



The Long (and Fun) Road That Led To DSMGA-II

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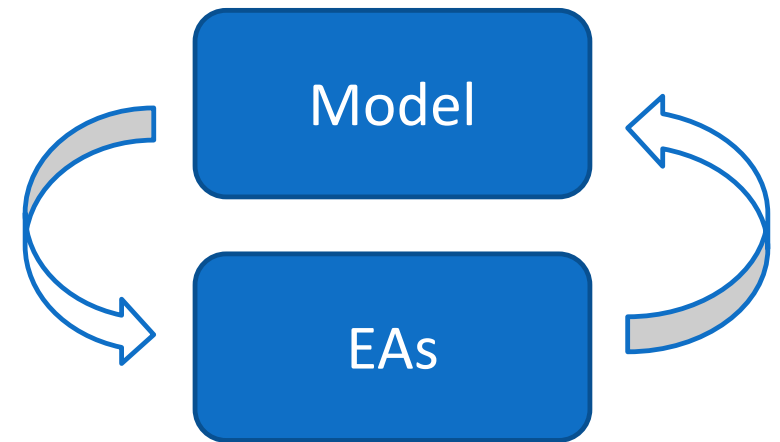
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Outline

- Dependency detection
- From fully separable to overlapping structures
- Linkage trees and optimal mixing
- DSMGA-II

EDAs and Model Based EAs

- Model based EAs
 - Not necessarily estimating the distribution
- A metric to detect dependencies
- A way to regularize the model complexity
- A way to utilize the model (effective and efficient mixing)



What is DSMGA (Yu *et al.*, 2003) ?

Pairwise Dependency
Detection

	A	B	C	D	E	F	G
A	●	●	●	●	●	●	●
B	●	●	●	●	●	●	●
C	●	●	●	●	●	●	●
D	●	●	●	●	●	●	●
E	●	●	●	●	●	●	●
F	●	●	●	●	●	●	●
G	●	●	●	●	●	●	●



Thresholding

	A	B	C	D	E	F	G
A	●	×			×	×	
B	×	●		×			×
C		×	●	×			×
D		×	×	●	×		×
E				×	●	×	
F	×			×	×	●	
G		×	×	×			●



Clustering

	A	F	E	D	B	C	G
A	●	×	×		×		
F	×	●	×				
E	×	×	●	×			
D			×	●	×	×	×
B				×	●		×
C				×	×	●	×
G				×	×	×	●



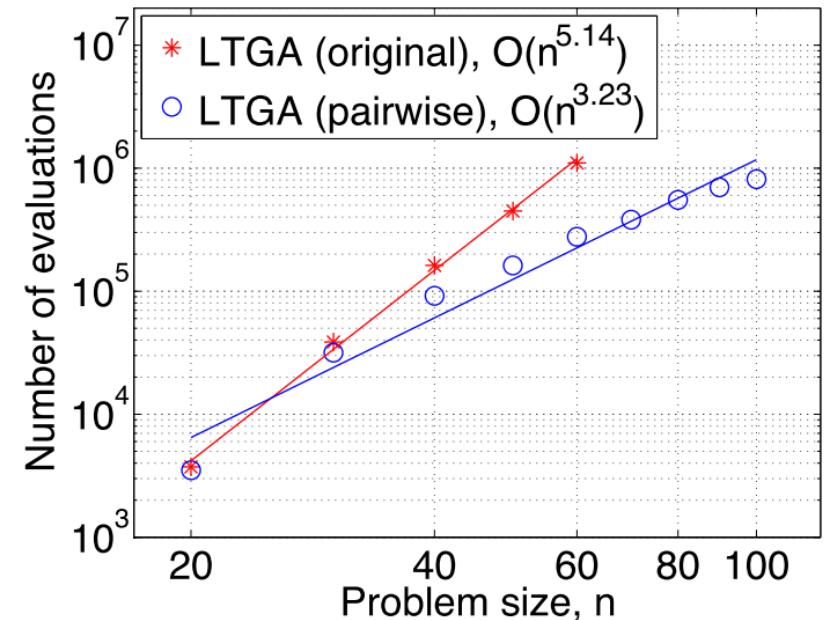
Recombination

[AFE]
[EDBCG]

- To me, DSMGA is more a decomposition of MBEAs than MBEA itself.
- So we can study each part independently.

