Cloud Computing Assignment: MPI on the cloud

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1 Settings

1.1 Parameters

The experiments have been done with the following values:

Size of the square matrix: 128, 256, 512, 1024. Number of nodes: 1, 2, 4, 8, 16 (except with 2 and 3 vCPUS VMs). This number is limited by the quota in vCPUs (32) or Memory (50GB) and finally the different flavors of virtual machines which have been used:

- 1 m1.tiny (512MB RAM, 1vCPU)
- 2 m1.small (2GB RAM, 1vCPU)
- 3 m1.medium (4GB RAM, 2vCPUs)
- 4 m1.large (8GB RAM, 4vCPUS)

When a virtual machine has more than 1 vCPU, the MPI flag -npernode is used, in order to execute one instance of the application per vCPU, otherwise. Otherwise, it is useless to choose high quality flavors.

1.2 Automation of the exectution

These tests have been run using only one command to avoid a maximum of human interactions. A ruby script has been developed to create the nodes cluster and to destroy it. When the cluster has been built, it automatically runs ansible in order to configure the nodes and install Open MPI on them.

Once the platform is ready, another script is able to run the experiments with the different kinds of matrix et get the output result back. (run_matrix_multiply.sh)

Finally a last script (run experiment.sh) takes care to:

- 1. Build a cluster with N nodes and flavor F
- 2. Run MPI experiments
- 3. Get the results
- 4. Shutdown the cluster
- 5. Restart with other parameters

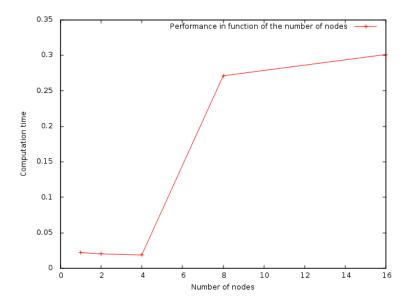
This last program takes care of everything, nothing else has to be done, except reading and interpreting the results!

2 Results

The results in this section come from VMs of flavor 1, which have only one CPU.

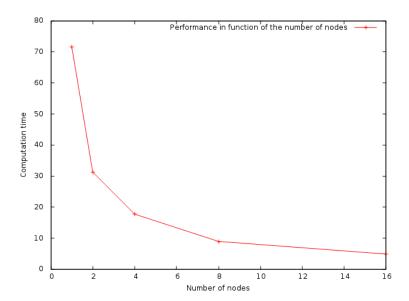
2.1 Small matrices computation

The raw data results can be read in Appendix A. First, for the small matrices (128 rows/columns), it is observable that the communication between the nodes becomes too heavy and even if the computation time is reduced, the overall duration increased hugely. Whatever is the flavor, the performance increases when the number of nodes is low, and then, they get strongly worse.



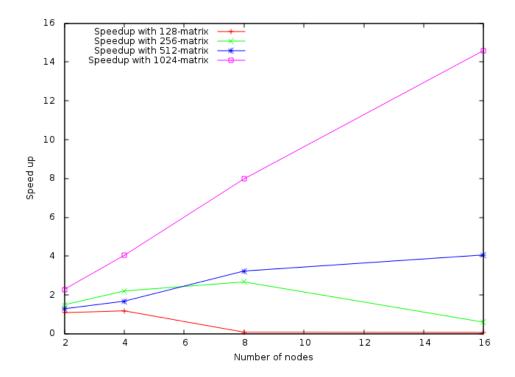
2.2 Large matrices computation

In the case of large matrices computation, the results are closer to what has to be expected, they are better and better but with a speed-up which is not linear.



2.3 Speedup Evolution

The two previous results can be illustrated in the following graph.



With a small matrix, the speed up is getting lower than 1 quickly. It means that the execution time is increasing when the number of nodse is increasing. However, when the matrix get bigger, the speed up increases when the number of VMs increases.

2.4 Cloud irregularity

We have to keep in mind that the hardware is shared among all the user, our tasks don't get dedicated CPUs for example. Consequently, some results may be inadequate. For instance we can se in the measure done with flavor 3 instances, it has been faster to compute the multiplication of a 1360×1360 matrix than a 1024×1024 matrix which is logically abnormal.

3 Pricing

The price for a single instance with one CPU on Amazon EC2 depends of the flavor:

- m1.tiny \$0.065 per hour
- m1.small \$0.130 per hour
- m1.medium \$0.260 per hour
- m1.large \$0.520 per hour

For small-scale experiments like this one, the price would be completely negligible, however for MPI operations, the lesser the communications, the faster the execution. As a result, it is more interesting to take high-CPUs instances. For instance, in the case of intensive computational MPI tasks, one m1.large is more interesting than four m1.tiny. The price per CPU is identical.

Conclusion

To conclude. we can see that MPI on the cloud is working well, however the communication overhead is more important than on a supercomputer and it's important to be aware of it. Furthermore the performance may be unstable, the instances may share their CPUs with other resource-consuming instances.

Appendices

A Results

	128	256	512	1024
1	0.0221	0.1784	1.6523	43.8849
2	0.0186	0.1260	0.9865	29.6860
4	0.2568	0.1083	0.8879	16.5539
8	0.2630	0.0826	0.5524	11.1385
16	0.5145	0.2203	1.1342	11.8693

Table 1: Computation time according to the number of nodes to the size of the matrix for VMs flavor 1

	128	256	512	1024
1	0.0221	0.1787	1.6726	71.6537
2	0.0203	0.1198	1.2978	31.3095
4	0.0187	0.0811	0.9964	17.7449
8	0.2711	0.0669	0.5174	8.9784
16	0.3011	0.2976	0.4117	4.9098

Table 2: Computation time according to the number of nodes to the size of the matrix for VMs flavor 2

	128	256	512	1024	1360
1	0.0137	0.0937	0.8426	34.9952	21.3736
2	0.0169	0.0758	0.6042	11.7280	11.0668
4	0.2616	0.0698	0.4093	6.0268	6.4729
8	0.3310	0.0868	0.4060	6.0640	6.0757

Table 3: Computation time according to the number of nodes to the size of the matrix for VMs flavor 3

	128	256	512	1024
1	0.0223	0.1798	1.7618	41.7868
2	0.0212	0.1254	1.8515	34.9107
4	0.0213	0.1011	0.7157	18.3058

Table 4: Computation time according to the number of nodes to the size of the matrix for VMs flavor 4