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Applications of Biased Randomized Algorithms in Rich Transportation & Logistics

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1. Introducing the IN3 @ Castelldefels (1/2)



1. Introducing the IN3 @ Castelldefels (2/2)

- Internet Interdisciplinary Institute
- Parque Tecnológico del Mediterráneo











2. Introducing ICSO @ IN3 (1/2)

- What is ICSO? → Internet Computing & Systems Optimization.
- Goals? → To develop Hybrid Intelligent Algorithms for Smart Decision Making & Analytics.
- Applications fields? → Transportation & Logistics, Internet Computing, Smart Cities, Finance, etc.
- Coordination?
 DPCS and SLP research groups at IN3 & UOC's Computer, Multimedia and Telecommunication Studies.

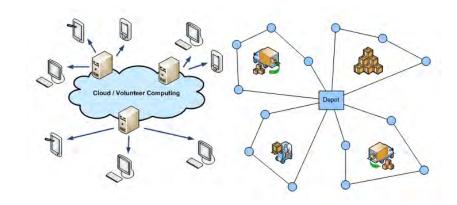






MINISTERIO DE CIÊNCIA E INNOVACIÓN





Topics of interest:

- Applied Optimization Simulation
- Internet Computing
- Biased Randomized Heuristics

&

- Metaheuristics & Simheuristics
- Logistics & Transportation
- Finance & Risk Analytics
- Manufacturing & Production
- Etc.

2. Introducing ICSO @ IN3 (2/2)































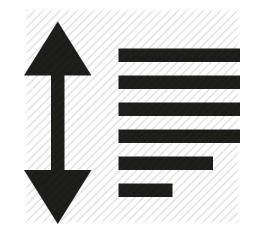


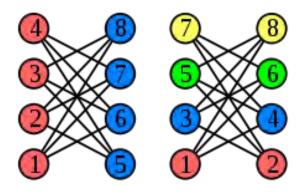
3. Biased-Randomization of Heuristics (1/4)

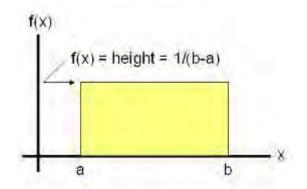
- Heuristics employ an iterative process. At each iteration, the next constructive movement is selected from a sorted list.
- Deterministic and iterative greedy procedures.

- if we randomize the order in which the elements of the list are selected, then a different output is likely to occur each time.
- However, a uniform randomization will basically destroy the logic behind the greedy behavior.

Juan, A.; Faulin, J.; Ferrer, A.; Lourenço, H.; Barrios, B. (2013): "MIRHA: multi-start biased randomization of heuristics with adaptive local search for solving non-smooth routing problems". TOP, 21: 109-132.







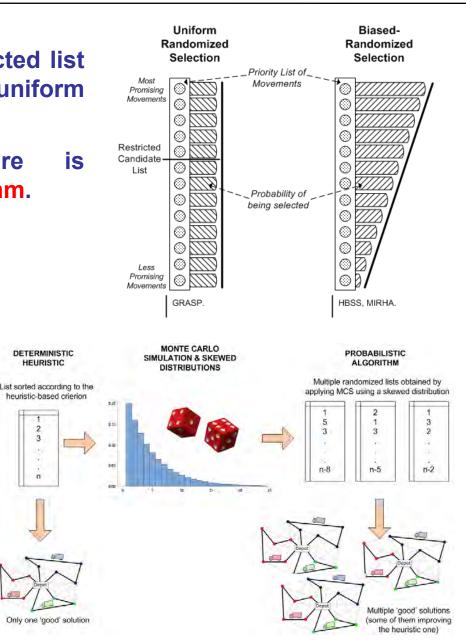
3. Biased-Randomization of Heuristics (2/4)

DETERMINISTIC

HEURISTIC

- **GRASP** proposes to consider a restricted list of candidates, and then apply a uniform randomization.
- deterministic procedure Thus. is transformed into a randomized algorithm.

