

Using Simulated Partial Dynamic Run-Time Reconfiguration to Share Embedded FPGA Compute and Power Resources across a Swarm of Unpiloted Airborne Vehicles

David Kearney and Mark Jasiunas

Reconfigurable Computing Laboratory, School of Computer and Information Science, University of South Australia, Mawson Lakes Boulevard, Mawson Lakes, South Australia 5095, Australia

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We show how the limited electrical power and FPGA compute resources available in a swarm of small UAVs can be shared by moving FPGA tasks from one UAV to another. A software and hardware infrastructure that supports the mobility of embedded FPGA applications on a single FPGA chip and across a group of networked FPGA chips is an integral part of the work described here. It is shown how to allocate a single FPGA's resources at run time and to share a single device through the use of application checkpointing, a memory controller, and an on-chip run-time reconfigurable network. A prototype distributed operating system is described for managing mobile applications across the swarm based on the contents of a fuzzy rule base. It can move applications between UAVs in order to equalize power use or to enable the continuous replenishment of fully fueled planes into the swarm.

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1. INTRODUCTION

The term swarm is usually identified with a group of living organisms who arrange themselves to cooperate to achieve a common task that no one of them could complete as an individual. For example, a swarm of birds may fly in a slipstream formation to save on energy or a swarm of ants will construct a shortest spanning tree path between a food source and their nest [1]. UAVs that cooperate to achieve a common task (such as geolocation) in an autonomous way (using agents) have been given by analogy the title of swarm in this paper.

Small UAVs (of weight less than 25 kg and wingspan less than 3 m) are often limited by their resources as compared with larger manned and unmanned planes. For example, some small UAVs rely on battery power for both the engine and electronics whilst others use conventional internal combustion engines with battery/generator system that allows energy conversion from fuel to electricity but small UAVs only require modest fuel inputs to maintain level flight and thus the power requirements of the computing resources can still consume a substantial amount of the available energy and reduce the range and endurance time of the plane.

In this paper, we introduce the concepts of sharing a single FPGA among different tasks that may not need to execute at the same time and allowing such tasks to migrate between members of the swarm either to share power across the swarm or provide for the replacement of members of the swarm who may need refuelling without stopping the execution of tasks critical to the swarms mission.

The paper is organized as follows. In Section 2, we review the literature on capabilities and applications of small UAVs and the compute platforms they might use. We examine publications that report the benefits swarms of UAVs. We show that whilst there have been many publications of swarm applications, there has been less attention to the resource sharing possibilities of swarms especially extensions to compute sharing and power sharing. In Section 3, we introduce a typical scenario where power and FPGA computer resource sharing could be beneficial in a swarm of UAVs performing a surveillance function.

Section 4 presents work showing how a single FPGA can be shared amongst several compute tasks that are relevant to UAV applications. This is the first time an operating system for reconfigurable computing has been implemented to execute practical embedded applications.

Section 5 introduces infrastructure for mobility of applications between UAVs. We explain why we have opted for agent-based decentralized control of mobility and fuzzy rules for the decision making. We describe check pointing of applications.

2. PREVIOUS WORK AND REVIEW OF LITERATURE

The review of literature first discusses the capabilities of and applications to which small UAVs have been applied. We describe the computing requirements for a small UAV performing these applications. We show from the literature that scarce resources for small UAVs include electrical power and high-performance computing capability. We give examples from the literature that show how power can be minimized and computing capability maximized on a single UAV by the use of FPGAs on UAVs in preference to more traditional software only embedded systems.

Next we investigate the advantages that a swarm of UAVs has over single platforms in overcoming small UAV limitations. We give examples of how a swarm can improve application performance in geolocation by using the diversity of sensor locations. We highlight that there is no literature of the use of a swarm to share the scarce resources that support these types of applications. In particular, there has been no investigation of the sharing of power and high-performance embedded computing resources across the swarm.

Next we review the literature on the sharing of the types of embedded FPGA compute resources that are used on small UAVs. Using our definition of partial dynamic run-time reconfiguration, we show how published operating systems for reconfigurable computing might allow the sharing of FPGA resources among many applications in UAVs applications. We note that the literature does not contain specific work on the extension of FPGA application sharing in a distributed sense across several FPGAs. These topics are the subject of this paper.

2.1. Capabilities and applications of single small UAVs

In this section, we describe how small UAVs have been used in civilian and defence roles. We illustrate both the advantages and limitations of small UAVs working alone.

Unmanned airborne vehicles are projected to become a major segment of the aviation industry over the next 20 years [2], primarily enabled by developments in computing, communications, and sensor technologies. An area where UAVs will likely make a major impact is in surveillance and remote data collection. Examples of applications include fire ground (active bushfire) surveillance, crop and vegetation surveying [3], emergency data communications and maintaining the security of people, and assets against terrorist-related threats [4]. Small UAVs (of gross mass less than 25 kg) will most likely perform these tasks, working together in closely co-located teams called swarms. This is because swarms can carry a range of sensors, and their diversity overcomes the limited field of view of a single small UAV flying at

a relatively low altitude. Swarms also provide increased reliability through redundancy.

The sensors used on small UAVs have in the past been confined to very light-weight devices. For example, video cameras and small RF sensors are quite practical on small UAVs. However, it is clear from studies conducted on large UAVs [5] and satellites that more complex sensors such as infrared imagers could provide a major improvement in the quality of information that can be gathered [6].

