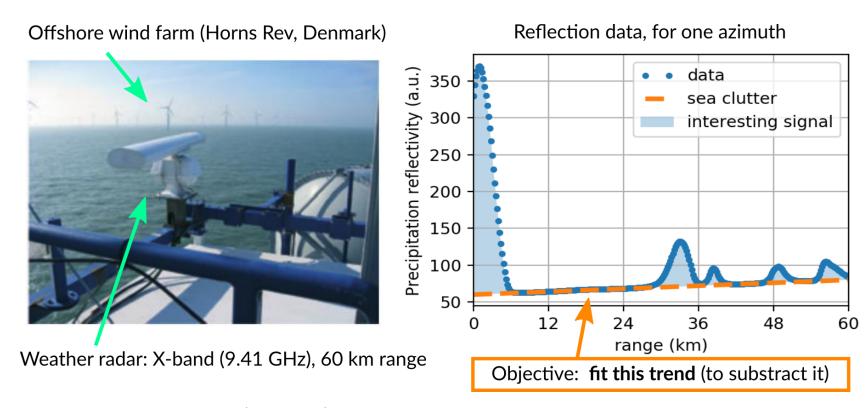
Example problem: Identifying the sea clutter in Weather Radar data.

- is a **robust regression** problem
  - is an optimization problem

An "IETR-colored" example, inspired by:

- Radar data: P.-J. Trombe et al., "Weather radars the new eyes for offshore wind farms?," Wind Energy, 2014.
- Regression methods: S. Boyd and L. Vandenberghe, Convex Optimization. Cambridge University Press, 2004. (Example 6.2)

## Weather radar: the problem of sea clutter



Given n data points  $(x_i, y_i)$ , fit a linear trend:

$$\hat{y} = a.x + b$$

An optimization problem with two parameters: a (slope), b (intercept)

## Regression as an optimization problem

The parameters for the trend (a,b) should minimize a criterion J which penalizes the residuals  $r_i = y_i - \hat{y} = y_i - a.x + b$ :

$$J(a,b) = \sum_i \phi(r_i)$$

where  $\phi$  is the *penaly function*, to be chosen:

- $\phi(r) = r^2$ : quadratic deviation  $\rightarrow$  least squares regression
- $\phi(r) = |r|$ : absolute value deviation
- $\phi(r) = h(r)$ : Huber loss

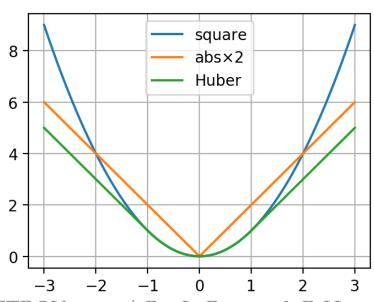
## Choice of penalty function

The choice of the loss function influences:

- the optimization result (fit quality)
  - e.g. in the presence of outliers
- the properties of optimization problem: convexity, smoothness

#### Properties of each function

- quadratic: convex, smooth, heavy weight for strong deviations
- absolute value: convex, not smooth
- Huber: a mix of the two



## How to solve the regression problem

#### **Option 1: specific tools**

a specific tool for each type of regression:

- "least square toolbox" (→ MultivariateStats.jl)
- "least absolute value toolbox" (→ quantile regression)
- "Huber toolbox" (i.e. robust regression  $\rightarrow$  ???)

