# An FPGA-specific Approach to Floating-Point Accumulation and Sum-of-Products

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#### **Abstract**

This article studies two common situations where the flexibility of FPGAs allows one to design applicationspecific floating-point operators which are more efficient and more accurate than those offered by processors and GPUs. First, for applications involving the addition of a large number of floating-point values, an ad-hoc accumulator is proposed. By tailoring its parameters to the numerical requirements of the application, it can be made arbitrarily accurate, at an area cost comparable to that of a standard floating-point adder, and at a higher frequency. The second example is the sum-of-product operation, which is the building block of matrix computations. A novel architecture is proposed that feeds the previous accumulator out of a floating-point multiplier whose rounding logic has been removed, again improving the area/accuracy tradeoff. These architectures are implemented within the FloPoCo generator, freely available under the LGPL.

## 1. Introduction

Most general-purpose processors have included floating-point (FP) units since the late 80s, following the IEEE-754 standard. The feasibility of FP on FPGA was studied long before it became a practical possibility [23, 17, 19]. As soon as the sizes of FPGAs made it possible, many libraries of floating-point operators were published (see [3, 16, 18, 22, 8] among other). FPGAs could soon provide more FP computing power than a processor in single precision [18, 22], then in double-precision [25, 10, 7]. Here single precision (SP) is the standard 32-bit format consisting of a sign bit, 8 bits of exponent and 23 bits of significand (or

mantissa), while double-precision (DP) is the standard 64-bit format with 11 bits of exponent and 52 significand bits. Since then, FPGAs have increasingly been used to accelerate scientific, financial and other FP-based computations. This acceleration is essentially due to massive parallelism [24], as basic FP operators in an FPGA are typically slower than their processor counterparts by one order of magnitude.

Most of the aforementionned applications are very close, from the arithmetic point of view, to their software implementations. They use the same basic operators, although the internal architecture of the operators may be highly optimised for FPGAs [17, 20, 20, 12]. For such applications, it is expected that GPUs will soon outperform FPGAs.

However, FPGAs are more flexible than that. For example, most published FP libraries are fully parameterisable in significand length and exponent length [16, 8], but applications that exploit this flexibility are rare [22, 25].

The FloPoCo project<sup>1</sup> studies how the flexibility of the FPGA target can be better exploited in the floating-point realm. In particular, it looks for operators which are radically different from those present in microprocessors. In the present article, such an operator is presented for the ubiquitous operation of floating-point accumulation, and applied to sums of products.

These operators are demonstrated in the FloPoCo core generator. One goal of FloPoCo is to fine-tune the architectural parameters to the target hardware (currently Virtex-4 and Stratix II) and to an objective frequency. This is also demonstrated.

#### 2. Floating-point accumulation

Summing many independent terms is a very common operation. Scalar product, matrix-vector and matrix-matrix products are defined as sums of products. Numerical in-

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<sup>1</sup>www.ens-lyon.fr/LIP/Arenaire/Ware/FloPoCo/

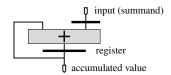


Figure 1. Iterative accumulator

tegration usually consists in adding many elementary contributions. Monte-Carlo simulations also involve sums of many independent terms. Many other applications involve accumulations of floating-point numbers, and some related work will be surveyed in section 6.

If the number of summands is small and constant, one may build trees of adders, but to accommodate the general case, it is necessary to design an iterative accumulator, illustrated by Figure 1.

It is a common situation that the error due to the computation of one summand is independent of the other summands and of the sum, while the error due to the summation grows with the number of terms to sum. This happens in integration and sum of products, for instance. In this case, it makes sense to have more accuracy in the accumulation than in the summands.

A first idea is to use a standard FP adder, possibly with a larger significand than the summands. The problem is that FP adders have long latencies: typically l=3 cycles in a processor, up to tens of cycles in an FPGA (see Table 1). This is explained by the complexity of their architecture, illustrated on Figure 2.

This long latency means that an accumulator based on an FP adder will either add one number every l cycle, or compute l independent sub-sums which then have to be added together somehow. This will add to the complexity and cost of the application, unless at least l accumulation can be interleaved, which is the case of large matrix operations [26, 4].

