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Different goals in multiscale simulations and how to reach them

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Summary. In this paper we sum up our works on multiscale programs, mainly simulations. We first start with describing what multiscaling is about, how it helps perceiving signal from a background noise in a flow of data for example, for a direct perception by a user or for a further use by another program. We then give three examples of multiscale techniques we used in the past, maintaining a summary, using an environmental marker introducing an history in the data and finally using a knowledge on the behavior of the different scales to really handle them at the same time.

Key words: Multiscale, clustering, dynamic graphs, adaptation

1 Introduction: What this paper is about, and what it's not

Although we delved into different applications and application domains, the computer science research goals of our team has remained centered on the same subject for years. It can be expressed in different ways that we feel are, if not exactly equivalent, at least closely connected. It can be defined as managing multiple scales in a simulation. It also consists in handling emergent structures in a simulation. It can often also be seen as dynamic heuristic clustering of dynamic data³. This paper is about this theme, about why we think

³ We will of course later on describe in more details what we mean by all this.

it is of interest and what we've done so far in this direction. It is therefore akin to a state of the art kind of article, except more centered on what we did. We will allude to what others have done, but the focus of the article is presenting our techniques and what we're trying to do, like most articles do, and not present an objective description of the whole field, as the different applications examples could make think : we're sticking to the same computer science principles overall. We're taking one step back from our works to contemplate them all, and not the three steps which would be necessary to encompass the whole domain, as it would take us beyond the scope of this book.

2 Perception: filtering to make decisions

I look at a fluid flow simulation but all I'm interested in is where does the turbulence happen, in a case where I couldn't know before the simulation [Tranouez 2005a]. I use a multi-participant communication system in a crisis management piece of software and I would like to know what are the main interests of each communicant based on what they are saying [Lesage 1999]. I use an Individual-Based Model (IBM) of different fish species but I'm interested in the evolution of the populations, not the individual fish [Prevost 2004]. I use a traffic simulation with thousands of cars and a detailed town but what I want to know is where the traffic jams are (coming soon).

In all those examples, I use a piece of software which produces huge amounts of data but I'm interested in phenomena of a different scale than the raw basic components. What we aim at is helping the user of the program to reach what he is interested in, be this user a human (Clarification of the representation) or another program (Automatic decision making). Although we're trying to stay general in this part, we focused on our past experience of what we actually managed to do, as described in "Some techniques to make these observations in a time scale comparable to the observed", this is not gratuitous philosophy.

2.1 Clarification of the representation

This first step of our work intends to extract the patterns on the carpet from its threads [Tranouez 1984]. Furthermore, we want it to be done in "real (program) time", meaning not a posteriori once the program is ended by examining its traces [Servat 1998], and sticking as close as possible to the under layer, the one pumping out dynamic basic data. We don't want the discovery of our structures to be of a greater time scale than a step of the program it works upon.

How to detect these structures? For each problem the structure must be analyzed, to understand what makes it stand out for the observer. This implies knowing the observer purpose, so as to characterize the structure. The

answers are problem specific, nevertheless rules seem to appear.

In many situations, the structures are groups of more basic entities, which then leads to try to fathom what makes it a group, what is its inside, its outside, its frontier, and what makes them so.

Quite often in the situation we dealt with, the groups members share some common characteristics. The problem in that case belongs to a subgenre of clustering, where the data changes all the time and the clusters *evolve* with them, they are not computed from scratch at each change.

The other structures we managed to isolate are groups of strongly communicating entities in object-oriented programs like multiagent simulations. We then endeavored to manage these cliques.

In both cases, the detected structures are emphasized in the graphical representation of the program. This clarification lets the user of the simulation understand what happens in its midst. Because modeling, and therefore understanding, is clarifying and simplifying in a chosen direction a multi-sided problem or phenomenon, our change of representation participates to the understanding of the operator. It is therefore also a necessary part of automating the whole understanding, aiming for instance at computing an artificial decision making.

