

2015 IEEE Congress on Evolutionary Computation Competition on: Large Scale Global Optimization

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Introduction

- Numerous meta-heuristic algorithms have been developed;
- Performance deteriorates rapidly as the dimensionality of a problem increases, i.e., **curse of dimensionality**;
- Many real-world problems exhibit such large-scale property;
- **What makes large scale optimization problems hard?**
 - Search space grows exponentially as the number of decision variables increases;
 - Properties of the search space may change;
 - Evaluations are usually expensive;
 - Interaction between variables;

Large Scale Global Optimization Benchmarks

- IEEE CEC 2008 LSGO benchmark suite: simple test functions.
- IEEE CEC 2010 and CEC 2012 benchmark suites: aim to provide a suitable evaluation platform for testing and comparing large-scale global optimization (LSGO) algorithms.
- **IEEE CEC 2013 benchmark suite***: extended upon CEC 2010 LSGO benchmark functions to better capture real-world problem characteristics; these functions pose new challenges to large scale black-box optimization algorithms.
- **Key features in CEC'2013 benchmark suite:**
 - Non-uniform subcomponent sizes;
 - Imbalance in the contribution of subcomponents;
 - Functions with overlapping subcomponents;
 - New transformations to the base functions: Ill-conditioning; Symmetry breaking; and Irregularities.

*Note that the CEC 2013 LSGO benchmark suite was used for the CEC 2015 LSGO competition.

Large Scale Global Optimization Challenge

- **Category 1:** Fully-separable functions;
- **Category 2:** Two types of partially separable functions:
 - (a) Partially separable functions with a set of non-separable subcomponents and one fully-separable subcomponents;
 - (b) Partially separable functions with only a set of non-separable subcomponents and no fully separable subcomponent.
- **Category 3:** Functions with overlapping subcomponents: the subcomponents of these functions have some degree of overlap with its neighbouring subcomponents. There are two types of overlapping functions:
 - (a) Overlapping functions with conforming subcomponents;
 - (b) Overlapping functions with conflicting subcomponents: 4. Fully-nonseparable functions.
- **Category 4:** Fully-nonseparable functions.

15 test functions (1000D) in total.

Experimental settings

- **Problems:** 15 minimization problems;
- **Dimensions:** $D = 1000$;
- **Number of runs:** 25 runs per function;
- **Maximum number of fitness evaluations:** $\text{Max FE} = 3 \times 10^6$;
- **Termination criteria:** when Max FE is reached.
- **Boundary Handling:** All problems have the global optimum within the given bounds.
- **Solution quality** for each function when the FEs counter reaches:
 - $\text{FEs}_1 = 1.2e+5$
 - $\text{FEs}_2 = 6.0e+5$
 - $\text{FEs}_3 = 3.0e+6$
- The best, median, worst, mean, and standard deviation of the 25 runs should be recorded

Experimental results

1000D		f_1	f_2	f_3	f_4	f_5	f_6	f_7	f_8
1.2e5	Best	x.xxe+xx							
	Median								
	Worst								
	Mean								
6.0e5	Best								
	Median								
	Worst								
	Mean								
3.0e6	Best								
	Median								
	Worst								
	Mean								
1.2e5	Best	x.xxe+xx							
	Median								
	Worst								
	Mean								
6.0e5	Best								
	Median								
	Worst								
	Mean								
3.0e6	Best								
	Median								
	Worst								
	Mean								

Median is used to assign points for ranking all comparing algorithms, according to the Formula 1 point system¹:

