On the Design and Implementation of a Lossless Data Compression and Decompression Chip

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Abstract—A lossless data compression and decompression (LDCD) algorithm based on the notion of textual substitution has been implemented in silicon using a linear systolic array architecture. This algorithm employs a model where the encoder and decoder each have a finite amount of memory which is referred to as the dictionary. Compression is achieved by finding matches between the dictionary and the input data stream whereby a substitution is made in the data stream by an index referencing the corresponding dictionary entry. The LDCD system is built using 30 application-specific integrated circuits (ASIC's) each containing 126 identical processing elements (PE's) which perform both the encoding and decoding function at clock rates up to 20 MHz.

I. INTRODUCTION

LOSSLESS data compression is the process of encoding a body of data into a smaller body that can be uniquely decoded at a later time back to the original data. Data that arise in practical applications typically contain redundancy of which compression algorithms can take advantage. Data compression can be applied in communication systems to increase the effective bandwidth of the channel, and in data storage systems to increase the effective capacity of the storage device. The processing speed of modern data processing systems such as disk controllers and data transmitters requires that data compression should be executed at very high speeds. To meet these stringent requirements, it is necessary to minimize the execution time of the compression algorithms. This can be only accomplished by defining parallel architectures capable of executing the defined data compression algorithms and implementing them in hardware systems. The most widely used class of data compression algorithms is the Lempel-Ziv method [1], [2]. The most computationally expensive step of this class of algorithms is the searching of the reconstruction vocabulary. During this step the compression processor has to search the entire vocabulary to identify any entries that match the input string. One of the primary methods that is used to reduce this high computational requirement is hashing [4]. A data compression chip set that utilizes hashing has been proposed in [5]. The disadvantage of such systems is that they require a two-level external memory to implement the hash tables. This implies that two memory accesses are required to find each key. In addition, the hash tables generally require large memories, which often experience slower access times. Because of these reasons the execution performance of such systems is rather limited. The reported speed in [5] is 7.5 MHz.

The textual substitution method that has been implemented in this architecture can be viewed as a practical realization of the universal data compression algorithms developed by Lempel and Ziv [1], [2]. These algorithms are used extensively in UNIX® systems (compress command) [3]. Textual substitution is a class of data compression algorithms capable of compressing text by identifying repeated substrings and replacing them by references to other copies defined in a dictionary D. The encoder repeatedly finds a match between the incoming characters of the input stream and the dictionary, deletes these characters from the input stream, transmits the index of the corresponding dictionary entry, and updates the dictionary with some method that depends on current contents of the dictionary and the input stream. A detailed description of the general class of textual substitution algorithms can be found in [6].

The basic idea of this method is to build a dictionary D to store a set of variable-length strings. The encoder (a finite state machine that compresses the data) repeatedly searches the incoming characters to identify a match with the dictionary entries based on a match heuristic (MH). This heuristic attempts to identify the longest possible match between the input stream and the dictionary entries to maximize the compression rate.

The encoder and the decoder maintain identical copies of the dictionary D which is augmented progressively by concatenating the previous match with new data. We will refer to this process as the update heuristic (UH). Finally, when there is no longer enough room to update the dictionary, an entry is deleted based on a delete heuristic (DH). Although, the least recently used (LRU) replacement policy is a good candidate for this purpose, it cannot be efficiently implemented in hardware. The SWAP heuristic [6], a simpler method with lower compression performance, can be used as an alternative solution.

II. PARALLEL IMPLEMENTATION

Parallelization of the textual substitution algorithms can be accomplished by simultaneously comparing the input data with multiple dictionary entries. This implies an associative

search between all vocabulary entries and the input data. This search can be efficiently implemented using content addressable memories. Although this scheme is very attractive in terms of implementation efficiency because of the high integration levels that we can achieve, it requires large data buffers to store data during the decompression phase. This is necessary since multiple decompressions of the same data token are possible. To eliminate the large buffering requirements, the searching process can be parallelized by distributing the reconstruction vocabulary in a number of processing elements, and having each element responsible for identifying entries within its local vocabulary. This fine-granularity parallelism can be efficiently exploited using systolic-type parallel architectures. The developed data compression system is based on this systolic approach, consisting of a number of identical processing elements capable of performing basic data compression operations, namely, comparison, multiplexing, and storing. The cells are connected in a linear pipe so that each cell communicates data and instructions with its immediate neighboring cells. The dictionary D is distributed among the processing elements so that each vocabulary entry is processed only by the corresponding element. The current implementation utilizes two vocabulary entries for each processing element.

