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ASAM
Automatic Architecture Synthesis and Application Mapping

D3.2: Instruction Set Synthesis

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1 General introduction

This document is aimed at surveying some promising existing Instruction Set (IS) synthesis approaches and at defining a preliminary proposal of an IS synthesis method for the purpose of the Automatic Architecture Synthesis and Application Mapping (ASAM) project.

The main aim of the new IS synthesis method proposed in this report is to provide an automatic support for the design of Instruction Sets in customizable Application Specific Instruction Set Processors (ASIPs), such as provided by Silicon Hive technology. IS synthesis is one of the typical steps of the overall ASIP design-flow: it is preceded by the more abstract synthesis of the general computation, communication and storage architectures and it is followed by the generation of the final hardware and software ASIP platform before compilation. It involves complex multi-objectives optimization problems whose solution requires an exploration of many possible different alternatives.

Most of the works present in literature focus on a specific case of the IS synthesis, being the Instruction Set Extension (ISE), that extends an existing (in most cases RISC-type) processor instruction set with extra multi-RISC-type instructions implemented in external acceleration hardware realized as an Application Specific Integrated Circuit (ASIC) or on Field Programmable Gate Array (FPGA). ISE has three main phases: the identification of promising instruction patterns for ISE, the selection of one or more of these promising patterns and the corresponding instruction hardware generation.

Contrary to most of the works discussed in the existing literature, the ASAM IS synthesis has also to handle the generation of the initial (extensible) instruction set. Furthermore, contrary to most of the existing works, the ASAM ISEs are realized as internal custom functional units placed within the issue slots of the ASIP data path and not with FPGA accelerator or external ASIC parts. Moreover, ASIP considered in ASAM are complex highly parallel VLIW processors, and not simple sequential RISC-type processors. This imposes different constraints on the instruction identification and selection problems for customizable ASIP than those typically observed in the past research reported in the literature. As a consequence, a different new method is required for the ASIP IS synthesis.

This document is organized as follows: the first part introduces the context of the problem of Instruction Set Synthesis and Extension and formulates the related requirements in the ASAM project; the second part analyzes and reviews the existing related literature on ISE and points out the advantages and limitations of the presented solutions from literature with respect to our aims in ASAM; the third part proposes an IS synthesis method accounting for the ASAM requirements and the final part concludes the document.
2 Context: ASIP and ASAM project

Many of the modern embedded applications are complex, heterogeneous and demand for high resolving throughput or short reaction time, low energy consumption and other physical and economic characteristics. For their adequate implementation, they require significantly high quality computing platforms to address the growing computational demand and ensure a low power consumption and area occupancy. The platform are often demanded to be highly flexible to enable adaptability, design re-use and to reduce the development and manufacturing costs. Finally, to adequately cope with their restrictive requirements and complexity their design has to be supported with well organized design flows, as well as effective and efficient design tools.

2.1 The Application-Specific Instruction Set Processors (ASIP)

An application-specific instruction set processor (ASIP) is a software-programmable processor whose architecture and instruction set can be optimized (at design time) to a specific application or application domain. The term ASIP exists since late 1980s and designate a processor that provides a high degree of flexibility and, due to its application-specific instructions, is substantially more efficient than a general purpose processor.

In recent years, many researches and commercial experiences have revealed the numerous advantages of design reuse and (re)-configurability in design of systems involving ASIPs [1-8]. Due to the application-specific architecture and instruction set tuning, configurable ASIPs can achieve performances and efficiency comparable to hardwired ASICs. The configurable ASIPs often include a minimum static Instruction Set Architecture (ISA) that can be extended by custom-defined instructions executed on a configurable hardware. The custom-defined instructions are inferred from the application in order to overcome eventual computational bottlenecks and are realized through a regular chip synthesis process.

Re-targetable compilers, such as Coware Processor Designer, Expression, Mescal, ASIPMeister, Tensilica’s compiler or HiveCC, are used to schedule and map a high-level application specification onto the optimized configurable ASIP platform. The aim of re-targetable compilers is both to allow for architecture-independent software designs and to ensure the efficiency of the architecture-design compilation; but many compilation optimizations depend on the input application specification, as for example the SIMD optimization that require an exploration of the possible intrinsic parallelism of the application and are usually achieved by adding pragmas to the application specification.

Due to their high degree of flexibility, effectiveness and efficiency, configurable ASIPs represent an adequate computing platform technology for the implementation of the modern complex, heterogeneous and highly-demanding embedded applications.

2.2 Open issues in Configurable ASIP

Many problems related to the ASIP automatic design still represent hot research topics. The major general challenge is the hardware and software co-design tuned for a specific application. Due to its complexity the co-design process requires new better design methodologies and automatic tools, including methods and tools for application analysis, ASIP micro-architecture design exploration and construction, application code optimizations and compilation of the optimized code onto the custom generated hardware platform.

The existing approaches for ASIP development can be generally sketched as in Figure 1.b (1): unlike in a standard compilation flow (Figure 1.a), an ASIP design flow includes a re-targetable compiler that takes as
input a source code and a description of the target ASIP machine (Figure 1.b). Compiler for customizable processors should at least generate by itself the best possible ASIP description for a given application (Figure 1.c).

Two main approaches exist to specify an ASIP description: the template-based and language-based approaches. The template-based approach uses dedicated parameterized architecture templates. In this approach, the architecture exploration and modification is very limited, mainly to the precise instruction and data formats, as well as selection and limited extension of operations. The main processor architecture remains unchanged. On the other hand, the compiler and simulator generation is easy. This approach is used for example for Xtensa [9] and Jazz [10]. The language-based approach exploits specific architectural description languages (ADL) [11-17]. In this approach the main processor architecture can be changed to some degree, and the architecture modification through changes in ADL specification is quite easy, but it is limited by the features of a particular ADL and often time consuming. This approach is used for instance in Chess (used in Target technology for ASIP (re-)configuration) [18,19], HiveLogic [20], and ASIPMeister [21].

The limitations of these two approaches are that the first one offers a low level a customizability that limits the achievable effectiveness and efficiency, and the second one has very broad customization abilities extremely difficult to be explored without an automatic process, supported with effective and efficient application analysis and restructuring tools, as well as, ASIP architecture and instruction-set exploration tools.

For both these approaches, the definition of an optimized application-specific machine is a very complex and error-prone task and should be handled by the compiler itself (Figure 1.c).

Figure 1. (a) Standard compiler for a specific machine. (b) Re-targetable compiler. It reads a machine description and compiles the code on it. (c) Compiler for customizable processors. It generates by itself the best machine for a given application. (1).

In the effort of developing a compiler including the automatic generation of the target machine, very many researches [1,2,5,7-10,18,20-28] have recently focused on the automatic generation of instruction set extensions. This solution provides significant improvements in a template-based approach, but can be improved to a large extend, in an ADL approach. Moreover, several more aspects of the ASIP configuration have to be automatized, in order to support automatic software and hardware optimized compilation.

Three very important aspects are:

1) The data transfer and storage micro-architecture (DTSM) design of an ASIP.
2) The application code parallelization and customization of the ASIP architecture corresponding to the most promising parallel application version.

3) The basic IS synthesis, before the instruction set extension.