In addition, an accumulator built out of a floating-point adder is inefficient, because the significand of the accumulator has to be shifted, sometimes twice (first to align both operands and then to normalise the result, see Figure 2). These shifts are in the critical path of the loop of Figure 1.

In this paper, we suggest to build an accumulator of floating-point numbers which is tailored to the numerics of each application in order to ensure that 1/ its significand never needs to be shifted, 2/ it never overflows and 3/ it eventually provides a result that is as accurate as the application requires. We also show that it can be clocked to any frequency that the FPGA supports. We show that, for many applications, the determination of operator parameters ensuring the required accuracy is easy, and that the area can be much smaller for a better overall accuracy. Finally, we

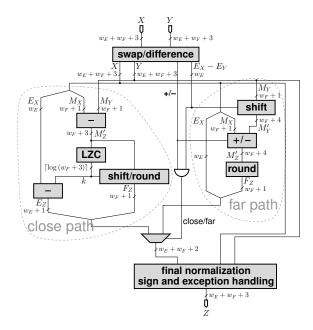


Figure 2. A typical floating-point adder ( $w_E$  and  $w_F$  are the exponent and significand sizes)

combine the proposed accumulator with a modified, errorless FP multiplier to obtain an accurate application-specific dot-product operator.

# 3. A fast and accurate accumulator

This section presents the architecture of the proposed accumulator. Section 4 will discuss the determination of its many parameters in an application-specific way.

#### 3.1. Overall architecture

The proposed accumulator architecture, depicted on Figure 3, removes all the shifts from the critical path of the loop by keeping the current sum as a large fixed-point number. Only the registers on the accumulator itself are shown. The rest of the design is combinatorial and can be pipelined arbitrarily. There is still a loop, but it is now a fixed-point addition for which current FPGAs are highly efficient. Specifically, the loop involves only the most local routing, and the dedicated carry logic of current FPGAs provides good performance up to 64 bits. For instance, a Virtex-4 with speed grade -12 runs such a 64-bit accumulator at more than 220MHz, while consuming only 64 CLBs. Section 3.3 will show how to reach even larger frequencies and/or accumulator sizes.

For clarity, some details are not shown on this figure.

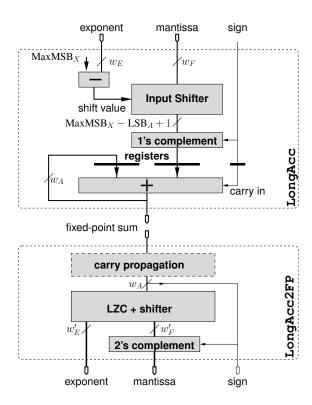


Figure 3. The proposed accumulator (top) and post-normalisation unit (bottom).

In particular, LongAcc also outputs three sticky bits (input overflow, input underflow, and accumulator overflow), and both operators manage exceptional cases (infinities and Not-a-Number).

Figure 4 illustrates the accumulation of several floatingpoint numbers (represented by their significands shifted by their exponent) into such an accumulator.

The shifters now only concern the summand (see Figure 3), and, being combinatorial, can be pipelined as deep as required by the target frequency.

As seen on Figure 3, the accumulator stores a two's complement number while the summands use a sign/magnitude representation, and thus need to be converted to two's complement. This can be performed without carry propagation: If the input is negative, it is first complemented (fully in parallel), then a 1 is added as carry in to the accumulator. All this is out of the loop's critical path, too.

#### 3.2. Parameterisation of the accumulator

Let us now introduce, with the help of Figure 4, the parameters of this architecture.

 w<sub>E</sub> and w<sub>F</sub> are the exponent size and significand size of the summands

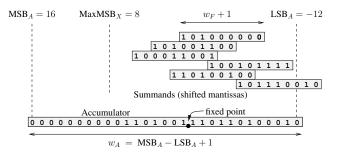


Figure 4. Accumulation of floating-point numbers into a large fixed-point accumulator

- MSB<sub>A</sub> is the position of the most-significant bit (MSB) of the accumulator. If the maximal expected sum is smaller than 2<sup>MSB<sub>A</sub></sup>, no overflow ever occurs.
- LSB<sub>A</sub> is the position of the least-significant bit of the accumulator. It will determine the final accuracy as Section 4 will show.
- For simplicity we note  $w_A = MSB_A LSB_A$  the width of the accumulator.
- MaxMSB<sub>X</sub> is the maximum expected position of the MSB of a summand. MaxMSB<sub>X</sub> may be equal to MSB<sub>A</sub>, but very often one is able to tell that each summand is much smaller in magnitude than the final sum. In this case, providing MaxMSB<sub>X</sub> < MSB<sub>A</sub> will save hardware in the input shifter.

