

IMPLEMENTATION OF LOW-POWER APPROXIMATE UNSIGNED MULTIPLIERS WITH CONFIGURABLE ERROR RECOVERY

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Abstract: Inexact circuits have been considered for applications that can endure some deficiency of precision with further developed execution or potentially energy effectiveness. Multipliers are key number juggling circuits in a considerable lot of these applications including DSP. In this paper, a clever estimated multiplier with a low power utilization and a short basic way is proposed for elite execution DSP applications. This multiplier uses a recently planned estimated addition that restricts it convey spread to the closest neighbors for quick halfway item gathering. Various degrees of exactness can be accomplished by utilizing either OR entryways or the proposed surmised addition in a configurable blunder recuperation circuit. The approximate multipliers using these two error reduction strategies are referred to as AM1 and AM2, respectively. Both AM1 and AM2 have a low mean error distance, i.e., most of the errors are not significant in magnitude. Compared with a Wallace multiplier optimized for speed, an 8×8 AM1 using four most significant bits for error reduction shows a 60% reduction in delay (when optimized for delay) and a 42% reduction in power dissipation (when optimized for area).

Index Terms— Approximate computing, multiplier, adder, error recovery, low-power, image processing.

1. Introduction

We have an abundance of new information, not just from big, efficient research and business machines, but also from billions of low-power devices of different sorts. While conventional workloads like transactional or database analysis continue to develop modestly, a variety of systems are explosively computerized to achieve greater insight in large quantities of structured and

unstructured data. Traditional computing implies an accuracy which is not required in most types of information. Nevertheless, though, these computational systems stay on high-precision (and reliable) frameworks for general purposes (and accelerator). The goal of approximate computing is to loosen these restrictions in order to achieve substantial computational performance improvements-while preserving reasonable output.

A main research aim in estimated calculation is to define the estimation thresholds within various layers of the machine stack (from computers to circuits and electronic components) to enable for an appropriate but probably specific outcome from the results obtained using precise calculation. Approximate computational techniques researched by different researchers centred mainly on improving one component of the device stack. In this paper we explored how the use of several solution approaches spanning upwards of one layer of the device stack added value and whether such composite effects became widely available across various implementation areas.

We focused on three types of approximations to provide a practical demonstration: missing measurements, approximating numerical computations and approximating coordination between computing elements. We test loop perforation, decreased arithmetic precision and relaxing of synchronization as representations of each division. We have chosen applications which are computer-priced which can affect our lives dramatically if they are inexpensive and widespread. Digital image processing, robots and machine intelligence is our research fields. Our business. Our findings show in all of the applications tested that in the applications tested we were in a

position to perforate hot loops by an average of 50%, thus that the total execution time proportionately while also delivering reasonable results efficiency. Furthermore, the data used to calculate the current 32 bits or even 64 bits could be reduced to 10-16 bits, with potential significant quality and power benefits. Truncated out the simultaneous applications we analyzed; the execution period was decreased by 50 percent by partial deletion of overhead synchronization.

Eventually, our results show that the advantages of these methods become amplified when used together. In other terms, the use of different strategies does not substantially decrease one another's efficacy. Seeing that the advantages of approximation computation are not confined to a particular class of programmers, these findings enable one to reconsider the design of the general function of the system such that various forms of approximations may be taken into account more effectively.

Rounding is one of the most effective methods to pack data input before processing. This approach will boost circuit features such as energy and usage, speed and range which are appropriate for estimated computation. Approximate computer technology works well with most of computer vision , image

processing, design recognition, image analysis, computer science, and machine learning applications that are resilient to errors. Work into these fields has created plenty of work resources Truncatedout the past few years. A multiplier is a basic computing block one of the most commodities consuming. Along this point, we are seeing endless studies along accuracy and energy loss dramatically compromised. Partial product production, partial product reduction and packaging are the fundamental building blocks of the converter. This paper suggests rounding as a modern input frame approach until partial product output. The precision curve plays a vital function in managing and raising the number of errors.

Guess it depends on requests, considerable. Specific algorithms on varying types of multiplier frames are applied. The input block has a 16bit and 32bit rounding technique based on precision levels. The intermediate results generated are categorized into active or partial products that are inactive. Both unused sections are nil and thus no compressors can be taken into account in the reduction process.

2. Design of proposed multiplier

A. The Approximate Adder

In this part, the plan of another inexact addition is introduced. This addition works on a bunch of pre-handled information

sources. The information pre-handling (IPP) depends on the commutativity of pieces with similar loads in various addends. For instance, think about two arrangements of contributions to a 4-cycle addition: I) $A = 1010, B = 0101$ and ii) $A = 1111, B = 0000$. Unmistakably, the augmentations in I) and ii) produce a similar outcome. In this cycle, the two info bits $A_i B_i = 01$ are comparable to $A_i B_i = 10$ (with I being the piece file) because of the commutativity of the relating pieces in the two operands. The fundamental principle for the IPP is to switch A_i and B_i if $A_i = 0$ and $B_i = 1$ (for any I), while keeping different blends (i.e., $A_i B_i = 00, 10$ and 11) unaltered. Thusly, more 1's are normal in A_n and more 0's are normal in B . Assuming that $A_i B_i$ are the i th bits in the pre-handled data sources, the IPP capacities are given by:

$$A_i = A_i + B_i, \tag{1}$$

$$B_i = A_i B_i. \tag{2}$$

Table 1: Truth table of the inexact addition cell. 'x' addresses that no such a mix happens due to the ipp

S_i/E_i	$\hat{B}_i \hat{B}_{i-1}$			
	00	01	11	10
00	0/0	X	X	X
01	0/0	1/0	X	X
11	1/0	1/1	1/0	0/0
10	1/0	X	X	0/0