DTSM synthesis, that includes the communication structure and the memory hierarchy exploration, is a fundamental part of any processor design [29]. It has been largely studied in the literature. DTSM can be inferred from the application analysis. Balasa et al. [30], survey works inferring storage requirement estimation from application analysis and optimization. [31,32] propose a method to explore the hierarchy structure limited to the usage of cache memories, that are too power consuming and costly to be embedded in portable systems. The authors of [33] propose a method to automatically tune storage requirements, communication and memory structures to a specific predictable application. They use local memories with a “pre-computed pre-fetching” that are cheaper than caches. This solution does not include instruction set synthesis and software and hardware code generation.

To achieve adequate application parallelization, in both template-based and ADL-based approach, the application specification has to include technology specific Application Programmer Interfaces (API) and technology specific instructions, called intrinsics [9]. This renders the parallel application code specification a very hard task that is currently performed manually. A possible solution is to automatically generate both the application code and the hardware platform by using application analysis. The proposed coarse design flow is sketched in Figure 2.

![Diagram](https://via.placeholder.com/150)

**Figure 2. HW/SW co-tuning.**

Many research works [34-44] exist on methods to map a high-level specification of an application onto a fixed or partial configurable parallel architecture. They apply loop transformations [45] to enhance data locality and allow for a better parallelism exploration. Few works [33,46] use loop transformations to tune both software and DTSM to a specific application by using loop transformations. But there is a lack of a comprehensive framework able to achieve optimized software/hardware application specific tuning.

One of the final ASIP configuration aspects is the IS synthesis. In the literature, the problem is mostly addressed for a single pipeline, while in an ADL ASIP specification more parallel pipelines can be used. In the target ASIP technology, a limitation on the single sequencer in a single ASIP still holds. Thus, parallel pipelines are synchronized at the same time, even if they have different latencies (cf. Figure 22). Furthermore in the literature the problem of IS synthesis is usually treated as a partial IS customization,
where a single large promising instruction set extension (ISE) is found to be implemented onto (in most cases) external specific hardware platforms (FPGA or ASIC). In the case of ADL-based ASIP customization approach as of HiveLogic, it is possible to achieve a higher efficiency with a full IS configuration. In addition, the construction of the initial IS requires simple instructions to allow for re-use. Finally, if an ISE is needed to remove specific bottlenecks, it is realized as a custom internal ASIP instruction implemented inside the ASIP data path.

For the ASIP architecture exploration and code generation, partitioning, scheduling, retiming and binding methods can be used similar to those presented in [47-57]. More information on (configurable) ASIPs, their design and its automation can be found in [1,2,5,7-10,18,20-28].

### 2.3 ASAM micro-architecture design flow

In order to provide the missing automated framework for configurable ASIP design and optimization the ASAM project proposes an ASIP design method as presented in Figure 3.

![Figure 3. The overall micro-architectural synthesis](image)

The ASAM method goes through three optimization phases:
1) Phase 1, which optimizes the initial application specification and infers from it the maximum achievable parallelism level. This parallelism is only due to data and control dependences intrinsic to the application and does not depend on the used high level specification language. The parallelism exploration is mainly achieved through loop and straight code transformations as presented in deliverable D3.1.

2) Phase 2, which takes as input a graph specification of the maximum parallel version of the application as computed in phase 1 and explores the possible ASIP hardware realizations. This exploration is performed at an abstract level and mostly focuses on the data transfer and storage micro-architecture. The exploration is performed by using loop transformations and abstract architectural model of the final ASIP platform.

3) Phase 3, which takes as input a few Pareto solutions of the exploration problem of phase 2 and for each one of them selects the best possible instruction sets, generates the files specifying the optimized software and hardware ASIP platform, compiles and simulates the platform with the Silicon Hive tools.

Phase 3 of this method is the context of the ASAM Instruction set synthesis problem presented in this deliverable.

A more in depth and complete presentation of the ASAM method for automatic ASIP design and optimization is presented in deliverable D3.3.
3 Instruction Set Synthesis state of the art

3.1 Overview on Instruction Set synthesis
In general, the processor micro-architecture design can be sub-divided into the following two main parts: the data transfer and storage micro-architecture (DTSM) design and the Instruction Set design [29,58]. This document focuses on the instruction set design and information on the DTSM design can be found in (D3.3).

The IS synthesis is one of the issues of the micro-architecture synthesis, and as such, it has to be performed using the issue decision model, being a partial and abstract quality model extracted from the model of the micro-architecture synthesis. It represents a complex optimization problem whose solutions have to satisfy the constraints and optimize the objectives of the issue decision model. The main general objectives of this issue are to achieve a high-performance and power efficiency, with a limited amount of hardware resources.

To perform the IS customization, the original High Level Language application specification (e.g. in C or C++) is usually first converted into a graph-based representation, typically DFG, CDFG, HCDG or similar. This graph-based application representation is analyzed, parallelized, and if needed, scheduled and assigned. It can be generated with any of the many existing compilers [18,20,59-63] etc.

3.1.1 Full and partial instruction set synthesis
The instruction set customization can be performed as a construction of a whole new application-specific instruction set [19,64] or as modification of an existing one through adding a new instruction sub-set [64-90]. In the first case, it is referred to as full customization, and in the second case as partial customization.

In full customization, the CDFG is analyzed and partitioned in disjoint cuts that cover the graph and give the maximum advantage for throughput, workload balancing, power dissipation and resource requirements. In a full customization, an important objective is the hardware sharing, achieved by re-using the same hardware parts to jointly realize different simple instructions. Although a fully customized instruction set can be very effective, the cost and time of handcraft designing a whole new processor for each application is usually excessive. Consequently, most of the existing works focus on extension of an existing instruction set.

In a partial instruction set customization, only the critical sub-graphs, responsible for performance bottlenecks, are usually extracted as potential custom instructions. Then they are adequately implemented as application-specific hardware accelerators to minimize execution time or maximize throughput, under hardware resources constraints. The hardware accelerators implementing the extra instructions are added to the existing processor/core and their customization involves both the hardware synthesis and the mapping of the application partially onto the newly synthesized hardware and partially onto the pre-existing processor/core. As a consequence, the instruction set customization and application mapping is a special case of the HW/SW partitioning problem [47-52].

Here a remark has to be made that such accelerators can also be exploited without any explicit changes to the instruction set architecture of a processor, for instance as discussed in [91,92].
3.1.2 Complexity of instructions and type of processors

Before considering the instruction set customization, several remarks have to be made on basic instruction types, and instruction execution in processors of different types.

From the granularity viewpoint, the instructions can be sub-divided into: fine-grain and coarse-grain. The fine-grain instructions implement small groups of basic operations (e.g. multiply accumulate). The coarse-grain instructions implement large blocks of basic operations, as whole (nested) loops or procedures (e.g. codecs, discrete filters or transforms). Let consider the example in Figure 4 that is a vector instruction and could represent a parallel implementation of a loop nest.

The fine-grain instructions can be reused in more applications or independent parts of an application than the coarse-grain, but the speed-up and the energy reduction of the coarse-grain instructions can be much higher.

In general, the fine-grained approach is adapted to specialize processors for a large application class, while the coarse-grain approach is adapted to specialize processors to a particular single application or small class of applications. Nevertheless, a high speed-up and energy gain can be expected if the promising ISEs represent the most frequently executed and critical operation patterns of a particular application.