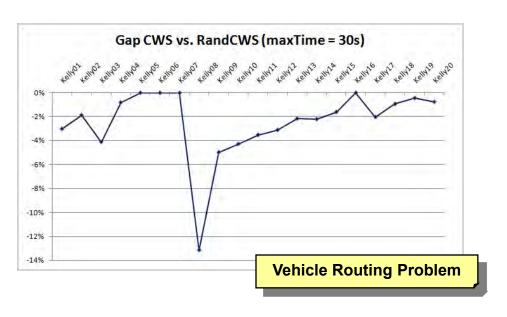
- BRAs: instead of restricting the list of candidates, assign a different probability of being selected to each potential movement in the sorted list.
- Thus, elements at the top of the list receive more probabilities of being selected.
- biased probability Idea: use distribution! Geometric, (e.g. Triangular, etc.)

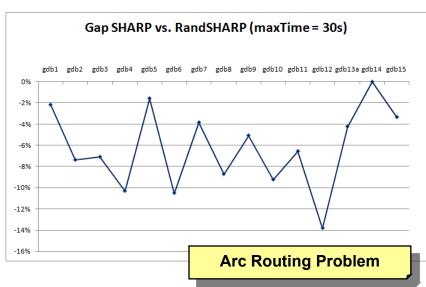


3. Biased-Randomization of Heuristics (3/4)

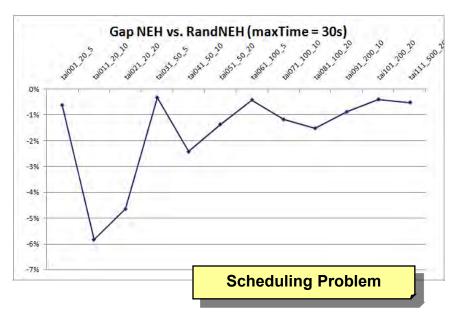
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Biased Randomization
baseSol = RandNEH // Randomized NEH solution
baseSol = localSearch(baseSol) // Classical local search
bestSol = baseSol
while endCondition is false // e.g. elapsedTime < maxTime
   currentSol = enhancedSwap(baseSol) //perturbation
   currentSol = localSearch(baseSol)
   delta = currentSol - baseSol
   if delta < 0 // acceptance criterion
      baseSol = currentSol
      if currentSol < hestSol
         bestSol = currentSol
      end if
   else if delta > 0 and lastMovementWasAnImprovement and delta <= lastImprovement
      baseSol = currentSol
   end if
end while
                            Grasas, A.; Juan, A.; Faulin, J.; Ramalhinho, H. (s-150729): "A survey on
return bestSol
                            Biased Randomization of Heuristics with applications to Logistics,
                            Transportation, and Production". JORS.
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3. Biased-Randomization of Heuristics (4/4)





- If you have a heuristic for a complex decision-making problem (in any field), you can use BR to improve it in a fast and natural way.
- Also, you can use BR to generate a set of alternative solutions (Pareto frontier).



4. Selected published applications (1/2)

Vehicle and Arc Routing Problems: CVRP, ARP, Heterogeneous, Green, Multi-Depot, Horizontal Collab. 2L-VRP, ...

Juan, A.; Faulin, J.; Jorba, J.; Riera, D.; Masip; D.; Barrios, B. (2011): "On the Use of Monte Carlo Simulation, Cache and Spliting Techniques to Improve the Clarke and Wright Savings Heuristics". JORS, Vol. 62, pp. 1085-1097.

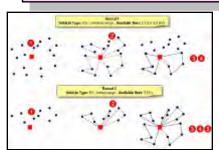
González, S.; Juan, A.; Riera, D.; Castellà, Q.; Muñoz, R.; Pérez, A. (2012): "Development and Assessment of the SHARP and RandSHARP Algorithms for the Arc Routing Problem". Al Communications, 25: 173-189

Juan, A.; Faulin, J.; Caceres, J.; Barrios, B.; Martinez, E. (2014): "A Successive Approximations Method for the Heterogeneous Vehicle Routing Problem: analyzing different fleet configurations". European J. of Industrial Engineering, 8(6): 762-788.

