



Ethane: A Heterogeneous Parallel Search Algorithm for Heterogeneous Platforms

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Introduction

- Metaheuristics provide a **fast and efficient** way for **solving problems**, and in many cases **parallelism** is necessary. There exists many parallel models.
- In a common modern lab we can usually find many **different hardware architectures**.
- It has been proven that **heterogeneous resources** can be used efficiently with **standard parallel models**, but few works exist about the design of a **specific parallel model** for such a heterogeneous environment.
- We present **Ethane**, a **heterogeneous parallel search** algorithm designed for its execution on **heterogeneous hardware** environments.
- We also propose **HydroCM** (HydroCarbon inspired Metaheuristics), a general schema based in hydrocarbons for shaping a **family of parallel heterogeneous metaheuristics**.



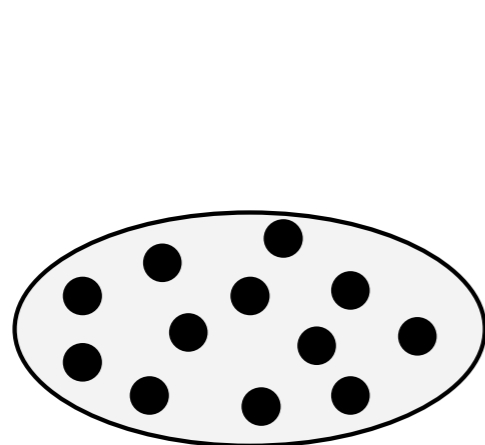
Decentralized and Heterogeneous PMs

- There exist **many parallel implementations** of metaheuristics.
- We will focus in two of the more common families of metaheuristics: **Evolutionary Algorithms** (EAs) and **Local Search Metaheuristics** (LSMs):
 - **Evolutionary Algorithms**: EAs are **population based** methods; a random population is enhanced through a **Nature-like evolution** process. EAs provide a good exploration of the search space, they are also called **exploration-oriented** methods.
 - **Local Search Metaheuristics**: LSMs use only **one candidate solution** which is enhanced by moving through its neighborhood replacing the candidate solution. LSMs are able to find local optima so they are called **exploitation-oriented** methods.
- Many different parallel models have been proposed for each method, we present the more representative ones.

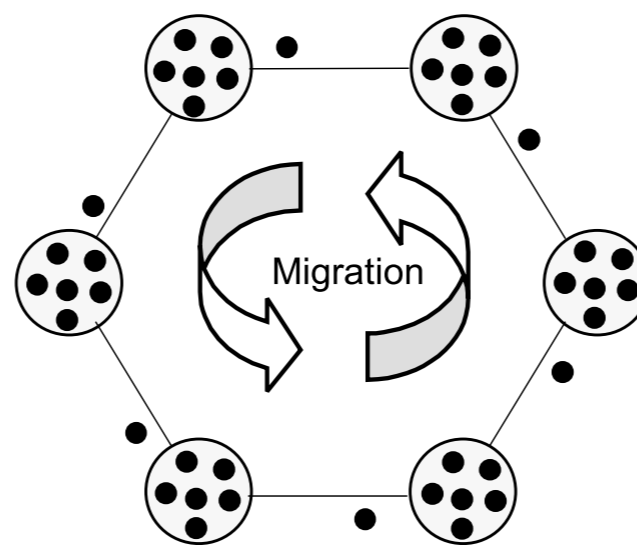


Parallel EA Models

- There exist many parallelization schemas for EAs, but we are going to focus in **structured populations**, which leads to a distinction between:
 - **Distributed EAs**: the population is divided into a number of **islands** running an **isolated instance of the EA**. Some individual are exchanged following a **migration** scheme. Few sub-algorithms, loose coupling.
 - **Cellular EAs**: there exists **only one population** structured into neighborhoods. An individual can only **interact with its neighbors**. The structure of the neighborhood leads to different behaviors. Many sub-algorithms, tight coupling.

