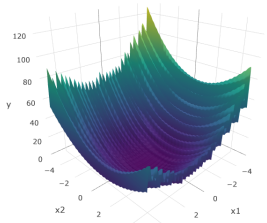


Optimization in Machine Learning

Multi-Start Optimization



Learning goals

- Multimodal functions
- Basins of Attractions
- Simple multi-start procedure

MOTIVATION

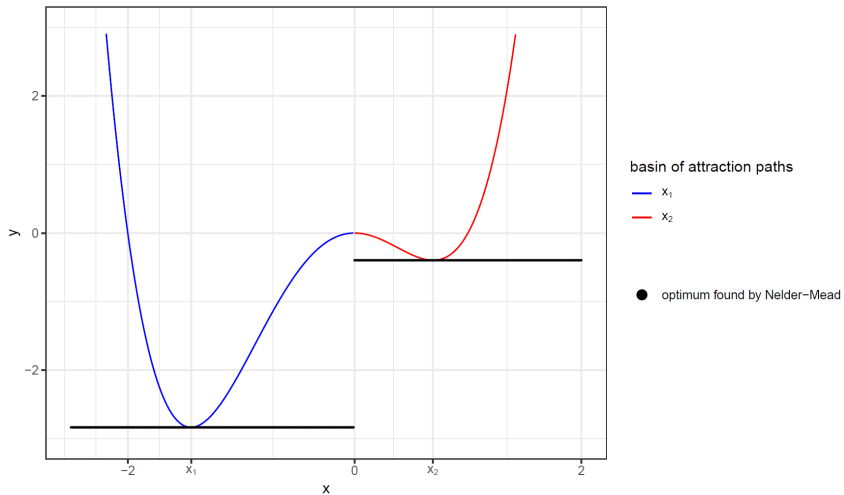
- So far: optimization method for unimodal objective function (exception: simulated annealing)
- If the objective function is actually multimodal, our procedures converge to **local minima**.
- Found optima may be different for different starting values $\mathbf{x}^{[0]}$

⇒ “**attraction areas**”

Let f_1^*, \dots, f_k^* be local minimum values of $f(\cdot)$ with $f_i^* \neq f_j^* \quad \forall i \neq j$.
Then the set \mathcal{A} is called basin of attraction of f_i^* for algorithm A .

$$\mathcal{A}(f_i^*, A) = \{\mathbf{x} \in S \subseteq \mathbb{R} : f(A(\mathbf{x})) \rightarrow f_i^*\}$$

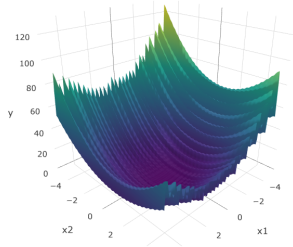
ATTRACTION AREAS



MULTI-STARTS

$$f(\mathbf{x}) = \sin^2(3\pi x_1) + (x_1 - 1)^2[1 + \sin^2(3\pi x_2)] + (x_2 - 1)^2[1 + \sin^2(2\pi x_2)]$$

- Global minimum: $f(\mathbf{x}^*) = 0$ at $\mathbf{x}^* = (1, 1)$
- We optimize the Levy function using the BFGS method with a random starting value of $-2 \leq x_1 \leq 2, -2 \leq x_2 \leq 2$ and note the objective function value of the result after optimization
- We repeat this 100 times



Distribution of the 100 optimization results (y values):

##	Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
##	0.0000	0.1099	0.5356	2.4351	1.9809	18.3663

MULTI-STARTS

Idea: use multiple starting points $\mathbf{x}^{[1]}, \dots, \mathbf{x}^{[k]}$ for algorithm A

Algorithm Multistart optimization

```
1: Given: optimization algorithm  $A(\cdot)$ ,  $f : \mathcal{S} \mapsto \mathbb{R}$ ,  $x \mapsto f(x)$ 
2:  $k = 0$ 
3: repeat
4:   Draw starting point  $x^{[k]}$  from  $\mathcal{S}$  (e.g. uniform if  $\mathcal{S}$  is of finite volume)
5:   if  $k = 0$  then  $\hat{x} = x_0$ 
6:   end if
7:   Initialize algorithm with start value  $x^{[k]} \Rightarrow \tilde{x} = A(x^{[k]})$ 
8:   if  $f(\tilde{x}) < f(\hat{x})$  then  $\hat{x} = \tilde{x}$ 
9:   end if
10:   $k = k + 1$ 
11: until Stop criterion fulfilled
12: return  $\hat{x}$ 
```

MULTI-STARTS

BFGS with Multistart gives us the true minimum of the Levy function:

```
iters = 20 # number of starts
xbest = c(runif(1, -2, 2), runif(1, -2, 2))

for (i in 1:iters) {
  x1 = runif(1, -2, 2)
  x2 = runif(1, -2, 2)
  res = optim(par = c(x1, x2), fn = f, method = "BFGS")
}

if (res$value < f(xbest)) {
  xbest = res$par
}

xbest
## [1] 1 1
```