Consequently, the type of instructions should not be decided a priori, but it should be decided after a careful analysis of the application or application class and related design requirements. In many cases, a mixed fine-/coarse-grain approach is the most appropriate one [93,94].

Moreover, the instruction type selection is related to the processors types. In general, the simpler is the processor and the more similar to the basic instructions are the ISEs, the easier is their integration into the processor. For instance, in scalar or superscalar processors, independently of the instruction character, the application-specific instruction integration does not need complex analysis to guarantee the instruction synchronization that is usually realized in hardware as an handshake protocol: the processor can perform some other unrelated computations while waiting for the completion application specific instruction on the accelerator [95,96].

The VLIW processors have usually instructions with global synchronization points. In consequence, the extra instructions should also have the same synchronization points. Instruction having very different latencies cause stalls in a VLIW pipeline and may degrade its performance.

For more complex ISEs, with critical path delay substantially longer than the VLIW clock cycle, this problem can be resolved through a multi-cycle instantiation implementation involving several pipeline stages. This problem does not exist for superscalar processors, where the hardware accelerator implementing an
application-specific instruction can be used the same way as a regular functional unit of the superscalar. Unfortunately, super-scalar processors often use complex hardware to dynamically identify parallel processing opportunities, and in consequence, are less suitable for embedded applications. In contrary, VLIW processors exploit compilers for finding instructions for parallel execution, and consequently, have a much simpler hardware.

3.1.3 Open issues in instruction set synthesis
The major challenge of the instruction set customization is the lack of adequate design automation tools that enable an efficient automated application analysis and instruction set customization. In the current engineering practice, application profiling and analysis are quite well supported by compilers and other automatic tools, but instruction customization is often limited, and either performed manually or only partly automated for some specific configurable ASIP architectures (see e.g. [9][21][13][14][19][97]).

In the next sections we discuss the main existing methods for ISE identification and selection and hardware instruction generation.

3.2 Pattern Identification
Custom instruction identification consists of analysis of a graph-based application representation to find some critical and repeating sub-graphs (operation patterns) that are good candidates to be converted into single custom instructions.

The search for the candidate patterns (instructions) is an optimization problem. Thus the number of its possible solutions is limited by exploration constraints and the search of the optimal solutions is guided by quality metrics. The exploration method can be exhaustive or based on heuristics. In the following we give brief information on the exploration constraints, quality metrics and exploration methods used in the existing works on pattern identification.

3.2.1 Exploration constraints
The exploration constraints are summarized in Table 1 and are usually imposed by the type of architecture, under the consideration of the targeted implementation technology or guarantees of a proper scheduling (e.g. related to the number of inputs and outputs, patterns convexity, operation type, etc.). The aim of these constraints is to facilitate the identification of the most promising candidate patterns in the application graph and to support pruning of the set of possible patterns during the instruction identification, if the solution space is too large.

<table>
<thead>
<tr>
<th>Architecture type</th>
<th>Proper Scheduling</th>
</tr>
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<tbody>
<tr>
<td>Extra instruction execution time in VLIW</td>
<td>Nbr. of instruction I/O ports for data access scheduling</td>
</tr>
<tr>
<td>Technology type</td>
<td>Patterns convexity</td>
</tr>
<tr>
<td>nbr. of instruction I/O ports for realization on an external FPGA</td>
<td>Instruction granularity</td>
</tr>
<tr>
<td></td>
<td>Extra instruction with Internal states</td>
</tr>
</tbody>
</table>

Table 1. Usual constraints of the optimization problem of promising patterns identification
3.2.2 Quality metrics

The quality metrics express the effectiveness and efficiency of a pattern and are related to various characteristics of the hardware implementing the individuated instructions, as the execution time, power/energy consumption, used hardware resources, etc, but often describe the properties at a more abstract level. Their aim is to compare the quality of the considered solutions to each other and provide selection criteria for the most promising solutions. Table 2 summarizes some of the most commonly used quality metrics.

The instruction identification methods proposed in [98-103] are aimed at maximizing the spatial reuse of patterns and are mainly based on statistics of pattern occurrences in the application graph. This often results in quite small and simple patterns that are often worthless to be realized onto external specific hardware because they do not guarantee a sufficient execution speedup. Several works show that larger and more complex patterns result in higher speedups (e.g. [104-106]) and different metrics have to be used to achieve them as for example in [6]. For higher speedups, not the spatial reuse of patterns, but rather their temporal reuse should be maximized.

Table 2. Quality metrics of the identification problem.

<table>
<thead>
<tr>
<th>Objectives</th>
<th>Methods</th>
<th>Papers</th>
</tr>
</thead>
<tbody>
<tr>
<td>Minimized the spatial re-use (of small and simple patterns)</td>
<td>Statistics of pattern occurrence in the CDFG</td>
<td>[98-103]</td>
</tr>
<tr>
<td>SpeedUp brought by the larger and complex ISE (maximize the temporal re-use)</td>
<td></td>
<td>[104-106] [6]</td>
</tr>
<tr>
<td>Reduce the average execution time</td>
<td>Identify the most frequently executed patterns</td>
<td>[105,107]</td>
</tr>
<tr>
<td>Reduce the worst-case execution time (for hard-real time application)</td>
<td>Identify the time consuming instructions on the critical path</td>
<td>[104,108]</td>
</tr>
<tr>
<td>Reduce the dynamic reconfiguration time for reconfigurable ASIP</td>
<td></td>
<td>[109-111]</td>
</tr>
<tr>
<td>Reduce the extra cycle needed to share hardware between the processor and the extension</td>
<td></td>
<td>[112]</td>
</tr>
</tbody>
</table>

To reduce the average execution time, the most frequently executed patterns should be identified, as e.g. patterns on the most frequently executed paths [105,107]. To reduce the worst-case execution time and satisfy hard real-time constraints, the patterns most frequently occur on the critical and near-critical paths are extracted [104,108]. In [108] the worst-case execution time is used as a metrics to guarantee satisfaction of the real-time constraints. In [109-111], the dynamic reconfiguration cost is accounted for,
and in [112] the extra clock cycles to move data between the register files and hardware units implementing the new instructions, as well as hardware sharing among the hardware units.

### 3.2.3 Pattern Identification methods

Some simple methods for instruction pattern identification are listed in Table 2. In addition to these methods, Table 3 gives summarizes some more complex ones, that are also the most frequently used in the existing literature.

Some other approaches [99,100,103,107] use specific heuristics to identify some promising patterns while discarding some less promising ones. The method proposed in [107] and several other methods find possible custom instruction candidates, while pruning the search space based on the input or output constraint violation by the candidate sub-graphs, operation type, convexity, etc.

Table 3. Most used methods for instruction pattern identification.