Dependency-Detection Metrics

- Multivariate vs. Pairwise
 - Multivariate: eCGA, BOA, ...
 - Pairwise: DSMGA, current GOMEA, DSMGA-II
 - Pairwise prevents from overfitting, especially when the population is small.
 - If you need 30 samples for good statistics, detecting dependencies among 5 bits in the multivariate way requires 1000 chromosomes.



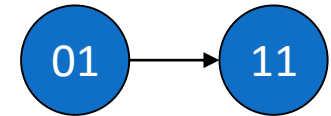
Pelikan *et al.*, 2011

Dependency Detection Metrics

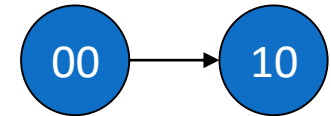
- Nonlinearity

- LINC, LIMD

$$|f_{00} + f_{11} - f_{10} - f_{01}|$$



- Simultaneity



- Differential mutual complement

$$p_{00} \cdot p_{11} \pm p_{01} \cdot p_{10}$$

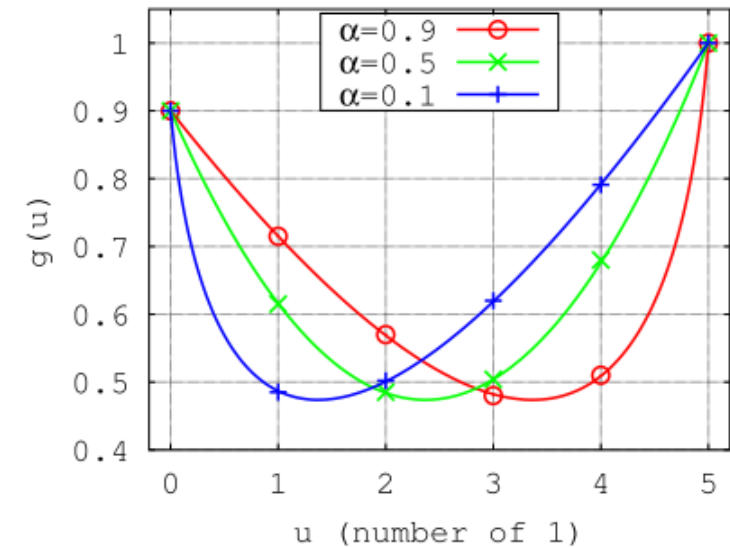
- Entropy-based (mutual information)

$$\sum_{xy} p_{xy} \cdot \log \frac{p_{xy}}{p_x p_y}$$

- ecGA, BMDA, BOA, GOMEA, DSMGA-I/II

How Do Metrics Differ?

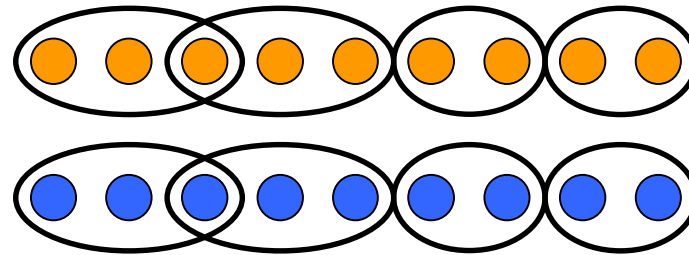
- Our work (Fan and Yu, 2011) indicated that all metrics are similar qualitatively.
- With optimum thresholds, all metrics performs equally on fully separable problems.
- Most MBEAs adopt MDL for thresholding, but not optimum.