Place	Points
1	25
2	18
3	15
4	12
5	10
6	8
7	6

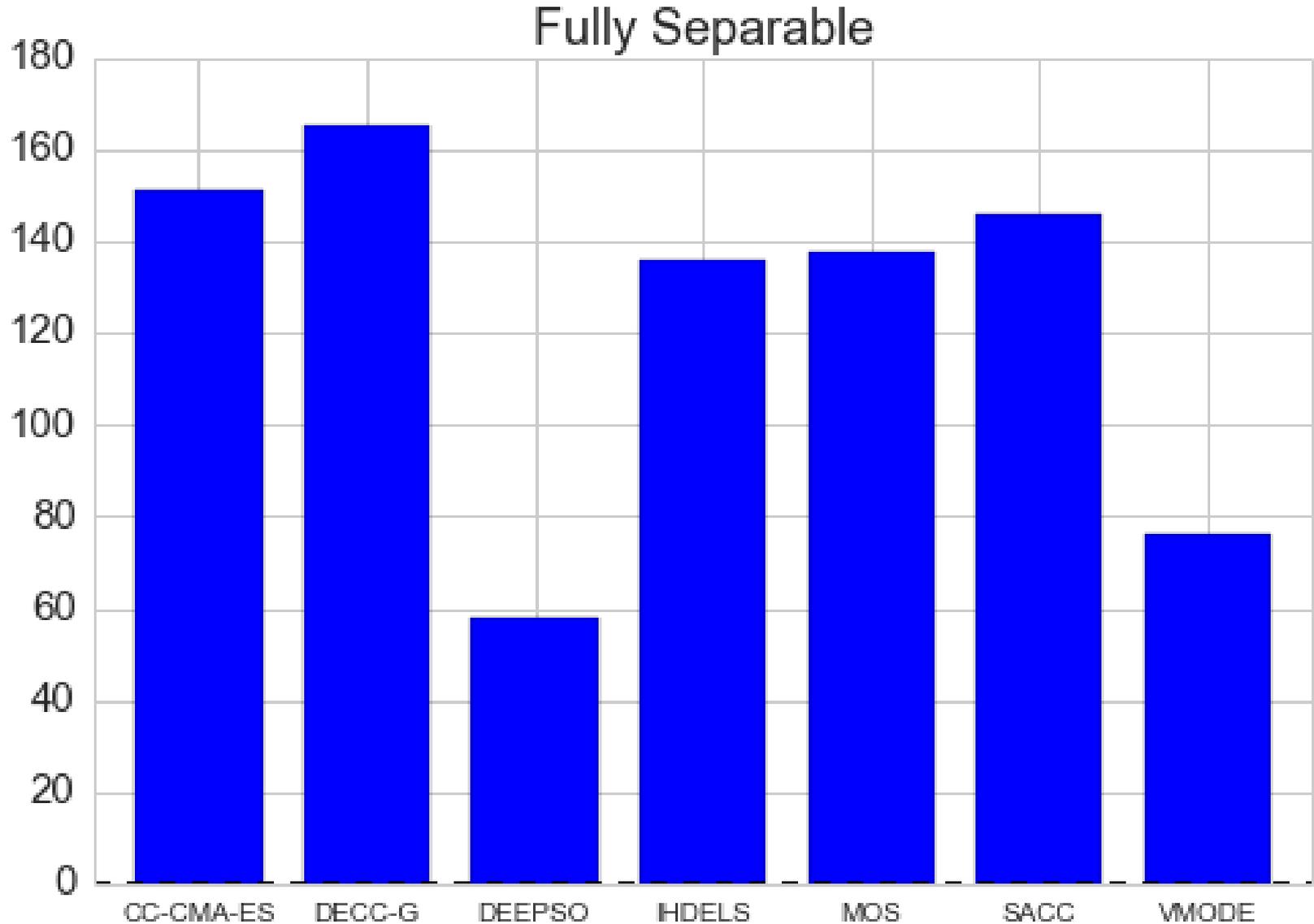
¹ URL: http://en.wikipedia.org/wiki/Formula_One_regulations

2015 entries to the competition

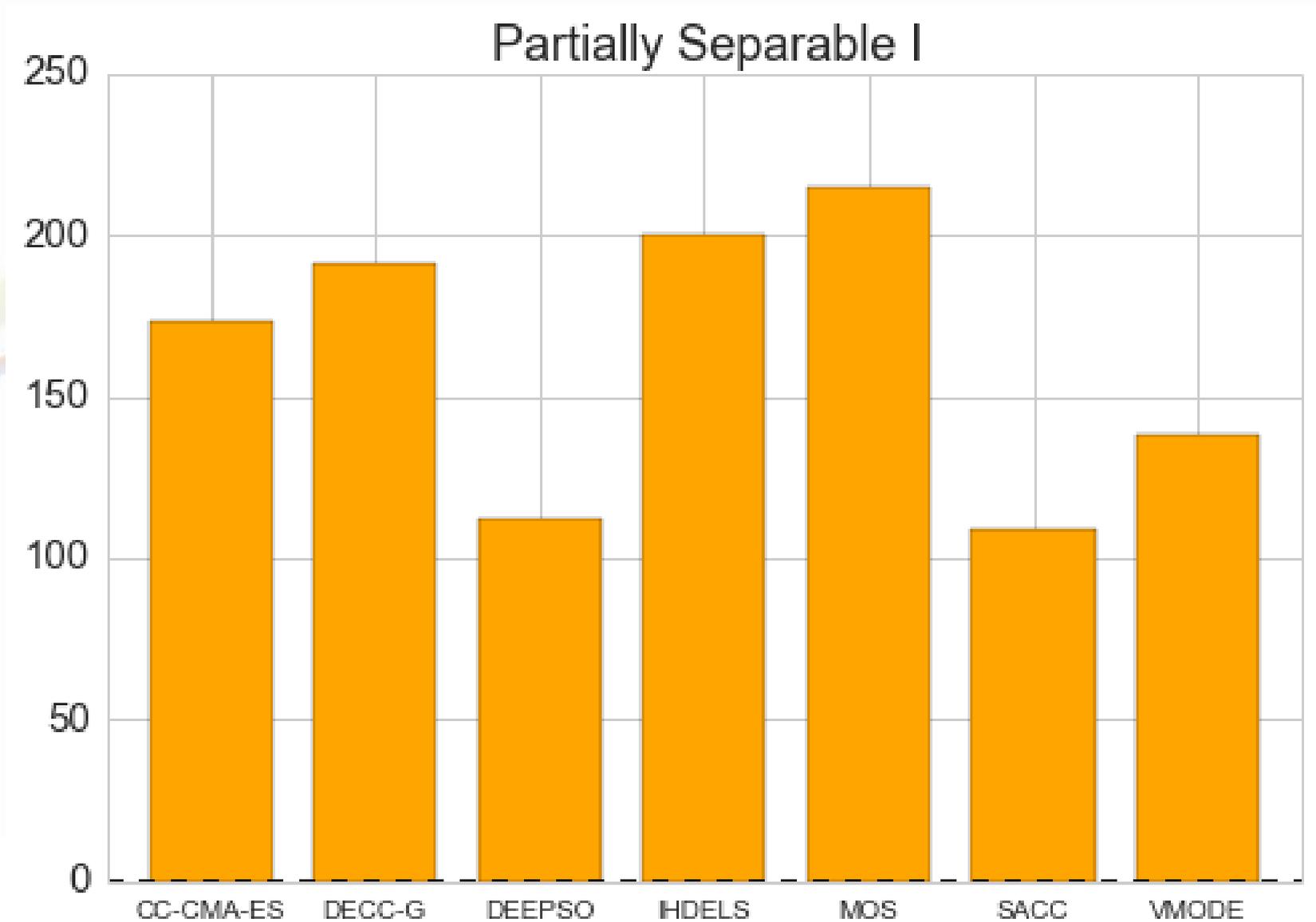
- CC-CMA-ES: Jinpeng Liu and Ke Tang
- **DECC-G: baseline model**, by Zhenyu Yang, Ke Tang and Xin Yao
- DEEPSO: Carolina G. Marcelino, Leonel M. Carvalho, Elizabeth F. Wanner, Paulo E. M. Almeida, and Vladimiro Miranda
- IHDELS: Daniel Molina and Francisco Herrera
- MOS: Antonio LaTorre, Santiago Muelas, Jose-Maria Pena
- SACC: Fei Wei, Yuping Wang, Yuanliang Huo
- VMODE: Ernesto Díaz López

