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NUMERICAL INVESTIGATION OF SWIRLING FLOWS STABILITY USING MATLAB DISTRIBUTED COMPUTING SERVER ON A WINDOWS OPERATING SYSTEM ENVIRONMENT

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Abstract: This paper presents an evaluation of the parallel processing for meshless algorithm for solving Navier-Stokes problems in fluid dynamics. The fluid dynamics existing solvers take use of finite element methods which are cumbersome and very exhaustive in computational resources. Meshless algorithms uses a different approaches using high level eigenvalue solvers. In order to evaluate the parallel processing performance, a cluster using Matlab development environment was build. Using the Matlab parallel and distributed toolbox it was possible to perform a profiling of the algorithm and to observe the eventual bottlenecks and flaws of the algorithm. Using this approach we obtained a very effective tool for processing the numerical stability investigation of the swirling flow in the Francis hydropower turbine. **Keywords:** Cluster technology, parallel computational algorithms, fluid dynamics spatial stability, swirling flow, meshless method

CONSIDERATIONS ABOUT PARALLEL COMPUTING

Parallel programming and the design of efficient parallel programs have been well established in high-performance, scientific computing for many years. The simulation of scientific problems is an important area in natural and engineering sciences of growing importance. More precise simulations or the simulations of larger problems need greater and greater computing power and memory space. In the last decades, high-performance research included new developments in parallel hardware and software technologies [1-4] and a steady progress in parallel high-performance computing can be observed. Popular examples are simulations of weather forecast based on complex mathematical models involving partial differential equations or crash simulations from car industry based on finite element methods [5].

Other examples include drug design and computer graphics applications for film and advertising industry [6]. Depending on the specific application, computer simulation is the main method to obtain the desired result or it is used to replace or enhance physical experiments. A typical example for the first application area is weather forecast where the future development in the atmosphere has to be predicted, which can only be obtained by simulations [7]. In the second application area, computer simulations are used to obtain results that are more precise than results from practical experiments or that can be performed with less financial effort. An example is the use of simulations to determine the air resistance of vehicles [8]. Compared to a classical wind tunnel experiment, a computer simulation can give more precise results because the relative movement of the vehicle in relation to the ground can be included in the simulation. This is not possible in the wind tunnel, since the vehicle cannot be moved. Crash tests of vehicles are an obvious example where computer simulations can be performed with less financial effort.

Computer simulations often require a large computational effort. A low performance of the computer system used can restrict the simulations and the accuracy of the results obtained significantly. In particular, using a high-performance system allows larger simulations which lead to better results. Therefore, parallel computers have often been used to perform computer simulations. Today, cluster systems built up from server nodes are widely available and are now often used for parallel simulations. To use parallel computers or cluster systems, the computations to be performed must be partitioned into several parts which are assigned to the parallel resources for execution. These computation parts should be independent of each other, and the algorithm performed must provide enough independent computations to be suitable for a parallel execution. This is normally the case for scientific simulations. To obtain a parallel program, the algorithm must be formulated in a suitable programming language. Parallel execution is often controlled by specific runtime libraries or compiler directives which are added to a standard programming language like C, Fortran or Java [9-13].

MATLAB [14] is currently the dominant language of technical computing with approximately one million users worldwide, many of whom can benefit from the increased power offered by widely available multicore processors and multinode computing clusters. MATLAB is also an ideal environment for learning about parallel computing, allowing the user to focus on parallel algorithms instead of the details of the implementation.

Higham & Higham [15] and Moler [16] offer an introduction to MATLAB in their surveys. In [17] Jeremy Kepner introduces the theory, algorithmic notation, and an "under the hood" view of distributed array programming. He discusses metrics for evaluating performance and coding of a parallel program. A selected survey of parallel application analysis techniques are presented in this book, with a particular emphasis on how the examples used in the book relate to many wider application domains. In [18], Rauber and Runger take up the new development in processor architecture by giving a detailed description of important parallel programming techniques that are necessary for developing efficient programs for multicore processors as well as for parallel cluster systems or supercomputers. Both shared and distributed address space architectures are covered.