The 2002 NASA project used the solar power pathfinder UAV to demonstrate crop monitoring over the coffee plantations in Hawaii [3]. This UAV is capable of extremely long loitering times which were used to map weed invasions as well as irrigation and fertilization irregularities. This project also demonstrated how UAVs can plan flight paths to avoid obstructed view of the ground by cloud cover. NASA has also used APV-3 UAVs to survey vineyards in Monterey California where up to \$12.5 million in produce is lost annually due to frost damage [7]. The UAV collected hyper-spectral imagery which was relayed to ground stations where data was combined with information gathered from ground sensors.

2.2. FPGAs as compute platforms for small UAVs

A reconfigurable computer is a processing platform consisting of a general purpose processor interfaced to memory and a programmable logic device PLD [8]. The most widely used PLD is a field programmable gate array (FPGA) [9]. An FPGA is an array of logic cells connected via programmable routing. Each logic cell can be configured to perform logic functions allowing complex circuits to be constructed. FPGAs are ideal for implementing common types of algorithms on UAVs [10–14].

Sharing an FPGA amongst several applications dynamically is a relatively new concept in the reconfigurable computing field. This was first proposed by Wigley and Kearney [15] who defined the basic required components, being allocation, partitioning, placement, and routing. Allocation, partitioning and placement algorithms have been further explored in [16–18], and routing and on-chip networks in [19, 20].

2.3. Advantages of swarms of UAVs

In this section, we describe the advantages of small UAVs. It is shown using example applications how swarms can increase the capabilities of such UAVs.

Small inexpensive UAVs have been found useful in military roles. They can be considered somewhat expendable, allowing swarms to operate in closer proximity to threats where sensors and effectors are more effective and operate using less power [21]. One such area of research is electronic warfare where the goal is to gather information and suppress the enemy's information gathering using electronic sensors and effectors (jamming). For example, several UAVs can be used to geolocate the position of radar emitters for suppression [22]. A UAV can fly much closer to a radar emitter making jamming possible at very low power. While the prospect

of armed UAVs in combat roles has been explored, the current focus remains on intelligence, surveillance, and reconnaissance missions [23].

Geolocation is a good example of the benefits of swarms. It requires the cooperation and exchange of information between several UAVs. Geo-location works by taking a directional bearing of an object from a number of different locations and combining them to determine the objects' exact position. Finn et al. describe how a group of 6 sensors can reduce the location error by more than 80% (Figure 1) [21].

2.4. Sharing resources in a swarm: a typical scenario

The missions of UAV swarms can be divided into two classes. In the single mission, we have a swarm requiring N planes each with different capabilities to perform the swarm function. We have just N planes available. We deploy these planes and attempt to arrange their computing tasks so that all planes run out of fuel at the same time. Allowing for fuel to return to base (assumed the same for each plane) we end the deployment when each plane has just this much fuel left. The aim is to maximize the time that the swarm is deployed over the target area doing useful work.

In the continuous mission scenario N planes are required to form the swarm but we assume that we have $N + 1$ or more planes available. Thus it is possible to maintain a continuous mission by retiring planes from the swarm that are running low on fuel and replacing them with other planes with a full fuel load. The objective in this case is for example to maintain continuous surveillance over the target area. Task mobility is essential in the continuous mission scenario. In the following we describe why this is the case.

If the computing tasks that the swarm must execute are stateful applications like tracking [6] the continuous mission is only feasible if task state can be migrated from the members of the swarm that are running low on fuel to those that are replacing them. Thus task mobility is required for this type of mission to be feasible. In the single mission case task mobility is not strictly necessary for feasibility. Tasks can be loaded on each member of the swarm. The swarm will then remain aloft till the first plane in the swarm losses power. Then the whole swarm must return to base. It might seem possible therefore to plan so that each plane has exactly the fuel loaded for the tasks needed to perform if you know in advance the workload that the swarm will encounter. However, we do not know in advance the workload of the swarm in many practical situations. For example, imagine that the task of the swarm is to perform surveillance. This application consists of a continuous task of scanning the seas below. UAV1 looks for an object using a low power visible CMOS camera. When the object is identified, then a high power periodic task is invoked to gain an alternative image of the object using an IR sensor on UAV2. The relative power consumption depends on how often the IR sensor is used during the mission. Because we cannot predict how many objects will be detected on the mission, we cannot predict the relative power consumption between the UAV1 and UAV2 due to the difference in the power required to operate the sensors.

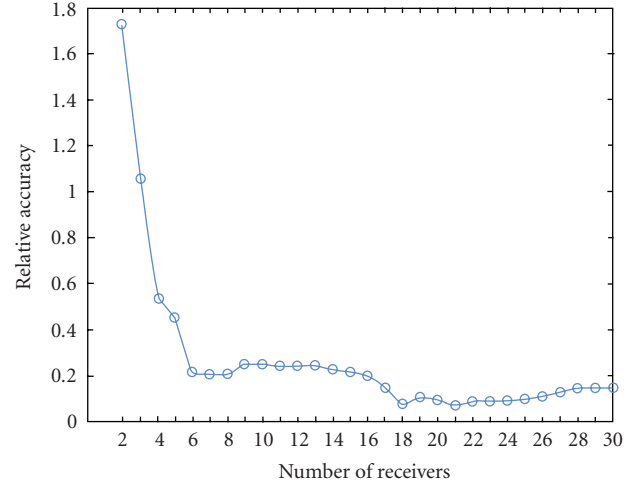


FIGURE 1: Reduction in the location error margin (Y-axis) with the number of sensors (X-axis) used to determine the location [21].