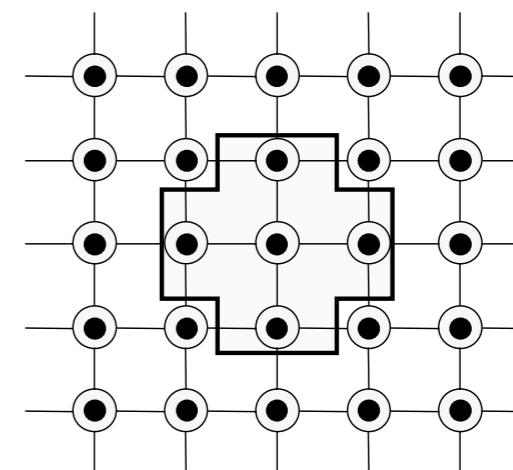


(a)



Workers

(b)

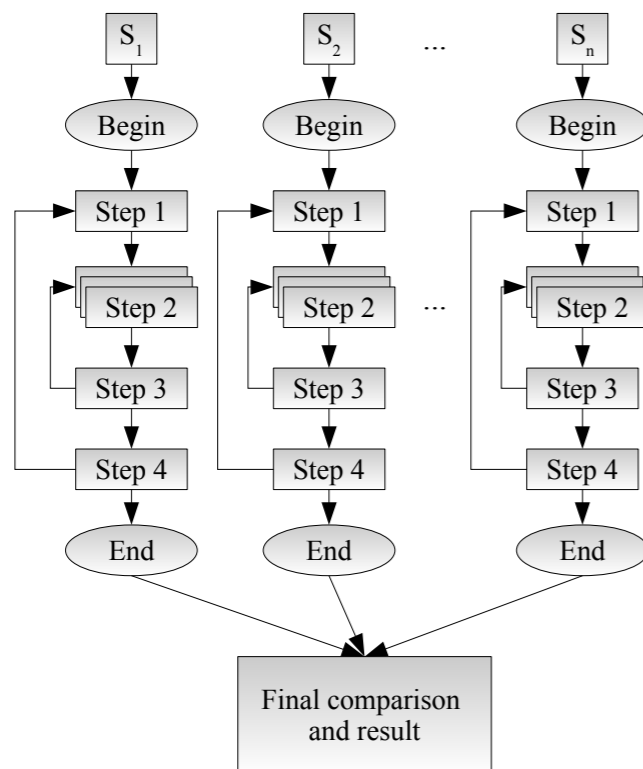


Workers

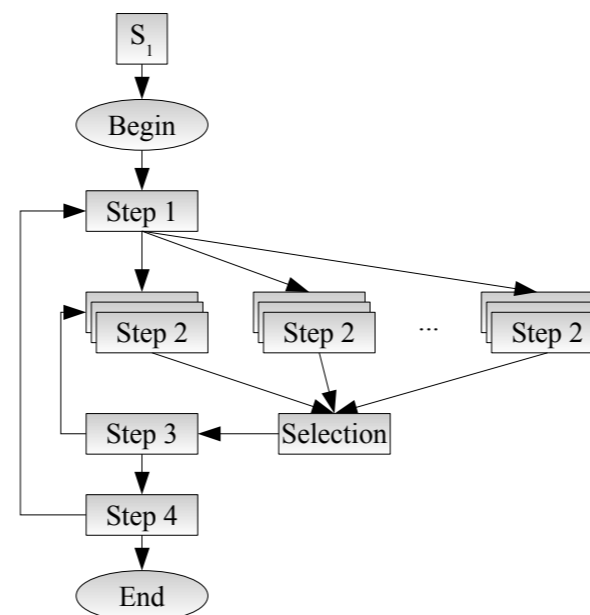
(c)