2.2 Automatic decision making

Just like the human user makes something of the emerging phenomena the course of the program made evident, other programs can use the detected organizations.

For example in the crisis management communication program, the detected favorite subject of interest of each of the communicant will be used as a filter for future incoming communications, favoring the ones on connected subjects. Other examples are developed below, but the point is once the structures are detected and clearly identified, the program can use models it may have of them to compute its future trajectory. It must be emphasized that at this point the structures can themselves combine into groups and structures of yet another scale, recursively. We're touching there an important component of complex system [Simon 1996]. We may hope the applications of this principle to be numerous, such as robotics, where perceiving structures in vast amounts of data relatively to a goal, and then being able to act upon these accordingly is a necessity.

We're now going to develop these notions in examples coming from our past works.

3 Some techniques to make these observations in a time scale comparable to the observed

The examples of handling dynamic organization we chose are taken from two main applications, one of a simulation of a fluid flow, the other of the simulation of a huge cluster of computed processes, distributed over a dynamic network of computing resources, such as computers. The methods titled “Maintaining a summary of a simulation” and “Reification: behavioral methods” are theories from the hydromechanics simulation, while “Traces of the past help understand the present” refers to the computing resources management simulation. We will first describe these two applications, so that an eventual misunderstanding of what they are does not hinder later the clarity of our real purpose, the analysis of multiscale handling methods.

In a part of a more general estuarine ecosystem simulation, we developed a simulation of a fluid flow. This flow uses a particle model [Leonard 1980], and is described in details in [Tranouez 2005a] or [Tranouez 2005b]. The basic idea is that each particle is a vorticity carrier, each interacting with all the others following Biot-Savart laws. As fluid flows tend to organize themselves in vortices, from all spatial scales from a tens of angstrom to the Atlantic Ocean, this is these vortices we tried to handle as the multiscale characteristic of our simulation. The two methods we used are described below.

The other application, described in depth in [Dutot 2005], is a step toward automatic distribution of computing over computing resources in difficult conditions, as:

- The resources we want to use can each appear and disappear, increase or decrease in number.
- The computing distributed is composed of different object-oriented entities, each probably a thread or a process, like in a multiagent system for example (the system was originally imagined for the ecosystem simulation alluded to above, and the entities would have been fish, plants, fluid vortices etc., each acting, moving . . .)

Furthermore, we want the distribution to follow two guidelines:

- As much of the resources as possible must be used,
- Communications between the resources must be kept as low as possible, as it should be wished for example if the resources are computers and the communications therefore happen over a network, bandwidth limited if compared to the internal of a computer.

This the ultimate goal of this application, but the step we’re interested in today consists in a simulation of our communicating processes, and of a program which, at the same time the simulated entities act and communicate,

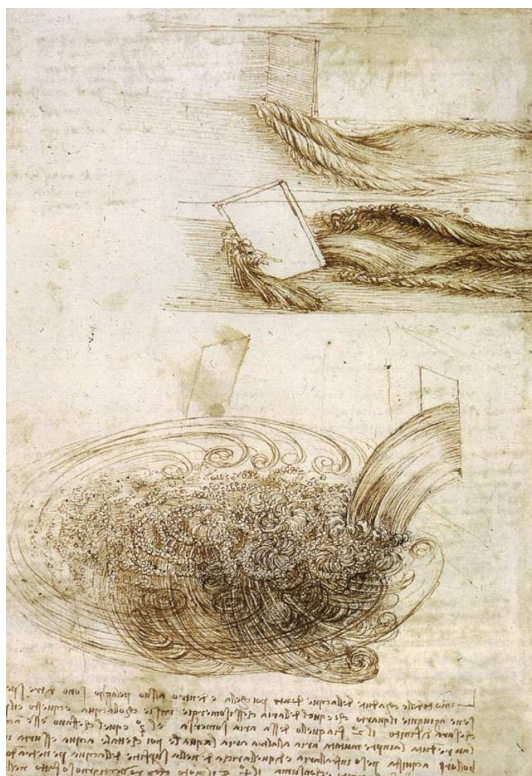


Fig. 1. Studies of water passing obstacles and falling by Leonardo Da Vinci, c. 1508-9. In Codex Leicester.

advises how they should be regrouped and to which computing resource they should be allocated, so as to satisfy the two guidelines above.