Parallel programming is an important aspect of high-performance scientific computing but it used to be a niche within the entire field of hardware and software products. However, more recently parallel programming has left this niche and will become the mainstream of software development techniques due to a radical change in hardware technology. Major chip manufacturers have started to produce processors with several power efficient computing units on one chip, which have an independent control and can access the same memory concurrently. Normally, the term core is used for single computing units and the term multicore [19, 20] is used for the entire processor having several cores. Thus, using multicore processors makes each desktop computer a small parallel system. The technological development toward multicore processors was forced by physical reasons, since the clock speed of chips with more and more transistors cannot be increased at the previous rate without overheating. Multicore architectures in the form of single multicore processors, shared memory systems of several multicore processors, or clusters of multicore processors with a hierarchical interconnection network will have a large impact on software development. In 2009, dual-core and guad-core processors are standard for normal desktop computers, and chip manufacturers have already announced the introduction of oct-core processors for 2010. It can be predicted from Moore's law [21] that the number of cores per processor chip will double every 18-24 months. According to a report of Intel, in 2015 a typical processor chip will likely consist of dozens up to hundreds of cores where a part of the cores will be dedicated to specific purposes like network management, encryption and decryption, or graphics [22].

As computational fluid dynamics (CFD) matures so rise the expectations of what it can or should deliver. Practitioners and designers are no longer content with qualitative statements on trends, but judge the utility and value of CFD by its ability to provide quantitatively accurate predictions for property fields and engineering parameters derived therefrom. In the extreme, theoretical fluid dynamicists expect fully-resolved simulations of turbulence to provide fundamental information on turbulence mechanics of greater accuracy and detail than can be derived from the most sophisticated experimental techniques.

In computational fluid dynamics, several software packages solving either the Euler or the Navier-Stokes equations around complex geometries have been developed and are currently used by aircraft or engine manufacturers. The use of unstructured meshes is one of the most efficient ways to solve complex problems, because they allow high flexibility in specifying a geometry. The possibility of using the modern parallel machines even for unstructured mesh based codes implies the necessity of dealing with two problems: the first one lies on the strategies and algorithms which can be used for the partitioning of the grid and the mapping of the subdomains among the processors; the second one is related to the structure of the code, which has to be designed is such a way that the features of the modern parallel machines are fully exploited.

The first efficient implementation of a 2D unstructured flow solver on MIMD parallel computers was presented by Venkatakrishnan et al [23] in 1992: they demonstrated that a good supercomputer performance can be reached and that a careful implementation of the message passing routines is a critical point, even for an explicit code. In 1995, Lanteri [24] developed a parallel version of an industrial code based on a mixed finite element/finite volume method. The parallelization strategy combines mesh partitioning and message passing programming model: basically, the same old serial code is going to be executed within every subdomain. Modifications occurred in the main loop in order to take into account the assembly phases of the subdomain results.

The literature on these topics can be considered exhaustive for two-dimensional applications and parallel machines of the old generation. The same cannot be said for three-dimensional complex flows and for the new machines in terms of optimization of the performances and assessment of algorithms and techniques (i.e. [25-28]).

CLUSTER CONFIGURATION

In this paper we have build and tested a small scale cluster based on a Matlab Parallel Processing Toolbox. Using the internal cluster manager from Matlab we were able to evaluate the algorithm

behaviour using a distributed process. In this situations we have perform the profiling and we have noticed a speed increase of the algorithm compared to a single computer run. The cluster was conceived using homogenous hardware: Dell Optiplex 755, Intel(R) Core(TM)2 Duo CPU, 2.66GHz 1.97 GHz, 1.95 GB of RAM.

MATLAB Distributed Computing Server (MDCS) is a toolbox that lets users solve computationally and data-intensive problems by executing MATLAB and Simulink based applications on a computer cluster.

For numerical investigation of swirling flows stability, a Windows operating system cluster was configured in Computer Aided Mathematics and Numerical Analysis Laboratory of the Engineering Faculty of Hunedoara (Figure 1).



Figure 1. Computer Aided Mathematics and Numerical Analysis Laboratory at the Engineering Faculty of Hunedoara.