Thus in the absence of task mobility it could be expected that one UAV would run of power sooner than the other. If we have task mobility, then we can equalize the power between the UAVs.

2.5. Agents mobility and mobile agents

The question now arises as to how we can arrange for this mobility to happen. We have decided to use the agent paradigm to express and control this mobility. It is generally accepted that an agent must possess at a minimum the properties of *autonomy*, *social interaction*, *reactivity*, and *proactiveness* [24]. *Mobile agents* are a special class of agents that are able to migrate between host computer systems while execution [25]. Mobile agents are not able to function without the support of an *agent environment* that executes on host systems and aids in the migration process. In the remainder of this section, the key properties of agents are examined in greater detail.

The autonomous operation of agents in dynamic systems are one of their most attractive features. An autonomous agent is entrusted to act and decide on courses of action without being specifically directed by the user [26]. This ability of agents is especially useful in dynamic environments where deterministic processes or agents would require constant instruction from the user. Milojevic et al. [27] defines the transfer of authority to act on a user behalf as the defining attribute of mobile agents when compared to other forms of mobile code and execution.

The agent paradigm implies a degree of interaction between agents and external entities. Social interactions are implemented by exchanging messages formatted in an agent communication language [28]. The messages can contain information or coordination of activities where agents are collaborating to achieve common goals. Through teambuilding, individual agents have the ability to increase their effectiveness by cooperative coordination in order to achieve

common goals [29]. In agent environments with restricted resources, selective teambuilding and coordination can maximize the usage of resources.

2.6. Conclusion

FPGAs are an appropriate platform for small UAVs because they have low power requirements yet can compute high complexity tasks such as image processing. Small UAVs are best arranged as swarms so that the limited capabilities of each member of the swarm can be combined. Once a swarm is established all members of the swarm need to be present to perform the task. We have shown that mobility of FPGA tasks between members of the swarm will allow a swarm to be active for a longer period of time or allow the continuous replacement of members of the swarm. The literature survey shows that there have been no examples of FPGA task mobility and within the context of embedded systems for UAVs there are no examples of the sharing of tasks even on a single UAV. These topics are the subject of the rest of this paper.

3. SHARING UAV COMPUTING TASKS ON A SINGLE FPGA

3.1. Introduction

Embedded applications designed for implementation on an FPGA have traditionally had exclusive use of the resources of the device. As FPGA devices get bigger it is now feasible to load many compute tasks onto a single FPGA. In UAV applications, however, not all tasks need to be active at once. Sharing the resource by not loading tasks till they are needed and removing them when complete can save power and improve the overall flexibility of the UAV as a compute platform. In the scenario defined above, compute tasks can be loaded onto the FPGA in a sequence that is not known at design time. This requirement fundamentally changes the way tasks must be designed because no task will know in advance exactly which FPGA resources are available when it begins to execute. An operating system [15] (or run time system) performs these resource allocations dynamically. Despite the extensive research performed on these systems [30–37], there has been no demonstration of practical UAV embedded computing tasks actually being controlled by such a run time system. In this section, we describe a practical demonstration of an embedded operating system for FPGAs working with tasks relevant to UAVs. Firstly, the basic elements of the operating system for embedded FPGAs are described. We then show why true partial dynamic run time reconfiguration for the practical tasks needed on a UAV cannot be achieved with current FPGA hardware. We describe how check pointing of applications together with caching of whole chip configurations can overcome most of these limitations. The implementation of the memory arbitration and run-time reconfigurable on-chip network required to support practical applications running under the operating system are then detailed. Compute tasks intended to execute under the operating system

have special requirements (including the ability to be check-pointed) and these are explained next. Finally, we give details of the actual tasks that we have demonstrated running under the operating system.

3.2. Basic elements of an embedded operating system for FPGAs

This section describes the basic elements of an operating system for reconfigurable computing that allows resources to be shared on a single FPGA and resource allocation decision to be made at run time. The unique phase of FPGA resource allocation is area allocation and Figure 2 provides a pictorial representation of this phase.

When the application is to begin, the circuit is given to the operating system. The operating system uses precompiled information about the circuit along with the current state of the FPGA to write a set of constraints which define what resources each circuit will be allocated to. These constraints along with the circuit descriptions are then used to generate a configuration for the FPGA. Resource allocation algorithms are required to determine the region of unoccupied resources that can be used to implement the hardware task. Being able to place tasks arbitrarily makes best use of the FPGA area as opposed to tile-based placement which generates internal fragmentation when small tasks are placed within large fixed tiles.

If the FPGA is considered a rectangular region, and hardware tasks are polygons defining an area which contain the resources necessary to run a particular task, the allocation problem can be reduced to a geometric packing problem. The aim of an allocation algorithm is to place the polygon tasks into a 2D area as efficiently as possible. The possible allocation algorithms vary greatly in efficiency, complexity, and functionality. The time critical nature of dynamic IP core placement means that many allocation algorithms widely used for offline placement, such as simulated annealing [38], cannot be used. Some candidates for online allocation algorithms are Best Fit, Bottom Left [39], Bazargan's fast template placement [40], and the Minkowski sum [17]. It has been shown that the Minkowski sum is the fastest algorithm in execution and has acceptable performance by not fragmenting the space on the FPGA.