Parallel LSM Models

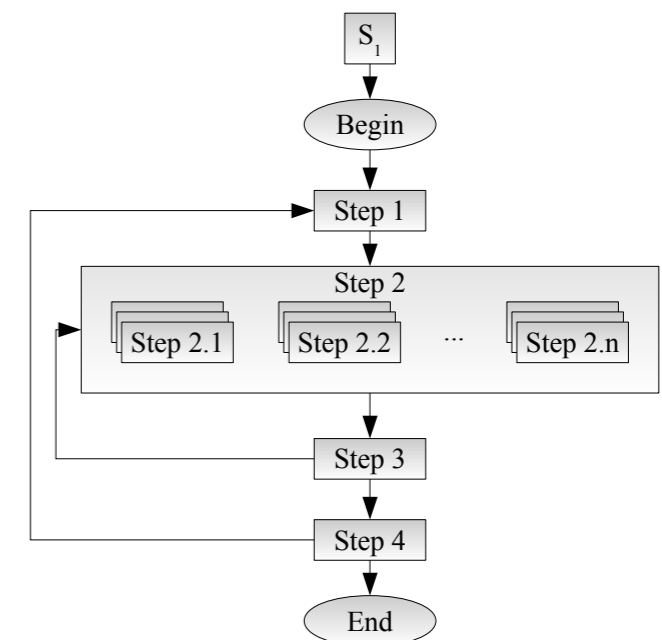
- Three parallel LSM models are specially extended in the literature:
 - **Parallel multistart model**: several **independent instances** are run simultaneously.
 - **Parallel moves model**: a kind of master-slave architecture where all the **slaves perform a move** and the **master selects one** of them.
 - **Move acceleration model**: The quality of each candidate is **evaluated in a parallel centralized way**.



(a)



(b)



(c)

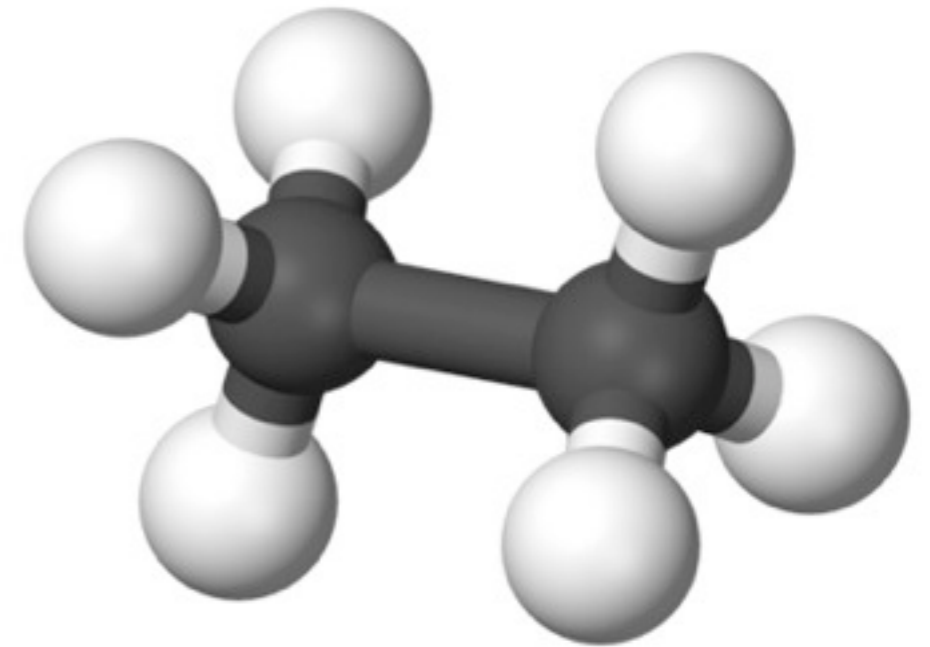


Achieving the Heterogeneity

- We can **modify the behavior** of an algorithm by changing the search features between sub-algorithms, obtaining what we call a **parallel heterogeneous metaheuristic**.
- We can classify parallel heterogeneous metaheuristics in **four levels** depending on the **source of heterogeneity**:
 - **Parameter level**: the same algorithm is used in each node, but **different parameters** are used in one or more of them.
 - **Operator level**: heterogeneity is achieved using **different mechanisms** for exploring the search space, such as different operators.
 - **Solution level**: a **different encoding** is used for the solutions in each node.
 - **Algorithm level**: each component can run a **different algorithm**.
- The algorithm proposed in our work is an **algorithm level parallel heterogeneous metaheuristic**, which is **based in ssGA and SA**.