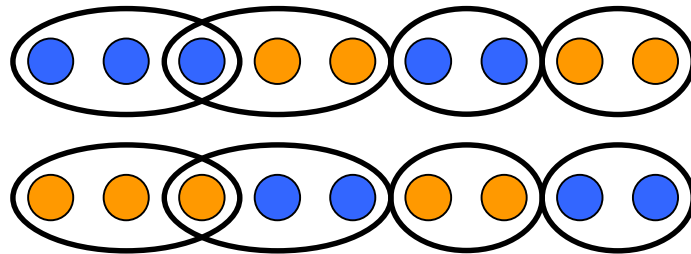
Bézier Trap

Mixing with Overlapping BBs

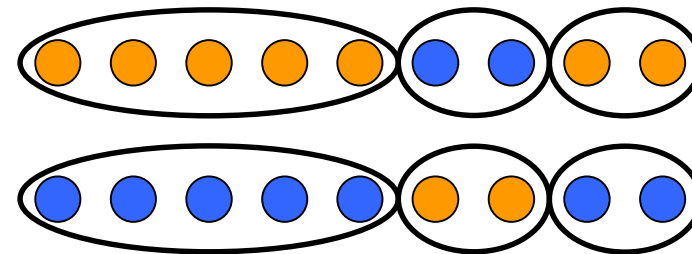
- To group or not to group?



Break one module



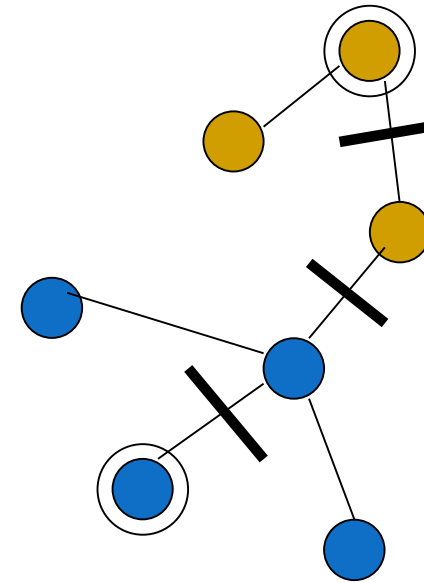
One big module



For problems with m BBs, you can at most break $O(m^{1/2})$ of them (Yu et al., 2004).

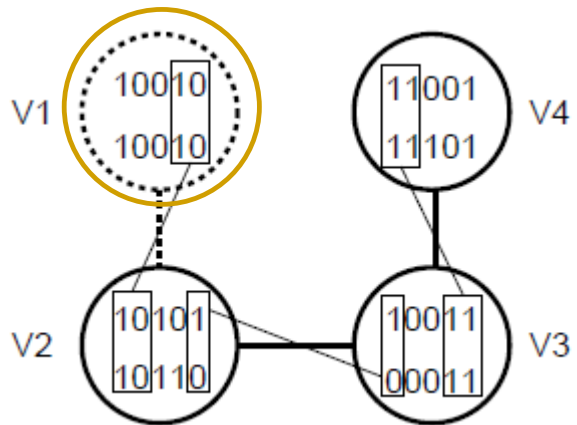
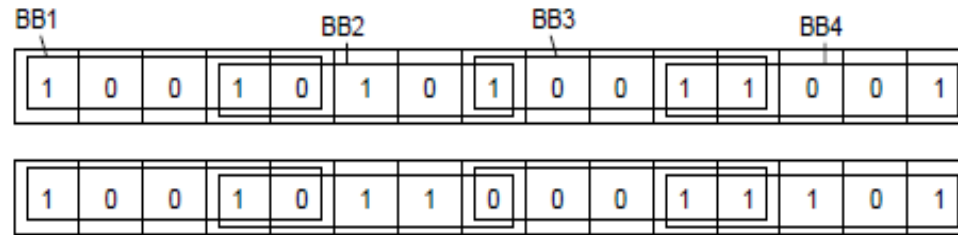
minCut *(Yu et al., 2005)*

- (1) Every BB needs to be recombined with non-zero probability.
 - (2) The recombination needs to be least disruptive.
- Build a BB graph.
 - Randomly choose two nodes and find the mincut that separates them.