The first step to set up the cluster configuration was to install the MDCS on a node called the head node. The license manager was also installed on the head node. After performing this installation, the MDCS was installed on the other cluster nodes, called worker nodes. Figure 2 shows the installations that was performed on our MDCS cluster nodes.

The MDCS service must be running on all machines being used for job managers or workers. This service manages the job manager and worker processes. One of the major tasks of the MDCS service is to recover job manager and worker sessions after a system crash, so that jobs and tasks are not lost as a result of such accidents. To run the MDCS, the license manager must be running on the head node. We made sure



Figure 2. MDCS cluster with local access configuration.

the license manager was running by performing a Status Enquiry. The next step in configuring the cluster was to start the job manager and workers. From the Admin Center, the setup and monitoring utility for the Matlab Distributed Computing Server, we select the list of the hosts in the text box. When Admin Center completed the update, the listing looked like the following Figure 3.



Figure 3. Nodes added in the configuration of the cluster.

At this point, the connectivity between nodes was tested. This assures that the cluster can perform the necessary communications for running MDCS processes. After the cluster configuration was completed, job managers have been started using the Job Manager module. In the Start Workers dialog box, we specify the number of workers to start on each host, a value that cannot exceed the total number of licenses we have. A number of two workers were set for each host. The Connectivity Testing dialog box shows the results of the last test, reported in Figure 4.



Figure 4. The connectivity test of the configuration of the cluster passed.

⁰ 0.2 0.4 0.6 0.8 1 1.2 1.4 Dimensionless radius 1 1.2 1.4 Figure 5. Axial and circumferential velocity profiles

Axial velocity vortex rope model Swirl velocity vortex rope model

PARALLEL AND DISTRIBUTED INVESTIGATION OF THE VORTEX ROPE MODEL

The helical vortex breakdown (also known as precessing vortex rope) is a self-induced instability of a swirling flow, encountered in the draft tube cone of hydraulic Francis turbines operated far from the best efficiency. Figure 5 shows the axial and the swirl velocity profiles of the vortex rope model used in our stability analysis.

Let us define the eigenvalue problem governing the hydrodynamic stability in operator formulation

$$\left(k L^{[k]} + \omega L^{[\omega]} + L\right) \mathbf{u} = 0, \quad \mathbf{u} = \left(F \quad G \quad H \quad P\right)^{T} \quad (1)$$

where

$$L^{[k]} = \begin{pmatrix} r & 0 & 0 & 0 \\ 0 & U & 0 & 0 \\ 0 & 0 & rU & 0 \\ U & 0 & 0 & 1 \end{pmatrix}, \quad L^{[\omega]} = \begin{pmatrix} 0 & 0 & 0 & 0 \\ 0 & -1 & 0 & 0 \\ 0 & 0 & -r & 0 \\ -1 & 0 & 0 & 0 \end{pmatrix}, \quad L = \begin{pmatrix} 0 & 1+d_r & -m & 0 \\ 0 & mW/r & 2W/r & -d_r \\ 0 & W+rdW/dr & mW & m \\ mW/r & dU/dr & 0 & 0 \end{pmatrix}$$
(2)

The boundary relations are translated into equations that complete the computational model and can be transposed as

$$\sum_{1}^{N} (-1)^{k+1} g_k = \sum_{1}^{N} (-1)^{k+1} h_k = 0 , \qquad (3)$$

$$f_{2} \frac{2}{r_{\max}} + \sum_{\substack{k \text{ odd}}}^{N} f_{k} \frac{2(k-1)}{r_{\max}} \left[\sum_{\substack{r=k-1\\k \text{ even}}}^{2} (-2) \right] + \sum_{\substack{k \text{ even}}}^{N} f_{k} \frac{2(k-1)}{r_{\max}} \left[\sum_{\substack{r=k-1\\k \text{ odd}}}^{2} 2+1 \right] = 0,$$
(4)