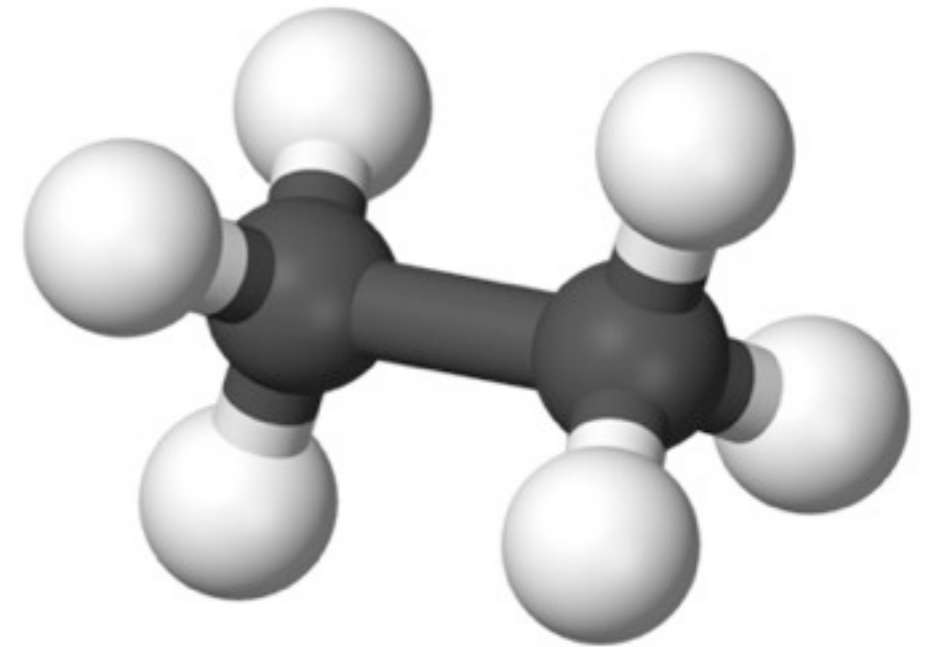
Description of the Model

- Ethane is a **Nature inspired heterogeneous parallel search algorithm** designed for its execution in a **heterogeneous hardware** platform.
- We propose a **communication schema** where most of the communication load lays over the fastest nodes, and the slowest ones are placed as *slaves*.
- Ethane is inspired by the **chemical compound** of the same name, which consists of 6 hydrogen atoms and 2 carbon atoms, where carbon atoms are bond together and 3 hydrogen atoms are bond to each of them.
- We propose the same schema using two basic **algorithms** resembling different **atoms**, and **communication channels** resembling **chemical bonds**.