3.1 Maintaining a summary of a simulation

The first method we would like to describe here relates to the fluid flow simulation. The hydrodynamic model we use is based on an important number of interacting particles. Each of these influences all the others, which makes n^2 interactions, where n is the number of particles used. This makes a great number of computations. Luckily, the intensity of the influence is inversely proportional to the square of the distance separating two particles. We therefore use an approximation called Fast Multipoles Method (FMM), which consists in covering the simulation space with grids, of a density proportional to the density of particles (see Figure 2-a). The computation of the influence of its colleagues over a given particle is then done exactly for the ones close enough, and averaged on the grid for those further. All this is absolutely monoscale.

As the particles are vorticity carriers, it means that the more numerous they are in a region of space, the more agitated the fluid they represent is. We would therefore be interested in the structures built of close, dense particles, surrounded by sparser ones. A side effect of the grids of the FMM, is that they help us do just that. It is not that this clustering is much easier on the grids, it's above all that they are an order of magnitude less numerous, and organized in a tree, which makes the group detection much faster than if the algorithm was ran on the particles themselves. Furthermore, the step by step management of the grids is not only cheap (it changes the constant of the complexity of the particles movement method but not the order) but also needed for the FMM.

We therefore detect structures on

- Dynamic data (the particles characteristics)
- With little computing added to the simulation,

which is what we aimed at.

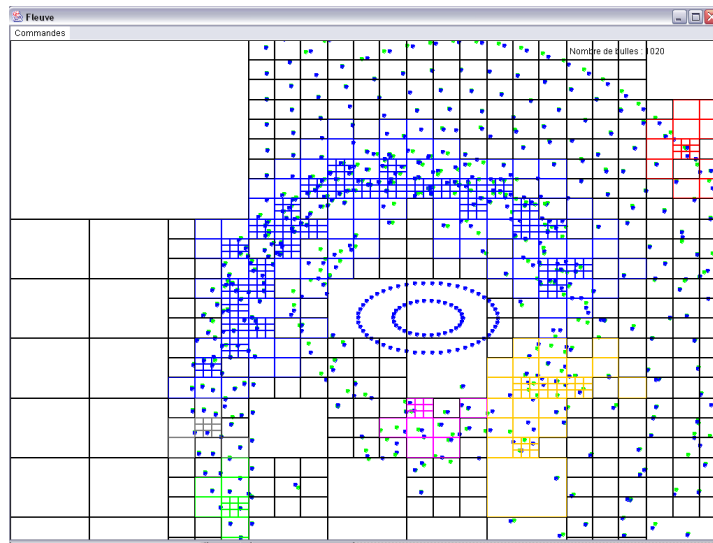
The principle here is that through the grids we maintain a summary of the simulation, upon which we can then run static data algorithm, all this at a cheap computing price.