The main claim of the present article is the following: For most applications accelerated using an FPGA, values of  $MaxMSB_X$ ,  $MSB_A$  and  $LSB_A$  can be determined a priori, using a rough error analysis or software profiling, that will lead to an accumulator smaller and more accurate than the one based on an FP adder. This claim will be justified in section 4.

This claim sums up the essence of the advantage of FP-GAs over the fixed FP units available in processors, GPUs or dedicated floating-point accelerators: We advocate an accumulator specifically tailored for the application to be accelerated, something that would not be possible or economical in a general-purpose FPU.

# **3.3.** Fast accumulator design using partial carry-save

If the dedicated carry logic of the FPGA is not enough to reach the target frequency, a partial carry-save representation allows to reach any arbitrary frequency supported by the FPGA. As illustrated by Figure 5, the idea is to cut the large carry propagation into smaller chunks of k bits (k=4

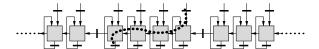


Figure 5. Accumulator with 4-bit partial carrysave. The boxes are full adders, bold dashes are 1-bit registers, and the dots show the critical path.

on the figure), simply by inserting  $\lfloor (\text{MSB}_A - \text{LSB}_A)/k \rfloor$  registers. The critical path is now that of a k-bit addition, and the value of k can therefore be chosen to match the target frequency. This is a classical technique which was in particular suggested by Hossam, Fahmy and Flynn [11] for use as an internal representation in processor FPUs. For k=1 one obtains a standard carry-save representation, but larger values of k are prefered as they take advantage of dedicated carry logic while reducing the register overhead. The FloPoCo implementation computes k out of the target frequency. For illustration, k=32 allows to reach 400MHz on Virtex-4 and StratixII. The additional hardware cost is just the few additional registers -1/4 more in our figure, and 1/32 more for 400MHz accumulation on current FPGAs.

Of course a drawback of the partial carry-save accumulator is that it holds its value in a non-standard redundant format. To convert to standard notation, there are two options. One is to dedicate  $\lfloor (MSB_A - LSB_A)/k \rfloor$  cycles at the end of the accumulation to add enough zeroes into the accumulator to allow for carry propagation to terminate. This comes at no hardware cost. The other option, if the running value of the accumulator is needed, is to perform this carry propagation in a pipelined way before the normalisation – this is the carry propagation box on Figure 3. The important fact is again that this carry propagation is outside of the critical loop.

#### 3.4. Post-normalisation unit, or not

Figure 3 also shows the FloPoCo LongAcc2FP postnormalisation unit, which performs the conversion of the long accumulator result to floating-point.

Let us first remark, using a few examples, that this component is probably much less useful than the accumulator itself.

In [6], the FPGA computes a very large integration – several hours – and only the final result is relevant. In such applications, it makes no sense to dedicate hardware to the conversion of the accumulator back to floating-point. FPGA resources will be better exploited at speeding up the computation as much as possible, and FloPoCo provides a small helper program to perform this conversion in software.

Another common case is that one needs one normalisa-

tion every N accumulations. For instance, a dot product of vectors of size N accumulates N numbers before needing to convert the result back to floating-point. Therefore, in matrix operations, one pipelined  $\mathtt{LongAcc2FP}$  may be shared between N dot product operators [26], at the cost of some multiplexers and routing. Alternatively, one may use N instances of  $\mathtt{LongAcc2FP}$  running at 1/N the frequency of the accumulator – they will be smaller. In both cases, it makes sense to provide  $\mathtt{LongAcc2FP}$  as a separate component, as on Figure 3. In the following we give separate synthesis results for the accumulators themselves and the post-normalisation unit.

Note that the same discussion holds for an accumulator based on an FP adder of latency l (that actually computes l intermediate subsums). If only the final sum is needed, it may be computed in software at no extra hardware cost. However, if the running sum is needed at each cycle, it will take l-1 additions to get it [26, 4].