Description of the Model

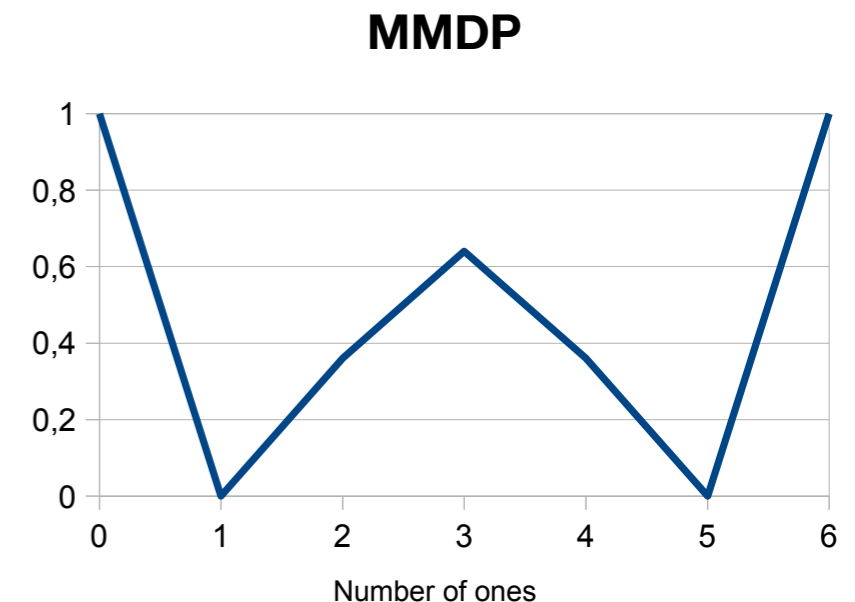
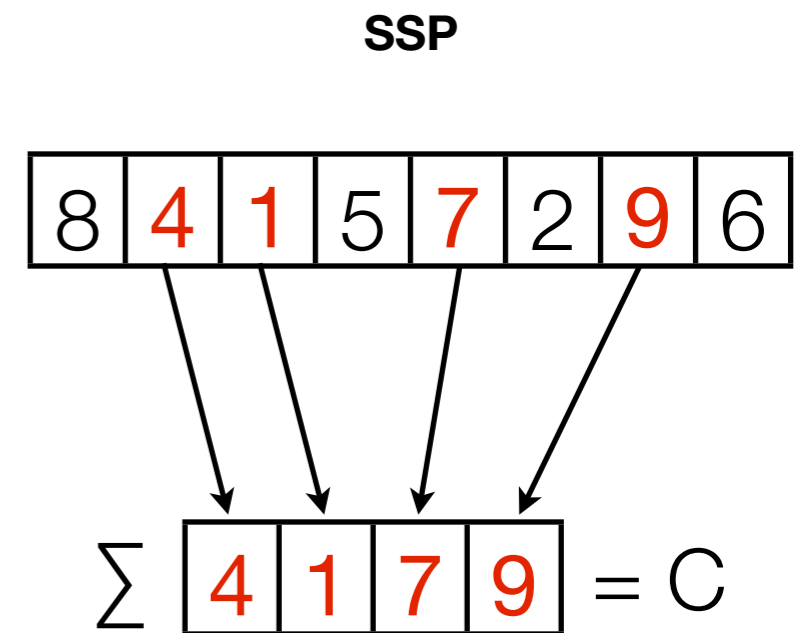
- We have implemented two **instances of Ethane**:
 - **Ethane G**: **ssGA** sub-algorithms are assigned to **central nodes** and SA islands are assigned to the *slave* nodes.
 - **Ethane S**: **SA** islands are placed in the **central nodes** while **ssGA** components are placed as *slaves*.
- With this schema the most of the communication falls on the **central nodes**, which **are provided with the fastest hardware**, moving some of the load out of the slowest nodes.
- From this inspiration it arises **HydroCM** as a **generic model** for a complete family of **parallel heterogeneous metaheuristics** based on the different structures of **hydrocarbons**.





Benchmark Problems

- We have used two problems in the analysis: the **Subset Sum Problem** (SSP) and the **Massively Multimodal Deceptive Problem** (MMDP) with 6 bits.
 - **SSP**: consist in **finding a subset of values** from a set of integers such that their sum **approaches a constant C without exceeding**. Our instance consists of 2048 random integers in the range $[0...10^4]$ from a Gaussian distribution.
 - **MMDP**: is a **deceptive problem** (make the algorithm converge to wrong regions of the search space). A binary string encodes **k 6 bit sub-problems** which contribute with a partial fitness depending on **its number of 1's** (unitation). We have used an instance with 150 bits so that the global optimum is $k = 25$.