The Minkowski sum is a useful geometric algorithm which can identify the perimeter of a region of space where a task can be successfully placed without interfering with other tasks. Once the Minkowski sum has been used to identify the area where a valid placement can be performed, the bottom-left most position is selected for allocation. The Minkowski algorithm has two major advantages over virtually all other allocation algorithms. First is that the algorithm can correctly allocate nonrectangular cores, whereas other algorithms must place non-rectangular shapes within a rectangle for placement causing further area fragmentation.

Secondly, the algorithm naturally handles holes in the free space which are commonly created when tasks end their execution. Finally, the Minkowski sum is linear in complexity for rectangular polygons, but increases in complexity for

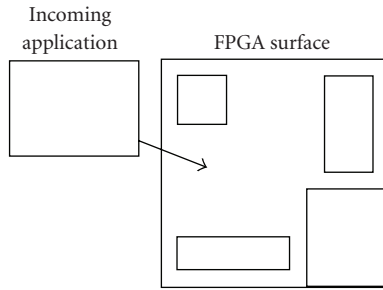


FIGURE 2: The allocation phase of an operating system. The incoming application must be placed on the FPGA so it does not contend with resources of existing applications.

polygons with more edges. The worst case complexity, when the cores to be placed are nonrectangular and concave, is $O(m^2n^2)$, where m is the number of vertices in the union of the placed cores and n is the number of vertices of the core to be placed—nonrectangular concave shapes are not common in real reconfigurable computing applications.

In this section, we have described the basic operations performed by an operating system for reconfigurable computing.

3.3. Dynamic partial reconfiguration and real FPGAs

The most general way an FPGA can be reconfigured is denoted in this paper as partial dynamic run-time reconfiguration. This term is defined to mean that an FPGA with tasks loaded and executed on it can have part of its area reconfigured without necessarily stopping the existing tasks. We note that whilst there have been publications reporting partial dynamic run-time reconfiguration on a limited scale [41], the current architecture FPGAs have numerous constraints which prevent this operation for practical size circuits. We will now describe why this is the case.

For the purpose of this discussion, the configuration process of the popular Xilinx Virtex family of FPGAs is adopted as an example, as the configuration of an FPGA is family dependent. Reconfiguration of Virtex chips is column-based with frames within each column being able to be reconfigured atomically. This means that reconfiguration of any part of a thin vertical column of a chip implies stopping any circuit that intersects with this column. Whilst one could wait for circuits to complete current tasks before triggering any reconfiguration operation this would add arbitrary latency to the start up of any task which is impractical in the real time applications that are commonly used on UAVs.

In order to be able to reconfigure parts of the FPGA without affecting circuits already executing that are not going to be reconfigured, the interconnects which communicate to logic within the area being reconfigured and logic elsewhere on the chip must be able to be hot swapped using a mechanism such as tristate buffers or LUT-based macros [42]. Typically the number and location of these types of tristate buffers severely constrains the way tasks can be configured

on the FPGA. The LUT-based macroapproach implies fixed tile-based layout which suffers from internal tile fragmentation since the maximum size of tasks placed is not known in advance.

The current absence of a practical mechanism allowing partial dynamic run-time reconfigurations of arbitrary-shaped regions of the device has led us to propose a compromise which we call simulated partial dynamic reconfiguration. In this situation, existing applications on the FPGA are checkpointed and the entire FPGA is reconfigured. After new tasks are added and old tasks removed, then all currently active tasks are started. Because checkpointing is a key to making simulated partial dynamic reconfiguration possible, the next section explains how checkpointing is achieved for practical UAV applications.

3.4. Checkpointing

The consequences of using simulated partial dynamic reconfiguration to load and remove hardware tasks is that every resource on the device is reprogrammed with new configuration data and in the process overwrites all currently executing tasks. During this process, the state of tasks executing is lost. A mechanism for preserving state during reconfiguration is required. We call this process checkpointing of applications.

There are two options for checkpointing. In the first, which we call cooperative checkpointing, the operating system tells tasks that a reconfiguration is required and waits for all tasks to reach their checkpoints. In the second, which we call preemptive checkpointing, tasks periodically do their own checkpointing allowing them to be restarted at that checkpoint even if reconfiguration is forced at an arbitrary time.

In cooperative checkpointing, the latency between when the operating system requests a reconfiguration and when the last task completes its checkpointing is unbounded and can not be known in advance. There is a chance that poorly designed tasks may never reach a checkpoint thereby freezing the operating system. There is an area overhead in cooperative checkpointing because extra circuitry must be provided to preserve the state of the circuit. For pre-emptive checkpointing, there is no latency for the operating system in requesting a reconfiguration because all tasks can be stopped immediately; a reconfiguration becomes necessary. There is also an area overhead in the pre-emptive checkpointing of applications which is the same as the area used in cooperative checkpointing. It might be imagined that pre-emptive checkpointing would slow applications down because of the time overhead of periodic saving of state. However, the periodic saving of state can be executed in parallel in many applications with the normal computation of the task and thus this overhead can be minimized. Pre-emptive checkpointing is easier for application developers to manage because there is no need to interface to a special reconfiguration interrupt coming from the operating system. For the reasons listed above, we have implemented the pre-emptive checkpointing detailed above. In the next paragraph, we detail how this has been implemented.

For pre-emptive checkpointing, the application is decomposed into groups of logic which represent atomic operations. These atomic operations then become states in a state machine. Each time the machine transitions into a new state, the variables that make up the state of the application are stored in external memory. The application performs processing within a loop. At each iteration this state is updated. At any point the application can thus be terminated. When the application resumes execution it will restart from the beginning of the last checkpoint.

