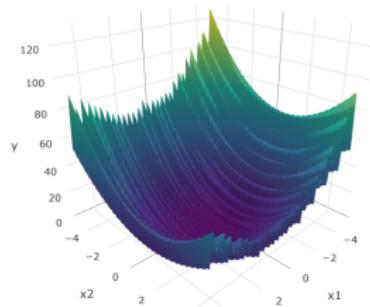


# Optimization in Machine Learning

## Multi-Start Optimization



### Learning goals

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

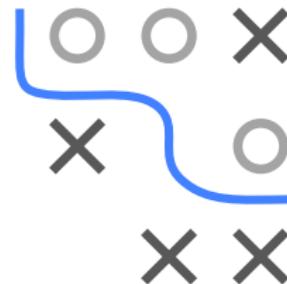


# MOTIVATION

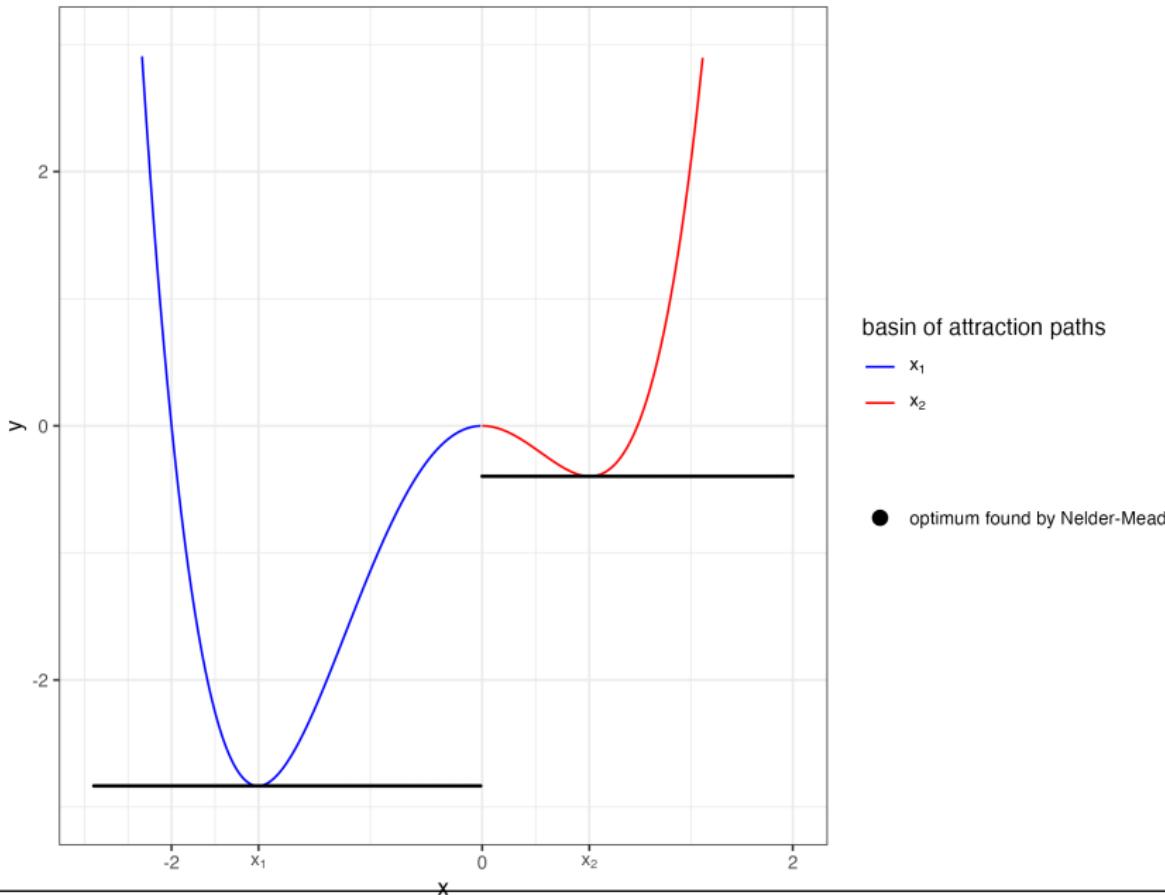
- So far: derivative-free methods for unimodal objective function (exception: simulated annealing)
- With multimodal objective functions, methods converge to local minima.
- Optimum found may differ for different starting values  $\mathbf{x}^{[0]}$
- Attraction areas:
- Let  $f_1^*, \dots, f_k^*$  be local minimum values of  $f$  with  $f_i^* \neq f_j^* \quad \forall i \neq j$ .
- Notation:  $A(\mathbf{x}^{[0]})$  denotes result of algorithm  $A$  started at  $\mathbf{x}^{[0]}$
- Then: Set

$$\mathcal{A}(f_i^*, A) = \{\mathbf{x} : A(\mathbf{x}) = f_i^*\}$$

is called *attraction area/basin of attraction* of  $f_i^*$  for algorithm  $A$



# ATTRACTION AREAS

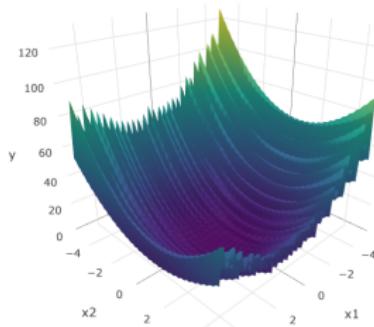


# MULTI-STARTS

Levy function:

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

- Global minimum:  $f(\mathbf{x}^*) = 0$  at  $\mathbf{x}^* = (1, 1)^\top$
- Optimize  $f$  by BFGS method with random starting point in  $[-2, 2]^2$  and collect result
- Repeat 100 times



Distribution of 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

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## Algorithm Multistart optimization

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

---



# 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
```

