MAKING SENSOR NETWORKS INTELLIGENT

Peter Sapaty  
Institute of Mathematical Machines & Systems, National Academy of Sciences  
Glushkova Ave 42, Kiev 03187, Ukraine

Masanori Sugisaka  
Department of Electrical and Electronic Engineering, Oita University  
700 Oaza Dannoharu 870-1192 Japan

Joaquim Filipe  
Departamento Sistemas e Informática, Escola Superior de Tecnologia de Setúbal  
Setúbal 2910-761, Portugal

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Abstract: A universal solution for management of dynamic sensor networks will be presented, covering both networking and application layers. A network of intelligent modules, overlaying the sensor network, collectively interprets mission scenarios in a special high-level language that can start from any nodes and cover the network at runtime. The spreading scenarios are extremely compact, which may be useful for energy saving communications. The code will be exhibited for distributed collection and fusion of sensor data, also for tracking mobile targets by scattered and communicating sensors.

1 INTRODUCTION

Sensor networks are a sensing, computing and communication infrastructure that allows us to instrument, observe, and respond to phenomena in the natural environment, and in our physical and cyber infrastructure (Culler at al., 2004; Chong, Kumar, 2003). The sensors themselves can range from small passive microsensors to larger scale, controllable platforms. Their computation and communication infrastructure will be radically different from that found in today’s Internet-based systems, reflecting the device- and application-driven nature of these systems.

Of particular interest are wireless sensor networks, WSN (Wireless; Zhao, Guibas, 2004) consisting of spatially distributed autonomous devices using sensors to cooperatively monitor physical or environmental conditions, such as temperature, sound, vibration, pressure, motion or pollutants, at different locations. WSN, however, have many additional problems in comparison to the wired ones. The individual devices in WSN are inherently resource constrained--they have limited processing speed, storage capacity, and communication bandwidth. These devices have substantial processing capability in the aggregate, but not individually, so we must combine their many vantage points on the physical phenomena within the network itself. In addition to one or more sensors, each node in a sensor network is typically equipped with a radio transceiver or other wireless communications device, a small microcontroller, and an energy source, usually a battery. The size of a single sensor node can vary from shoebox-sized nodes down to devices the size of grain of dust. Typical applications of WSNs include monitoring, tracking, and controlling. Some of the specific applications are habitat monitoring, object tracking, nuclear reactor controlling, fire detection,
traffic monitoring, etc. In a typical application, a WSN is scattered in a region where it is meant to collect data through its sensor nodes. They could be deployed in wilderness areas, where they would remain for many years (monitoring some environmental variable) without the need to recharge/replace their power supplies. They could form a perimeter about a property and monitor the progression of intruders (passing information from one node to the next). At present, there are many uses for WSNs throughout the world.

In a wired network like the Internet, each router connects to a specific set of other routers, forming a routing graph. In WSNs, each node has a radio that provides a set of communication links to nearby nodes. By exchanging information, nodes can discover their neighbors and perform a distributed algorithm to determine how to route data according to the application’s needs. Although physical placement primarily determines connectivity, variables such as obstructions, interference, environmental factors, antenna orientation, and mobility make determining connectivity a priori difficult. Instead, the network discovers and adapts to whatever connectivity is present.

Fig. 1 shows what we will mean as a sensor network for the rest of this paper.

![Figure 1: Distributed sensors and their emergent network.](image)

It will hypothetically consist of (a great number of) usual sensors with local communication capabilities, and (a limited number of) those that can additionally transmit collected information outside the area (say, via satellite channels). Individual sensors can be on a move, some may be destroyed while others added at runtime (say, dropped from the air) to join the existing ones in solving cooperatively distributed problems.

The aim of this paper is to show how any imaginable (or even so far unimaginable) distributed problems can be solved by dynamic self-organized sensor networks if to increase their intelligence as a whole, with a novel distributed processing and control ideology and technology effectively operating in computer networks.

2 THE DISTRIBUTED MANAGEMENT MODEL

The distributed information technology we are using here is based on a special Distributed Scenario Language (DSL) describing parallel solutions in computer networks as a seamless spatial process rather than the traditional collection and interaction of parts (agents). Parallel scenarios in DSL can start from any interpreter of the language, spreading and covering the distributed space at runtime, as in Fig. 2.

![Figure 2: Runtime coverage of space by parallel scenarios.](image)

The overall management of the evolving scenarios is accomplished via the distributed track system providing hierarchical command and control for scenario execution, with a variety of special echo messages. We will mention here only key features of DSL, as the current language details can be found elsewhere (Sapaty et al., 2007), also its basics from the previous versions (Sapaty, 1999, 2005; Sapaty et al., 2006).