In total seven entries to the CEC'15 LSGO competition, including 2 CEC'15 papers, plus 5 direct entries (without CEC'15 papers). The same CEC'13 LSGO benchmark suite was used for this CEC'15 LSGO competition.

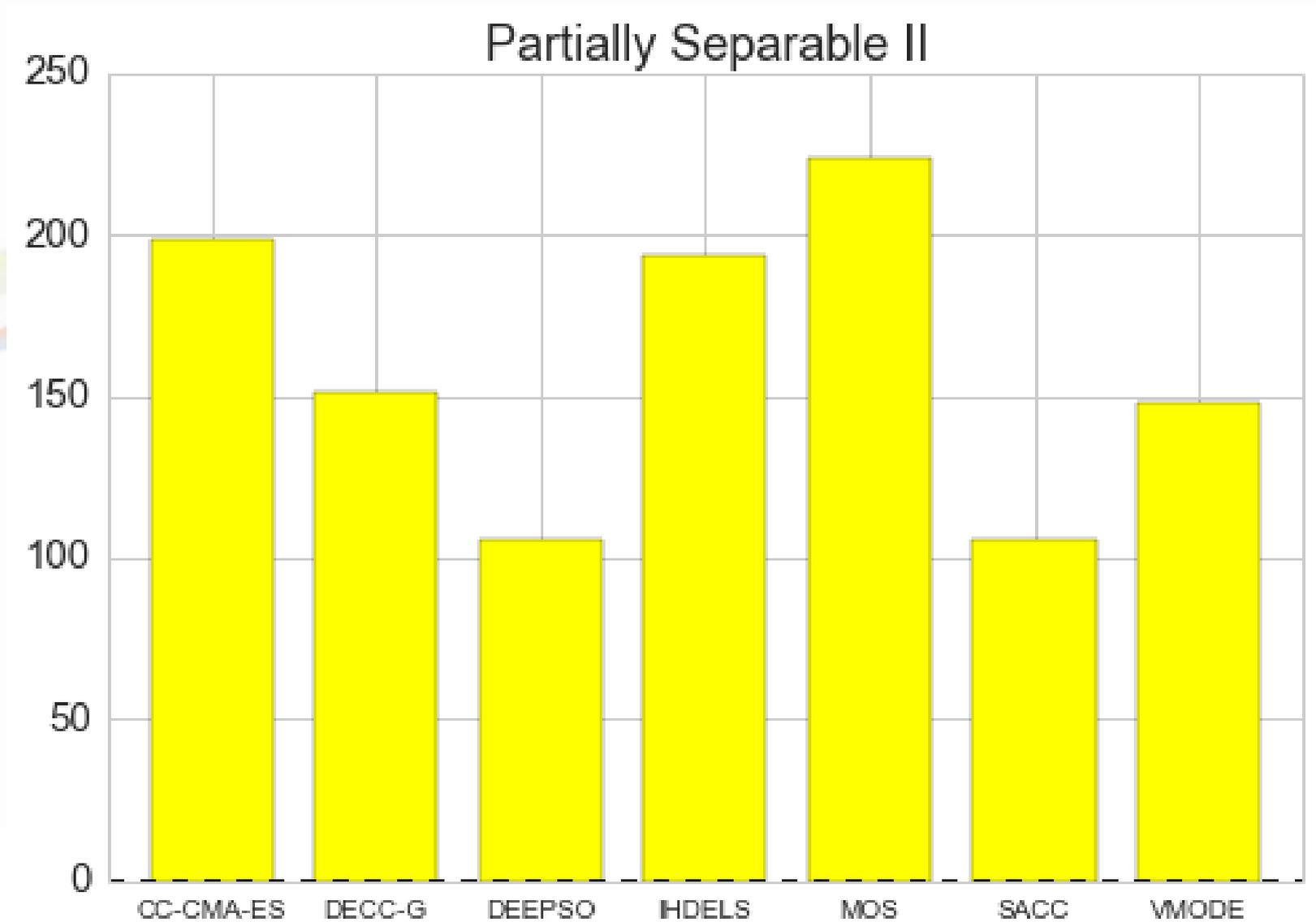
Category 1



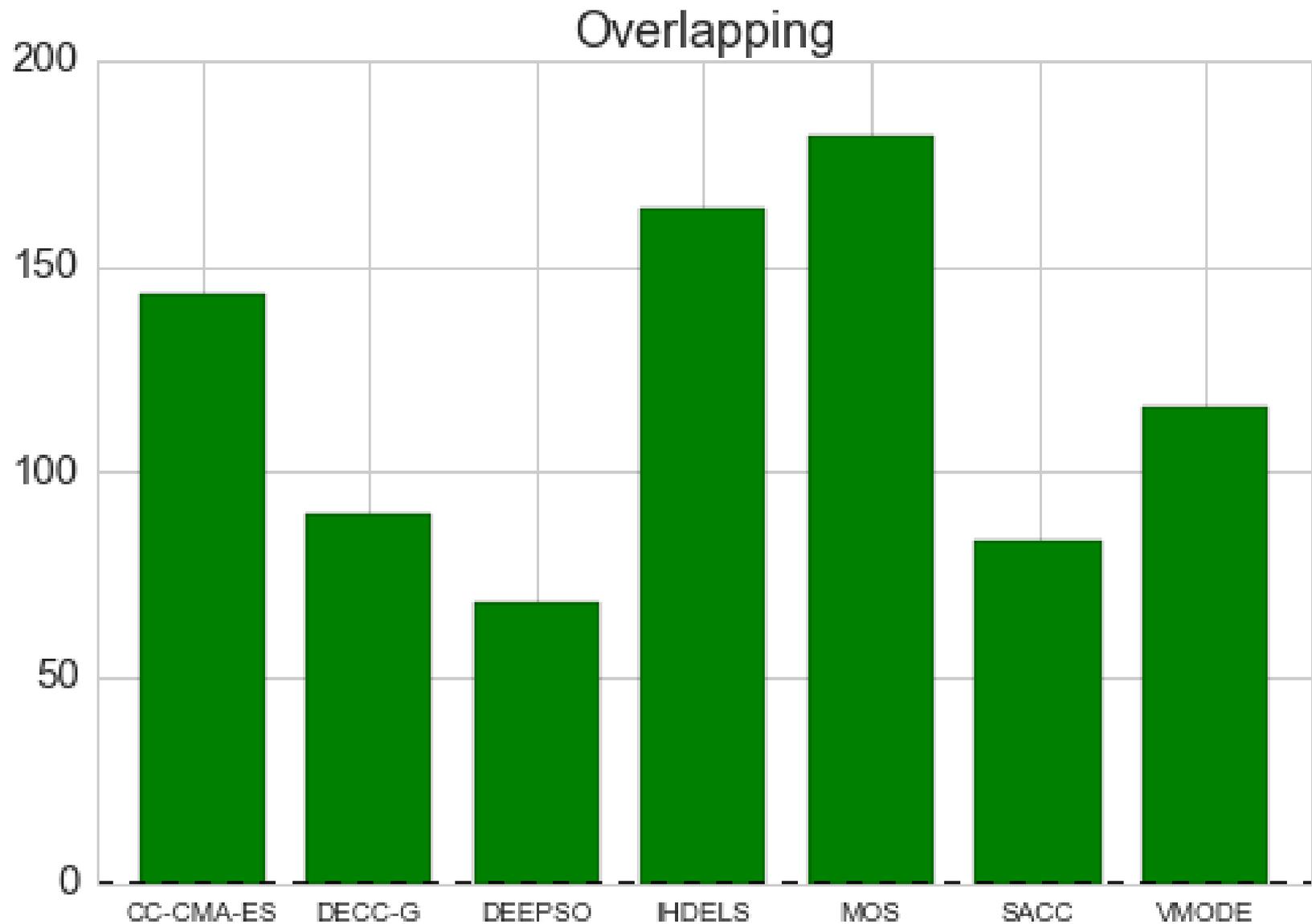
Category 2 (a)