Back to LongAcc2FP, it mostly consists in leading-zero/one counting and shifting, followed by conversion from 2's complement to sign/magnitude, and rounding. If the accumulator holds a partial carry-save value, the carries need to be propagated. This simply requires  $\lceil w_A/k \rceil$  pipeline levels, each consisting of one k-bit adder and  $\lceil w_A/k \rceil - 1$  registers of k bits, and it can actually be merged with the 2's complement conversion. Again, all this may be performed at each cycle and pipelined arbitrarily.

# 3.5. Synthesis results

All the results in this article are synthesis results (before place-and-route) obtained for Virtex-4, speedgrade -12, using ISE10.1. Post place-and-route results will depend on the FPGA occupation and floorplanning. Very similar results have been obtained for Altera Stratix II.

Table 1 illustrates the performance of the proposed accumulator compared to one built using a floating-point adder from the Xilinx CoreGen tool. These operators are not functionnaly equivalent. The proposed accumulator is more accurate (Section 4.3 will study this quantitatively), but does not return a normalised result as the accumulator based on an FP adder.

For each summand size, we build accumulators of twice the size of the input significand (MSB $_A=w_E$ ), LSB $_A=-w_E$ ) for two configurations: a small one where MaxMSB $_X=1$ , and a larger one where MaxMSB $_X=MSB_A=w_E$ . Again, these results are for illustration only: An accumulator should be built in an application-specific way. As section 4 will show, a typical accumulator will be between these two configurations.

Table 2 provides results for the LongAcc2FP postnormalisation unit. These results are preliminary: the target frequency of 400MHz is not reached, due to limits of the

summand $(w_E, w_F)$	CoreGen FP adder $(w_E, w_F)$	$2w_F$ accumulator, $MaxMSB_X = 1$	$2w_F$ accumulator, $MaxMSB_X = MSB_A$
(7,16)	304 slices + 1 DSP, 12 cycles @ 359 MHz	129 slices, 8 cycles @ 472 MHz	176 slices, 9 cycles @ 484 MHz
(8,23) SP	317 slices + 4 DSP, 16 cycles @ 450 MHz	165 slices, 8 cycles @ 434 MHz	229 slices, 9 cycles @ 434 MHz
(10,37)	631 slices + 1 DSP, 14 cycles @ 457 MHz	295 slices, 10 cycles @ 428 MHz	399 slices, 11 cycles @ 428 MHz
(11,52) DP	771 slices + 3 DSP, 15 cycles @ 366 MHz	375 slices, 11 cycles @ 414 MHz	516 slices, 12 cycles @ 416 MHz

Table 1. Compared synthesis results for an accumulator based on FP adder, versus proposed accumulator with  $MSB_A = w_E$ ,  $LSB_A = -w_E$ , all targetted for 400MHz on a Virtex-4.

current pipelining framework of FloPoCo.

# 4. Application-specific accumulator design

Let us now justify the claim, made in 3.2, that the few parameters of the proposed accumulator are easy to determine on a per-application basis. We acknowledge that the main purpose of floating-point is to free the designer from the painful task of converting a computation on real numbers to fixed-point. Indeed, the proposed accumulator is definitely a floating-point operator, and we hope to convince the reader that the effort it requires to set up is minimal.

### 4.1. A performance/ accuracy tradeoff

First note that a designer has to provide a value for  $MSB_A$  and  $MaxMSB_X$ , but these values do not have to be accurate. For instance, adding 10 bits of safety margin to  $MSB_A$  has no impact on the latency and very little impact on area. Now from the application point of view, 10 bits mean 3 orders of magnitude. For most applications, it is huge. A designer in charge of implementing a given computation on FPGA is expected to understand it well enough to bound the expected result with a margin of 3 orders of magnitude. An actual example is detailed below in 4.2. As another example, consider a Monte Carlo simulation where the accumulation computes an estimate of the value of a share. No share will go beyond, say, \$100,000 before something happens that makes the simulation invalid anyway.

