

# KL-Regularized Least Squares

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```
Precompiling KLLS
✓ ArrayInterface
✓ OffsetArrays
✓ OffsetArrays → OffsetArraysAdaptExt
✓ ArrayInterface → ArrayInterfaceStaticArraysCoreExt
✓ ArrayInterface → ArrayInterfaceReverseDiffExt
✓ StaticArrayInterface
✓ StaticArrayInterface → StaticArrayInterfaceStaticArraysExt
✓ StaticArrayInterface → StaticArrayInterfaceOffsetArraysExt
✓ CloseOpenIntervals
✓ LayoutPointers
✓ VectorizationBase
✓ SLEEFPirates
✓ LoopVectorization
✓ LoopVectorization → SpecialFunctionsExt
✓ LoopVectorization → ForwardDiffExt
✓ Octavian
    Info Given KLLS was explicitly requested, output will be shown live
WARNING: using NLPModels.grad in module KLLS conflicts with an existing
identifier.
WARNING: using NLPModels.hess in module KLLS conflicts with an existing
identifier.
✓ KLLS
✓ Octavian → ForwardDiffExt
18 dependencies successfully precompiled in 21 seconds. 202 already precompiled.
1 dependency had output during precompilation:
└─ KLLS
  │ [Output was shown above]
  └─
```

## Load physics data

The file `PhysicsData.npz` contains the matrix and vectors generated by the script `PhysicsData.py`, where we get the output from `generate_data`:

- `A` is the first output argument `A`
- `b` is the second output argument `data0`

- $x_0$  is the third input argument, which seems to be the ground truth distribution.

```
data = npzread("PhysicsData.npz", ["A", "b", "x0"])
@unpack A, b, x0 = data
klprob = KLLSData(A, b)
```

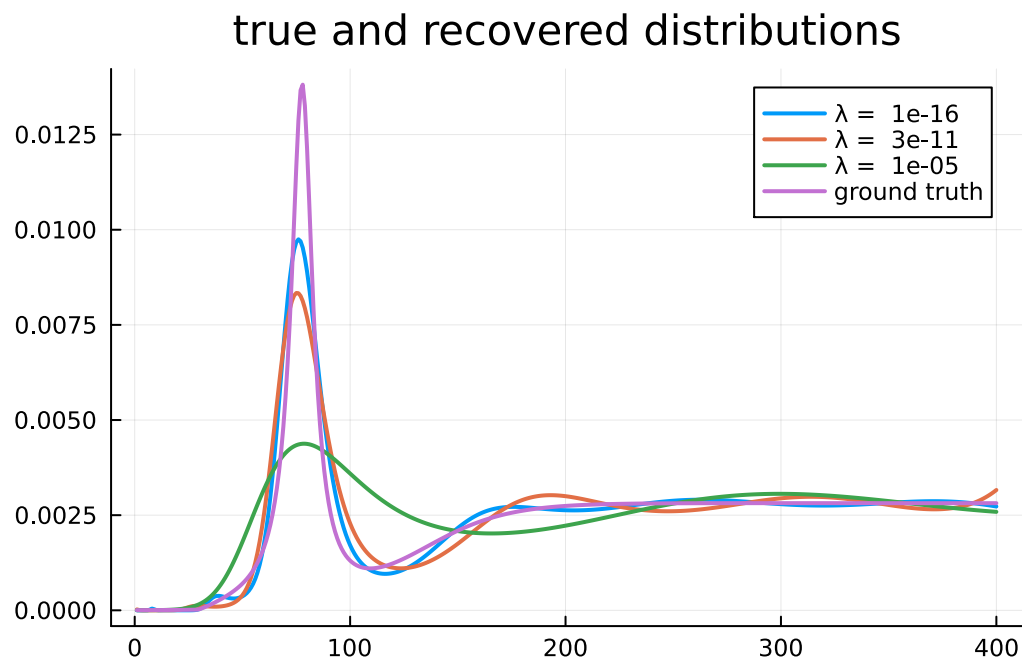
## Solve over a range of regularization parameters

Solve the problem for a range of logarithmically spaced regularization parameters  $\lambda$  between  $10^{-8}$  and 1.

```
stats = map(exp10.(range(-16, stop=-5, length=3))) do  $\lambda$ 
    klprob. $\lambda$  =  $\lambda$ 
    p, y, stats = newtoncg(klprob)
    ( $\lambda$ = $\lambda$ , p=p, iters=stats.iter,  $\nabla$ dNrm=stats.dual_feas)
end;
```

Plot the recovered distributions for each value of  $\lambda$  and overlay the ground truth distribution  $x_0$ .

```
lab = hcat([@sprintf(" $\lambda$  = %6.0e",  $\lambda$ ) for  $\lambda$  in getfield.(stats, : $\lambda$ )]...)
default(lw=2, title="true and recovered distributions")
plot(getfield.(stats, :p), label=lab)
plot!(x0, label="ground truth")
```



The curve corresponding to the smallest parameter  $\lambda$  ( $1e-8$ ) best approximates the modes of the ground-truth distribution, but smaller values of  $\lambda$  don't help.

These tests use a uniform prior. Does the data generator make a prior available? If so, this could be used to improve the results.

## Accuracy of the solution

Now try to solve the problem as accurately as possible. Set  $\lambda = 0$  and the tightest tolerance reasonable.

```
klprob. $\lambda$  = 0.0  
p, y, stats = newtoncg(klprob, atol=1e-12, rtol=1e-12, verbose=0)
```

```
m, n = size(A)  
@printf("%20s: %11.2e\n", "rms(p-x0)", norm(p - x0)/ $\sqrt{n}$ )  
@printf("%20s: %11.2e\n", "rms(Ap-b)", norm(A*p - b)/ $\sqrt{m}$ )  
@printf("%20s: %11f\n", "Solve time", stats.elapsed_time)  
@printf("%20s: %11d\n", "Number of iterations", stats.iter)  
plot([p x0], lab=["p" "x0"], title="true vs recovered distributions")
```

```
      rms(p-x0):      3.35e-04  
      rms(Ap-b):      2.52e-10  
      Solve time:    30.000023  
Number of iterations:    196578
```