We have investigated pre-emptive checkpointing in the three applications that were implemented in our operating system. The first application was feature tracking. In this application, video scenes are searched for a collection of adjacent pixels with common characteristics. If such a collection of pixels is located, the coordinates are calculated as output. For this application, checkpointing is not necessary because there is no state retained between one video frame and the next. This means that if reconfiguration is initiated in the middle of a frame, the data will be lost and the data from the next available frame will be calculated. The second application we investigated was Sobel edge enhancement. In this application, a buffer of frames is processed by the algorithm to generate the output. Checkpointing only requires the recording of which frame is required to be analyzed. If the edge detection of the required frame is interrupted, it is only necessary to go back and recover the input frame from a buffer and restart the edge detection from this frame. The final application we implemented a data encryption algorithm. Checkpointing this application is similar to Sobel application because it is only necessary to remember what block was being processed before processing was interrupted by the reconfiguration. More complex application such as a correlation tracker [6] will require more checkpoints as the data is processed iteratively to produce the result. In such an application, checkpointing after each iteration is required.

3.5. Sharing resources amongst applications

It has just been shown that constraint files and geometric allocation algorithms can be used to confine the logic resources of hardware circuits to mutually exclusive regions and that checkpointing can enable simulated partial dynamic reconfiguration so that circuits can be swapped on and off an FPGA. In this section, the interconnection and arbitration between these logic circuits are considered.

There are three components required for multiple circuits to access shared external (off-chip) memory; a network, an arbitrator, and a memory partitioning policy. The network specifies the interface that tasks must connect to in order to communicate with the arbitrator. The network specifications include both wiring definitions and protocols for read and write requests. The design of the network itself is one of the most influential components of the operating system as far as performance is concerned, as the design will determine the data throughput between the memory banks and the processing circuit. The on-chip network connects the applications to the arbiter which controls the access to memory

and resolves contention. The memory partitioning policy determines how the applications share the available memory. These components and their implementations are now discussed.

3.5.1. Memory network

An on-chip network is used to connect the memory arbiter to the applications. Six network topologies that are candidates for implementation, bus, star, mesh, ring, tree, and fat tree, are described by Kearney and Veldman [19]. Each is investigated for its suitability for implementation specifically for the UAV swarm environment. In evaluating the topologies, the following criteria are considered.

Ease of implementation

How difficult is this topology to implement natively on an FPGA given that the network must be dynamically reconfigured?

Wire routing cost

How expensive is it to route wires to a new application in the topology? Some topologies require many wires to be run over large distances on the chip, which is a very expensive operation in an FPGA environment.

Concurrency

How well does this topology support concurrency? The topology should allow, for example, multiple memory banks that are connected to the FPGA to be accessed simultaneously.

Latency

What is the latency and how does it vary as applications join the network?

Scalability

How does this topology scale for large numbers of applications? How does the latency or wire routing cost complexity increase as more applications are added to the network?

Impact on area allocation

The network must work in an environment where cores arrive and must be dynamically placed on the FPGA. How does the topology constrain the locations possible for a new application? Allocation algorithms suggested in [17] favor locations that minimize the amount of area fragmentation because fragmented area is not available for new applications. How will the new network topology interact if we allow the allocator to favor locations that need shorter and therefore cheaper routes to the network and reduce the fragmentation of area to a minimum?

TABLE 1: Evaluation of network topologies + means favourable – unfavourable +/- neutral.

| | East of implementation | Wire routing cost | Concurrency | Latency | Scalability |
|----------|------------------------|-------------------|-------------|---------|-------------|
| Bus | ++ | – | -- | – | -- |
| Star | ++ | -- | ++ | ++ | -- |
| Mesh | -- | -- | + | + | + |
| Ring | ++ | + | – | +/- | +/- |
| Tree | +/- | ++ | +/- | + | + |
| Fat tree | – | – | + | + | + |

In the wire complexity criteria when a new application is added, the star is not favoured because it requires new global routes to the arbiter. The arbiter will be near the edge of the chip because of the need for access to wide memory busses so these new routes may need to cross the chip. The bus is better than the star because only new global arbitration lines must be added to the arbiter; the remainder of the bus can just be extended. The ring and the tree are particularly easy to extend. The fat tree may require new bandwidth at its root for the addition at some locations which may precipitate further reorganization in a dynamic environment. The concurrency criterion favours the more complex topologies such as mesh and fat tree and directly conflicts with the recommendations of ease of implementation and wire complexity. This means that to use a bus (and to a lesser extent a ring) there may be a need to duplicate channels to maintain a reasonable level of concurrency.

The latency criterion does not strongly favor any topology although the predictability of the latency varies quite markedly for some solutions like the mesh depending on the number of hops between the source and destination of the packets. The scalability results are also more uniform. The bus suffers from poor latency scalability.

The impact on the area allocation is quite varied. For the bus, placing applications somewhere near an existing bus on the chip is favourable. This is a simple distance metric. For the star new applications must be placed so as not to block future applications from reaching the memory arbiter. It is expected that this means starting allocation at the largest distance away from the memory arbiter which is straightforward to calculate. A minimization of distance and number of hops to the memory arbiter in the mesh option could be used to guide allocation. The ring is similar to the bus, finding a location near an existing ring and if needed extending the ring outwards is straightforward for the allocator. With trees there is a complex tradeoff between putting new applications as few hops from the arbitrator as possible and avoiding congestion at the root. The tree is thus quite hard to interface to the allocator and the interactions will be more complex than the star. A summary of common allocation algorithms is shown in Table 1.

