# Finding the best network for laying renewable energy based solar panel roads: A GPU parallel algorithm implemented on CUDA

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*Abstract*— The objective of this paper is implementation of ACO algorithm on GPU to combat real life problems of road network identification along with an application focusing on renewable energy. GPUs are specialized microprocessors that accelerates graphics operation. Parallel processing is required when we consider a heavy code with so much of similar iterations. CUDA is NVIDIA's architecture for parallel computing that is used for extensive parallel computing and increases the performance by employing the GPU (Graphical Processing Unit). We have Ant colony optimisation algorithm implementation that is a bit different than others. Also, we compare it with the sequential code and the results are that it is very fast as compared to sequential code. To deal with the execution of optimisation we will propose two different approaches, one will be the serial approach of the ACO algorithm to generate the network and other will be GPU / CUDA based approach. We will compare the execution time in both the cases and then find out the speed up. An applicability of this approach is for generating the best possible road network for city coordinates where we try to get the network with least cost. This is of immense applicability for developing countries where road networks are upcoming.

Keywords— CUDA, GPU, Parallel Processing, travelling salesman problem, Road network identification

# I. INTRODUCTION

With growing population and ever-increasing energy demands it is very important that we start generating enough energy for our needs without dependency on the fossil fuels to preserve our future, thus there is a lot of scope for the renewable energy. Large expanse of land available for China and India along with abundance of sunlight throughout the year in the subcontinent, there is opportunity for developing a road network for all the cities and small towns especially in India, with these roads being of special solar panels which have been tested successfully on one mile stretch in China. Thus these roads can be a source of electricity for us.

The scope of the paper is limited to the computational and algorithmic aspects of this idea. For finding the best route possible between the cities as explained above for the road network of solar panels which will allow the transportation over it, we need to find the route joining all the points of interest. Thus, the problem can be reduced to that of a basic case of travelling sales man problem. Now there can be various approaches to solve the travelling salesman problem, one such approach is Ant colony optimisation.

Using this optimisation, we will be able to find the network we desire.

To deal with the execution of the optimisation we will propose two different approaches, one will be the serial approach of the ACO algorithm to generate the network and other will be GPU / CUDA based approach. We will compare the execution time in both the cases and then find out the speed up for both the cases.

Section II of this paper gives details about the related work which has been done on this topic, Section III explains the methodology we adopted and the algorithm used, Section IV mentions the results and the sections V gives the conclusion and also the future work which can be done on this topic.

# **II. RELATED WORK**

# AN EFFICIENT GPU IMPLEMENTATION OF ANT COLONY OPTIMISATION FOR THE TRAVELLING SALESMAN PEROBLEM

This paper talks about the sophisticated and clean ACO approached on GPU. They have presented an efficient method for selecting random cities by number of ants. The method uses iterative random trial which is useful to find next city in highly efficient and with low computation cost. They have considered many programming issues of the architecture of GPU such as shared memory conflicts. Also, they have introduced a new method with stochastic trial in the roulette-wheel selection.