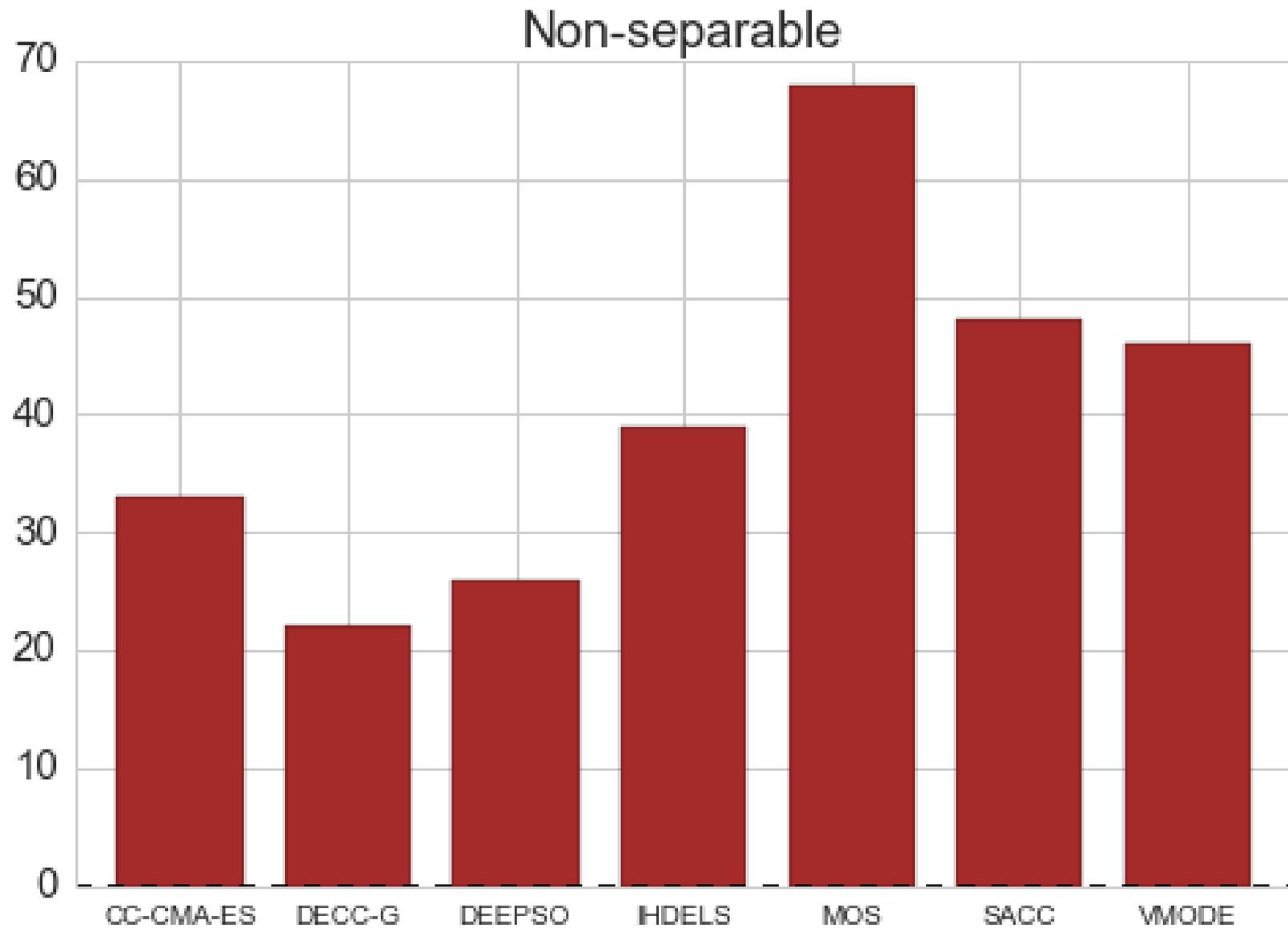
Category 2 (b)