A DSL program, or wave, is represented as one or more constructs called moves (separated by a comma) embraced by a rule, as follows:

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wave → rule ( { move , } )
```

Rules may serve as various supervisory, regulatory, coordinating, integrating, navigating, and data processing functions, operations or constraints over moves.
A move can be a constant or variable, or recursively an arbitrary wave itself:

\[
\text{move} \rightarrow \text{constant} | \text{variable} | \text{wave}
\]

Variables classify as nodal, associated with space positions and shared by different waves, frontal, moving in space with program control, and environmental, accessing the environment navigated. Constants may reflect both information and physical matter.

Wave, being applied in a certain position of the distributed world, can perform proper actions in a distributed space, terminating in the same or in other positions. It provides final result that unites local results in the positions (nodes) reached, and also produces resultant control state. The (distributed) result and the state can be subsequently used for further data processing and decision making on higher program levels. Parallel waves can start from different nodes in parallel, possibly intersecting in the common distributed space when evolving in it independently.

If moves are ordered to advance in space one after the other (which is defined by a proper rule), each new move is applied in parallel in all the nodes reached by the previous move. Different moves (by other rules) can also apply independently from the same node, reaching new nodes in parallel. The functional style syntax shown above can express any program in DSL, but if useful, other notations can be used, like the infix one. For example, an advancement in space can use period as operator (separator) between successive steps, whereas parallel actions starting from same node can be separated by a semicolon. For improving readability, spaces can be inserted in any places of the programs—they will be automatically removed before execution (except when embraced by quotes).

The interpreter may have its own physical body (say, in the form of mobile or humanoid robot), or can be mounted on humans (like in mobile phones). A network of the interpreters can be mobile and open, changing its volume and structure, as robots or humans can move at runtime. We will be assuming for the rest of this paper that every sensor has the DSL interpreter installed, which may have a software implementation or can be a special hardware chip.

In the following sections we will show and explain the DSL code for a number of important problems to be solved by advanced sensor networks, which confirms an efficiency of the proposed distributed computational and control model.

3 ELEMENTARY EXAMPLE

3.1 The Task

An elementary task to be programmed in DSL may look like follows. Let it needs to go to the physical locations of a disaster zone with coordinates (using x-y pair here for simplicity) x25_y12, x40_y36, and x55_y21, measure temperature there, and transmit its value, together with exact coordinates of the locations reached, to a collection center. The corresponding program in DSL will be as follows:

\[
\text{Hop}(x25 \_y12, x40 \_y36, x55 \_y21).
\text{Transmit(Temperature \& Location)}
\]

The program moves independently to the three locations given, and in every destination reached measures temperature using special environmental variable Temperature. Using another environmental variable Location, it attaches to the previous value exact coordinates of the current physical position (which, by using GPS, may differ from the initially intended, rough coordinates). The two-value results are then independently transmitted from the three locations to a collection center.

This program reflects semantics of the task to be performed in a distributed space, regardless of possible equipment that can be used for this. The latter may, for example, be a set of sensors scattered in advance throughout the disaster zone, where hopping by coordinates may result in a wireless access of the sensors already present there—not necessarily moving into these points physically. As another solution, this program may task mobile robots (single or a group) to move into these locations in person and perform the needed measurement and transmission upon reaching the destinations.

3.2 Single-Robot Solution

Let us consider how the previous program will be executed with only a single robot available (let it be Robot 1, assuming other robots not reachable). After its injection into the robot’s interpreter (see Fig. 3), the first, broadcasting statement:

\[
\text{Hop}(x25 \_y12, x40 \_y36, x55 \_y21)
\]

will be analyzed first. It naturally splits into three independent hops, but only one can be performed at the start by a single robot.
Hop(x25_y12, x40_y36, x55_y21).
Transmit(Temperature & Location)

Figure 3: Single-robot solution.

The interpreter hops virtually into the point x25_y12 ordering robot to move into the corresponding physical location. Upon arrival, the second statement:

Transmit(Temperature & Location)

will be executed by measuring temperature, attaching coordinates, and transmitting the result via channels available. The rest of the program, represented as:

Hop(x40_y36, x55_y21).
Transmit(Temperature & Location)

will be analyzed again, with hop to x40_y36 extracted, robot moved into the second location, and measurement result transmitted as before. The program’s remainder, now as:

Hop(x55_y21).
Transmit(Temperature & Location)