<table>
<thead>
<tr>
<th>Methods</th>
<th>Papers</th>
</tr>
</thead>
<tbody>
<tr>
<td>Using heuristics (ex. Constraints violation or dominated quality metrics) to discard less promising patterns</td>
<td>[99,100,103,107]</td>
</tr>
<tr>
<td>Growing patterns around a seed node while considering the quality metrics of the pattern.</td>
<td>[94] and [106]</td>
</tr>
<tr>
<td>Template matching and sub-graph isomorphism (find the most occurrent sub-graphs that match pattern templates in an existing library of possible ISEs)</td>
<td>[64,106] [4,6,69,70,74,78,82,84,86,88,89,93,98,100,103,106,112-122]</td>
</tr>
<tr>
<td>Template generation (ca be used as a front-end to the template matching to form the template library)</td>
<td>[4,29,70,86,88,90,93,99,101,103,121,123]</td>
</tr>
<tr>
<td>Exhaustive exploration</td>
<td>[102] [124,125][107][6]</td>
</tr>
</tbody>
</table>

The methods proposed in [94] and [106] incrementally grow patterns when observing performance gains and penalties related to the input or output constraint violation. A commonly used concept in the custom instruction identification is this of a template. Template is an operation pattern known or assumed to be a promising candidate for a custom instruction. Custom instruction identification can be performed as template matching or template generation. Template matching assumes the existence of a template library and consists of finding the number of occurrences in the application graph or the number of repetitive executions of particular existing templates from the template library (e.g. [64,106]). The most frequent templates are then implemented as custom instructions. This problem is similar to the sub-graph isomorphism problem [4, 6, 69, 70, 74, 78, 82, 84, 86, 88, 89, 93, 98, 100, 103, 106, 112-122], and it is known that the directed sub-graph isomorphism problem is NP-complete [126].
Template generation consists of creating new templates (e.g. [4, 29, 70, 86, 88, 90, 93, 99, 101, 103, 121, 124]). Usually it starts with selection of a particular node or a larger pattern to be a seed, and gradually grows the seed through absorbing some neighboring nodes, when observing the influence of the pattern growth on its parameters included in the constraints and objectives of the pattern quality metrics that guide the search. The pattern optimizing the quality metrics is accepted as a new template. After constructing one or more new templates, the number of template occurrences in the application graph or the number of their repetitive executions is checked to prune the less frequent templates or to accept the most frequent once. Some approaches that combine the template matching with generation have also been proposed (e.g. [93,102]). In [127] it has been demonstrated that the number of different prevalent data-flow patterns in popular multimedia benchmarks is very limited (approximately 10 patterns). In [128,129] this has been experimentally proven for the second time. It has been demonstrated that a relatively small number of predefined templates, called morphable structures, is needed for a near-optimal instruction set customization for the relatively narrow multimedia application class, and a rapid custom instruction generation method is presented based on this fact. The same is proven for the third time in [130]. A similar idea of the custom instruction generation speedup through only considering the major blocks of CDFG is presented in [131].

In general, the problem of custom instruction identification is of exponential complexity, because the set of possible new custom instructions grows exponentially with the number of the application graph nodes. In the past, exhaustive enumeration, several dynamic programming-based algorithms (e.g. [102]) and Integer Linear Programming algorithms (e.g. [124,125] ) have been proposed to solve the problem, but these approaches are not efficient for larger general problem instances [107]. Consequently, to efficiently solve this problem for large instances, only some easier to process specific application graphs and/or sub-graphs should be considered, or adequate heuristic algorithms have to be used. [6] proposes an exhaustive method (in the sense that the whole solution space is explored) that includes pruning techniques that largely reduce the exploration time and cost with the consequence that the optimal solutions are found in a reasonable time.

3.2.4 The most often used simplifications for the identification problem

Table 4. Used simplification in pattern identification problem.

<table>
<thead>
<tr>
<th>Simplifications</th>
<th>Papers</th>
</tr>
</thead>
<tbody>
<tr>
<td>Acyclic graph</td>
<td>All read papers</td>
</tr>
<tr>
<td>Connected graph</td>
<td>[102,106,108,132][133]</td>
</tr>
<tr>
<td>The non-overlapping templates</td>
<td>[107,133]</td>
</tr>
<tr>
<td>Number of I/O output (MAXMISO, valid sub-graphs)</td>
<td>[107,134][135][134,136][137].</td>
</tr>
</tbody>
</table>

The solution difficulty of the instruction identification problem depends on the kind of application graphs and sub-graphs (templates, operation patterns, instructions) considered. For this reason simplifications, such as summarized in Table 4, are used to handle the IS synthesis problem. In particular, since cyclic graphs cannot be easily sorted, acyclic graphs are considered in most cases. Although a cyclic graph can be
transformed into an acyclic one (e.g. through unrolling the cycles), this significantly increases the graph complexity.

Also, only connected graphs are considered in most cases (e.g. [102,106,108,132][133] [109]), and disconnected graphs are processed in parts through processing their connected components, despite the fact that the direct consideration of the disconnected graphs makes possible a more effective parallelism exploitation (e.g. [70,104,106,133]).

Moreover consideration of multi-output templates and overlapping templates during the custom instruction generation and selection is difficult [107,133]. During instruction generation the disjoint templates are usually considered, and the nodes absorbed into a template are immediately removed from the application graph. Although the overlapping templates consideration can potentially produce better results, their consideration drastically increases the problem difficulty, and additionally, the costs related to replication of the common nodes of the overlapping templates may sometimes exceed the performance gains due the overlapping template consideration.

Regarding the number of outputs of a sub-graph (template, operation pattern, instruction) the following two types of sub-graphs can be distinguished: multiple inputs single output (MISO) and multiple inputs multiple outputs (MIMO). MISO sub-graphs of maximal size are called MAXMISO.

The type of patterns or instructions considered directly relates to the instruction identification problem complexity. The exhaustive enumeration of MISO patterns is exponential, as it is strictly related to the sub-graph enumeration problem which is known to be exponential [107,134]. However, the exhaustive enumeration of MAXMISO patterns is linear in the number of nodes [135], because the intersection of MAXMISO patterns is empty.

Since the identification of MIMO instructions may result in more significant performance gains, some algorithms combine MAXMISO instructions per levels in order to obtain MIMO instructions, i.e. several MAXMISO instructions of the same level of a reduced graph are combined into one convex MIMO instruction. Works based on this idea are presented in [134,136] and a frame work for the automatic generation and selection of convex MIMO instructions in [137].

In [70] the identification of convex MIMO instructions is presented through clustering of MAXMISO instructions to maximally exploit the MAXMISO-level parallelism. In this algorithm, the convexity is guaranteed by construction. Through extension of this algorithm, a heuristic linear complexity algorithm has been constructed for identification of convex MIMO instructions [134].

Some other papers present sub-graph enumeration algorithms limited to only the so-called legal patterns which are the convex sub-graphs that satisfy some architectural constraints, as number of I/O operands, pipeline depth, and other constraints [109] [67,99,100,105]. While the number of all sub-graphs is exponential, the number of legal sub-graphs is polynomial [138].

### 3.2.5 Detailed examples of pattern identification methods

In this section we give some detailed examples of existing methods for instruction pattern identification. The idea is to give a more insight into the problem and its existing solutions. Most of the presented pattern identification methods are included in some longer frameworks for ASIP customization. We also briefly present the corresponding frameworks.
The DURASE system and its pattern identification

The DURASE system is a framework to generate instruction extensions for application-specific reconfigurable processors. Its input consists of a C-written application description, a model of the target architecture and the basic instruction set to be extended. Its output consists of an FPGA-implemented processor extension and the instructions to access this extension. The analysis part involves a code optimization front-end based on the GECOS compiler and including polyhedral transformations for data parallelism [26].

In this section, we discuss identification of computational patterns from [82], but the DURASE design process also involves selection of specific patterns that speed up application and will be presented in section 3.3.3.