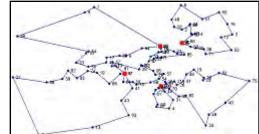
Juan, A.; Goentzel, J.; Bektas, T. (2014): "Routing Fleets with Multiple Driving Ranges: is it possible to use greener fleet configurations?". Applied Soft Computing, 21: 84-94.

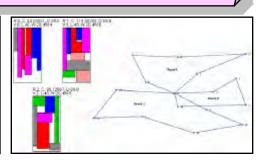
Juan, A.; Pascual, I.; Guimarans, D.; Barrios, B. (2014): "Combining Biased Randomization with Iterated Local Search for solving the Multi-Depot Vehicle Routing Problem". ITOR, DOI: 10.1111/itor.12101.

Dominguez, O.; Juan, A.; Nuez, I.; Ouelhadj, D. (2015): "An ILS-Biased Randomization algorithm for the Two-dimensional Loading HFVRP with Sequential Loading and Items Rotation". Journal of the Operational Research Society, doi: 10.1057/jors.2015.48.









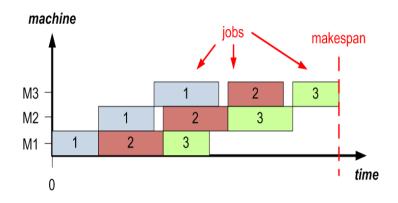
4. Selected published applications (2/2)

Networking & Scheduling Problems: Routing in sensor networks, PFSP, ...

Vilajosana, X.; Llosa, J.; Pacho, C.; Vilajosana, I.; Juan, A.; Vicario, J.; Morell, A. (2010): "ZERO: Probabilistic Routing for Deploy and Forget Wireless Sensor Networks". Sensors, Vol. 10, Issue 10, pp. 8920-8937.

Juan, A.; Lourenço, H.; Mateo, M.; Luo, R.; Castella, Q. (2014): "Using Iterated Local Search for solving the Flow-Shop Problem: parametrization, randomization and parallelization issues". Int. Transactions in Operational Research, 21(1): 103-126.



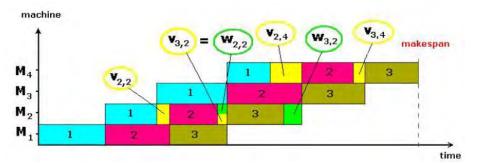


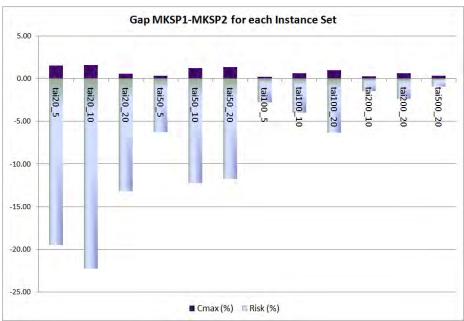
5. COPs with non-smooth objective functions

Scheduling problems with non-smooth objective functions

Ferrer, A.; Guimarans, D.; Lourenço, H.; Juan, A. (rs-150915): "Solving non-smooth flow-shop problems with failure-risk penalties using a biased-randomized iterated local search". Expert Systems with Applications.

TCost :=
$$C_{\text{mksp}} + \sum_{k \in \mathcal{M}} \Psi^k(v_{1,k}, \dots, v_{n,k}),$$

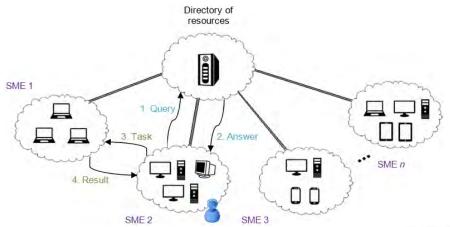




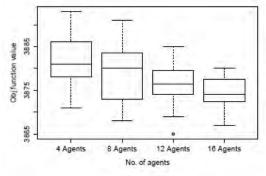
6. Multi-agent & web-based solving approaches

Multi-agent solving approaches to COPs

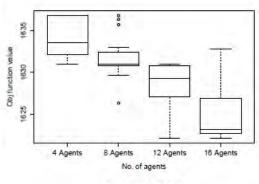
Martin, S.; Ouelhadj, D.; Beullens, P.; Ozcan, E.; Juan, A. (2nd review): "Multi-agent based Cooperative Search in Combinatorial Optimisation". European Journal of Operational Research.