Title	Author	Journal	an	d Key concepts	Advantages	Disadvantages	Future
		DATE					enhancement
Improving	Lawrence	IEEE	congres	s Implementation	New parallel	Roulette wheel	A more compact
Ant Colony	Dawson	2013	-	of tour	implementation	selection method	and enhanced
Optimization	Iain Stewart			construction	executed upto	cannot be used	version of
performance				using roulette	82x faster while	on minimization	roulette.
on the GPU				wheel selection.	not changing the	problems.	
using CUDA				Proposed a new	quality of tour	1	
0				parallel	constructed. And		
				implementation	almost 8.5 times		
				of roulette wheel	more than GPU		
				which DS	existing parallel		
				roulette which	implementation		
				reduces running	mprementation		
				time of tour			
				construction			
The GPU-based	Rafal	2016		Proposal of three	Parallel ACS on	Selective	Algo should be
Parallel Ant	Skinderowicz	2010		parallel versions	Nvidia Kepler	pheromone is	tested by using
Colony System				of Ant colony	GK104 is able to	costly and	new generation
colony system				system (which is	obtain speed up	complex.	of GPU's
				similar to ACO	of $25x$ vs	eompieni	01 01 0 0.
				and MMAS).	sequential ACS		
				The two of them	while in case of		
				uses standard	selective		
				pheromone	pheromone it		
				memory and the	comes about to		
				third one uses	be 17x		
				selective			
				approach.			
				upprouen			
Accelerating	Laurence	IEEE co	ongress	Implements a	First parallel	Result in some	Parallel
ant colony	Dawson Jain	2014	JII 51 000	novel data parallel	implementation	drawbacks like	implementation
optimization-	A Stewart	2014		approach that maps	of an ACO based	broken edges	of other such
based edge	11 Stewart			individual ants to	edge detection	bioken euges	thing based on
detection on				thread wrans GPU	eage detection.		
the GPU				is used to reduce			neo
using CUDA				number of			
using CODIT				iterations			
Novel Method to	Mr A P	Internationa	1	Implementation of	narallel	Implemented	Improvement in
Improve $\Delta CO$	Pandel Mr	Iournal	u of	data-parallel GPU	accomplishment	hoth the	SS required since
Performance on	B S Patil	Innovative	01	execution of the	executes up to 8-	construction of	Schedule
the GPU Using	$Mr \qquad A I$	Research	in	ACO algorithm to	12x faster than	Schedule and	construction
CLIDA for Nurse	Patil	Computer	and	solve nurse roster	sequential	pheromone	nhase uses a new
Roster Scheduling		Communice	ation	scheduling	execution at the	undate phases on	efficient
Problem		Engineering	,	nrohlem	same time as	the GPU	execution of
		Lingineering		P. Solom.	preserving the		roulette wheel

quality of

Schedules formation.

the

# Table 1: Papers on ACO and CUDA

March 2016

selection.

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A REVIEW ON	Stephy Johny	International Journal	idea is to	The renewable	As it requires	Solar roadways
SOLAR	and Keerthi	of Recent Innovation	replace the	energy	huge initial	will solve the
ROADWAYS: THE	Susan John	in Engineering and	asphalt roads	generated by	investment, it	problems of usage
FUTURE OF		Research	with solar	solar road	would be	of fossil fuels and
ROADS			roadways on	panels will	difficult to	energy
		Volume: 02 Issue: 03	our streets,	replace the	install solar	consumption.
		March- 2017	highways	current need for	roadways in	The future work
		(IJRIER)	parking lots	fossil fuel	developing	involves making
			and sidewalks	which is used	countries.	it possible in real
			that collect	for generation		life.
			solar energy to	of electricity		
			be used by our	which in turn		
			homes and	can reduce the		
			businesses.	greenhouse		
				gases nearly to		
				half.		

# COMPUTE UNIFIED DEVICE ARCHITECTURE (CUDA)

CUDA is NVIDIA's architecture for parallel computing that is used for extensive parallel computing and increases the performance by employing the GPU (Graphical Processing Unit). CUDA is identifying its use in various branches which includes image and video processing, simulation of fluid dynamics, seismic analysis, ray tracing and many more.

CUDA programming has a hierarchy of thread groups called block, grid and thread. A grid is divided into some number of blocks and in each block, there consist an equal number of threads.

CUDA C extends C language by allowing the developer to declare C function. They are called as kernels. When kernel is involved all blocks, which are there in the grid are allocated to the running processor and threads in each block is executed by the cores present in the running processor.

Steps:

1. Initialise the ants and cities, get the coordinates of the cities.

**III.** METHODOLOGY

- 2. Initialise required matrices
- 3. Random assignment of cities to ants
- For (pheromone update < max iterations) {
  - 4. Tour construction for each ant
  - 5. While (tour construction == TRUE)

{

6. Minimum distance travelled by an ant taken

Calculate order of cities visited

Calculate the distance of tour

7. Update the delta matrix taking in to account the optimisation for each ant

8. More ants on route, higher delta value, higher pheromone value

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9. Update pheromone definite times

Steps explanation:

