



# Optimisation Algorithms in Statistics II – Spring 2021

## Assignment 2

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Perform the solutions individually and send your report **until April 26** to me. Try to keep this deadline. However, if you have problems with it, there will be a final deadline on August 31 for all assignments.

Please send me one **pdf-file** with your report (alternatively, Word is ok, too), and additionally, please send me your code in one separate **plain-text file** (an R-markdown, `.rmd`, is possible but not required).

### Problem 2.1

Consider a unidimensional minimisation problem where the minimum is attained at 0. Consider further a particle in the PSO algorithm with  $x^{(1)} = 0$  and  $x^{(2)} = -2$  ( $v^{(2)} = -2$ ). For this particle, we have  $p_{\text{best}}^{(t)} = g_{\text{best}}^{(t)} = 0$  for all  $t$  (the starting value happened to identify already the minimum; the stagnation assumption is fulfilled here). We consider a standard PSO with parameters  $w, c_1 = c_2 =: c$  which generates the sequence  $x^{(t)}, t = 1, 2, \dots$  for this particle.

- Compute the sequence  $E[x^{(t)}]$  for iteration number  $t = 1, 2, \dots, 40$  and plot  $E[x^{(t)}]$  versus  $t$  for different combinations of  $w$  and  $c$ . Use the pairs  $(0.721, 1.193)$  (default in R-package `psa`),  $(0.9, 1.193)$ ,  $(0.721, 2.2)$ ,  $(0.2, 3)$  for  $(w, c)$  and at least one further pair of your choice.
- Simulate the sequence  $x^{(t)}, t = 1, 2, \dots, 40$ , around 1000 times. Compute the Monte Carlo estimate for  $\text{Var}(x^{(t)})$  for each  $t = 3, \dots, 40$  (which is simply the variance of the say 1000 simulated values for  $x^{(t)}$ ). Plot the estimated variance versus iteration number. Do this for the  $(w, c)$ -pairs which you have used in a.
- Based on your results from a. and the empirical results from b.: Can you confirm the theoretical results about order-1 and order-2 stability?

## Problem 2.2

An iterative algorithm is given with the iteration rule

$$x^{(t+1)} = x^{(t)} + R_{t+1}(x^* - x^{(t-1)}) + S_{t+1}(x^* - x^{(t-2)}),$$

where  $R_i$  are independent  $\text{Unif}[0, r]$ -distributed random variables ( $r \geq 0$ ),  $S_i$  are independent  $\text{Unif}[0, s]$ -distributed random variables ( $s \geq 0$ ), and  $x^{(1)}, x^{(2)}, x^{(3)}$  are given starting values and  $x^*$  is a fixed value.

- Consider for simplicity unidimensional values for  $x^{(t)}$  and  $x^*$ . Determine (with help of  $\mathbf{R}$ ) the pairs  $(r, s)$  ( $0 \leq r \leq 2, 0 \leq s \leq 2$ ) which ensure an order-1 stable sequence  $(x^{(t)})$  and show them in a plot.
- What does the results from a. say for the multidimensional case? It could be either when

$$\mathbf{x}^{(t+1)} = \mathbf{x}^{(t)} + R_{t+1}(\mathbf{x}^* - \mathbf{x}^{(t-1)}) + S_{t+1}(\mathbf{x}^* - \mathbf{x}^{(t-2)}),$$

or when

$$\mathbf{x}^{(t+1)} = \mathbf{x}^{(t)} + \mathbf{R}_{t+1} \otimes (\mathbf{x}^* - \mathbf{x}^{(t-1)}) + \mathbf{S}_{t+1} \otimes (\mathbf{x}^* - \mathbf{x}^{(t-2)}),$$

where  $\otimes$  denotes the component-wise product and the components of the random vectors  $\mathbf{R}_i, \mathbf{S}_i$  are all independent, uniformly distributed,  $\text{Unif}[0, r], \text{Unif}[0, s]$ , respectively. Discuss it briefly; no derivations or programming are expected for this Part b.