Ruiz, X.; Calvet, L.; Ferrarons, J.; Juan, A. (2015): "SmartMonkey: a web browser tool for solving combinatorial optimization problems in real time". Proceedings of the 2015 Int. Conf. of the Forum for Interdisciplinary Mathematics (FIM2015). November 18-20. Barcelona, Spain.





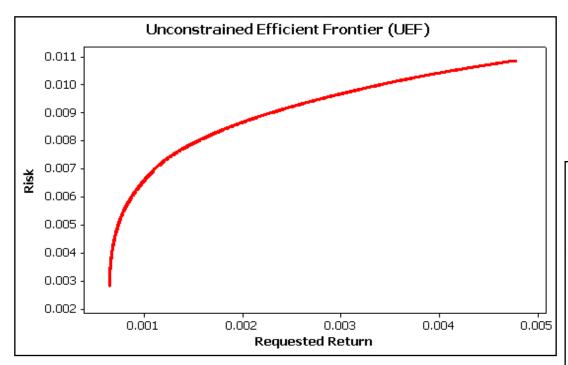


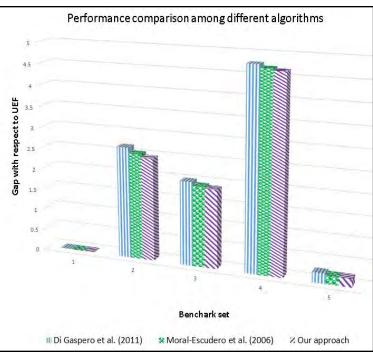
(b) A-n63-k9

7. Financial models

Financial Models: Rich Portfolio Optimization Problems

Kizys, R.; Juan, A.; Sawik, B. (s-150322): "ARPO: An Iterated Local Search Algorithm for Portfolio Optimization under Realistic Constraints". Quantitative Finance.

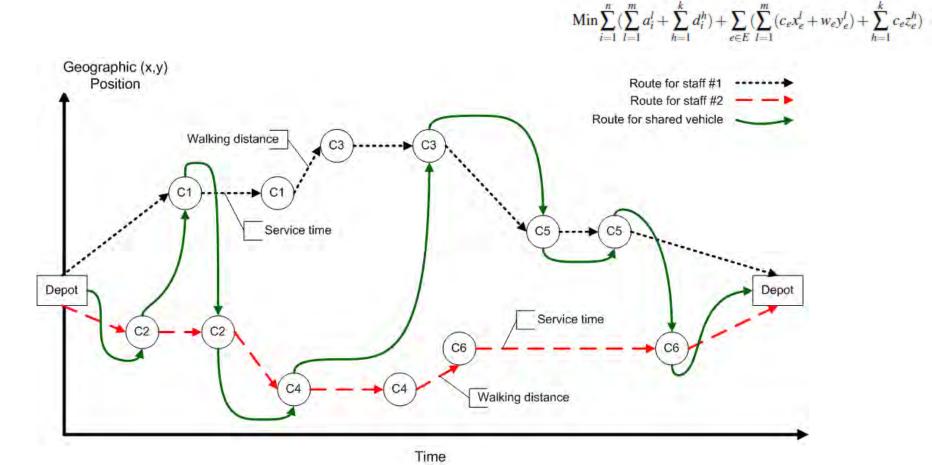




8. Synchronized VRPs

Rich VRPs with dynamic behavior.