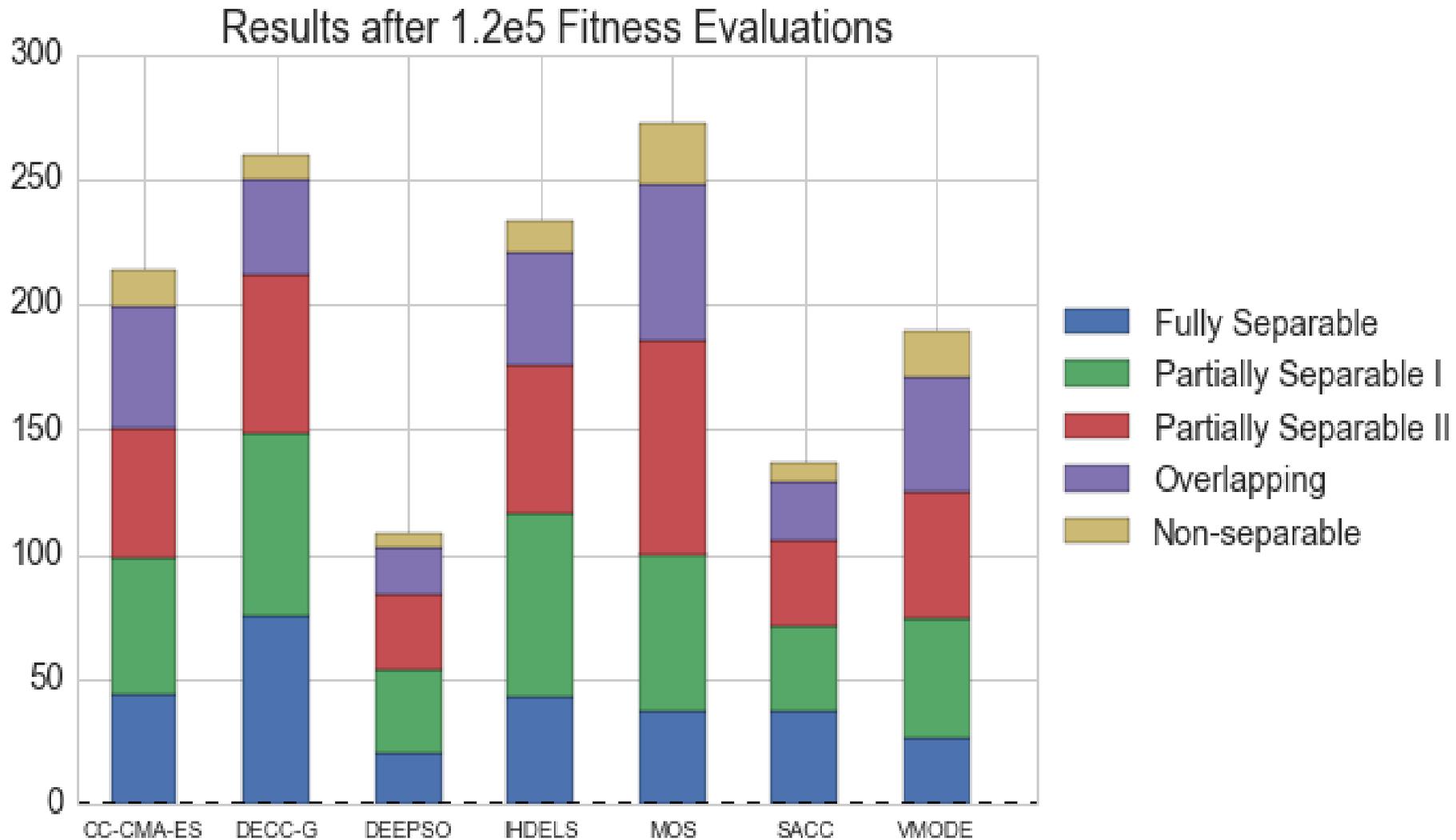
Category 3



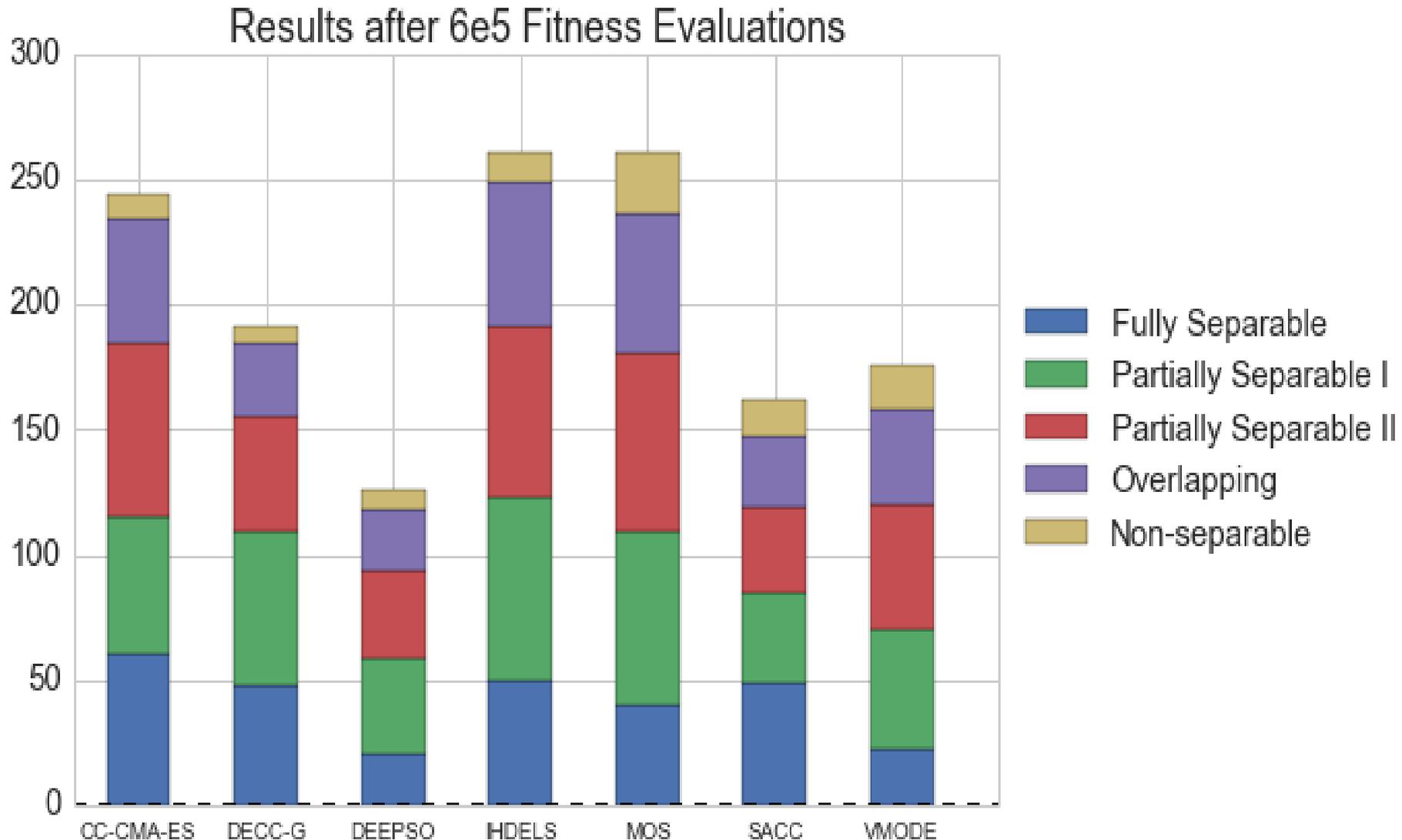
Category 4



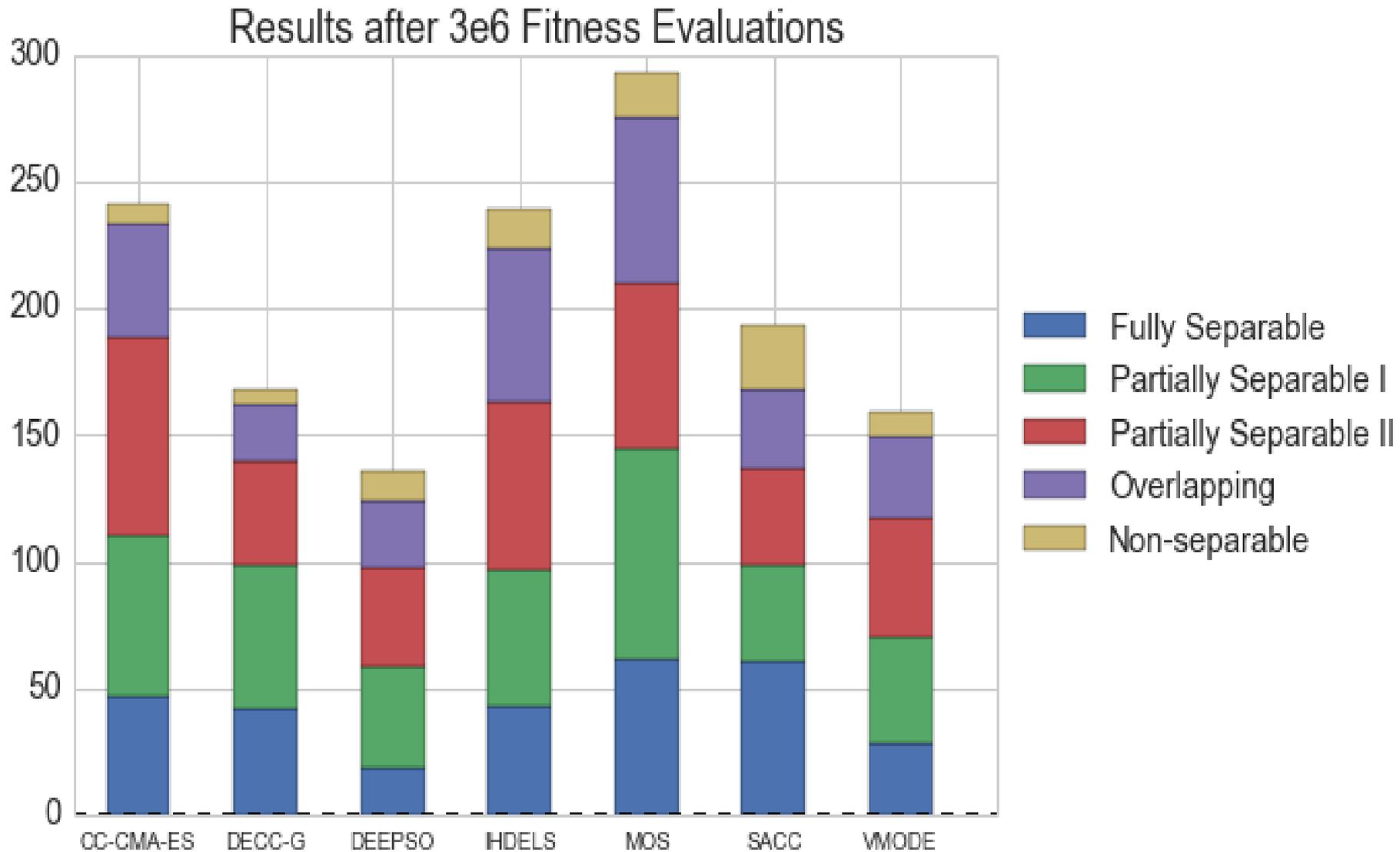
Results at 1.2e5 FEs



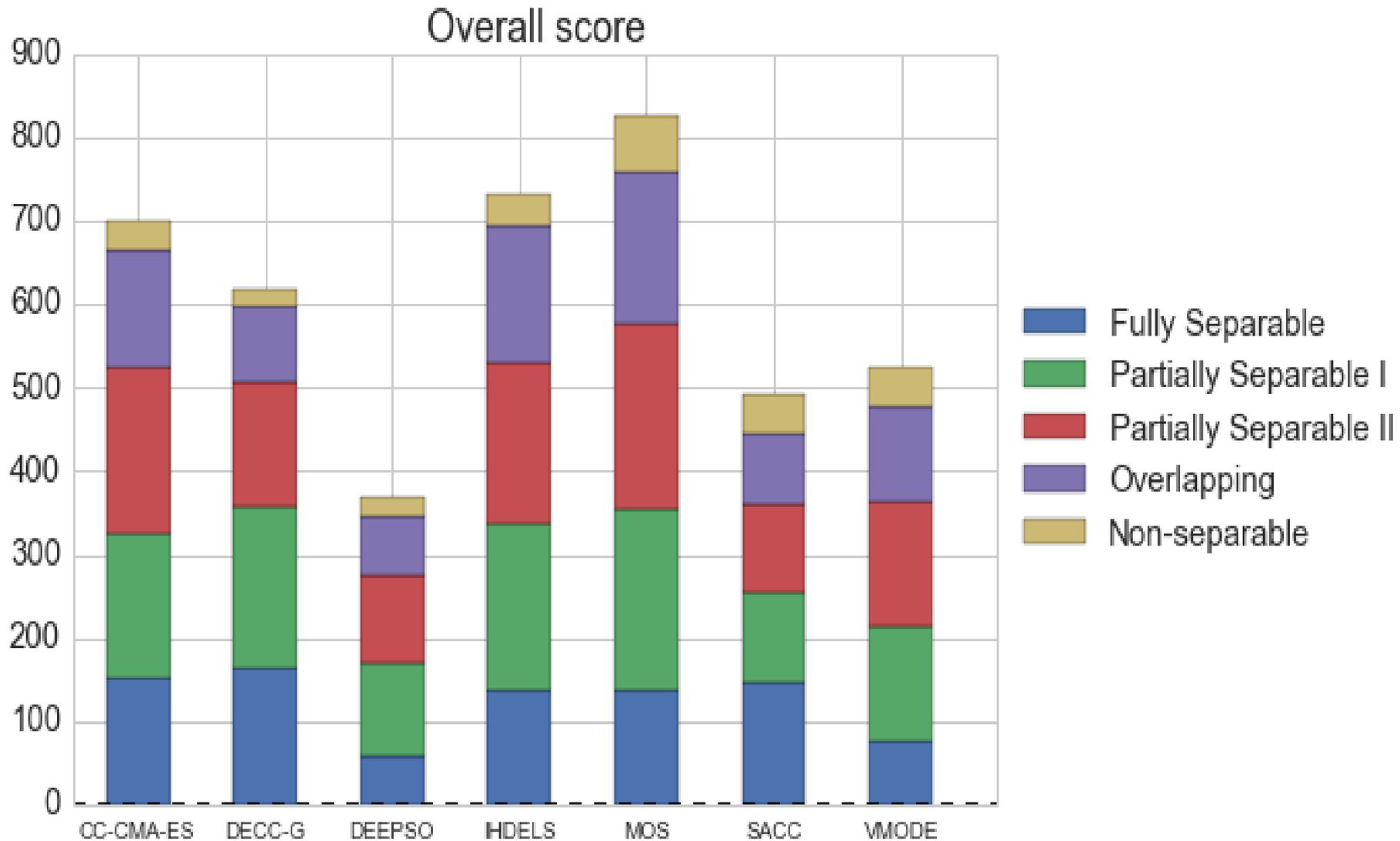
Results at 6.0e5 FEs



Results at 3.0e6 FEs



Overall Scores



Winners

- **First place: MOS (827 points)**
- **Second: IHDELS (733)**
- **Third: CC-CMA-ES (699 points)**
- **Fourth: DECC-G (619 points)**

Summary

- Seven entries including 2 CEC'15 papers, plus 5 result entries only;
- Combining different meta-heuristics;
- Strong local search;
- Decomposition has a cost; Some trade-offs between decomposition cost and optimization.
- Clear winner: **MOS** (Multiple Offspring Sampling): MOS-based Hybrid Algorithms (also the winner for 2013 LSGO competition).

Questions?