will cause movement and operation in the third location x55_y21. The program terminates after shrinking to:

Transmit(Temperature & Location)

All these steps are depicted in detail in Fig. 3.

3.3 Multiple-Robot Solution

Let us consider now the case where other robots can be requested and accessed from the robot into which we injected our program (let it be Robot 1 again), see Fig. 4. After analyzing the first statement, splitting it into individual hops and attaching to each of them the replicated rest of the program (i.e. second statement) the interpreter in Robot 1 will produce the following three independent programs:

Hop(x25_y12).
Transmit(Temperature & Location)

Hop(x40_y36).
Transmit(Temperature & Location)

Hop(x55_y21).
Transmit(Temperature & Location)

Leaving one of them (say, the first) for execution in itself, Robot 1 requesting other available robots (let them be Robot 2 and Robot 3) by a wireless channel sends electronically the obtained other two programs to them. After this, all three programs will be operating in the three mentioned robots independently and in parallel.

Each robot first executes the hop statement, moving into the location given by physical coordinates, and upon reaching the destination, executes the second statement measuring temperature and detecting exact coordinates, subsequently transmitting these to the collection center, exactly as for the previous case with a single robot. All this is depicted in detail in Fig. 4.

Figure 4: Multiple-robot solution.

The shown above were only elementary examples of DSL code and rules of its execution. In other, more complex cases DSL allows us to dynamically form networked knowledge arbitrarily distributed between dynamic resources, also providing hierarchical control of distributed processes. DSL can provide description of system missions on a semantic level, with telling what, where, and why to do rather than how and who should do this, effectively abstracting from the resources (computer networks, robots,
humans) that can implement them, which can be emergent and not known in advance. These programs can be executed by any number of technical or biological processors, where organization of resources, their communication and synchronization are delegated to efficient automatic interpretation, with strict and simple implementation rules.

Application programs in DSL are often hundreds of times shorter than in conventional languages (like C or Java). The DSL programs are interpreted rather than compiled, adjusting to the execution environment at runtime, which is especially important when system components may be changed or damaged during the execution. The approach is based on the new, universal, formal system within which any distributed and parallel algorithms can be expressed and implemented.

4 COLLECTING EVENTS THROUGHOUT THE REGION

Starting from all transmitter nodes, the following program regularly (with interval of 20 sec.) covers stepwise, via local communications between sensors, the whole sensor network with a spanning forest, lifting information about observable events in each node reached. Through this forest, by the internal interpretation infrastructure, the lifted data in nodes is moved and fused upwards the spanning trees, with final results collected in transmitter nodes and sent in parallel outside the system using rule Transmit (See Fig. 5).

\[
\text{Hop(all transmitters). Loop(}
\frac{\text{Sleep(20). IDENTITY = TIME. Transmit(}}\text{Fused(data))}
\]

Globally looping in each transmitter node (rule loop), the program repeatedly navigates (rule repeat) the sensor set (possibly, in competition with navigation started from other transmitters), activating local space observation facilities in parallel with the further navigation.

The resultant forest-like coverage is guaranteed by allowing sensor nodes to be visited only once, on the first arrival in them. The hierarchical fusion rule fuse, collecting the scattered results, also removes record duplicates, as the same event can be detected by different sensors, leaving only most credible in the final result. To distinguish each new global navigation process from the previous one, it always spreads with a new identity for which, for example, current system time may be used (using environmental variables IDENTIFY and TIME of the language).

5 CREATING HIERARCHICAL INFRASTRUCTURES

In the previous program, we created the whole spanning forest for each global data collection loop, which may be costly. To optimize this process, we may first create a persistent forest infrastructure, remembering which nodes were linked to which, and then use it for a frequent regular collection and fusion of the scattered data. As the sensor neighborhood network may change over time, we can make this persistent infrastructure changeable too, updating it with some time interval (much larger, however, than the data collection one), after removing the previous infrastructure version. This can be done by the following program that regularly creates top-down oriented links named infra starting from the transmitter nodes (as shown in Fig. 6).

\[
\text{Hop(all transmitters). Loop(}
\frac{\text{Sleep(120). IDENTITY = TIME. Repeat(}}\text{Fused(data))}
\]

Figure 5: Parallel navigation and data collection.
first come). Remove links(all). Stay(create link(-infra, BACK)))

Figure 6: Runtime creation of hierarchical infrastructure.

This infrastructure creation program provides competitive asynchronous spatial processes, so each time even if the sensors do not change their positions, the resultant infrastructure may differ, as in Fig. 7.