Parameters and Platform

- For every **ssGA** population, the parameters are: a size of **64** individuals, a crossover probability of **0.8** and a mutation probability of **4** divided by the chromosome length.
- For **SA** we used the same mutation probability.
- We have chosen a **migration frequency of 50** for all the models, migrating **1 individual** in all cases.
- We have used **8 different machines**:
 - **2** Intel Core 2 Quad Q9600 @ 2.66GHz with 4GB of RAM.
 - **6** Intel Pentium 4 @ 2.4GHz with 1GB of RAM.
- All of them are managed by a **GNU/LINUX OS**, but different distributions. Since the algorithms are **implemented in JAVA**, all of them run the 1.6.0_0 version of the Java Virtual Machine (JVM).



Tests and Analysis

- For our analysis we have used the presented instances of **Ethane**, a **ssGA unidirectional ring** (with the computers placed in a kind of MaxSumSort), and the **panmictic** versions of **SA and ssGA**.
- We have analyzed three performance measures, being **numerical effort**, total **run time** and **speedup**, as well as the **evolution of the fitness**.
- As we do not know the statistical distribution of the data, they have been statistically compared with **Mann-Whitney U test**.



Numerical Effort

Algorithm	SSP		MMDP6	
	Average	Std. Deviation	Average	Std. Deviation
Ethane G	146418	174433	1572735	919691
Ethane S	202815	198696	708231	430353
ssGA Ring	214824	239125	786583	805837
Panm. ssGA	179792	175177	*	*
Panm. SA	81737	93627	*	*

- Ethane performed **better than the panmictic algorithms** for both problems. For the SSP, both Ethane versions performed numerically **better than the reference** ssGA ring, and one of the instances (Ethane S) performed better even for the MMDP.
- Also the standard deviation is better in our algorithms and so their behavior is **more robust**.
- All the differences in behavior are **statistically significant**.



Total Run Time

Algorithm	SSP		MMDP6	
	Average	Std. Deviation	Average	Std. Deviation
Ethane G	5318	6226	9195	4942
Ethane S	7155	6922	3052	1546
ssGA Ring	7453	8107	3194	3380
Panm. ssGA	30008	29387	*	*
Panm. SA	13300	15443	*	*

- Ethane clearly **reduced the time** needed to find the solution with respect to the **panmictic algorithms** for both problems. For the SSP, both Ethane versions also performed **better than the reference** ssGA ring; for the MMDP Ethane S outperformed the reference model.
- Again, the standard deviation is better in our algorithms, showing a **more stable** behavior.
- All the differences in the behavior are **statistically significant**.



Speedup

- The **speedup** represents the ratio between sequential and parallel average execution times. The algorithms run in the **single and multi-processor** platform must be exactly the same, and the reference processor is **the fastest one** (standard defined for heterogeneous systems).

Algorithm	SSP		MMDP6	
	Avg. time	Speedup	Avg. time	Speedup
Ethane G	15995	3.00x	41943	4.56x
Ethane S	17817	2.49x	20627	6.76x
ssGA Ring	18137	2.43x	21227	6.64x