In the specific case of UAV swarms where typically a small number of high throughput real-time applications share a set

of memory banks, the key attributes are concurrency and latency. For this reason, the star is the favoured topology. The relatively low wire routing cost and poor scalability is not expected to affect the systems performance since few applications are expected to be executing concurrently on small UAVs.

3.5.2. Memory allocation and arbitration policies

The task of a memory arbiter is to control access from several applications to shared external memory. A variety of different policies to deal with contention can be implemented and are discussed in [19].

Memory allocation can be done either statically or dynamically. In static allocation, the available memory is divided into partitions which are allocated to the tasks. In dynamic allocation, memory is assigned to tasks as needed resulting in more efficient use of the memory resources. Although arguably advantageous, dynamic allocation is significantly more complex in hardware environments, and to date there has been little research in this field.

3.5.3. Implementation of resource arbitration

A memory arbiter was developed as part of the prototype of the operating system for UAVs. This was run on a reconfigurable computer consisting of a Celoxica RC1000 development board fitted to a low power PC motherboard. The RC1000 board has 4 memory banks each with 2 MB of memory connected to the FPGA device. Each bank of memory can be read/written by either the host or the FPGA after the memory bank has been requested. The memory controller used by the operating system uses static allocation which means that it divides each memory bank into fixed 1 MB blocks each of which is allocated to a separate application. This allows 8 tasks to run concurrently on the FPGA. There are two primary functions that the memory controller must perform. First, read/write requests to common memory banks must be arbitrated, and second, local addresses must be converted to global addresses. The components of the memory network are shown in Figure 3.

The arbiter implemented a round-robin algorithm to arbitrate read and write requests from the applications. Figure 4 shows a diagram of the memory arbitrator and applications connected in a star topology.

The on-chip network interface includes a data bus, an address bus, a command bus for specifying read, write, or stream operations, a clock line which is used to provide applications access to the FPGAs clock and several control lines.

3.6. Experience running the applications under the OS

The operating system for reconfigurable computing has been tested for its suitability for UAV applications by implementing a scenario that will put the operating system under similar loads to what is expected if it were mounted in a UAV. The application scenario has three stages of execution, each time running a different set of algorithms on the FPGA. These

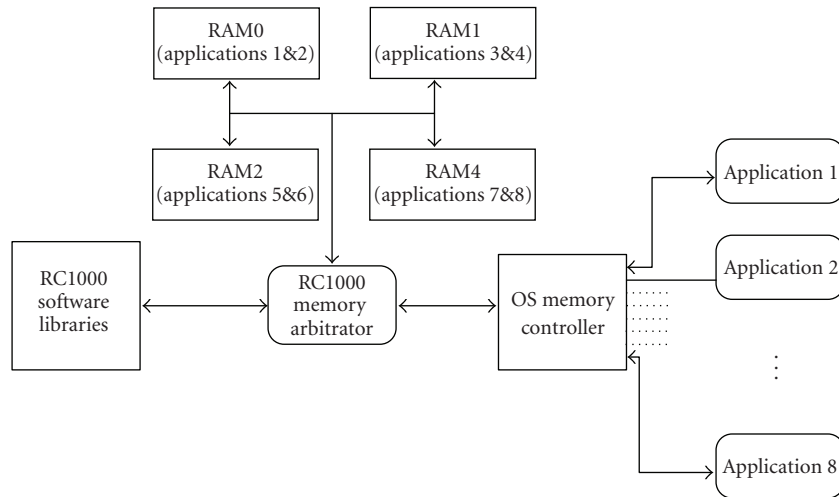


FIGURE 3: Components of the memory network.

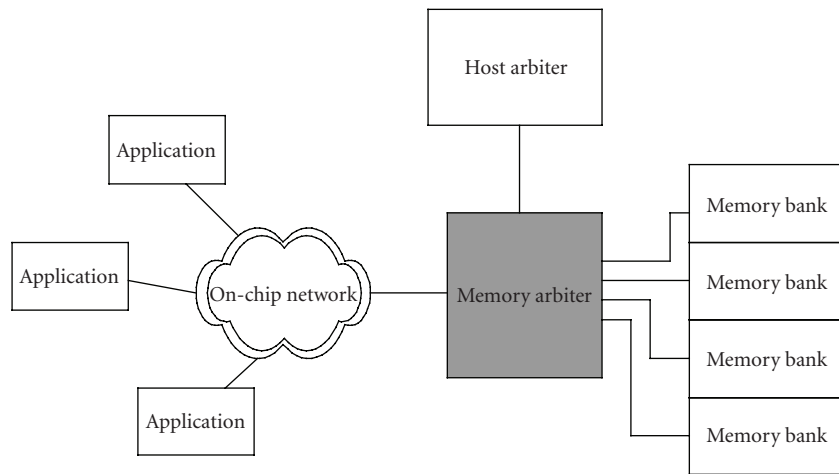


FIGURE 4: A star network configuration is used to implement the on-chip network for use in UAV swarms for its ease of implementation, support of concurrency, and low latency. The poor scalability of this topology is not expected to become an issue due to the small number of concurrent applications executing on UAVs.

algorithms have been selected as typical of the sort useful on UAVs.

