APPENDIX A

TEST PROBLEMS IN OPTIMIZATION

In order to validate any new optimization algorithm, we have to validate it against standard test functions so as to compare its performance with well-established or existing algorithms. There are many test functions, so there is no standard list or set of test functions one has to follow. However, various test functions do exist, so new algorithms should be tested using at least a subset of functions with diverse properties so as to make sure whether or not the tested algorithm can solve certain type of optimization efficiently.

In this appendix, we will provide a subset of commonly used test functions with simple bounds as constraints, though they are often listed as unconstrained problems in literature. We will list the function form f(x), its search domain, optimal solutions x_* and/or optimal objective value f_* . Here, we use $x = (x_1, ..., x_n)^T$ where n is the dimension.

Ackley's function:

$$f(\mathbf{x}) = -20 \exp\left[-\frac{1}{5}\sqrt{\frac{1}{n}\sum_{i=1}^{n}x_i^2}\right] - \exp\left[\frac{1}{n}\sum_{i=1}^{n}\cos(2\pi x_i)\right] + 20 + e, \quad (A.1)$$

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where n = 1, 2, ..., and $-32.768 \le x_i \le 32.768$ for i = 1, 2, ..., n. This function has the global minimum $f_* = 0$ at $\mathbf{x}_* = (0, 0, ..., 0)$.

De Jong's functions: The simplest of De Jong's functions is the so-called sphere function

$$f(\mathbf{x}) = \sum_{i=1}^{n} x_i^2, \qquad -5.12 \le x_i \le 5.12, \tag{A.2}$$

whose global minimum is obviously $f_* = 0$ at (0, 0, ..., 0). This function is unimodal and convex. A related function is the so-called weighted sphere function or hyper-ellipsoid function

$$f(\mathbf{x}) = \sum_{i=1}^{n} ix_i^2, \quad -5.12 \le x_i \le 5.12,$$
 (A.3)

which is also convex and unimodal with a global minimum $f_* = 0$ at $x_* = (0, 0, ..., 0)$. Another related test function is the sum of different power function

$$f(x) = \sum_{i=1}^{n} |x_i|^{i+1}, \qquad -1 \le x_i \le 1,$$
 (A.4)

which has a global minimum $f_* = 0$ at (0, 0, ..., 0).

Easom's function:

$$f(x) = -\cos(x)\cos(y)\exp\left[-(x-\pi)^2 + (y-\pi)^2\right],$$
 (A.5)

whose global minimum is $f_* = -1$ at $x_* = (\pi, \pi)$ within $-100 \le x, y \le 100$. It has many local minima. Xin-She Yang extended in 2008 this function to n dimensions, and we have

$$f(x) = -(-1)^n \left(\prod_{i=1}^n \cos^2(x_i) \right) \exp\left[-\sum_{i=1}^n (x_i - \pi)^2 \right], \tag{A.6}$$

whose global minimum $f_* = -1$ occurs at $\boldsymbol{x}_* = (\pi, \pi, ..., \pi)$. Here the domain is $-2\pi \le x_i \le 2\pi$ where i = 1, 2, ..., n.

Equality-Constrained Function:

$$f(\boldsymbol{x}) = -(\sqrt{n})^n \prod_{i=1}^n x_i,$$
(A.7)

subject to an equality constraint (a hyper-sphere)

$$\sum_{i=1}^{n} x_i^2 = 1. (A.8)$$

The global minimum $f_* = -1$ of f(x) occurs at $x_*(1/\sqrt{n},...,1/\sqrt{n})$ within the domain $0 \le x_i \le 1$ for i = 1, 2, ..., n.

Griewank's function:

$$f(\mathbf{x}) = \frac{1}{4000} \sum_{i=1}^{n} x_i^2 - \prod_{i=1}^{n} \cos(\frac{x_i}{\sqrt{i}}) + 1, \qquad -600 \le x_i \le 600, \tag{A.9}$$

whose global minimum is $f_* = 0$ at $\boldsymbol{x}_* = (0, 0, ..., 0)$. This function is highly multimodal.

Michaelwicz's function:

$$f(\boldsymbol{x}) = -\sum_{i=1}^{n} \sin(x_i) \cdot \left[\sin(\frac{ix_i^2}{\pi})\right]^{2m},$$
 (A.10)

where m = 10, and $0 \le x_i \le \pi$ for i = 1, 2, ..., n. In 2D case, we have

$$f(x,y) = -\sin(x)\sin^{20}(\frac{x^2}{\pi}) - \sin(y)\sin^{20}(\frac{2y^2}{\pi}), \tag{A.11}$$

where $(x, y) \in [0, 5] \times [0, 5]$. This function has a global minimum $f_* \approx -1.8013$ at $\boldsymbol{x}_* = (x_*, y_*) = (2.20319, 1.57049)$.