In the DURASE pattern identification the two following methods are used: a graph covering method using constraints programming (CP), to find some promising covers of the application graph with the identified patterns and to ensure the validity of identified patterns with respect to the architectural and technological constraints, and the graph matching, to identify patterns with the highest occurrence. The entry point is an acyclic graph that can be inferred by unrolling a cyclic graph.

The pattern identification has three steps:

1) During the first step, a CP-based covering of the initial graph identifies all possible computational patterns that respect three kinds of constraints: the convexity of the pattern, the number of I/Os of
the individuated sub-graph (i.e. the number of incident or exiting edges of the sub-graph) and a coarse estimation of the implemented ISE delay.

2) The second step of the identification, prunes the space of the identified patterns by considering only the non-isomorphic ones.

3) The third step further prunes the space of identified patterns by preserving only the patterns that have a number of occurrences in the original application graph that is comparable to those of their single nodes.

This algorithm identifies small and recurrent patterns that improve the pattern re-use but does not ensure finding the higher efficiency larger application-specific patterns. Its strength is in the formal definition and usage of architectural and technological constraints that help finding a number and kind of patterns ensuring a high coverage of the initial graph.

Such an approach can be re-used in a full IS customization to determine an initial basic IS already tuned to the application. In this case, the pattern generation should also include a constraint on the availability of the individuated sub-graphs in a pre-existing technology-related library.

**ISA Customization based on LANCE-compiler and generating LISA-based ASIP descriptions**

The method [73] proposes a generic adaptable flow to design Application specific ISE either for full custom ISEs or for ISA adaptation.

The ASIPs are usually composed of a base processor and several ISEs, that can be implemented as custom functional units on co-processors tightly coupled with the base-processor core. They consider an ASIP design flows based on reconfigurable/configurable processors that are extendable.

They distinguish three phases in an ASIP design:

1) The application profiling, that finds the bottlenecks and other hotspots of an application
2) ISE identification that analyzes the DFG of the hotspots and combines together some arithmetic, logic and data-transfer operations into a single specific instruction.
3) Verification and integration of ISEs into ASIP. This step consists in:
   a. Conversion of the DAG in a data-path
   b. Re-targeting ASIP compiler
   c. IS simulation & verification

This method is intended to find larger, not-reusable, but optimized instruction set extensions to be implemented on an external co-processor.

Figure 6 presents the proposed ISA customization flow. The input is a C-code that is transformed in an intermediate representation (three-address code IR). Then the ISE identification transforms this IR in a CDFG by using information from the micro-profiler and architectural constraints. Here there is the first loop of optimization. Finally the back-end takes the annotated/optimized CDFG and produces either partial extension for a configurable processor or a whole ADL design of an ASIP. In the back-end there is another optimization loop (DSE) to infer the ADL design.

They have two methods for ISE identification. The first method is based on ILP; it iteratively adds new nodes (instructions) to an ISE and maximizes an objective function while respecting architectural
constraints. The second method uses HLS technique to pipeline the instructions of the hotspots on several ISEs and then it generates ADL through LISAték. This method establishes a set of forbidden instructions, such as “processor state update”, “load”, “store”, etc., that can only run on the base-processor. In these two methods, it is possible to have architectural constraint, such as the number of I/O ports of the ISE (i.e. on the number of general purpose registers used for the communication between the basic processor and the ISEs) and the number of used HW resources and scratch-pad memories.

The back-end of the ISA customization produces a transformed C and RTL (Verilog). The C can be integrated in Coware corXpert for ESL system-IP generation or in LISAték for ASIP design and ADL generation.

In [76] another pattern identification method for custom instructions is presented. The main contribution of this work is twofold: it takes into account the bandwidth limitation, due to the limited number of the General Purpose Registers (GPRs) used in the communication between the core processor and the external custom ISE, and it considers the possibility to overcome this limitation by using Internal Registers (IRs).

The proposed pattern identification method is exemplified in Figure 7. It accounts for three types of constraints and is executed in two steps. The used constraints are: data-flow related, i.e. the convexity and schedulability of the individuated pattern; constraints on area and latency of the possible custom instructions and the architectural constraints on the number of available GPRs.

The two steps of the custom pattern identification use Integer Linear Programming (ILP) to locally optimize the two problems of first iteratively finding an optimal partitioning of the initial DFG representing the application hotspot to be speeded-up, and then maximize the number of communications using GRP; while still allowing for communications between the core processor and the extension based on IRs.

**Figure 6. ISA customization flow.**

The main concern (and architectural constraint) of this method is the limitation on the number of GPR for the communication between the basic processor and the ISEs. In the case when custom instructions do not need to be realized on external hardware support but can be realized as internal instructions (HIVElogic) this limitation is less important, because the communication can always happens through IRs. This method does not have an explicit phase of selection because the number of identified IS extensions is limited by construction in the searching algorithm.
After the identification, the found partitioning goes through a phase of scheduling and register allocation before the actual hardware implementation.

**Enumer07 and newenum and maximal valid sub-graph methodologies**

In [67,85], the authors propose an improvement for the existing methods to extract ISEs by adding new pruning criteria. They classify the existing methods with respect to the method used to solve the identification problem: reducing the exploration space, heuristics or genetic, exhaustive branch-and-bound, single-output small blocks, clusterizing and connected graph. They improve the Enumer07 that is a pattern identification method that handles disconnected multi-output graphs. The problem statement is the following: find valid patterns in an input DFG so that: patterns are convex, they have a limited number of I/O and do not contain invalid operations. As the original one, the improved enumer07 has 3 steps: *node_select*, *unite* and *split*. In the improved enumer07 the used optimization criteria are improved and the identification is more efficient. *Node_select* starts from an empty pattern and iteratively selects a valid new node from the input DFG. *Unite* treats node that can be added to the constructed pattern and *split* treats node that cannot be added by splitting the input DFG in disconnected graphs. In this paper they propose new criteria to pruning the exploration space in select and unite. The first criterion is the violation of I/O constraints (if this violation cannot be solved by expanding the node) the corresponding branch of the graph is pruned. The second criterion is the single-output constraint to decide if a node is connected to the constructed pattern or not.

**Maximal valid sub-graph method**

The maximal valid sub-graph is a convex sub-graph that does not contain invalid nodes. A sub-graph is said to be convex, if all the paths connecting its nodes are contained in the sub-graph itself and the forbidden instructions are user-defined and usually include load, store, branch, jump, etc. Based on the definition of Maximal Valid Sub-graph (MVS), a fast method to enumerate (or identify) promising patterns is presented in [79].
Figure 8. The Maximal Valid Sub-graph method.

On the contrary of other methods generating patterns by merging neighbor nodes, this method solves the problem in a top-down manner. They propose a formally defined and correct by construction method. The method iteratively considers and eliminates from the graph the invalid nodes. During a given iteration of the successive eliminations, it enumerates all the possible maximal convex patterns. If these patterns do not contain invalid instructions, they are marked as identified instruction patterns. On the base of this definition, we exemplify the method with respect to Figure 8. The graph G in the figure has three invalid nodes (1, 4 and 8). At the first iteration, node 4 is eliminated and a single convex sub-graph G1 still containing invalid instruction is created. From G1 and by eliminating node 8, two possible sub-graphs G21 and G22 are inferred. Note that G21 is not connected but convex and does not contain invalid instruction, thus it is selected as promising instruction pattern. This method selects large patterns and is appropriate for ISE on dedicated external hardware.