3.2 Traces of the past help understand the present

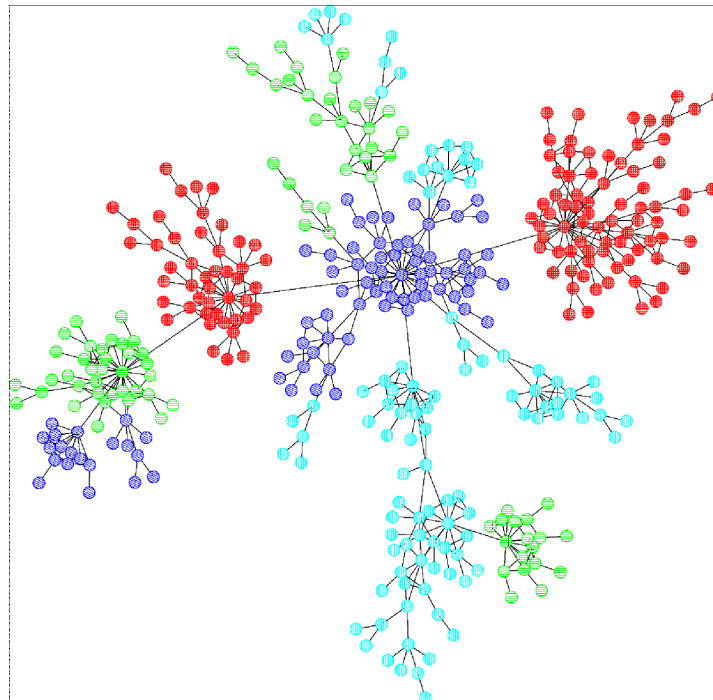
The second method relates to the detection of communication clusters inside a distributed application. The applications we are interested in are composed of a large number of object-oriented entities that execute in parallel, appear, evolve and, sometimes, disappear. Aside some very regular applications, often entities tend to communicate more with some than with others. For example in a simulation of an aquatic ecosystem, entities representing a species of fish may stay together, interacting with one another, but flee predators. Indeed organizations appear groups of entities form. Such simulations are a good example of applications we intend to handle, where the number of entities is often too large to compute a result in an acceptable time on one unique computer.

To distribute these applications it would be interesting to both have approximately the same number of entities on each computing resource to balance the load, but also to avoid as much as possible to use the network, that costs significantly more in terms of latency than the internals of a computer. Our goal is therefore to balance the load and minimize network communications. Sadly, these criteria are conflicting, and we must find a tradeoff.

Our method consists in the use of an ant metaphor. Applications we use are easily seen as a graph, which is a set of connected entities. We can map



a - Each color corresponds to a detected aggregate



b - Each color corresponds to a computing ressource

Fig. 2. Detection of emergent structures in two applications with distinct methods

entities to vertices of the graph, and communications between these entities to the edges of the graph. This graph will follow the evolution of the simulation. When an entity appear, a vertex will appear in the graph, when a communication will be established between two entities, an edge will appear between the two corresponding vertices. We will use such a graph to represent the application, and will try to find clusters of highly communicating entities in this graph by coloring it, assigning a color to each cluster. This will allow to identify clusters as a whole and use this information to assign not entities, but at another scale, clusters to computing resources (Figure 2-b).

For this, we use numerical ants that crawl the graph as well as their pheromones, olfactory messages they drop, to mark clusters of entities. We use several distinct colonies of ants, each of a distinct color, that drop colored pheromones. Each color corresponds to one of the computing resources at our disposal. Ants drop colored pheromones on edges of the graph when they cross them. We mark a vertex as being of the color of the dominant pheromone on each of its incident edges. The color indicates the computing resource where the entity should run.

To ensure our ants color groups of highly communicating entities of the same color to minimize communications, we use the collaboration between ants: ants are attracted by pheromones of their own color, and attracted by highly communicating edges. To ensure the load is balanced, that is to ensure that the whole graph is not colored only in one color if ten colors are available, we use competition, ants are repulsed by the pheromones of other colors.

Pheromones in nature being olfactory molecules, they tend to evaporate. Ants must maintain them so they do not disappear. Consequently, only the interesting areas, zones where ants are attracted, are covered by pheromones and maintained. When a zone becomes less interesting, ants leave it and pheromone disappear. When an area becomes of a great interest, ants colonize it by laying down pheromones that attract more ants, and the process self-amplifies.