Perm Functions:

$$f(\mathbf{x}) = \sum_{j=1}^{n} \left\{ \sum_{i=1}^{n} (i^{j} + \beta) \left[\left(\frac{x_{i}}{i} \right)^{j} - 1 \right] \right\}, \qquad (\beta > 0),$$
 (A.12)

which has the global minimum $f_* = 0$ at $x_* = (1, 2, ..., n)$ in the search domain $-n \le x_i \le n$ for i = 1, ..., n. A related function

$$f(x) = \sum_{j=1}^{n} \left\{ \sum_{i=1}^{n} (i+\beta) \left[x_i^j - (\frac{1}{i})^j \right] \right\}^2, \tag{A.13}$$

has the global minimum $f_* = 0$ at (1, 1/2, 1/3, ..., 1/n) within the bounds $-1 \le x_i \le 1$ for all i = 1, 2, ..., n. As $\beta > 0$ becomes smaller, the global minimum becomes almost indistinguishable from their local minima. In fact, in the extreme case $\beta = 0$, every solution is also a global minimum.

Rastrigin's function:

$$f(\mathbf{x}) = 10n + \sum_{i=1}^{n} \left[x_i^2 - 10\cos(2\pi x_i) \right], \quad -5.12 \le x_i \le 5.12, \quad (A.14)$$

whose global minimum is $f_* = 0$ at (0, 0, ..., 0). This function is highly multimodal.

Rosenbrock's function:

$$f(\mathbf{x}) = \sum_{i=1}^{n-1} \left[(x_i - 1)^2 + 100(x_{i+1} - x_i^2)^2 \right],\tag{A.15}$$

whose global minimum $f_* = 0$ occurs at $x_* = (1, 1, ..., 1)$ in the domain $-5 \le x_i \le 5$ where i = 1, 2, ..., n. In the 2D case, it is often written as

$$f(x,y) = (x-1)^2 + 100(y-x^2)^2,$$
 (A.16)

which is often referred to as the banana function.

Schwefel's function:

$$f(\mathbf{x}) = -\sum_{i=1}^{n} x_i \sin\left(\sqrt{|x_i|}\right), \quad -500 \le x_i \le 500,$$
 (A.17)

whose global minimum $f_* \approx -418.9829n$ occurs at $x_i = 420.9687$ where i = 1, 2, ..., n.

Six-hump camel back function:

$$f(x,y) = (4 - 2.1x^2 + \frac{1}{3}x^4)x^2 + xy + 4(y^2 - 1)y^2,$$
 (A.18)

where $-3 \le x \le 3$ and $-2 \le y \le 2$. This function has two global minima $f_* \approx -1.0316$ at $(x_*, y_*) = (0.0898, -0.7126)$ and (-0.0898, 0.7126).

Shubert's function:

$$f(\boldsymbol{x}) = \left[\sum_{i=1}^{n} i \cos\left(i + (i+1)x\right)\right] \cdot \left[\sum_{i=1}^{n} i \cos\left(i + (i+1)y\right)\right], \quad (A.19)$$

which has 18 global minima $f_* \approx -186.7309$ for n = 5 in the search domain $-10 \le x, y \le 10$.

Xin-She Yang's functions:

$$f(\boldsymbol{x}) = \left(\sum_{i=1}^{n} |x_i|\right) \exp\left[-\sum_{i=1}^{n} \sin(x_i^2)\right],\tag{A.20}$$

which has the global minimum $f_* = 0$ at $x_* = (0, 0, ..., 0)$ in the domain $-2\pi \le x_i \le 2\pi$ where i = 1, 2, ..., n. This function is not smooth, and its derivatives are not well defined at the optimum (0, 0, ..., 0).

A related function is

$$f(\mathbf{x}) = -\left(\sum_{i=1}^{n} |x_i|\right) \exp\left(-\sum_{i=1}^{n} x_i^2\right), \quad -10 \le x_i \le 10,$$
 (A.21)

which has multiple global minima. For example, for n = 2, we have 4 equal minima $f_* = -1/\sqrt{e} \approx -0.6065$ at (1/2, 1/2), (1/2, -1/2), (-1/2, 1/2) and (-1/2, -1/2).