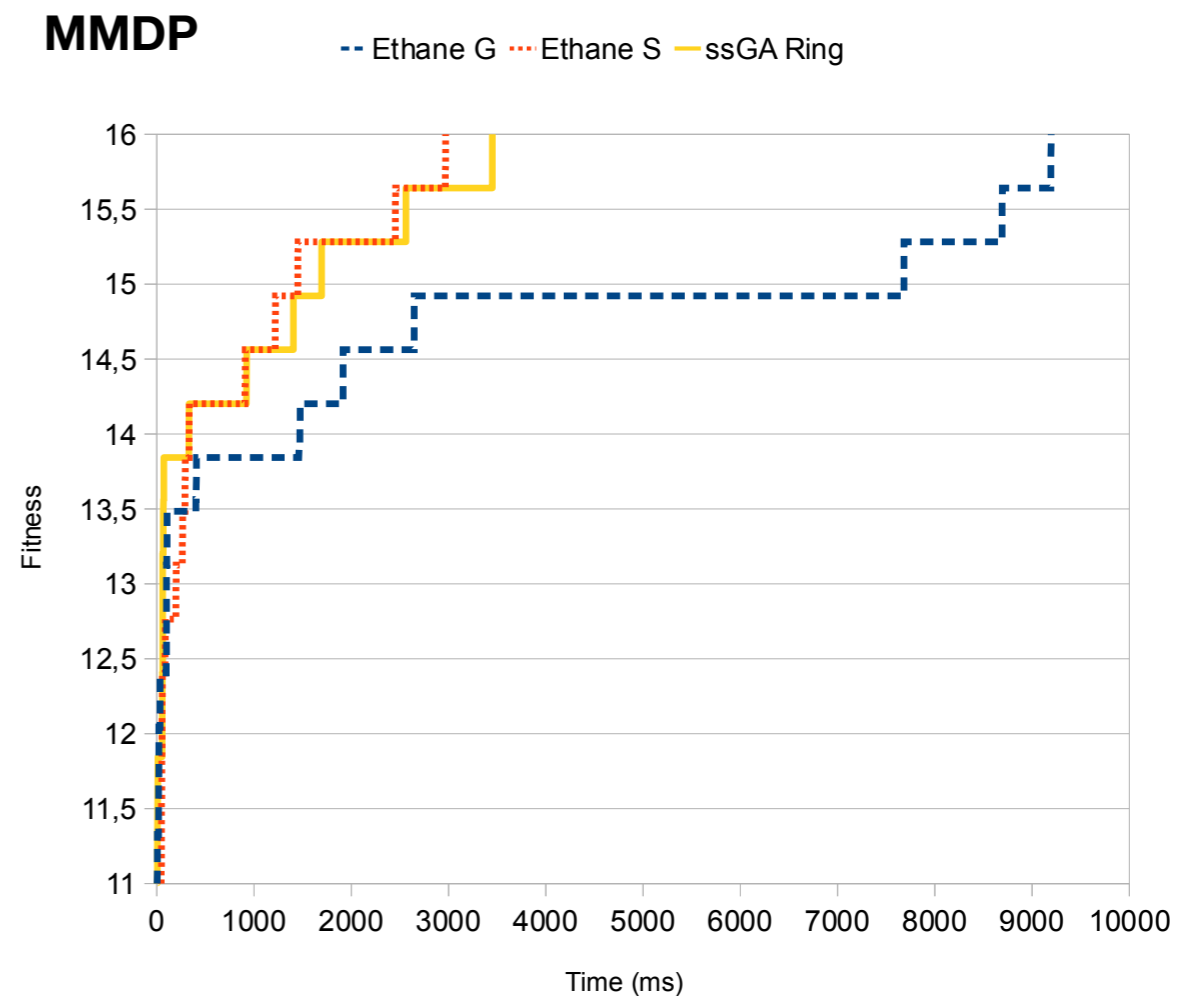
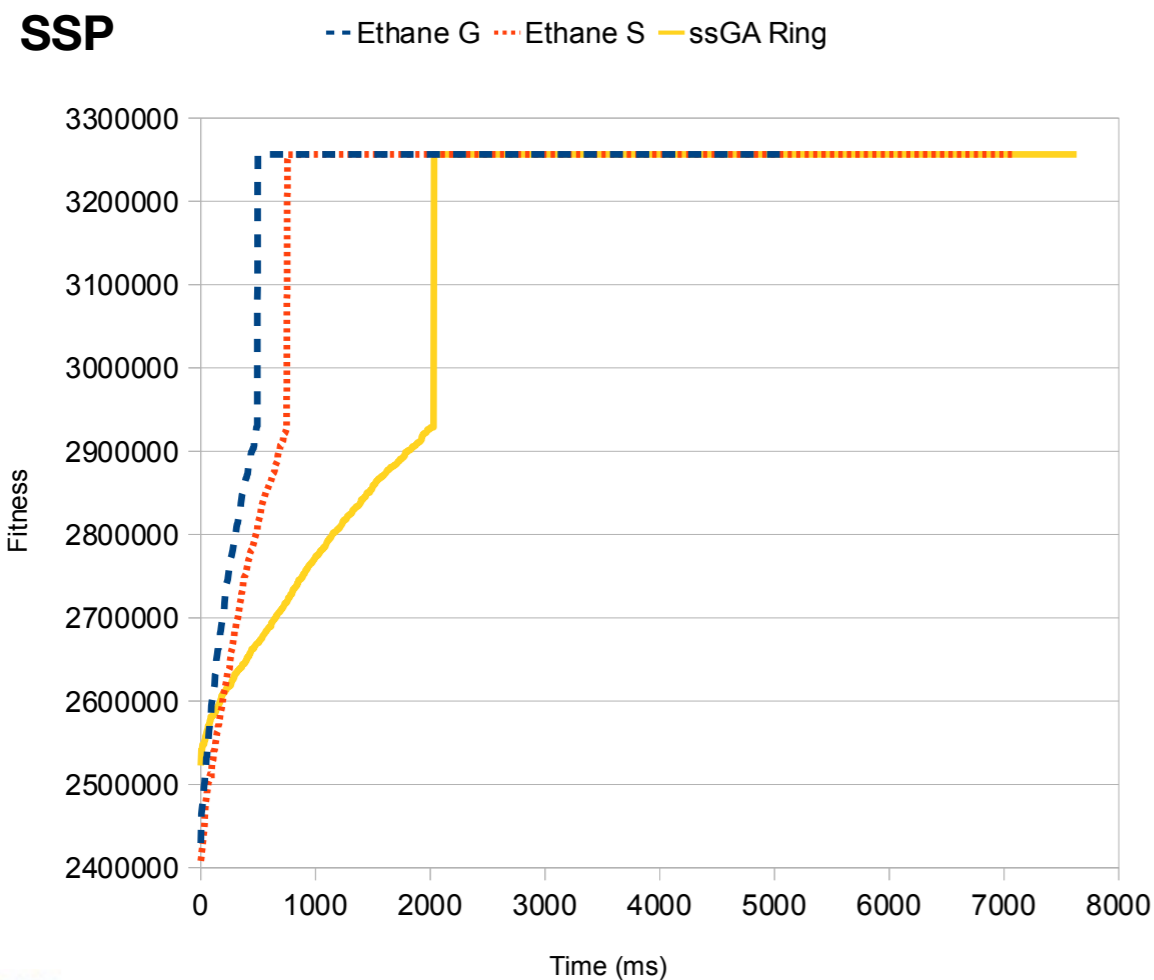
- Both versions of Ethane have obtained a **better speedup** than the ssGA for the SSP, but only Ethane S has achieved a better speedup for the MMDP.
- The speedup for the MMDP was **quite good** although lineal speedup was not achieved. In the case of SSP, Ethane G and S have **not showed a very good speedup**, and ssGA has showed even a **worse speedup**. This fact could be explained by the **huge difference** among the power of the different hardware configurations used.





Evolution of the Fitness

- In the case of **SSP** both versions of Ethane clearly **outperforms the ssGA ring**.
- For **MMDP**, Ethane S performed better than Ethane B, and was even able to **converge faster than the ssGA ring**.





Conclusions and Future Work

- We have presented a new **heterogeneous parallel search algorithm** based on the structure of **ethane**.
- We have also shaped a **general heterogeneous model** based on the structure of **hydrocarbons**.
- The tests have shown that our model **outperforms the reference ssGA ring** in terms of **run time** and **numerical effort**, and is even able to find solutions in a **more robust manner**.
- As future work we propose to **extend the HydroCM model**, as well as to **assess its performance** with different configurations and real-life applications.
- Our goal is to turn an often undesired heterogeneous platform into a **useful platform for (specialized) algorithms**.



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Thanks for your attention. **Questions?**