We respect the metaphor here since it brings us the very interesting property of handling the dynamics. Indeed, our application continuously changes, the graph that represents it follows this, and we want our method to be able to discover new highly communicating clusters, while abandoning vertices that are no more part of a cluster. As ants continuously crawl through the graph, they maintain the pheromone color on the highly communicating clusters. If entities and communications of the simulation appear or disappear, ants can quickly adapt to the changes. Colored pheromones on parts where a cluster disappeared evaporate and ants colonize new clusters in a dynamic way. Indeed, the application never changes completely all the time; it modifies itself smoothly. Ants lay down “traces” of pheromones and do not recompute the

color of each vertex at each time, they reuse the already dropped pheromone therefore continuously giving a distribution advice at a small computing price, and adapting to the reconfigurations of the underlying application.

3.3 Reification: behavioral methods

This last example of our multiscale handling methods was also developed on the fluid flow simulation (Figure 3). Once more, we want to detect structures in a dynamic flow of data, without getting rid of the dynamicity by doing a full computation on each step of the simulation. The idea here is doing the full computation only once in a good while, and only relatively to the unknown parts of our simulation.

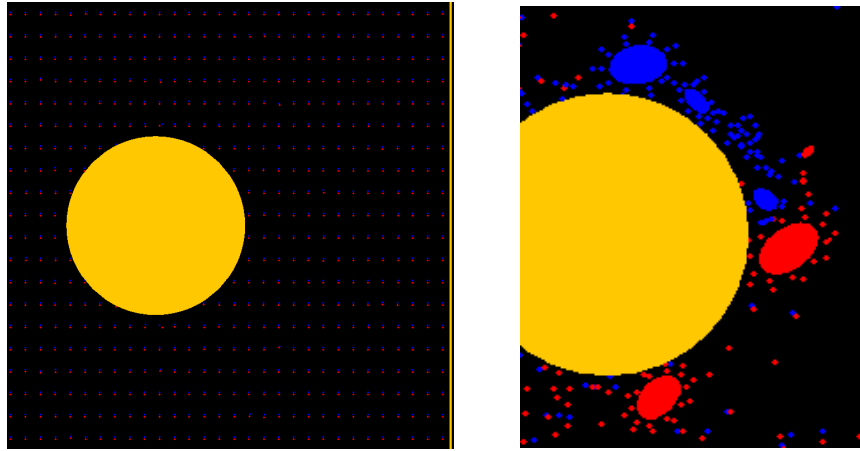


Fig. 3. Fluid flow around an obstacle. On the left, the initial state. On the right, a part of the flow, some steps later (the ellipses are vortices)

We begin with detecting vortices on the basic particles once. Vortices will be a rather elliptic set of close particles of the same rotation sense. We then introduce a multiagent system of the vortices (Figure 3-right). We have indeed a general knowledge of the way vortices behave. We know they move like a big particle in our Biot-Savard model, and we model its structural stability through social interactions with the surrounding basic particles, the other vortices and the obstacles, through which they can grow, shrink or die (be dissipated into particles). The details on this can be found in [Tranouez 2005a]. Later on, we occasionally make a full-blown vortex detection, but only on the remaining basic particles, as the already detected vortexes are managed by the multiagent system.

In this case, we possess knowledge on the structures we want to detect, and we use it to build actually the upper scale level of the simulation, which

at the same time lightens ulterior structures detection. We are definitely in the category described in Automatic decision making.

4 Conclusion

Our research group works on complex systems and focuses on the computer representation of their hierarchical/holarchical characteristics [Koestler 1978], [Simon 1996], [Kay 2000]. We tried to illustrate that describing a problem at different scales is a well-spread practice at least in the modeling and simulating community. We then presented some methods for handling the different scales, with maintaining a summary, using an environmental marker introducing a history in the data and finally using knowledge on the behavior of the different scales to handle them at the same time.

We now believe we start to have sound multiscale methods, and must focus on the realism of the applications, to compare the sacrifice in details we make when we model the upper levels rather than just heavily computing the lower ones. We save time and lose precision, but what is this trade-off worth *precisely*?

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