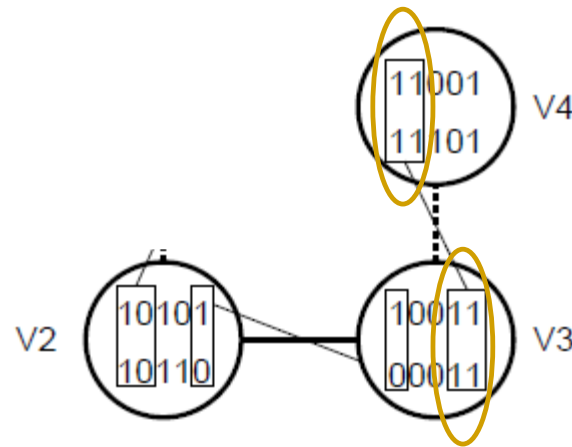


minCut+ (Tsuji *et al.*, 2006)

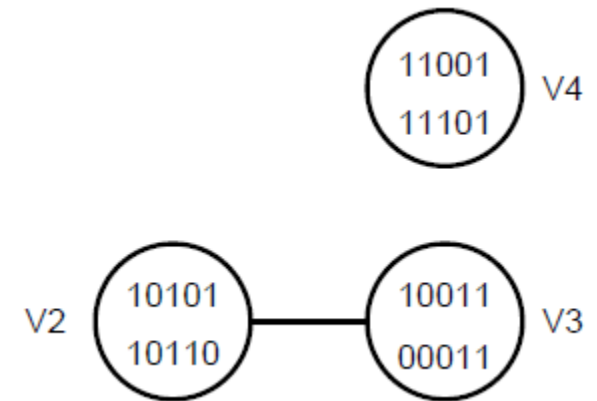
parents



Remove unnecessary nodes
(alleles in BBs are identical)



Remove unnecessary edges
(alleles in overlaps are identical)



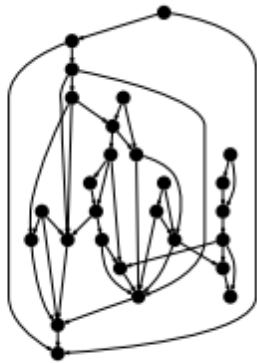
resulted graph

The Threshold and Bayesian Ways...

- For problems with overlapping structures, using thresholds is dangerous.
- Bayesian way is not easily comprehensive for humans.