Yang also designed a standing-wave function with a defect

$$f(\mathbf{x}) = \left[e^{-\sum_{i=1}^{n} (x_i/\beta)^{2m}} - 2e^{-\sum_{i=1}^{n} x_i^2} \right] \cdot \prod_{i=1}^{n} \cos^2 x_i, \qquad m = 5, \quad (A.22)$$

which has many local minima and the unique global minimum $f_* = -1$ at $x_* = (0, 0, ..., 0)$ for $\beta = 15$ within the domain $-20 \le x_i \le 20$ for i = 1, 2, ..., n. He also proposed another multimodal function

$$f(\mathbf{x}) = \left\{ \left[\sum_{i=1}^{n} \sin^2(x_i) \right] - \exp\left(-\sum_{i=1}^{n} x_i^2\right) \right\} \cdot \exp\left[-\sum_{i=1}^{n} \sin^2 \sqrt{|x_i|} \right], \quad (A.23)$$

whose global minimum $f_* = -1$ occurs at $x_* = (0, 0, ..., 0)$ in the domain $-10 \le x_i \le 10$ where i = 1, 2, ..., n. In the 2D case, its landscape looks like a wonderful candlestick.

Most test functions are deterministic. Yang designed a test function with stochastic components

$$f(x,y) = -5e^{-\beta[(x-\pi)^2 + (y-\pi)^2]} - \sum_{j=1}^K \sum_{i=1}^K \epsilon_{ij} e^{-\alpha[(x-i)^2 + (y-j)^2]}, \quad (A.24)$$

where $\alpha, \beta > 0$ are scaling parameters, which can often be taken as $\alpha = \beta = 1$. Here ϵ_{ij} are random variables and can be drawn from a uniform distribution $\epsilon_{ij} \sim \text{Unif}[0,1]$. The domain is $0 \leq x, y \leq K$ and K = 10. This function has K^2 local valleys at grid locations and the fixed global minimum at $\boldsymbol{x}_* = (\pi, \pi)$. It is worth pointing that the minimum f_{\min} is random, rather than a fixed value; it may vary from $-(K^2 + 5)$ to -5, depending α and β as well as the random numbers drawn.

Furthermore, he also designed a stochastic function

$$f(\boldsymbol{x}) = \sum_{i=1}^{n} \epsilon_i \left| x_i - \frac{1}{i} \right|, \qquad -5 \le x_i \le 5,$$
 (A.25)

where ϵ_i (i = 1, 2, ..., n) are random variables which are uniformly distributed in [0, 1]. That is, $\epsilon_i \sim \text{Unif}[0, 1]$. This function has the unique minimum $f_* = 0$ at $\boldsymbol{x}_* = (1, 1/2, ..., 1/n)$ which is also singular.

Zakharov's functions:

$$f(\mathbf{x}) = \sum_{i=1}^{n} x_i^2 + \left(\frac{1}{2} \sum_{i=1}^{n} i x_i\right)^2 + \left(\frac{1}{2} \sum_{i=1}^{n} i x_i\right)^4, \tag{A.26}$$

whose global minimum $f_* = 0$ occurs at $x_* = (0, 0, ..., 0)$. Obviously, we can generalize this function as

$$f(x) = \sum_{i=1}^{n} x_i^2 + \sum_{k=1}^{K} J_n^{2k},$$
 (A.27)

where K = 1, 2, ..., 20 and

$$J_n = \frac{1}{2} \sum_{i=1}^{n} i x_i. (A.28)$$

REFERENCES

- D. H. Ackley, A Connectionist Machine for Genetic Hillclimbing, Kluwer Academic Publishers, 1987.
- C. A. Floudas, P. M., Pardalos, C. S. Adjiman, W. R. Esposito, Z. H. Gumus,
 S. T. Harding, J. L. Klepeis, C. A., Meyer, C. A. Scheiger, Handbook of Test Problems in Local and Global Optimization, Springer, 1999.
- 3. A. Hedar, Test function web pages, http://www-optima.amp.i.kyoto-u.ac.jp/member/student/hedar/Hedar_files/TestGO_files/Page364.htm
- M. Molga, C. Smutnicki, "Test functions for optimization needs", http://www.zsd.ict.pwr.wroc.pl/files/docs/functions.pdf
- X. S. Yang, "Firefly algorithm, Lévy flights and global optimization", in: Research and Development in Intelligent Systems XXVI, (Eds M. Bramer et al.), Springer, London, pp. 209-218 (2010).
- 6. X. S. Yang and S. Deb, "Engineering optimization by cuckoo search", Int. J. Math. Modeling and Numerical Optimization, 1, No. 4, (in press) (2010).
- 7. X. S. Yang, "Firefly algorithm, stochastic test functions and design optimization", Int. J. Bio-inspired Computation, 2, No. 2, 78-84 (2010).