Upak system

In [88,89], the authors present a method to automatically generate an application specific reconfigurable HW accelerator.

The ASIP design includes: the identification of frequent computational patterns, the selection of a sub-set of these patterns for which the mapping and scheduling reach the maximal coverage of the application graph. The flow is composed of three steps and is presented in Figure 9.

During the first step, some promising computational patterns are identified from a hierarchical input application graph. This step is described later in this section. During the second step, the identified patterns are used to explore which patterns will be used in a particular scenario for application execution. This is done during the scheduling and pattern selection phase using specific methods built on the top of the constraint programming solver JaCoP. The input to this phase contains the identified patterns, the ASIP architecture model and the specific scheduling constraints (e.g., execution time limit or resource constraints). This is described in section 3.3.3. The output of the system is a set of the selected patterns and
a corresponding schedule for execution of the application using the selected patterns and processor instructions.

![Diagram](image)

**Figure 9. Upak flow.**

The first step of the method generates a set of promising instruction patterns, by having all the nodes of the application graph as seeds. Subsequently, it is checked if the generated patterns are isomorphic to each other and then the patterns having a maximal coverage are selected (i.e. the patterns having the highest number of matches within the application graph are selected). Thus, for each generated pattern a graph matching algorithm is used. It checks if the type, the I/O structure and the neighborhood structure of two graphs match with each other. At the end of this iterative process, a Definitively Identified Pattern Set is selected.

All the patterns in the DIPS are merged to form a unique HW cell. To merge the patterns, they first construct a compatibility graph, where the compatible operations (i.e. operations having the same type and structure) are mapped to the same node or edge. Then, they find the maximum weighted clique to realize the merged patterns. This clique is then scheduled with time or resources constraints, by using 2D rectangles that represent the time during which each node (instruction) is alive.
Maximal-clique-based methodology

The paper [87] proposes a method to identify ISE by solving maximal clique problem and by serializing I/O of the custom functional unit. The proposed method clusterizes the nodes of the input CDFG in equivalence class. Then, it constructs a Cluster Graph and finds the maximal cliques (connected graphs) that are potential optimal ISEs. Finally it serializes I/O on the found ISEs and selects the ones having the maximal speed-up and smallest area. To limit the number of considered ISE, they define a set of forbidden instructions (load, store and jump) that cannot be executed on the extension.
They use a time model to define an objective function of the ISE identification. In fact, they characterize each node with two metrics (SW and HW) giving the number of cycles needed to execute the node in software or in hardware. The objective function to be optimized is the overall sum of cycles needed to execute software instructions minus the overall sum to execute hardware instructions.

### 3.3 Instruction Set Selection

Custom instruction selection consists in selecting the most promising sub-set of custom instructions from the set of custom instructions constructed in the process of custom instruction identification or already existing in a given library. As the identification problem the selection is an optimizing process based on quality metrics and using some exploration methods. Next sub-sections describe the metrics and methods most frequently used in published research.

#### 3.3.1 Quality metrics

The selection is realized using some quality metrics involving evaluation of an instruction sub-set in terms of performance, area, power consumption, etc. after its realization in hardware. These quality metrics are similar to those used for the instruction identification, but at this time the metrics should express the effectiveness and efficiency of an instruction set and not the quality of particular instructions. Table 5 summarizes the most frequently used quality metrics.

[139] presents an algorithm that aims at selection of a minimal set of instructions that maximizes the number of covered nodes in the application graph. An interesting observation of the experimental research of this work is that increasing the number of different instructions (patterns) used to solve the coverage problem, results in a significant increase of the number of nodes covered only up to a certain level, above which, usage of more templates does not substantially influence the number of nodes covered. Consequently, a reasonably small number of well-designed and adequately selected custom instructions can result in a significant performance gain.
Table 5. Most used quality metrics to solve the selection problem

<table>
<thead>
<tr>
<th>Quality metrics for the selection problem</th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>Metric</td>
<td>Paper</td>
</tr>
<tr>
<td>maximizes the number of covered nodes</td>
<td>[139]</td>
</tr>
<tr>
<td>minimizing the number of distinct instructions used</td>
<td>[140]</td>
</tr>
<tr>
<td>maximizing the number of pattern occurrences</td>
<td>[91,141]</td>
</tr>
<tr>
<td>maximizing the execution frequency</td>
<td>[101,105,108][104]</td>
</tr>
<tr>
<td>Accounting for the occurrence of specific nodes</td>
<td>[93]</td>
</tr>
<tr>
<td>Optimizing resource sharing</td>
<td>[143,144]</td>
</tr>
</tbody>
</table>

Other methods aim at minimizing the number of distinct used instructions [140], maximizing the number of pattern occurrences [91,141], the execution frequencies [101,105,108][104], the occurrence of specific nodes [93] or the resource sharing [143,144].

3.3.2 Exploration methods
The instruction selection problem is a specific graph coverage problem that is known to be NP-hard [145]. Consequently, to solve this problem for large and complicated instances, some heuristic algorithms have to be used.

Table 6. Most used methods for the exploration of the optimal Instruction Set selection.

<table>
<thead>
<tr>
<th>Methods of the selection problem</th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>Method</td>
<td>Paper</td>
</tr>
<tr>
<td>Linear programming</td>
<td>[125,134,146]</td>
</tr>
<tr>
<td>Branch-and-bound</td>
<td>[112][147,148]</td>
</tr>
<tr>
<td>Heuristics</td>
<td>[149][99,119,132]</td>
</tr>
<tr>
<td>Constraint programming</td>
<td>[26,69,82]</td>
</tr>
</tbody>
</table>

In the past, several exact LP-based algorithms have been proposed [125,134,146] for instruction selection and some branch-and-bound-based algorithms [112], as well as, some branch-and-bound-based algorithms for general covering problems [147,148] and some effective and efficient heuristic algorithms for similar coverage problems [149].

Also, some other heuristic methods have been proposed for instruction selection (e.g. [99,119,132]). Especially the method discussed in [119], which is based on constraint programming and performing the final instruction selection during scheduling and mapping seems to be very promising.
Recently, new instruction set customization method, based on constraint programming and including an automatic tool-chain, was presented in [26,69,82]. This method is adapted to heterogeneous embedded multi-processor systems involving (re-)configurable embedded processors and reconfigurable hardware accelerators.

3.3.3 Detailed examples of instruction set selection methods

In this section we give some detailed example of existing methods for instruction set selection.

The DURASE system and its selection method

This sub-section gives information about the selection in the DURASE system previously presented in 3.2.5. This selection method presented in [69] uses constraint programming to model and solve instruction selection for processors that can be extended with a functionality mapped on reconfigurable cell fabric (FPGA). Starting from a hierarchical graph of the application, that captures data and control dependences, the DURASE system first performs a patterns identification phase, then it performs the patterns selection and instructions scheduling and mapping. As described in 3.2.5, the pattern identification is based on constraints programming: the search targets patterns that are not isomorphic and satisfy some architectural and technical constraints. The identified patterns are promising candidates for hardware realization. From the set of these patterns, the DURASE system selects those resulting on the best hardware performances when used in a unique application specific instruction set. To achieve this aim, the DURASE system solves concurrently the scheduling and the matching of the application graph onto a set of possible patterns candidates (also called matches) found during pattern identification. The method first constructs a matching (coverage) matrix between application nodes and matches. Then it solves a graph matching problem with resources and time constraints.