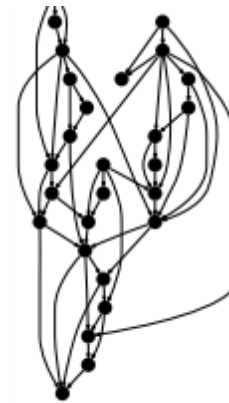
(a) Gen:1



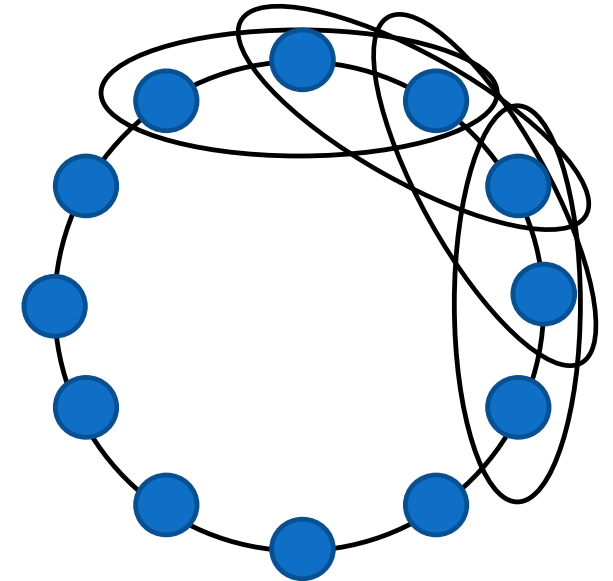
(b) Gen:3



(c) Gen:5



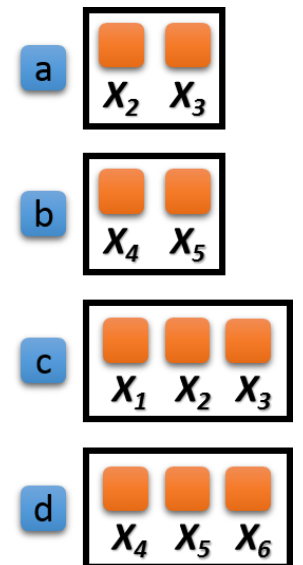
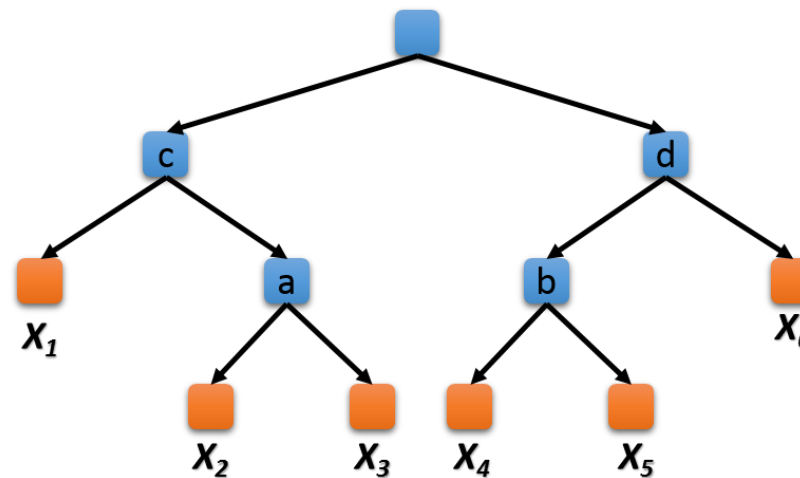
(d) Gen:7



In a cyclic topology, all genes depend on each other.

The Linkage Tree Way and OM (Thierens and Bosman *et al.* 2010~)

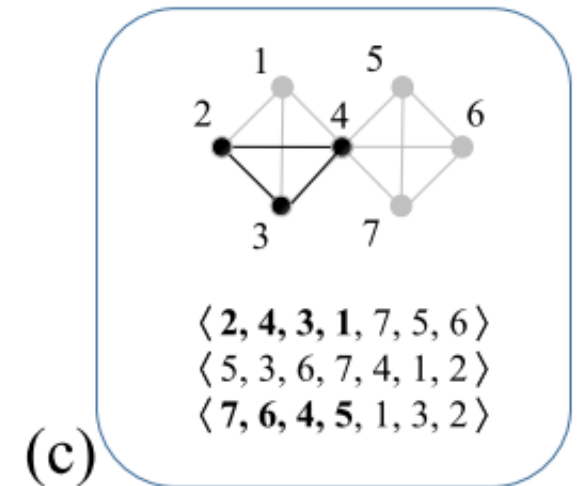
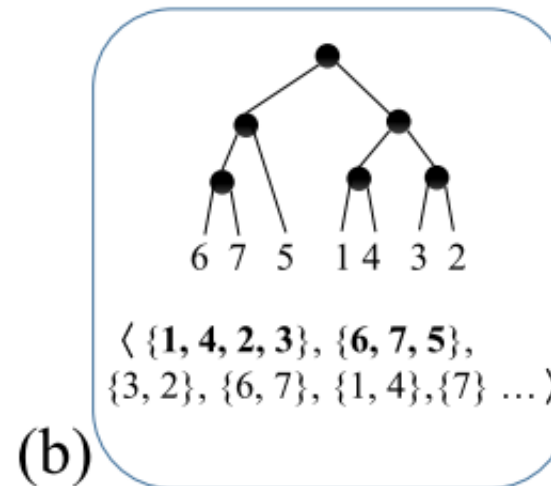
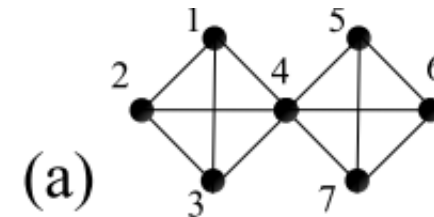
- Does not need any threshold when identifying BBs
- OM : noise-free decision making
- Population sizing
 - Supply: $2^k \log m$
 - Decision making: $2^k m^{0.5} \log m$
 - Model building: $2^{2k} m \log m$



Incremental Linkage Set (ILS)

$\{2\}, \{2,4\}, \{2,4,3\}, \{2,4,3,1\}, \{2,4,3,1,7\}, \dots$

- Starting from one gene, incrementally add the next most dependent gene one-by-one.
- More expressive