Figure 7: Another possible infrastructure.

Having created a persistent infrastructure, we can use it frequently by the event collection program, which can be simplified now as follows:

Hop(all transmitters).
Loop(
  Sleep(20).
  Transmit(
    Fuse(
      Repeat(
        Free(observe(events));
        Hop(+infra))))))

The global infrastructure creation program (looping rarely) and the event collection and fusion one (looping frequently) can operate simultaneously, with the first one guiding the latter on the data collection routes, which may change over time.

6 ROUTING LOCAL EVENTS TO TRANSMITTERS

We have considered above the collection of distributed events in the top-down and bottom-up mode, always with the initiative stemming from root nodes of the hierarchy--the latter serving as a parallel and distributed tree-structured computer. In this section, we will show quite an opposite, fully distributed solution, where each sensor node, being an initiator itself, is regularly observing the vicinity for the case an event of interest might occur.

Having discovered the event of interest, each node independently from others launches a spatial cyclic self-routing process via the infrastructure links built before, which eventually reaches the transmitter node, bringing with it the event information, as shown in Fig. 8. The data brought to the transmitters should be fused with the data already existing there.

Figure 8: Routing scattered events to transmitters.

The corresponding program will be as follows.

Hop(all nodes).
Frontal(Transfer).
Nodal(Result).
Loop(
  Sleep(5).
  Nonempty(
    Transfer = observe(events)).
  Repeat(hop(-infra)).
  Result = Result & Transfer)
The transmitter nodes accumulating and fusing local events, arriving from sensor nodes independently, can subsequently send them outside the system. Different strategies can be used here. For example, one could be waiting until there are enough event records collected in the transmitter before sending them, and the other one waiting for a threshold time and only then sending what was accumulated (if any at all). The following program combines these two cases within one solution, where arriving data from sensors is accumulated in nodal variable Result.

\[
\text{Hop}(\text{all transmitters}). \\
\text{Loop}(
\quad \text{Or}(
\quad \text{Quantity}(\text{Result}) \geq 100,
\quad \text{Sleep}(60). \text{Result} \neq \text{nil})). \\
\text{Transmit}(\text{Result})
\]

This program in every transmitter can work in parallel with the previous program collecting events and looping in every sensor (in transmitters as well, assumed to be sensors too), and also with the earlier program, starting in transmitters, for the regular infrastructure updates.

### 7 TRACKING MOBILE OBJECTS

Let us consider some basics of using DSL for tracking mobile (say ground or aerial) objects moving through a region controlled by communicating sensors, as shown in Fig. 9. Each sensor can handle only a limited part of space, so to keep the whole observation continuous the object seen should be handed over between the neighboring sensors during its movement, along with the data accumulated during its tracking and analysis.

The space-navigating power of the model discussed can catch each object and accompany it individually, moving between the interpreters in different sensors, thus following the movement in physical space via the virtual space (Sapaty, 1999). This allows us to have an extremely compact and integral solution unattainable by other approaches based on communicating agents. The following program, starting in all sensors, catches the object it sees and follows it wherever it goes, if not seen from the current point any more (more correctly: if its visibility becomes lower than a given threshold).

\[
\text{Hop}(\text{all nodes}). \\
\text{Frontal}(\text{Threshold}) = 0.1. \\
\text{Frontal}(\text{Object}) = \text{search}(\text{aerial}). \\
\text{Visibility}(\text{Object}) > \text{Threshold}. \\
\text{Repeat}(
\quad \text{Loop}(
\quad \text{Visibility}(\text{Object}) > \text{Threshold}). \\
\quad \text{Maximum destination}(
\quad \text{Hop}({\text{directly reachable}}). \\
\quad \text{Visibility}(\text{Object}) > \text{Threshold}))
\]

The program investigates the object’s visibility in all neighboring sensors in parallel and moves control along with program code and accumulated data to the neighboring sensor seeing it best (if visibility there exceeds the threshold given).

This was only a skeleton program in DSL, showing the space tracing techniques for controlling single physical objects. It can be extended to follow collectively behaving groups of physical objects (say, flocks of animals, mobile robots, or troops). The spreading individual intelligences can cooperate in the distributed sensor space, self-optimizing jointly for the pursuit of global goals.

### 8 AVERAGING PARAMETERS FROM A REGION

Let us consider how it could be possible to assess the generalized situation in a distributed region given, say, by a set of its border coordinates, in a fully distributed way where sensors located in the region communicate with direct neighbors only. Assume, for example, that the data of interest is maximum pollution level throughout the whole region (it may also be temperature, pressure, radiation level, etc.) together with coordinates of the location showing
this maximum.