![Figure 12. Example of a nodes and identified patterns matching.](image)

Figure 12 gives an example of matching matrix between nodes and identified promising patterns. A point in the matrix indicates a possible matching between a promising patterns \(m_i\) and a node \(n_j\). A black point indicates a selected specific matching.

To select the specific matches, the method uses an abstract model for scheduling and mapping as shown in Figure 13. The promising patterns \(m_i\) are represented as 2 dimensional rectangles whose height represents the number of used resources and whose width represents the time latency. The aim is to minimize the scheduling length under resource constraints. The rectangle heights and widths are the variables of a constraint programming problem defined over a finite domain.
Some other details are added to enrich the scheduling and mapping model. These details are:

1) The usage of both a single and multiple nodes matching model, the first one represents the mapping of tasks on the basic processor and the second one represents the mapping of tasks onto the parallel ISEs.

2) The categorization of the rectangles according to the kind of task that they represent, e.g. a task on an extension has to take into account the time to transfer data in input and in output, these tasks are classified in category C. Category A of tasks includes the launching of tasks on ISEs and the control on the status of the processor. Category B indicates the launching, data transfer and computation execution on ISEs.

---

**Optimal sub-graph covering for VLIW processors**

In [72], the authors propose a method to optimize the IS selection in the case of VLIW processors.

The instruction set selection is solved as a sub-graph covering problem minimizing the execution time of the realized task. Instead of solving the problem of minimizing the number of independent operators covering the operations in the DFG of the application (i.e. Minimum Number covering (MNC) problem), the authors solve the minimum length covering (MLC) problem, i.e. they minimize the length of critical path of the ISE. The idea is that the MLC problem better targeted to multi-issue/VLIW processors.

They solve the problem with an exact method using dynamic (recurrent) programming and prune the search space by applying several simplifications, such as:

1) **Divide and conquer** (only consider Single Input Single Output sub-graph to be explored with dynamic programming, for the MIMO they enumerate all possible covering sub-graphs).

2) **Sorting and pruning**, sub graphs covering more nodes of the critical path have a priority of selection and sub-graphs included in other (as fast) sub-graphs are pruned.

3) **They use lower bound for critical path (branch and bound)** and they prune non-convex solutions.
The analysis starts with a Boolean mapping matrix (Figure 14) and solves the coverage problem with the dynamic formulation given in Figure 15, which is equivalent to the Bellman-Ford algorithm that finds the length path “i->j” by recursively considering the length of the sub-paths “i->k” and “k->j”. The considered sub-paths are those identified by the Boolean matrix and the path length is replaced by the path delay $D[i,j]$.

For the CFU generation, the method produces a set of parameters to configure CFU template.

Rapid design of area-efficient custom instructions for reconfigurable embedded processing

The following is an example of a comprehensive method including pattern identification (generation) and selection.

In [74], the authors propose a rapid design space exploration framework for a first identification of a reduced set of profitable application specific instructions for a VLIW architecture that can be extended with an FPGA-based reconfigurable functional unit.

The proposed method involves the following three steps:

1) A phase of a pattern library construction Figure 16. This phase is a one-time effort aimed to provide a library of patterns that are frequent for a given application or application domain. It includes:

   a. A pattern enumeration step that is based on the application profiling and enumerates the most frequent patterns in the CDFG or a specific application of a group of GDFGs of applications in the same domain (Figure 16.a).

   b. A pattern grouping step that is based on the graph isomorphism: isomorphic sub-graphs are grouped together and they are classified according to their number of occurrences (frequency) in a pattern library (Figure 16.b).

After the phase 1) of pattern library construction has been executed once, the phases 2) and 3) are
iteratively repeated for each specific application.

![Pattern library generation diagram](image)

**Figure 16. Pattern library generation.**

2) The template Library selection phase (Figure 17) is aimed to select some of the patterns from the library constructed in phase 1) in order to have an optimal coverage of the application graph. This second phase includes:

   a. A template selection step (Figure 17.c) that takes as input the templates corresponding to the patterns in the library of phase 1). This step assigns a gain to each pattern. The gain considers only the speed-up that the assigned pattern brings to the execution time. Finally this step selects the templates with the highest gains and forms a library of templates.

   b. The selected library of templates is used by the next step: the pattern matching (Figure 17.d). This step computes the matching between the templates selected in phase 2).a and some sub-graphs of the specific application. The matching is computed through a conflict graph in which the nodes represent the matches between sub-graphs and templates, and the edges represent a possible conflict between matches, i.e. two matches linked by an edge realize common instructions. The optimal matching (Figure 17.e) is that having the maximum number of independent matches (i.e. the largest sub-set of non-adjacent vertices).

3) After finding the optimal matching, the method passes through a phase of hardware generation (Figure 17.f and Figure 17.g). This is done by clusterizing the nodes of the output graph of the matches selection into groups of instructions that can be realized on the same hardware resources. The individuated clusters are realized on parallel ISEs. The optimization criterion of the cluster generation and selection is the area occupancy of the selected clusters.
Selecting profitable custom instructions for reconfigurable processors

In [78], the authors propose a method to select candidates for custom-instruction sets. They model the application as a graph of instructions and they find valid sub-set of instructions that satisfy the hardware constraints. They group the sub-graphs into “instances” of isomorphic sub-graphs. Then they solve the knapsack problem (chose the most useful things to bring in a limited-spaced sack) to find the most promising sub-graphs to be realized as custom instructions.

They use two criteria: a ratio gain to area and a ratio gain to re-use. They solved the problem with a greedy algorithm, with a taboo search and a branch and bound.

A problem is that the chosen criteria only account for re-use. Furthermore, the composition of weighted criteria, also if it can be used to express a trade-off between criteria, does not represent any physical property of the device and gives less efficient results than a multi-objective exploration.

Methodology to derive context adaptable architectures for FPGAs

In [86], the authors propose a method to design context adaptable architectures on partial and dynamically reconfigurable FPGAs. The target applications are Data-flow graphs embedded in control graphs. They take as input different candidates DFGs. They found common sub-graphs (CSGs), which are frequently repeated sub-graphs inside the initial DFG of the application. For a given set of resources, they compute the force direct schedule (FDS) for the found CSGs. Then they substitute these CSG with a custom single instruction into the initial DFG, which now contains both CSGs (as custom instructions) and non-repeated sub-graphs.

Finally the method iteratively selects the non-repeated sub-graphs of the initial DFGs and computes the FDS for these non-common sub-graphs and consequently up-date the resource set if needed. The time and resources constraints on the CSG realization also constrains the scheduling (and resource utilization) of the Non-Common Sub-Graphs realization.
The authors of [71] distinguish two phases in ASIP design: IS design, and processor structure, RF, communication design. ASIPs usually adopt VLIW architectures and thus their architecture can be optimized at compilation time with an adequate exploration process. The authors use architectures with several parallel pipelines to achieve high performance systems. They can customize the architecture with respect to the number of pipelines, instructions realized by the pipes and the forwarding paths, i.e. direct links that forward the results from the previous instruction to the instruction that is currently executed in the EX stage of a pipeline, this reduces the data hazards. The authors are concerned with the improvement of forwarding paths when having parallel pipelines.

