CMA–VNS: A Short Description

Fan Xue and Geoffrey Q.P. Shen*
{fan.xue, bsqpshen}@polyu.edu.hk

July 8, 2015

1 Introduction

The CMA-VNS (Covariance Matrix Adaptation Variable Neighborhood Search) solver is a competitor for the first Combinatorial Black-Box Optimization Competition (CBBOC 2015\(^1\)). CMA-VNS is a hyper-heuristic which employs a CMA-ES (Covariance Matrix Adaptation Evolution Strategy) [Hansen et al., 2003] evolution which is followed by an iterated local search. The reason of employing the CMA-ES is the efficient estimation of variables in CMA-ES. Hence the local search, variable neighborhood search in this context, can be facilitated against expensive (and/or limited) evaluations.

2 Main Procedure

Figure 1 shows the pseudo code of CMA-VNS. There are two main blocks in the pseudo code: CMA block (line 4-7) and VNS block (line 8-15). The proportion of the two block is controlled in a hyper-heuristic mechanism which tested the overall performance of a number of modes before-hand and trained a set of rules from the results.

The CMA block trains a bi-population CMA-ES\(^2\) to predict promising solutions in a given black-box problem. One of the most important parameter LAMBDA is determined by the pre-determined rules. If any new best known solutions are found, they can be appended into the elite set except for the premature solutions at very beginning.

The VNS block employs three mechanisms. First, a backbone (common bits of elite sets, see [Zhang and Looks, 2005]) is constructed to strengthen the intensification of searching. Then an estimated promising solution is generated by CMA-ES for diversification then combined with the backbone. At last, a VNS procedure tries to find improvements of the generated solution. If the new result is better than the best known value, the new result is appended to the elite set.

---

*Corresponding author. Tel.: (852) 2766 5817; Addr.: Department of Building and Real Estate, The Hong Kong Polytechnic University, Hunghom, Kowloon, Hong Kong S.A.R., China P.R.

\(^1\)See http://web.mst.edu/~tauritzd/CBBOC/.

\(^2\)From LIBCMAES, see https://github.com/beniz/libcmaes.
procedure CMA–VNS (dimension, evaluations) \hfill \triangleright \text{Note: Main procedure}

params ← configure_by_pre-trained_rules (dimension, evaluations)

elites ← ∅

repeat \hfill \triangleright \text{Note: CMAES goes first}

solutions ← BIPOP-CMAES (params.lambda)

elites += solutions.new_best_knowns

until params.cma_eval is met

repeat \hfill \triangleright \text{Note: Then VNS}

backbone ← common_bits (elites)

s_0 ← biopCMAES.predict with backbone \hfill \triangleright \text{Note: By CMAES and elites}

solution ← VNS (s_0) \hfill \triangleright \text{Note: With a flipping tabu list}

if solution is new best known then

elites += solution

end if

until params.vns_eval is met

end procedure

Figure 1: Pseudo code of CMA–VNS

3 Additional features

Solution cache A hashtable cache is employed to map bit vectors to objective function values. Before any evaluation request, the bit vector is tested first in the cache. A successful hit can save time and increase effective evaluations.

Tabu list of flipping In the procedure of VNS, a tabu list of flipping is used to avoid try bit flips which failed recently. When the list usually increases from empty set to all bits quickly, i.e. no bit can be However, the list will be cleared once until no improvements found since last clearance. Experiments show that the tabu list makes evaluation of solutions more effective.

Training free The parameters of CMA-VNS are determined by pre-trained rules. The dimension and maximum evaluations are the two main inputs for using the rules. In addition, the rules are found effective enough without re-running and re-training for each NK problem set.

References