During encoding, each cell compares two data tokens of the input string with the local vocabulary entries. If a match is found, the cell substitutes the two data tokens with a single token (the cell’s index). The one token is substituted by the index and the other token is marked as invalid (loaded with a nilpointer). When both data tokens are valid and there is no match, the data are released to the next processing element one at a time. During decoding, each cell compares the incoming compressed data with its index. If a match is found, the cell replaces the index with the data tokens that are stored in the local entries while one of the data tokens is released to the next cell. A “stop” signal is also sent to the preceding cell in the pipe indicating that the current processing element is not ready to receive any new data. Tokens are either compressed/decompressed, or passed unmodified to the next cell. Compression is achieved in a progressive fashion, where a large number of processing elements can contribute to the compression of an input string. One interesting feature of this architecture is that no distinction is made between valid data tokens and valid index pointers. This feature enables pointers and other tokens to be further compressed by succeeding cells. This progressive compression of data builds larger matches from smaller matches and allows one to replace large character strings by only a few pointers.

This architecture offers high system throughput since comparisons for identifying matches between incoming data and dictionary strings can be executed in parallel for all the vocabulary entries. The implemented algorithm employs an on-line model, where compression is achieved using only current and previous data. This implies that neither the transmitter nor the receiver needs a priori the entire data volume to perform data compression or decompression operations.

The dictionary consists of variable-length strings and is built using combinations of ASCII characters and indices that point to previously defined strings. The implemented textual substitution algorithm utilizes the identity heuristic [7]: each cell uses the first two valid tokens as the new vocabulary entries that are stored in the local registers. Once a cell has learned a pair of tokens, it switches automatically to encoding mode. This training algorithm guarantees that encoder and decoder operate in lock-step fashion, which is necessary to maintain identical dictionaries in both ends of the communication channel or the storage media. Each pair of vocabulary entries is represented by a 12-b token which is the index of the corresponding cell. The size of the token limits the vocabulary to 4096 entries. The first 256 entries of the vocabulary are implied and they represent the ASCII character set. It has been found through simulation that a pipe of 3780 cells is adequate for compressing various types of text data.

The on-line model requires that the encoder and the decoder should learn their data in reverse order: strings compressed by the rightmost cell of the encoder pipe should be decompressed by the leftmost cell of the decoder pipe (Fig. 1). This implies that the dictionary entries and the index of the leftmost cell of the encoder should be identical with those of the rightmost cell of the decoder. Similarly, the rightmost cell of the encoder should be identical with the leftmost cell of the decoder. This symmetry can be very easily exploited in the described systolic architecture.

III. ADDITIONAL MODES OF OPERATION

Besides the encoding and decoding modes that were previously described, the lossless data compression and decompression (LDCD) system can also operate in “flush” mode, where the data can be pushed out of the pipe without destroying the existing vocabulary. This mode is very useful in cases where input data may become temporarily unavailable, in which case the already processed data can be pushed out of the pipe while
the existing vocabulary remains intact. To compress multiple files in a successive manner, a sequence of CLEAR instructions can be used to clear the vocabulary entries in each cell. In this case, the already processed data are pushed out of the pipe, while data following the CLEAR instruction are used to construct the vocabulary of the new file. One final mode of operation consists of a special built-in test function for the data-path logic. The TEST instruction is used to perform offline testing of the data-path logic by shifting identical data to each data register and comparing them. Any mismatches among these data registers indicate a system failure.

IV. LDCC CHIP ARCHITECTURE

To implement the LDCC algorithm in silicon, a linear systolic array architecture consisting of identical processing elements is used to achieve high throughput. Each processing cell performs both the encode and decode operations on 12-b data operands and communicates control and data information only with its two neighboring cells. Due to the complexity of a single processing cell, only 126 cells can be placed on a chip given the chosen technology for implementation. System specifications required that the chip contains $2^n - 2$ cells such that the symmetry between the encoding and decoding indices was maintained and that sufficient addresses were available for mpointers (which are nonvalid indices). Based on the developed architecture, the LDCC system will be composed of 30 VLSI ASIC chips implementing a total pipe length of 3780 processing elements together with the necessary controller logic, data buffers, and host interface logic.

The architecture of each LDCC processing element (PE) consists of a bit-slice-oriented data-path block and a control logic block. The data-path module, shown in Fig. 2, consists of four 12-b static registers which store the dictionary entries (storeA, storeB) as well as the input data tokens (bufA, bufB), two 12-b equality comparators, and several 12-b multiplexers.

During the encoding operation, data tokens are passed into the buffer registers bufA and bufB where they are compared to the previously stored dictionary entries found in storeA and storeB. If the comparators determine that no match occurs between the two register pairs, the data are passed unchanged through the PE to its successor cell. If a match does occur between these two register sets, the PE replaces the two stored data tokens (representing some character string) by a single pointer representing the index of the PE.