It may be more difficult to evaluate  $MaxMSB_X$ . In doubt,  $MaxMSB_X = MSB_A$  will do, but in many cases

$(w_E,w_F)$	LongAcc2FP, $2w_F \rightarrow w_F$
(7,16)	124 slices, 5 cycles @ 343 MHz
(8,23) SP	158 slices, 5 cycles @ 338 MHz
(10,37)	325 slices, 7 cycles @ 267 MHz
(11,52) DP	413 slices, 7 cycles @ 266 MHz

Table 2. Preliminary synthesis results for a LongAcc2FP compatible with Table 1, rounding an accumulator of size  $2w_F$  to an FP number of size  $w_F$ .

application knowledge will help reduce it, hence reducing the input shifter size. For instance, in Monte Carlo simulations, probabilities are smaller than 1. Another option is profiling. A typical instance of the problem may be run in software, instrumented to output the max and min of the absolute values of summands. Again, the trust in such an approach comes from the possibility of adding 20 bits of margin for safety.

In some cases, the application will dictate  $MaxMSB_X$  but not  $MSB_A$ . In this case, one has to consider the number n of terms to add. Again, one will usually be able to provide an upper bound, be it the extreme case of 1 year running at 500MHz, or  $2^{53}$  cycles. In a worst-case scenario on such simulation times, this suggests the relationship  $MSB_A = MaxMSB_X + 53$  to avoid overflows. For comparison, 53 is the precision of a DP number, so the cost of this worst case scenario is simply a doubling of the accumulator itself, but not of the input shifter which shifts up to  $MaxMSB_X$  only. It will cost just slightly more than 53 LUTs in the accumulator (although much more in the post-normalisation unit if one is needed).

The last parameter,  $LSB_A$ , allows a designer to manage the tradeoff between precision and performance. First, remark that if a summand has its LSB higher than LSB<sub>A</sub> (case of the 5 topmost summands on Figure 4), it is added exactly, entailing no rounding error. Therefore, the proposed accumulator will compute exactly if the accumulator size is large enough so that its LSB is smaller than those of all the inputs. Conversely, if a summand has an LSB smaller than LSB<sub>A</sub> (case of the bottommost summand on Figure 4), adding it to the accumulator entails a rounding error of at most  $2^{LSB_A-1}$ . In the worst case, when adding n numbers, this error will be multiplied by n and invalidate the  $\log_2 n$ lower bits of the accumulator. A designer may lower LSB $_A$ to absorb such errors, an example is given below in 4.2. A practical maximum is again an increase of 53 bits for 1 year of computation at 500MHz.

Here we have only discussed the errors due to the accumulation process. In practice, even when a summand is added exactly, it is usually the result of some rounding, so it carries an error of the order of its LSB, which it adds to the accumulator. These summand errors, which are outside of the scope of this article (they can be reduced by increasing

 $w_F$ ), will typically dwarf the rounding errors due to the accumulator. This suggests that the previous worst-case analysis will typically lead to an accumulator that is much more accurate (and bulky) than the application actually requires.

All considered, it is expected that an accumulator will rarely need to be designed larger than 100 bits. Note that the fast carry chain of the smallest Virtex-4 already extends to 128-bit.

Finally, thanks to the sticky output bits for overflows in the summands and in the accumulator, the validity of the result can be checked a posteriori.

#### 4.2. A case study

In the inductance computation of [6], physical expertise tells that the sum will be less than  $10^5$  (using arbitrary units due to factoring out some physical constants), while profiling showed that the absolute value of a summand was always between  $10^{-2}$  and 2.

Converting to bit positions, and adding two orders of magnitude (or 7 bits) for safety in all directions, this defines  $\text{MSB}_A = \lceil \log_2(10^2 \times 10^5) \rceil = 24$ ,  $\text{MaxMSB}_X = 8$  and  $\text{LSB}_A = -w_F - 15$  where  $w_F$  is the significand width of the summands. For  $w_F = 23$  (SP), we conclude that an accumulator stretching from  $\text{LSB}_A = -23 - 15 = -38$  (least significant bit) to  $\text{MSB}_A = 24$  (most significant bit) will be able to absorb all the additions without any rounding error: No summand will add bits lower than  $2^{-38}$ , and the accumulator is large enough to ensure it never overflows. The accumulator size is therefore  $w_A = 24 + 38 + 1 = 63$  bits.

Remark that only  $LSB_A$  depends on  $w_F$ , since the other parameters ( $MSB_A$  and  $MaxMSB_X$ ) are related to physical quantities, regardless of the precision used to simulate them. This illustrates that  $LSB_A$  is the parameter that allows one to manage the accuracy/area tradeoff for an accumulator.

