Effective Programming Practices for Economists

Numerical Optimization

Derivative-Based Line Search Algorithms

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Basic Idea (optimagic docs)

- 1. Evaluate function at initial point
- 2. Use first derivative to get step direction
- 3. Guess initial step length based on (approximated) second derivative
- 4. Pick candidate step based on line search procedure (see next slide)
- 5. Accept the new parameter and go back to 1.

(ignore the case where we don't accept)

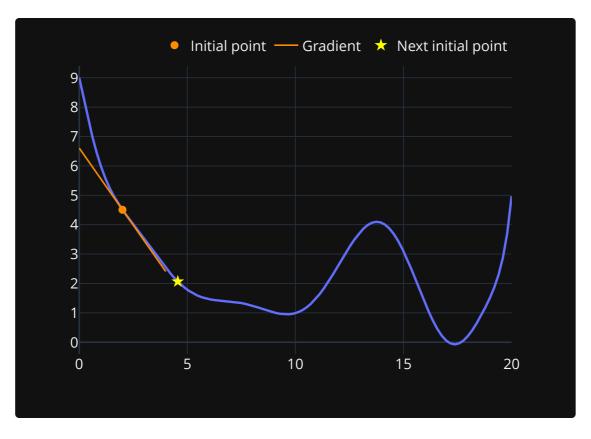
The candidate step

lacktriangle Compute the search direction p based on some x

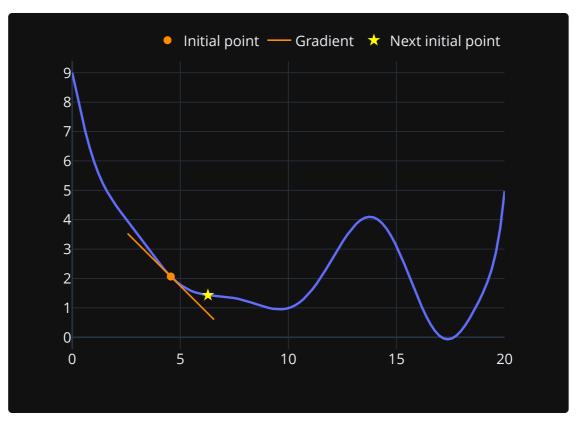
For gradient descent: p = -f'(x)

- lacktriangle Line search: Choose step length lpha to minimize f along the direction p
 - remains a 1d problem even with many parameters
 - only solved approximately
 - quite complicated if you really want to understand it
 - most of the time accepts the first guess
- lacktriangle The candidate step x_c is defined as: $x_c = x + \alpha \, p$

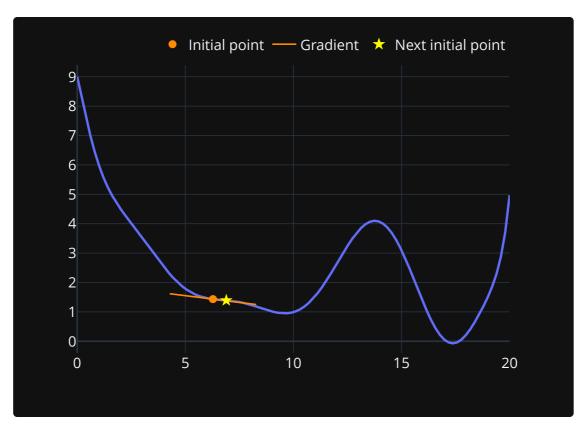
Initial Evaluation



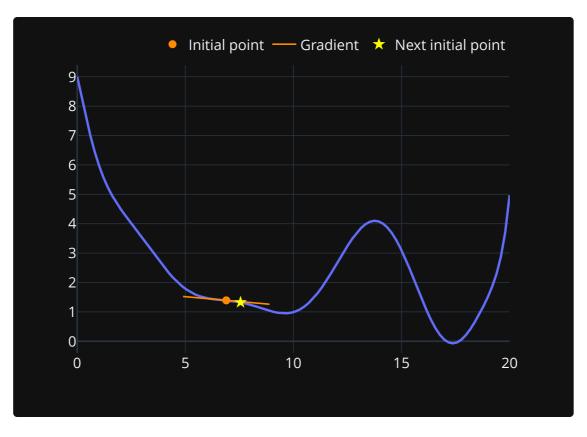
Large gradient, small curvature → Big step.