The application simulates a common reconnaissance role of a UAV. In such roles, UAVs are often used to acquire data that is used to help decision making on the ground. Because of the limited bandwidth between the sensors on the UAVs and ground stations, it is often desirable to reduce the quantity of data that is sent. For example, in a typical mission lasting several hours it is quite possible that a UAV will be tracking objects for a time period of only few minutes. It makes sense only to consider these few minutes of tracking to for relaying to ground stations.

The goal of this application then is to process an incoming stream of video and detect when objects of interest are

in the field of view. Once detected by a tracking algorithm which has been tuned to track just those objects of interest, the video stream is passed through an edge enhancement filter (Sobel filter) and then into a buffer. Once the buffer is full, it is encrypted and then placed in an output buffer ready for transmission to the ground station. Each of the algorithms is implemented as a reconfigurable computing algorithm managed by the operating system.

Input data was generated for the applications and the performance of the system in terms of application, and total system throughput was measured in two configurations. In the first case, each application had memory allocated in separate memory banks. In the second case, applications shared a memory bank. In the case of shared memory with the tracking

and Sobel algorithms running in parallel, the tracking algorithm suffered a 40% loss in throughput due to contention of the memory bank. With the tracking executing concurrently with encryption, tracking throughput was reduced by 8%. In both cases, however, the total throughput of the system was greater when multiple tasks are executing. Although it is clearly desirable to have a memory bank dedicated to each application, the performance loss due to contention is acceptable and applications remain able to perform their tasks. An example of the application and FPGA utilization is shown in Figure 5.

3.7. Conclusion

In this section, we have described the components that are required for the run-time loading and unloading of circuits on an FPGA using an operating system. Checkpointing has been used as a means to allow simulated partial dynamic reconfiguration in the absence of a practical partial dynamic reconfiguration mechanism. The Minkowski sum algorithm is used to identify locations of free resources for the execution of new circuits, which are then connected to external memory by an on-chip network and memory arbiter. This has been implemented and it has been shown capable of executing practical UAV applications.

4. SHARING FPGA COMPUTING AND POWER RESOURCES ACROSS A SWARM OF UAVs

Sharing a single FPGA among many embedded tasks, allowing them to be loaded at any time, is a necessary first step to making these tasks mobile across a swarm of UAVs each of which is fitted with an FPGA. In this section, we explain how the operating system is extended to support this mobility. In the next section, the autonomous agent-based design of the distributed operating system and the fuzzy rule base that controls task migration are described. An agent-based environment has been chosen for the swarm because it allows members of the swarm to be considered as disposable in a way that does not place the whole swarm in jeopardy. The behavior of each agent in an autonomous agent-based environments is usually governed by rules which are specific to each agent. We describe how we have adopted a fuzzy rule base for our agents.

4.1. Using agents for resource sharing

In this section, the justification for using agents is presented and the consequences for this choice on the swarm are explained. A swarm of UAVs is a collection of many different types of resources ranging from platforms, to sensors and effectors, to processing units. To best make use of these, they must be interconnected in such a way as to enable them to not only share the resource, but manage it responsibly. This requires coordination in resource allocation which involves balancing the needs of applications with other resources such as power and bandwidth. Although there are many ways in which this can be implemented, the nature of a swarm makes

any form of centralized control undesirable as it introduces a single point of failure in a system prone to unreliability.

Computing agents are a distributed computing paradigm that suits such environments. Agents are a subclass of computer programs that exhibit the properties of autonomy, social ability, reactivity, and proactiveness. The agents can be further categorized as mobile or static agents. A static agent may represent a resource such as a camera which is fixed to a platform whereas mobile agents represent applications that may move their execution between platforms. Unlike many other distributed computing paradigms, mobile agents allow the transfer of state, not just execution, between nodes. This is done under the agents own control, which allows applications to customize migration rules which can further enhance the advantages of the distributed system by taking advantage of application specific knowledge. The behavior of an agent is specified as a set of basic rules that govern its behavior. A static sensor, for example, might have rules which specify under what conditions it should share data with an application. The agent may rank connected applications in order of mission priority and throttle the bandwidth of trivial applications in favor of mission critical tasks.

When developing resources or applications as agents in our network, the implementation is connected to the network by an agent interface. A skeleton agent interface provides basic communication functionality allowing the resource to be visible and accessible by other networked agents. Agents are defined on the network by their location and abilities. These are used at the time of creation to create a unique tuple which identifies that agent on the node. The tuple is defined as *sequence number*, *home node*, *class*, *current node*, *ability list*, where “*sequence number*” is a unique identification number with respect to the “*home node*,” which is the node that the agent was created on. “*class*” is the type of agent, “*current node*” is the node that the agent currently executes on, and finally the “*ability list*” describes the agent’s abilities.

We have observed that the aggregation of agents will produce emergent behaviors which can be guided by rules to achieve some overriding objective such as equalizing the power available in the swarm.

4.2. The UAV swarm agent environment

In this section, the infrastructure that supports the agent environment is described.

In order for these agents to be useful they must exist in a networked environment that supports their basic requirements, which are

- (i) discovery of other agents,
- (ii) communication with other agents,
- (iii) providing information about other nodes,
- (iv) migration between nodes.

Further requirements of the environment are

- (i) transaction type migration—all or nothing,
- (ii) message routing and forwarding.