#### 4.3. Accuracy measurements

Table 3 compares for accuracy and performance the proposed accumulator to one built using Xilinx CoreGen in the context of the previous case study. To evaluate the accuracies, we computed the exact sum using multiple-precision software on a small run (20,000,000 summands), and the accuracy of the different accumulators was computed with respect to this exact sum. The proposed accumulator is both smaller, faster and more accurate than the ones based on FP adders. This table also shows that for production runs, which are 1000 times larger, a single-precision FP accumulator will not offer sufficient accuracy.

Table 4 provides other examples of the final relative accuracy, with respect to the exact sum, obtained by using an FP adder, and using the proposed accumulator with twice

	accuracy	area	latency
SP FP adder acc	$1.2 \cdot 10^{-3}$	317 sl, 4 DSP	16 @ 450 MHz
DP FP adder acc	$2.8 \cdot 10^{-15}$	771 sl, 3 DSP	15 @ 366 MHz
proposed acc	$2.0 \cdot 10^{-16}$	247 sl	10 @ 454 MHz

Table 3. Compared performance and accuracy of different accumulators for SP summands from [6].

sum size	rel. error for unif[0, 1]		rel. error for unif[-1, 1]	
	FP adder	long acc.	FP adder	long acc.
1000	-5.76e-05	1.05e-07	-1.59e-05	1.40e-04
10,000	-2.74e-04	1.07e-08	-3.04e-04	2.36e-04
100,000	-4.31e-04	1.07e-09	2.54e-03	-2.73e-04
1,000,000	-0.738	-3.57e-09	3.18e-03	-4.47e-05

Table 4. Accuracy of accumulation of FP(7,16) numbers, using an FP(7,16) adder, compared to using the proposed accumulator with 32 bits (MSBA = 20, LSBA = -11).

as large a significand. In the first column, we are adding nnumbers uniformly distributed in [0,1]. The sum is expected to be roughly equal to n/2, which explains that the result becomes very inaccurate for n = 1,000,000: As soon as the sum gets larger than  $2^{17}$ , any new summand in [0,1]is simply shifted out and counted for zero. This problem can be anticipated by using a larger significand, or a larger MSBA in the accumulator as we do. In the second column, numbers are uniformly distributed in [-1,1]. The sum grows as well (it is a random walk) but much more slowly. As we have taken a fairly small accumulator (LSB<sub>A</sub> = -11), for the first sums floating-point addition is more accurate: While the sum is smaller than 1, its LSB is smaller than -16. However, as more numbers are added, the sum grows. More and more of the bits of a summand are shifted out in the FP adder, but kept in the long accumulator, which becomes more accurate. Note that by adding only 5 bits to it (LSB<sub>A</sub> = -16 instead of -11), the relative error becomes smaller than  $10^{-10}$  in all cases depicted in Table 4: Again,  $LSB_A$  is the parameter allowing to manage the accuracy/area tradeoff.

We have discussed in this section only the error of the long fixed-point accumulator itself (the upper part of Fig. 3). If its result is to be rounded to an FP(7,16) number using the post-normalisation unit of Figure 3, there will be a relative rounding error of at most  $2^{-17}\approx 0.76\cdot 10^{-5}$ . Comparing this value with the relative errors given in Table 4, one concludes that the proposed accumulator, with the given parameters, always leads to a result accurate to the two last bits of an FP(7,16) number.

CoreGen, SP ×, SP +	484 sl + 8 DSP, 26 cycles @ 366 MHz
ours, $SP \times$ , $DP$ acc	319 sl + 4 DSP, 13 cycles @ 363 MHz
CoreGen, SP $\times$ , DP $+$	973 sl + 7 DSP, 26 cycles @ 366 MHz
CoreGen, DP $\times$ , DP $+$	1241 sl + 19 DSP, 37 cycles @ 366 MHz
ours, DP ×, 105-bit acc	1441 sl + 9 DSP, 23 cycles @279 MHz

Table 5. Preliminary results for Sum-Of-Products

#### 5. Accurate Sum-of-Products

We now extend the previous accumulator to a highly accurate sum-of-product operator. The idea is simply to accumulate the exact results of all the multiplications. To this purpose, instead of standard multipliers, we use exact multipliers which return all the bits of the exact product: For  $1+w_F$ -bit input significand, they return an FP number with a  $2+2w_F$ -bit significand. Such multipliers incur no rounding error, and are actually cheaper to build than the standard  $(w_E,w_F)$  ones. Indeed, the latter also have to compute  $2w_F+2$  bits of the result, and in addition have to round it. In the exact FP multiplier, results do not need to be rounded, and do not even need to be normalised, as they will be immediately sent to the fixed-point accumulator. There is an additional cost, however, in the accumulator, whose input shifter is twice as large.