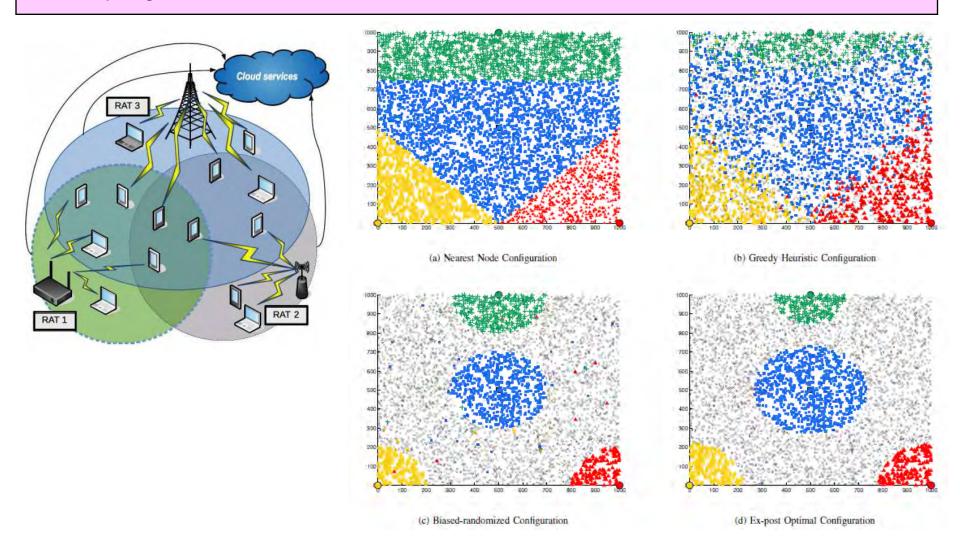
Fikar, C.; Juan, A.; Martinez, E.; Hirsch, P. (s-150518): "A discrete-event metaheuristic for dynamic home-service routing with synchronized ride-sharing". European Journal of Industrial Engineering.



9. Smart Cities

Cloud computing

Mazza, D.; Pages, A.; Tarchi, D.; Juan, A.; Corazza, G. (u.d.): "A social-aware biased-randomized algorithm for mobile cloud computing in smart cities".



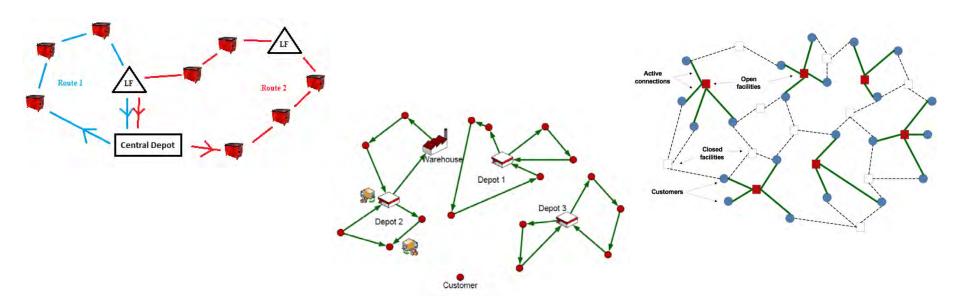
10. Other L&T problems

Waste collection management, rich IRPs, FLPs, ...

Gruler, A.; Fikar, C.; Juan, A.; Hirsch, P.; Contreras, C. (2015): "A Simheuristic for the Waste Collection Problem with Stochastic Demands in Smart Cities". Proceedings of the 16 ASIM Dedicated Conference on Simulation. Dortmund, Germany. September 23-25.

Pfeilsticker, L.; Juan, A.; Rabe, M. (2015): "Development of a Simheuristic Approach for Solving Realistic Inventory Routing Problems". Proceedings of the 16 ASIM Dedicated Conference on Simulation. Dortmund, Germany. September 23-25.

De Armas, J.; Juan, A.; Marques, J. (2015): "A Biased-Randomized Algorithm for the Uncapacitated Facility Location Problem". Proceedings of the 2015 Int. Conf. of the Forum for Interdisciplinary Mathematics (FIM2015). November 18-20. Barcelona, Spain.



11. Conclusions

- Biased randomization: by combining skewed probability distributions with direct Monte Carlo simulation, the logic behind the heuristic can be slightly randomized without losing it.
- BR allows to transform the deterministic heuristic procedure into a probabilistic algorithm that can be run several times.
- Advantages: easy-to-implement, few parameters, quiet flexible and efficient approach, real-time, ...
- Some applications to L&T, Production, Internet Computing, and Finances. The number of potential applications to these and other fields is really (really) large.











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THANK YOU!

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