}

- 1. Initialise the ants and cities, get the coordinates of the cities.
- 2. Initialise the distance matrix as distance between the cities, pheromone matrix as a highly negative value, delta matrix as zero, visited array and tour array for each ant as null/zero/empty.
- 3. Randomly assign a city to each ant as starting point, tour array will have that city as first element as well as mark that city as visited in the visited array
- 4. Now construct tour for each ant from the starting city selected above, each city to be visited next is selected randomly, no city can be visited more than once.
- 5. While tour construction (true) for all cities

for each ant

5.1 make next city as visited

5.2 save the order in which cities are visited in tour array

5.3 calculate the total distance traversed in the tour till that point

- 5.4 make next city as current city
- 6. Identify the Ant which traversed minimum distance while covering all the cities and reached back to the starting city.
- 7. Delta array will be updated by the rule

"for all ants

for two consecutive cities in the tour

Add (CONSTANT/distance

between consecutive cities in the tour) to the array elements in the delta matrix depicting the consecutive cities in the tour of a particular ant"

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8. This will make sure that if more ants take similar route to cities, higher the value of delta.

9. Pheromone will be updated definite number of times with each successive update affecting the pheromone value to a lesser extent.

10. For particular city pair

Pheromone update =pheromone current + delta value

11. Initialise the tour and visited array for each ant as empty, to account for successive iterations.

12. Use the maximum distance traversed in the tour and the pheromone matrix in the successive iterations.

#### **IV. RESULTS AND DISCUSSION**

Through this work we have tried to implement Ant colony optimisation in the sequential mode and in the parallel mode using CUDA. Then we applied this implementation on the travelling salesman problem to find the best route, we have taken the 'tsp' file of the coordinates of cities and applied it to the above scenario to obtain the best network for those points. The said points are the cities in developing country such as India and the said network is the road network which we want to develop.

Initially the algorithm was run in sequential fashion on different data points to get the results, however the time taken for the cities increases exponentially even though the increase in the number of cities is low.

As can be seen from the table above the time taken for 438 cities is 160 seconds, for 1002 cities its higher and it goes on increasing.

This results in a prohibitive picture for execution for dataset as large as 70009 cities which we want to consider. The time taken will be very large. Thus, we tried to implement the same on the parallel CUDA, where we found that the time taken was very much lower compared to the sequential execution.

In the parallel implementation first, we have equal number of ants and cities, the number of cities increased as 29, 48, 100, 200, 318, 438, 1000, 4000

The execution cities and parameters were kept the same as for sequential execution. The speedup results were found and the comparative graph has been shown above. In the case of 70009 cities, since we had no serial data due to prohibitive time requirements, we tried to test it on different number of ants considering different number of cities each time.

#### V. CONCLUSION AND FUTURE SCOPE

In this paper we have proposed an implementation of a parallel algorithm for finding best network for laying energy based solar panel roads. In our implementation we randomly assign city to each ant and start the tour and find the best tour possible by multiple iterations. This heavy task with so much iterations needs to be done in parallel for which CUDA is the best solution. We have successfully implemented the algorithm and find good results.

For 200 cities the parallel execution time was found to be 1.7389 seconds whereas serial execution time for the same was 36.172. Thus, GPU implementation attains the speedup factor of 20.80

For our future work, we are mainly going to focus on the accuracy and efficiency of the algorithm and will find some ways to optimise the solution.

Second it will be of some practical implication of the results we acquired more aligned with the real life scenarios with more parameters considered.

It can also be applied on different applications which we haven't considered in this paper requiring heavy computational tasks.

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Fig 1. NVIDIA visual profile for the code source: run on dell system nvidia GTX960M

NVIDIA Visual Profiler								
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1. CUDA Application Analysis						Select or highlight a si	ngle interval to see properties	
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Fig 2 Parallel vs Serial Execution Comparison Source: run on python2.7 plotted using matplotlib.pyplot on dell system nvidia GTX960M

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	Duration	4.8474 ms (4,847,396 ns)				
	Stream	Default				
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	Registers/Thread	32				
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	Theoretical	50%				
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Fig 4. Parallel execution over 70009 cities: time output source: run on python2.7 plotted using matplotlib.pyplot on dell system nvidia GTX960M

Table 2. serial vs parallel time	comparison Source: on dell
system nvidia	GTX960M

Cities	Serial	Parallel
29	0.812s	0.13s
48	2.11s	2.03s
100	8.987s	0.48s
127	13.63s	0.74s

200	36.17s	1.73s
318	92.91s	3.65s
438	160.94s	11.04s
1002	-	0.19s
4461	-	0.07s