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$$p_{2} \frac{2}{r_{\max}} + \sum_{k \text{ odd}}^{N} p_{k} \frac{2(k-1)}{r_{\max}} \left[\sum_{\substack{r=k-1\\k \text{ even}}}^{2} (-2) \right] + \sum_{k \text{ even}}^{N} p_{k} \frac{2(k-1)}{r_{\max}} \left[\sum_{\substack{r=k-1\\k \text{ odd}}}^{2} 2 + 1 \right] = 0,$$
(5)

$$\frac{2W_{r_{\max}}}{r_{\max}}\sum_{1}^{N}h_{k}-p_{2}\frac{2}{r_{\max}}-\sum_{\frac{3}{k_{odd}}}^{N}p_{k}\frac{2(k-1)}{r_{\max}}\left[\sum_{\substack{r=k-1\\k\,even}}^{2}2\right]-\sum_{\substack{k\\k\,even}}^{N}p_{k}\frac{2(k-1)}{r_{\max}}\left[\sum_{\substack{r=k-1\\k\,odd}}^{2}2+1\right]=0,$$
(6)

$$\sum_{1}^{N} g_{k} = 0, \ \left(kU_{r\max} - \omega\right) \sum_{1}^{N} h_{k} = 0, \ k \left(U_{r\max} \sum_{1}^{N} f_{k} + \sum_{1}^{N} p_{k}\right) - \omega \sum_{1}^{N} f_{k} = 0.$$
(7)

The numerical investigation employed both the positive and the negative modes. Let us define by the critical frequency that temporal frequency corresponding to a maximum growth rate for a given mode number. The numerical results are summarized in Table 1.

Table 1. The critical frequency and the maximum growth rates obtained for the investigated modes

Mode m	-3	-2	-1	0	1	2	3
Critical Frequency	0.1	0.3	0.3	0.3	0.3	0.3	0.35
Maximum Growth Rate	14.129	11.947	6.932	0.043	0.544	0.535	0.395

Figure 6 presents the three dimensional map of the growth rate as function of the frequency and mode number and the density map of the growth rate. The negative modes present an amplitude

increase depending on the mode number. However, these amplitude growth rates increase linear up to a frequency of 0.25 both for the positive and negative modes.



Figure 6. Plot of the growth rate as function of frequency and mode (left) and density map (right).

The convergence behavior of the collocation algorithm for m = -1 case is reported in Table 2 on few cluster configurations with six, four and two labs, respectively. It is noticeable that the time increase is relevant when the numerical experiments run on four labs configuration instead of 2 labs configuration. The time increase becomes irrelevant when the number of cluster nodes is increased at six labs, thus the conclusion that the cluster has no need to be extended for numerical improvements of our stability analysis.

Ν	Critical	Elapsed time	Elapsed time	Elapsed time			
	frequency	on 2 labs	on 4 labs	on 6 labs			
5	0.25	1.463658	0.196285	0.211601			
6	0.1	1.569234	0.225536	0.193199			
8	0.1	1.584621	0.210344	0.222434			
10	0.25	1.611874	0.328014	0.341408			
12	0.3	1.632951	0.345925	0.360558			
16	0.3	1.751369	0.431654	0.431196			
17	0.3	1.836951	0.460168	0.451093			
19	0.3	1.854693	0.519268	0.524420			
29	0.3	2.234852	1.005637	1.015383			
32	0.3	2.672955	1.479596	1.448778			
37	0.3	2.895647	1.682333	1.726657			
40	0.3	3.269854	2.064365	2.087408			
46	0.3	4.125965	2.942432	2.948638			
61	0.3	8.264985	6.453132	6.318485			

able 2. Elap	sed time (in se	econds) of the nun	nerical computation
for m	ode $m = -1$ o	n three cluster cou	ofigurations

CONCLUSIONS

The cluster technology represents the state of the art in computational technology. This extends the limit imposed by the single computer analysis, reducing in a large manner the amount of time required for solving complex mathematical models. Using a rather usual hardware (desktop PCs) it is possible to obtain a significant acceleration of the computation process. However, the acceleration depends on the algorithm structure and it is necessary to perform a profiling of the solving process (that means to analyze carefully the behavior of the computational process- the number of steps required, the mathematical operations involved, the size of matrix variables) in order to avoid the eventual bottlenecks in the algorithm.

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