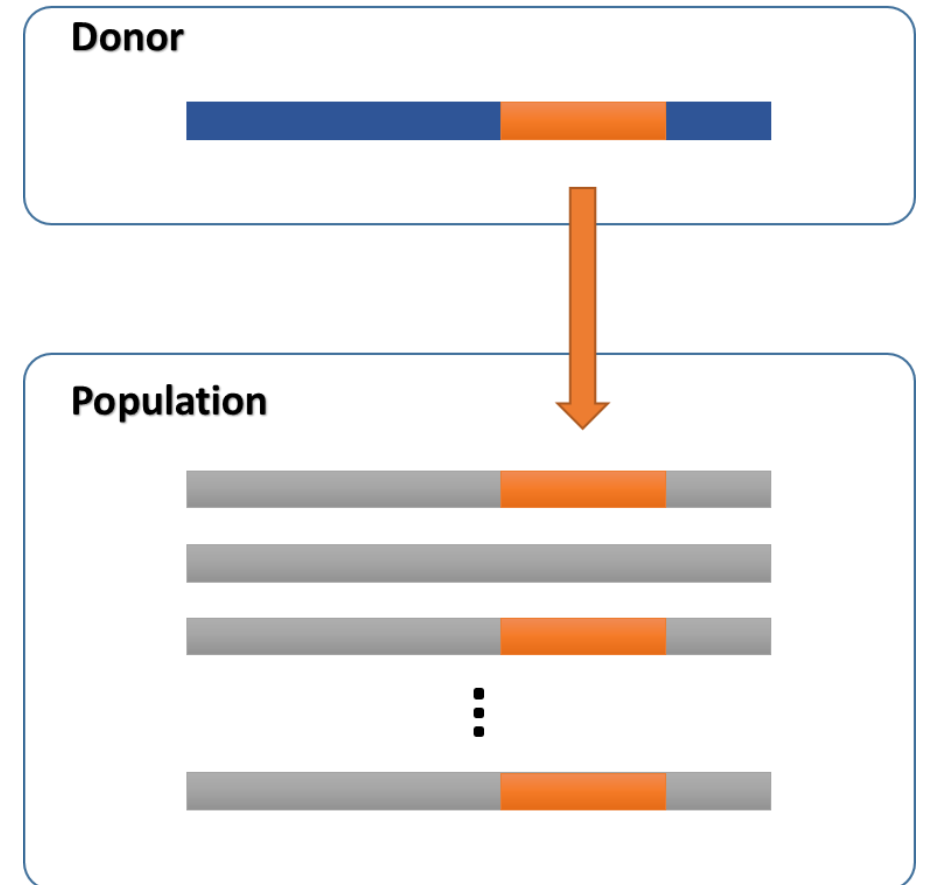


Restricted Mixing (RM)