The design starts from a c-code that is compiled on single-pipe architecture. Then a scheduling for a multi-pipe architecture is computed by supposing that all the FW paths are possible, finally the FW path are reduced on the bases of the existing data-path. It is possible to re-order the instructions in the same pipe or among different pipes without changing the data dependencies.

3.4 Instruction hardware (FU) construction and evaluation

Although the recent Mimosys Clarity tool [150] automatically identifies hardware accelerators from C code and automates the HDL generation and implementation of an application on the PowerPC and accelerators in XILINX FPGAs, it only delivers a single set of accelerators and does not enable any broader design space exploration for finding different sets of accelerators that could meet various constraints, optimize different objectives and realize different tradeoffs among the objectives. It is just a step in a good direction. Another related tool, Synfora’s PICO Express FPGA synthesizes (hierarchical) coarse-grain application-specific accelerators (instructions) for implementation in Xilinx Virtex and Spartan FPGAs. It performs algorithmic synthesis of C algorithms into their corresponding optimized RTL code that is further synthesized into the actual FPGA hardware with the tools of Synplicity and Xilinx. It makes possible a specific design space exploration, creation of multiple implementations characterized with area and performance estimates, and trade-off analysis [151]. The recent development in the reconfigurable ASIP field, the integrated development environment (IDE) of Stretch, partially automates the instruction set extension and application mapping on the Stretch families of S5000 and S6000 processors based on Xtensa and having an embedded reconfigurable instruction set extension fabric (ISEF) within the processor. The developers of systems based on the Stretch processors profile their applications expressed in C/C++, using the Stretch profiler, identify the parts of application code that have to be accelerated in ISEF, and appropriately annotate the C/C++ application code. These parts are then implemented as new instructions that are executed in a single cycle. From the annotated code, the Stretch compiler produces both the ISEF configuration and optimized application code, and configures the ISEF automatically [138].

3.5 Conclusion on the existing methodology for instruction set synthesis

Despite all the previous effort, no acceptable solution exists to the problem of fully automatic application analysis, (re)configurable ASIP instruction set customization, customized ASIP platform construction, and mapping of applications on such a customized platform. Since this problem is of high scientific interest and practical relevance, it represents a very hot research topic, and numerous research results related to this topic have been published recently [1,9,10,21,23-27,91-94]. An adequate full automation of the above processes is necessary due to many factors, including the growing complexity and requirements of application, designer productivity gap and short time to market requirement, as well as NP-hard character.
of the problems to be solved and complex tradeoffs to be resolved by the selection of an optimized custom instruction set from a usually huge set of candidate instructions and by the application mapping. As already mentioned, instruction set customization is usually performed when using a graph-based (e.g. DFG, CDFG, HCDG) representation of the application. Before starting the actual instruction set customization, various transformation and parallelization techniques are often applied to the graph-based application representation (e.g. [34-44]). Instruction set customization is usually performed in two steps, namely: custom instruction identification and custom instruction selection. In the literature related to (re-)configurable processors, instruction set customization is usually limited to instruction set extension, although in general, it should consider the elimination of less useful instructions and of a related hardware as well, of course, if this is at all possible. Furthermore, in the literature the custom instruction extending the initial instruction set are usually realized on external accelerators implemented on ASIC or FPGA. In consequence, one of the main requirements of their design is to minimize the communication between the basic processor and the accelerator, and particularly, the number of General Purpose Registers used for the communication between the basic processor and the external accelerating extension. In the case of the Silicon Hive VLIW ASIP technology used in the ASAM project, the custom instruction **extensions are realized as internal instructions** of the ASIP data-path being on the same chip and in the same issue slot as the basic instruction set. Even if the number of I/O of a custom instruction has to be taken into account, this problem is much less constraining than for the ISE identification and selection problems in the methods described in the literature. Moreover, it is not so much required, as in the case of external accelerators, to cluster all the found instruction extensions in a single MIMO complex instruction in order to reduce the number of external accelerators to be implemented. Indeed in Silicon Hive technology the instruction set extensions are treated, realized and used exactly as all the other instructions. With respect to the processor type, the ASAM project targets SIMD and MIMD VLIW ASIP processors, which often include vector instructions. For this reason, both the whole ASIP design and specifically the IS synthesis represent a much more complex task, and introduce new design challenges with respect to existing works that are mainly based on RISC extensible processors and only sometimes include some SIMD extensions. Another important challenge in the ASAM IS synthesis is the **selection of the initial IS** during which the exploration has to account for re-use. A further application specific IS extension step can **resolve the remaining bottlenecks**.
4 Instruction Set Synthesis for ASAM

This chapter is presented in the confidential section entitled Appendix: Instruction Set Synthesis for ASAM on page 55.
5 Conclusion

This document presented a survey of some promising existing Instruction Set (IS) synthesis approaches and defined an initial proposal of the ASAM IS synthesis method.

One of the aims of the ASAM project is to provide an automatic framework for the IS synthesis that is one of the fundamental steps of an ASIP design-flow.

Despite a large number of existing works on the IS extension and several works on the complete IS synthesis, no acceptable solution exists to the problem of automatic instruction set customization for complex customizable ASIPs, as those provided by Silicon Hive ASIP technology.

Instruction set customization is usually performed in two steps, namely: custom instruction identification and custom instruction selection. In the literature, the instruction set customization is usually limited to the instruction set extension, although in general, it should consider the selection of an initial extensible IS and elimination of less useful instructions. Moreover, in the existing published methods the ISEs are mostly realized as external accelerators implemented on FPGA or ASICs. Consequently, one of the main requirements of their design is to minimize the communication between the basic processor and the external accelerating extension. Adequately accounting for the I/O constraints is here one of the major problems. If more ISEs are found they are usually merged in a single complex ISE to be realized on a single external accelerator. The considered processors are often simple RISCs extended with RISC-based ISEs or (rarely) with SIMD accelerators.

In the case of the Silicon Hive VLIW ASIP technology used in the ASAM project, the custom instruction extensions are realized and used as all the other processor instructions. In consequence, although the constraints on the number of I/O of a custom instruction, as well as the number and kind of the selected instruction extensions have to be taken into account, they are much less constraining than in the case of the ISE identification and selection problems discussed in the literature. With respect to the processor type, the ASAM project targets SIMD and MIMD VLIW processors, which often include vector instructions. For this reason, both the whole ASIP design and, specifically, the IS synthesis represent a much more complex problem and introduce new design challenges comparing to existing works that are mainly focused on RISC extensible processors and only sometimes include SIMD extensions.

Another important difference comparing to the existing works is that the ASAM IS synthesis aims at selecting also the initial extensible IS. Only if necessary, it extends this set with some additional instruction extensions to resolve eventual bottlenecks. Thus two main fundamental problems are presented here: the initial application-specific IS selection and the application-specific instruction pool extension.

During the initial IS synthesis rather small re-usable instructions from the IS pool are used. For the instruction pool extension larger and more complex instructions are usually selected in order to increase the performance and efficiency brought by a particular instruction extension. Despite the mentioned substantial differences between the instruction set related methods presented in the literature and the methods needed for the VLIW ASIPs targeted in the ASAM project, when developing our new instruction set synthesis method for customizable VLIW ASIPs, we plan to re-use some parts of some existing efficient IS synthesis methods presented in the literature and adapt them to the case of the Silicon Hive technology.
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