

```
In [ ]: using Pkg
Pkg.add https://github.com/baggepinnen/Hyperopt.jl.git
Pkg.add https://github.com/mohamed82008/Nonconvex.jl.git
Pkg.add https://github.com/SciML/GalacticOptim.jl.git
```

```
In [33]: using Hyperopt, GalacticOptim
```

The function optimized should be calling another optimizer from GalacticOptim.

In the introduction of GalacticOptim, I find many "Local Gradient-Based Optimization", "Local Derivative-Free Optimization", "Local

Hessian-Based Second Order Optimization" "Local Hessian-Free Second Order Optimization", "Global Unconstrained Optimizers" methods.

And I also find a function in Hyperopt named "Hyperband" method can be used to call the local or global optimizer from GalacticOptim.

For example, we want to optimize the function f , and we can use `SimulatedAnnealing()` as our local optimizing function.

We adopt one local optimizer `ParticleSwarm()` as an example:

```
In [51]: using Optim
f(a;c=10) = sum(@. 100 + (a-3)^2 + (c-100)^2)
hohb = @hyperopt for i=100, sampler=Hyperband(R=50, η=3, inner=RandomSampler()), a = LinRange(0, 100, 100)
    if !(state === nothing)
        a, c = state
    end
    res = Optim.optimize(x->f(x[1], c=x[2]), [a, c], ParticleSwarm(), Optim.Options(f_calls_limit=1000))
    @show Optim.minimum(res), Optim.minimizer(res)
end

(Optim.minimum(res), Optim.minimizer(res)) = (3302.942420444705, [6.828991009511888, 43.535132619532504])
(Optim.minimum(res), Optim.minimizer(res)) = (5584.67295627623, [8.957450564377947, 26.18142686932223])
(Optim.minimum(res), Optim.minimizer(res)) = (6414.154374818459, [9.789082960369441, 20.828902196725352])
(Optim.minimum(res), Optim.minimizer(res)) = (2891.8375164503564, [9.549280406424725, 47.5696229022871])
(Optim.minimum(res), Optim.minimizer(res)) = (7446.940991777658, [6.640193600833616, 14.36303378606813])
(Optim.minimum(res), Optim.minimizer(res)) = (8000.911258703172, [3.908191906298828, 11.117569530505607])
(Optim.minimum(res), Optim.minimizer(res)) = (101.39011073103032, [3.881815320112066, 99.21736823968997])
(Optim.minimum(res), Optim.minimizer(res)) = (150.25254230465737, [3.999731122526228, 92.9819461396118])
(Optim.minimum(res), Optim.minimizer(res)) = (150.03263088697582, [2.6813397378587753, 92.9338067869391])
(Optim.minimum(res), Optim.minimizer(res)) = (100.61643567052181, [3.062635439455999, 99.21736823968997])
```

We can also call and optimize multiple optimizer to optimize one function and we use Hyperopt to optimize multiple local optimizer.

For example, in the code:

algorithm = [LBFGS(), SimulatedAnnealing(), ParticleSwarm(), NelderMead(), BFGS(),
NewtonTrustRegion()], we use three of them.

```
In [53]: hohb = @hyperopt for i=10, sampler=Hyperband(R=50, η=3, inner=RandomSampler()),
        algorithm = [LBFGS(), SimulatedAnnealing(), ParticleSwarm()],
        a = LinRange(1, 5, 1800),
        c = exp10. (LinRange(-1, 3, 1800))
        if !(state === nothing)
            x0, algorithm = state
        else
            x0 = [a, c]
        end
        println(i, " algorithm: ", typeof(algorithm).name.name)
        res = Optim.optimize(x->f(x[1], c=x[2]), x0, algorithm, Optim.Options(time_limit=i+1, show
        Optim.minimum(res), (Optim.minimizer(res), algorithm)
    end

5353314332 ... 954.9681399564245, 959.8698272584649, 964.7966740788197, 969.74880955697
17, 974.7263634952538, 979.7294663622529, 984.7582492962279, 989.8128441085481, 994.8933
83287148, 1000.0]), Any[([3.0000000520148795, 99.99999997527875], ParticleSwarm{Any} (Any
[], Any[], 0)), ([3.000000000010182, 100.00000000000081], LBFGS{Nothing, LineSearches.In
itialStatic{Float64}, LineSearches.HagerZhang{Float64, Base.RefValue{Bool}}}, Optim.var"#
17#19" (10, LineSearches.InitialStatic{Float64}
alpha: Float64 1.0
scaled: Bool false
, LineSearches.HagerZhang{Float64, Base.RefValue{Bool}}
delta: Float64 0.1
sigma: Float64 0.9
alphamax: Float64 Inf
rho: Float64 5.0
epsilon: Float64 1.0e-6
gamma: Float64 0.66
linesearchmax: Int64 50
psi3: Float64 0.1
display: Int64 0
mayterminate: Base.RefValue{Bool}
nothing Optim.var"#17#19" () Flat() true)) ([3.0066203040385835 100.0268464060837
```