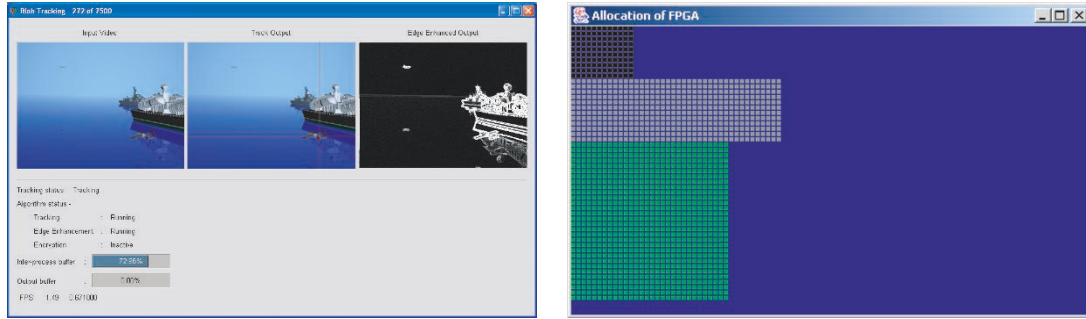


FIGURE 5: The simulation application showing the output of the tracking and Sobel algorithms is shown in (a) and FPGA utilization shown in (b). The memory arbiter (top polygon), tracking (middle polygon), and Sobel (bottom polygon) can clearly be seen in the utilization window.

If agents are to communicate and exchange information, they must first be able to find the location of other agents of interest. To facilitate this, each node maintains a list of agents currently at its locale. Nodes periodically exchange or update peers so that a global snapshot of agents is available at each node. When an agent wishes to communicate with another, the agent sends the message to the host on which it is executing along with the *sequence number/home node* key that identifies the host on the network. The sender's host then transmits the message to the recipient's host where it is passed to the receiving application. Should a node receive a message for a recipient that no longer exists on the network it must update its peers and handle the undelivered message? If the recipient has simply ended its execution, the message can be deleted and an update of the current agents exchanged. If the recipient has migrated, the message must be forwarded to the agent's new location and the agent table updated.

Mobile agents require the most support for the migration process. Performing a migration is an expensive process costing both power and throughput (due to downtime), and because of this, agents need as much information as possible available to make the best decisions possible. Information that a mobile agent may use include the CPU and memory usage on the target host platform, its physical location, the availability of reconfigurable computing resources, power usage, bandwidth to various other nodes, and the availability of other local resources. All these information are made available on request to mobile agents using messages passed directly to the target node.

If a mobile agent wishes to migrate to another node, a sequence of transactions takes place between itself and the target node to transfer its state and execution to the new location. First, the agent framework requires developers to write methods to extract agent state and allow itself to resume a state. When an agent is to migrate, it sends a request to the target host, which then invokes a new instance of that agent's class in a sleep state (so it is not performing any processing). The new instance is given a temporary identifier which is returned to the migrating node. When this is received, the agent stops performing its task, captures its state, and sends

it to the sleeping node which restores itself to this state. At this point, both nodes are notified and the original instance of the mobile agent ends its execution. The new instance is then free to resume its task in the new location. The thing to note here is that the developer of the mobile agent has not written code for migration, just recover and restore methods. The actual migration is performed upon a request to the agent environment.

4.3. Migration rules

While an application is performing its processing, the agent component is constantly examining the network for opportunities to increase its effectiveness through migration. The search for migrations is implemented in a separate thread so as not to directly affect the application. The objective for the application developer is to define a set of conditions where a migration is desirable. Fuzzy logic has been used thus far as the basis of expressing the desired behavior, although the framework allows the developer to use virtually means for expressing these conditions as rules.

Although the costs of migration in terms of resources can be modeled, the environment is dynamic and the advantage of migrating an application from one platform to another cannot be guaranteed. Fuzzy logic is used because it allows us to easily model this uncertainty. Consider the case of applications searching for targets within a subregion of the operating area of a swarm of UAVs. If there are many applications executing within this swarm, it may not be possible to control the flight of the UAV so the applications' rules must express its "desire" to migrate to planes that can focus sensors into this region. A high-level description of a rule that will exhibit the behavior is

"If (the visibility of sensors on this platform is LOW) AND (the visibility on another platform is HIGH) then (desire to migrate is HIGH)"

This rule may be combined with others that compare the power and bandwidth availability, the types of sensors and

utilization, as well as the cost of migrations and produce the desired behavior.

5. CONCLUSION

Swarms of small UAVs can benefit from the use of embedded reconfigurable computers. By extending an operating system for reconfigurable computing, we have constructed a distributed operating system that allows mobile agent enabled applications to migrate their execution between networked platforms based on fuzzy logic rules. This allows applications to not only migrate to increase performance or move closer to sources of data, but also allows power to be managed across a swarm to increase its overall mission time. The simulation of real applications shows that applications are still able to perform with acceptable performance when forced to share memory.

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David Kearney is the leader of the reconfigurable computing laboratory at the University of South Australia. This laboratory is home to the largest reconfigurable computer cluster in Australia. Kearney has more than ten years experience in research in electrical engineering and computer science. His research interests include the hardware Java compiler, operating systems for FPGAs and applications of reconfigurable computing to image processing for unpiloted airborne vehicles and bioinformatics. He has published more than 50 papers in international conferences and journals.



Mark Jasiunas completed his Bachelor of Computer Science (software engineering) and Honours at the University of South Australia before moving on to his current position as a Research Programmer and Ph.D. student in the reconfigurable computing laboratory. His current research interests include UAV autonomy, operating systems for FPGAs, and HW/SW mobile agents.