#### Option 2: a generic tool

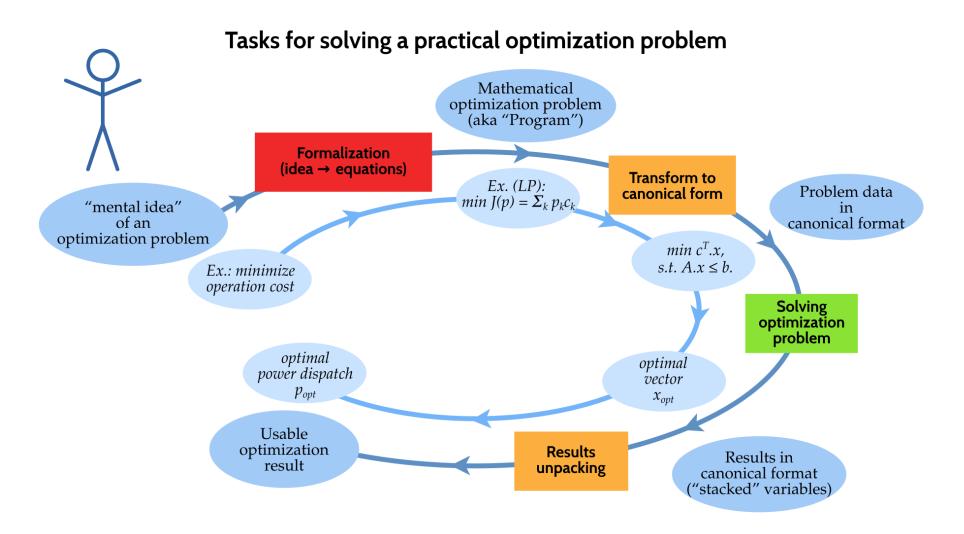
→ a Modeling Language for Optimization

+more freedom to explore variants of the problem

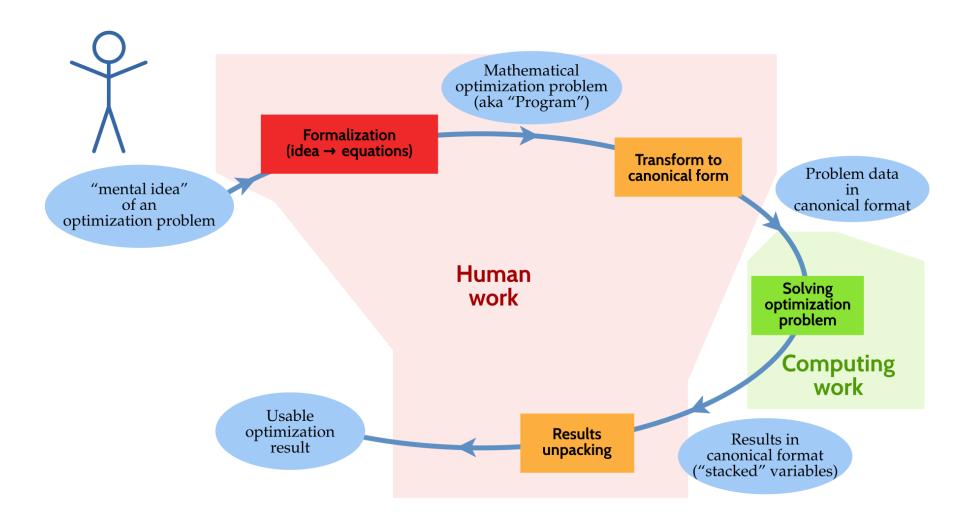
## Modeling Languages for Optimization

Purpose: make it easy to **specify** and **solve** optimization problems without expert knowledge.

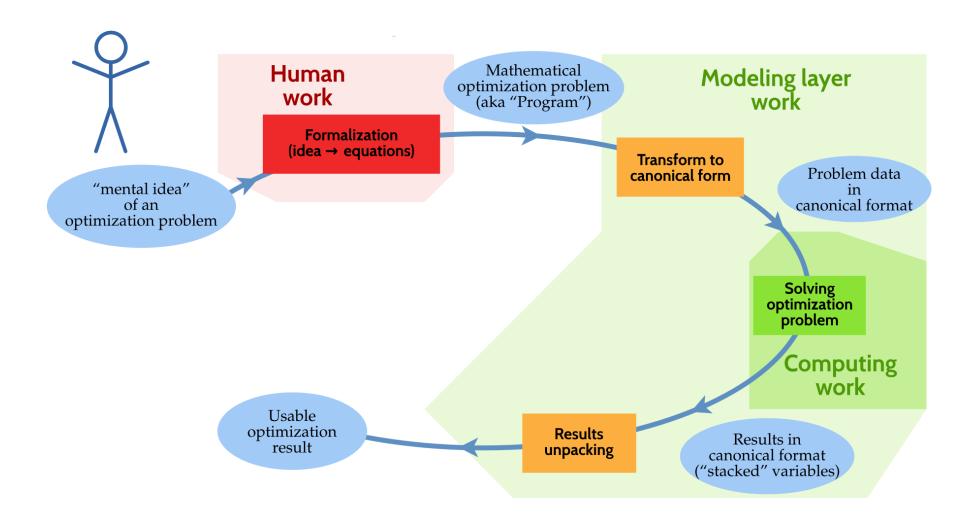
#### Tasks for solving a practical optimization problem



#### Tasks for solving a practical optimization problem



#### Tasks for solving a practical optimization problem



## JuMP: optimization modeling in Julia

The JuMP package offers a domain-specific modeling language for mathematical optimization.

JuMP interfaces to many optimization solvers: open-source (Ipopt, GLPK, Clp, ECOS...) and commercial (CPLEX, Gurobi, MOSEK...).

Other Modeling Languages for Optimization:

- Standalone software: AMPL, GAMS
- Matlab: YALMIP (previous seminar), CVX
- Python: Pyomo, PuLP, CVXPy

Claim: JuMP is **fast**, thanks to Julia's metaprogramming capabilities (generation of Julia code within Julia code).

## Regression with JuMP

Given x and y the 300 data points, common part:

```
m = Model(solver = ECOSSolver())
@variable(m, a)
@variable(m, b)
res = a*x .- y +b
```

res ("residuals") is an Array of 300 elements of type JuMP.GenericAffExpr{Float64, JuMP.Variable}, i.e. a semi-symbolic affine expression.

Now, we need to specify the penalty on those residuals

## Regression choice: least squares

$$\min \sum_i r_i^2$$

reformulated as a Second-Order Cone Program (SOCP):

 $\min j$ , such that  $||r||_2 \leq j$ 

```
@variable(m, j)
@constraint(m, norm(res) <= j);</pre>
@objective(m, Min, j)
```

 $(SOCP \rightarrow ECOS solver)$ 

## Regression choice: least absolute deviation

$$\min \sum_i |r_i|$$

reformulated as a Linear Program (LP)

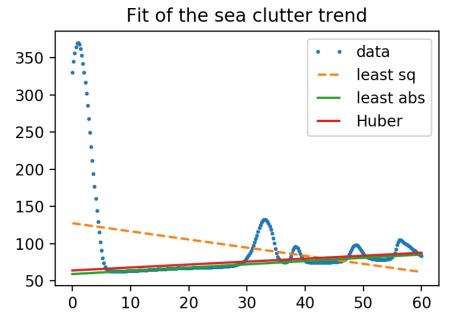
$$\min \sum_i t_i, \quad ext{such that } -t_i \leq r_i \leq t_i$$

```
@variable(m, t[1:n] )
@constraint(m, res .<= t)</pre>
@constraint(m, res .>= -t)
@objective(m, Min, sum(t));
```

## Solve! ♥<

```
julia> solve(m)
[solver blabla... ☒ ]
:Optimal # hopefully
```

```
> getvalue(a), getvalue(b)
(-1.094, 127.52) # for least squares
```



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