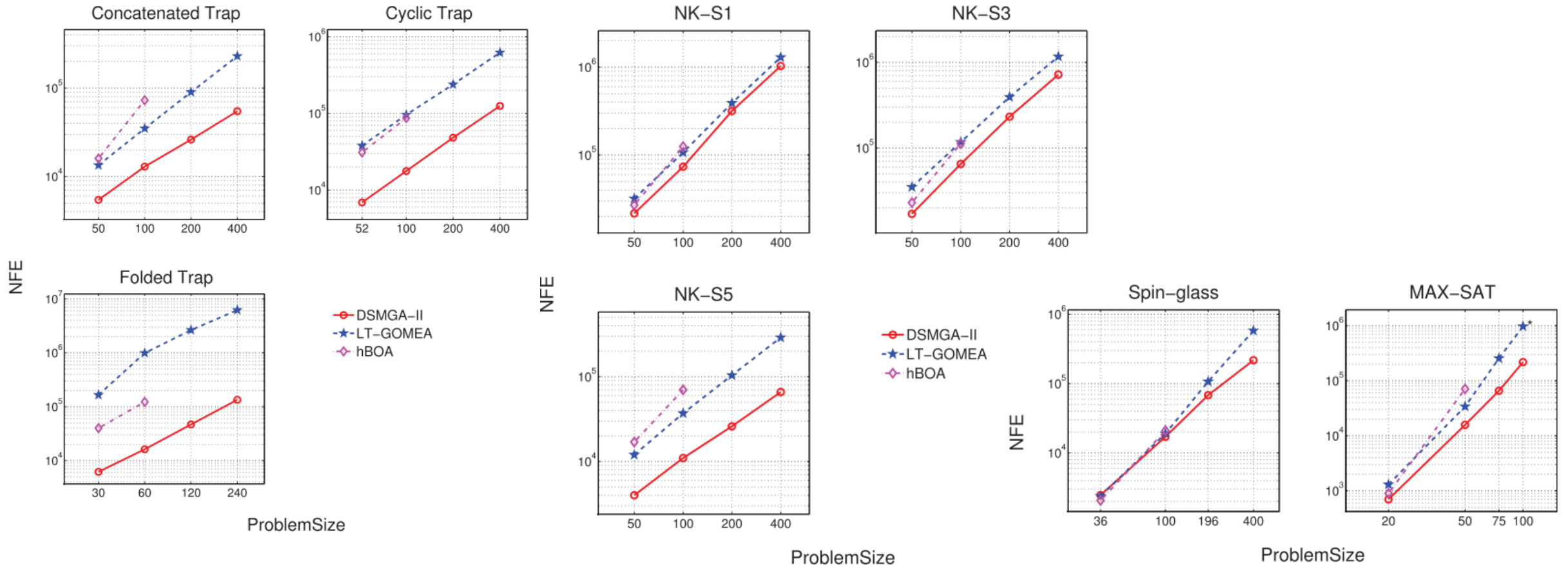
- All genes in the mask are altered
 - Same effect as restrictedly choosing the donor
- RM does not explicitly require a donor. Instead, it adopts **supply check**:
 - Masks stop growing when the corresponding patterns are not in the population.
- More efficient
 - Every different mask results in different chromosome.

Back Mixing (BM)

- Successful RM → good pattern
- Use that pattern to OM all chromosomes in the population
- Losing diversity quickly
- Seems risky, but actually safe
- Refining DSM

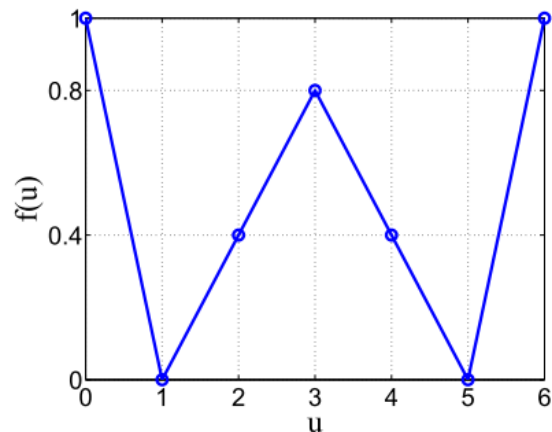


DSMGA-II Performances (Hsu and Yu, 2015)

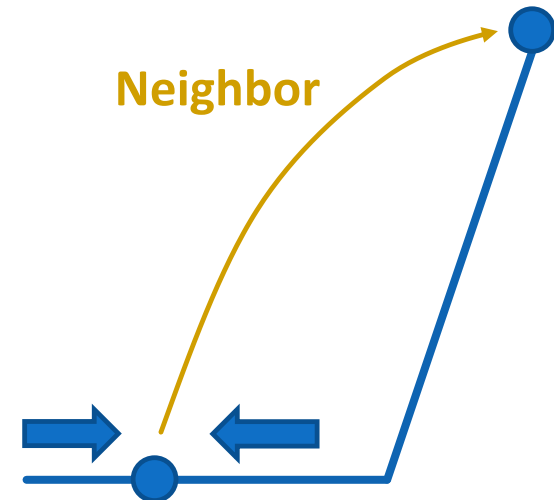


Problems with Plateau

- DSMGA-II does not actually explore plateau.
- Instead, back mixing shrinks the plateau into one point, and then neighborhood is established from that point to the optimum.



Folded Trap



Conclusion

- Pairwise detection is more robust than multivariate.
- Using threshold is dangerous for problems with overlapping structures.
- More expressive linkage models may result in more function evaluations without proper ways to reduce trials.
- Losing diversity is not always bad for EAs.
- OM seems nothing more than a hill climber at the first glance, but with adaptive neighborhood, it is powerful.
- Finally, we need more theoretical supports.