When compressed data are decoded, the incoming data pointers are placed in the bufA where they are compared to the PE’s index. If a match occurs, the pointer is replaced by the two vocabulary entries that were previously “learned” by the PE. If no match is found, the data are passed unchanged through the PE.

Control of each PE is governed by its own local 4-b finite state machine, a 2-b register containing status information called markA and markB, four control signals from the two neighboring cells, and the two match signals from the data-path comparators as shown in Fig. 3. In addition to these, there are two nonsystolic global signals that affect the PE’s operation. These two global signals are the set signal which simultaneously initializes all data and control registers on the chip, and the hold signal which simultaneously freezes all cells on the chip.

As shown in Fig. 3, a 4-b instruction called state_in is input to each PE from the preceding cell. Depending upon the current state of the cell and the control inputs, the state multiplexer is enabled to either accept the instruction directly into the state registers on the next clock edge or to accept a 4-b state instruction computed by one of the internal combinational logic blocks. The processor of each PE is a 16-state finite state machine. Of those states, six are reserved for encoding data, seven are required for decoding data, and one state is needed for the resetting of the processing element.

Two states in the finite state machine were used to implement a built-in self-test function on the data-path logic. In this self-test scheme, the user sends the TEST instruction along with a stream of input data to the LDCC array. This data is first stored into the bufA and storeA registers and is then passed to the bufB and storeB registers. By utilizing the comparators, it is possible to check if the data in these registers are an exact match. If a problem occurs in the data-path portion of a chip such that the comparator flags a mismatch in one of the processing cells, then that cell is instructed to output its index as the output data token along with the ERROR instruction. To propagate this error indication to the output, each subsequent processing cell is instructed to adopt the ERROR state and to propagate the index of the failed cell to its output at the next clock edge. In this manner, a fault in the data path of the array cannot only be detected by the system but can be isolated to the failed chip and even to the processing element.
where the failure occurred. This method provides a convenient mechanism for detecting and isolating data-path failures in the LDCD system to the component responsible for the failure, thereby minimizing system downtime and thus increasing overall system availability.

The combinational logic blocks are also responsible for generating an output instruction to the next cell (state.out), generating the four output control signals (mark.out, wait.out, stop.out, and lead.out), and creating the select signals for the various multiplexers which are present in each PE. Each cell also has a 12-b index which is divided into a 7-b cell address (which is hardwired internally) and a 5-b chip address which is provided externally through the Index pins. Although the 5-b chip index assignments are programmable, it is necessary to impose certain constraints in the chip addressing scheme when constructing the LDCD array in order to maintain the symmetry between the encode and decode modes of operation. A block diagram of the LDCD processing cell along with its I/O is shown in Fig. 4.

V. CHIP DESIGN AND IMPLEMENTATION

In parallel to the design procedure for the control logic, the data-path logic was implemented using a bit-slice approach. Using custom layout for maximum area utilization, a 1-b slice was developed that contains an instance of each of the four data registers along with two comparator cells and the assorted multiplexer cells. Portions of this bit-slice cell were also used to create the state and mark registers. The 12-b data comparator was implemented as two 6-b ripple comparators operating in parallel in order to achieve the necessary circuit performance. Upon completing the layout of the basic processing element, a generator program was developed to place and route the individual cells to form the core logic. Due to the layout floorplan of the PE, most I/O connections between cells were made through simple abutment. At the same time, the 7-b index of each PE was hardwired to its appropriate address through the application of the generator code. In addition to the array of processing cells in the chip logic, high-drive buffers were developed and manually integrated to distribute the clock and other global signals to the internal array.

A prototype of the LDCD chip, containing 126 processing cells, was designed and fabricated. The chip was designed in the 1.0-μm n-well Hewlett-Packard CMOS process with double metal interconnects and was packaged in a 88-pin ceramic leadless chip carrier. Some important characteristics of the design are shown in Table I. Upon receiving the packaged prototypes from the foundry, the chips were tested using a set of 1820 fault-graded test vectors. Fault coverage for these vectors was found to be in excess of 94%. The built-in test function utilizing the TEST instruction provides 100% fault coverage for data-path faults which occupy 63.3% of the total device count. Since this instruction also tests a small portion of the control logic, this off-line test strategy can detect and isolate approximately 70% of the single stuck-at faults in the entire LDCD array.
system can be used in real-time applications to increase the bandwidth of communication systems and can also be used to effectively increase the amount of mass storage available to computer systems. During the design of the LDCCD chip we have facilitated a design methodology that includes both custom and semicustom design styles. Custom design was used to maximize circuit density in highly repetitive logic blocks thus reducing the overall die size and maximizing system performance. Semicustom design, assisted by the available silicon compilation tools, was used during logic minimization and implementation of the control circuitry to maintain a quick design.

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REFERENCES

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