The following program, starting in all sensors located in the region, regularly measures the pollution level in its vicinity, updates local maximum and, by communication with direct neighbors, attempts to increase the recorded maximum there too. Eventually, after some time of this purely local communication activity all sensors will have the same maximum value registered in them and corresponding to the maximum on the whole region (see the overall organization in Fig. 10).

\[ \text{Nodal}(\text{Level, Max, Region}). \]
\[ \text{Frontal}(\text{Transfer}). \]
\[ \text{Region} = \text{region definition}. \]
\[ \text{Hop}(\text{all nodes, Region}). \]
\[ \text{Loop(} \]
\[ \text{Or parallel(} \]
\[ \text{Loop(} \]
\[ \text{Sleep(5)}. \]
\[ \text{Level} = \text{measure(pollution)}. \]
\[ \text{Stay(Level > Max. Max=Level)}. \]
\[ \text{Hop(directly reachable,Region)}. \]
\[ \text{Transfer > Max. Max=Transfer).} \]
\[ \text{sleep(120)).} \]
\[ \text{Level == Max.} \]
\[ \text{Transfer} = \text{Max & WHERE.} \]
\[ \text{Repeat(hop(-infra)).} \]
\[ \text{Transmit(Transfer))} \]

As there may be many sensors located in the region of interest, we will need forwarding only a single copy of this resultant maximum value to a transmitter for an output. This can be achieved by delegating this task only to the sensor whose measured local value is equal to the accumulated maximum in it, which corresponds to the overall region’s maximum.

Having discovered that it is the leader (after a certain time delay), such a sensor organizes repeated movement to the nearest transmitter via the earlier created virtual infrastructure, carrying the resultant maximum value in frontal variable Transfer, and sending it outside the system in the transmitter reached, as shown in Fig. 10.

Similarly, fully distributed, organization may be introduced for finding averaged values, or even for assembling the global picture of the whole region with any details collected by individual sensors (the latter may be costly, however, with a more practical solution skeleton shown in the next section).

9 ASSEMBLING FULL PICTURE OF THE REGION

To collect details from some region via local sensors and merge them into the whole picture could, in principle, be possible via local single-level exchanges only, as in the previous section, but the amount of communications and data transfer as well as time needed may be unacceptably high. We were finding only a single value (maximum) via frequent internode communications, with minimal code length.

But for obtaining the detailed global picture of the region or of some distributed phenomenon, we may have to gradually paint (grow) this picture in every sensor node simultaneously, with high communication intensity between the nodes. Also, there may be difficult to determine the completeness of this picture staying in local sensors only. A higher integrity and hierarchical process structuring may be needed to see a distributed picture via the dispersed sensors with limited visual capabilities and casual communications.

Different higher-level approaches can be proposed and expressed in DSL for this. We will show only a possible skeleton with spanning tree coverage of the distributed phenomenon and hierarchical collection, merging, and fusing partial results into the global picture. The latter will be forwarded to the nearest transmitter via the previously created infrastructure (using links infra), as in Fig. 11.

\[ \text{Hop(random, all nodes,} \]
\[ \text{detected(phenomenon)).} \]
\[ \text{Loop(} \]
\[ \text{Frontal(Full) = fuse(} \]

Figure 10: Distributed averaging with active routing.
Repeat(
  Free(collect(phenomenon));
  Hop(directly reachable, first come, detected(phenomenon)));
Repeat(hop(-infra)).
Transmit(Full))

In the more complex situations, which can be effectively programmed in DSL too, we may have a number of simultaneously existing phenomena, which can intersect in a distributed space. We may also face combined phenomena integrating features of different ones. The phenomena (like flocks of birds, manned or unmanned groups or armies, spreading fire or flooding) covering certain regions may change in size and shape, they may also move as a whole preserving internal organization, etc.

In the previous language versions (Sapaty, 1999, 2005; Sapaty et al., 2006), a variety of complex topological problems in computer networks were investigated and successfully programmed in a fully distributed and parallel manner, which included connectivity, graph patterns matching, weak and strong components like articulation points and cliques, also diameter and radius, optimum routing tables, etc., as well as topological self-recovery after indiscriminate damages (Sapaty, 1999).

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10 CONCLUSIONS

We have presented a universal and flexible approach of how to convert distributed sensor networks with limited resources in nodes and casual communications into a universal spatial machine capable of not only collecting and forwarding data but also solving complex computational and logical problems as well as making autonomous decisions in distributed environments.

The approach is based on quite a different type of high-level language allowing us to represent system solutions in the form of integral seamless spatial processes navigating and covering distributed worlds at runtime. This makes parallel and distributed application programs extremely short, which may be especially useful for the energy saving communications between sensors.

The code compactness and simplicity are achieved because most of traditional synchronization and data or agent exchanges (which are also on a high level, with minimum code sent) are shifted to efficient automatic implementation, allowing us concentrate on global strategies and global solutions instead.

REFERENCES