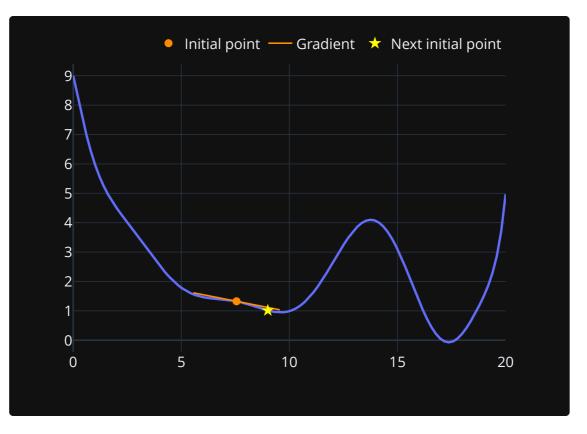
Large gradient, large curvature → Small step.



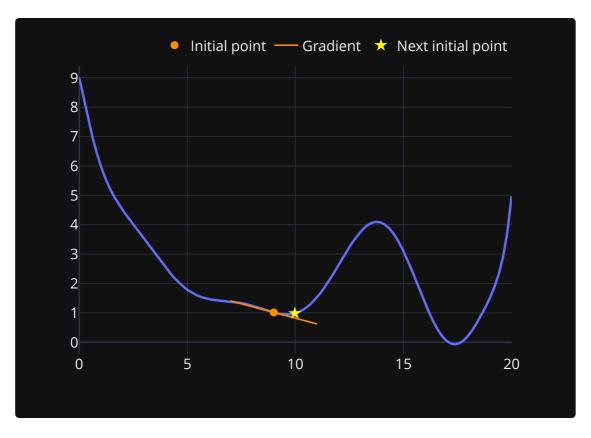
Small gradient, small curvature → Small step.



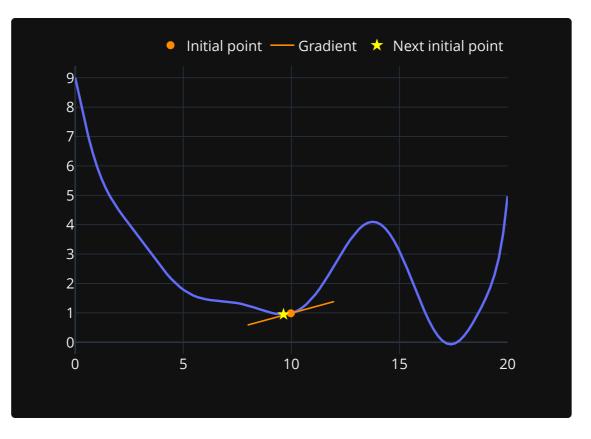
Small gradient, small curvature → Small step.



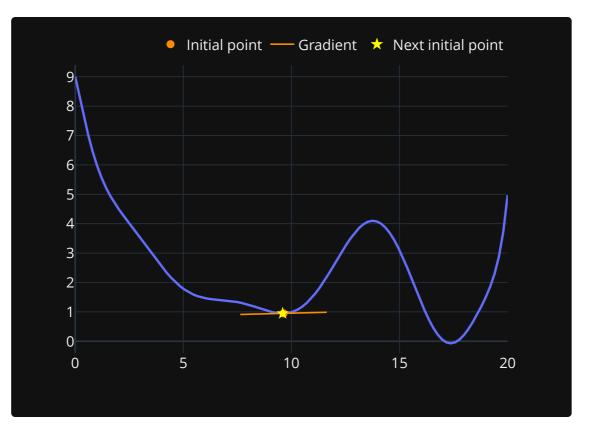
Medium-sized gradient, small curvature → Medium-sized step.



Small gradient, small curvature → Small step.



Reverse direction because gradient switches sign.



Converge because gradient is approximately zero.

Some Remarks

- No tuning parameters: Big advantage over other types of algorithms.
- Standard gradient descent would always use the same step length what we showed converges in fewer steps.
- Nevertheless, standard gradient descent can be computationally better in very high dimensional problems (Hessian becomes too large!).

A Real Algorithm: L-BFGS-B