This idea was advocated by Kulisch [14, 13] for inclusion in microprocessors, but a generic DP version requires a 4288 bits accumulator, which manufacturers always considered too costly to implement. On an FPGA, one may design an application-specific version with an accumulator of 100-200 bits only. This is being implemented in FloPoCo, and Table 5 provides preliminary synthesis results for single and double precision input numbers, with the same  $2w_F$  accumulators as in Table 1. Some work is still needed to bring the FloPoCo multiplier generator to CoreGen level.

# 6. Comparison with related work

Much research has been dedicated to converting floating-point computations to fixed-point. When an input vector is to be multiplied by a constant matrix (as happens in filters, FFTs, etc), one may use block floating-point (BFP), a technique known since the 50s and recently applied to FP-GAs [1, 2]. It consists in an initial alignment of all the input significands to the largest one (bringing them all to the same exponent), after wich all the computations (multiplications by constants and accumulation) can be performed in fixed-point. The proposed accumulator could be used as a building block for BFP, however it was designed for a much larger class of application, and with a motivation of accuracy inspired by Kulisch's work [14, 13].

The group-alignment based floating-point accumulation technique of He et al [12] applies BFP to arbitrary accumulation. The inputs are first buffered into blocks (called groups here) of size m (with m=16 in the paper). The numbers in a group are added using BFP. Then, these partial sums are fed to a final stage of FP accumulation that may run at 1/m the frequency of the first stage, and may therefore use a standard unpipelined FP adder. This is a very complex design (for SP, 443 slices without the last stage, 716 with it). Besides, the frequency of the BFP accumulator will not scale well to higher precisions without resorting to techniques similar to our partial carry save.

Luo and Martonosi [21] have described an architecture for the accumulation of SP numbers that uses two 64-bit fixed-point adders. It first shifts the input data according to the 5 lower bits of the exponent, then sends it to one of the fixed-point accumulators depending on the higher exponent bits. If these differ too much, either the incoming data or the current accumulator is discarded completely, just as in an FP adder. The critical path of the accumulator loop includes one 64-bit adder and a 3-2 compressor. The main problem with this approach (besides its complexity) is that it is a fixed design that will not scale beyond single-precision. Another one is that the detection of accumulator overflow may stall the operator, leading to a variable-latency design. The authors suggest a workaround that imposes a limit on the number of summands to add.

Zhuo and Prasanna [26], then Bodnar et all [4] have described high-throughput matrix operations using carefully scheduled standard FP adders. Performance-wise, this approach should be comparable to ours. Still, the proposed accumulator is more generic and exposes a finer control of the accuracy-performance tradeoff.

#### 7. Conclusion and future work

The accumulator design presented in this article perfectly illustrates the philosophy of the FloPoCo project: Floating-point on FPGA should make the best use of the flexibility of the FPGA target, not re-implement operators available in processors. The proposed accumulator is deliberately application-specific. In addition it may be tailored to be arbitrarily faster and arbitrarily more accurate than a naive floating-point approach, without requiring more resources.

This approach requires the designer to provide bounds on the orders of magnitudes of the values accumulated. We have shown that these bounds can be taken lazily. In return, the designer gets not only improved performance, but also a provably accurate accumulation process. We believe that this return is worth the effort, especially considering the overall time needed to implement a full floating-point application on an FPGA.

FloPoCo already includes many other operators, from constant multipliers [5] to elementary functions generators [9]. In addition to developping new operators, the challenge is now to integrate them in the emerging C-to-FPGA compilers [15]. In parallel, FloPoCo should be extended with tools that help a designer set up a complete floating-point datapath, managing synchronisation issues but also accuracy ones. Such tools are still at the drawing board.

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