Conditions (1) and (2) register the proliferate and produce signals utilized in

an equal addition, for example, the convey lookahead addition (CLA). The proposed addition can deal with information in equal by cutting the convey spread chain. Allow A_n and B to indicate the two info double operands of an addition, S be the total outcome, and E address the blunder vector. Simulated intelligence, B_i , S_i and E_i are the i th least huge pieces of A , B , S and E , separately. A convey proliferation chain begins at the i th bit when $B^i = 1$, $A^{i+1} = 1$, $B^{i+1} = 0$. In an exact addition, $S_{i+1} = 0$ and the convey spreads to the higher piece. In any case, in the proposed Truncated addition, S_{i+1} is set to 1 and aerror signal is produced as $E_{i+1} = 1$. This forestalls the convey signal from spreading to the higher pieces. Consequently, a convey signal is created simply by the produce signal, i.e., $C_i = 1$ just when $B^i = 1$, and it just spreads to the following higher piece, i.e., the $(i + 1)$ th position. Table I shows reality table of the surmised addition, where A^i , B^i and B^{i-1} are the contributions after IPP. The error signal is used for blunder pay purposes as talked about in a later segment. For this situation, the surmised addition is like an excess number framework [24] and the sensible elements of Table I are given by

$$S_i = B^{i-1} + B^i A^i, \quad (3)$$

$$E_i = B^i B^{i-1} A^i. \quad (4)$$

By supplanting A^i and B^i utilizing (1) and (2) individually, the rationale capacities regarding the first information sources are given by

$$S_i = (A_i \oplus B_i) + A_{i-1}B_{i-1}, \quad (5)$$

$$E_i = (A_i \oplus B_i)A_{i-1}B_{i-1}, \quad (6)$$

where I is the piece record, i.e., $I = 0, 1, \dots, n$ for a n -digit addition. Let $A^{-1} = B^{-1} = 0$ when I is 0, consequently, $S_0 = A_0 \oplus B_0$ and $E_0 = 0$. Likewise, $E_i = 0$ when A_{i-1} or B_{i-1} is 0. Consider a n -cycle addition, the data sources are given by $A = A_{n-1} \dots A_1 A_0$ and $B = B_{n-1} \dots B_1 B_0$, the specific total is $S = S_{n-1} \dots S_1 S_0$. Then, at that point, S_i can be figured as $S_i + E_i$ and hence, the specific amount of A and B is given by

$$S = S + E. \quad (7)$$

In (7) '+' signifies the expansion of two paired numbers rather than the 'OR' work. The error E is consistently non-negative and the surmised total is consistently equivalent to or more modest than the precise total. This is a significant component of this addition on the grounds that an extra addition can be utilized to add the blunder to the surmised total as a pay step. While this is instinctive in a addition plan, it is an especially valuable element in a multiplier plan as just a single extra addition is expected to lessen the blunder in the eventual outcome

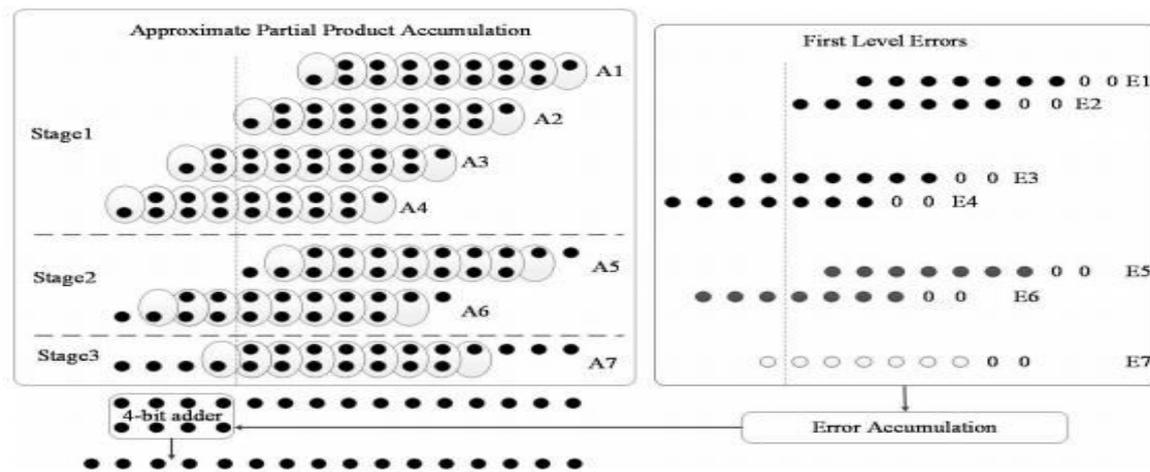


Fig. 1. A Truncated multiplier with incomplete blunder recuperation utilizing 5 MSBs of the error vector: a halfway item, aggregate or a blunder bit created at the primary stage; : a error bit produced at the subsequent stage; : a blunder bit created at the last stage.

B. Proposed Approximate Multiplier

A distinctive component of the proposed surmised multiplier is the effortlessness to utilize inexact adders in the fractional item gathering. It has been shown that this might prompt low precision [14], on the grounds that blunders might amass and it is hard to address errors utilizing existing surmised adders. In any case, the utilization of the recently proposed surmised addition conquers this issue by using the error signal. The subsequent plan has a basic way postpone that is more limited than a regular the slightest bit full addition, on the grounds that the new n-cycle addition can handle information in equal. The surmised addition has a fairly high error rate, however the component of creating both the total and blunder flags simultaneously diminishes blunders in the end result. A addition tree is used for

fractional item amassing; the errorsignals in the tree are then used to repay the blunder in the result to create an item with a superior precision. The engineering of the proposed estimated multiplier is displayed in Fig. 1. In the proposed plan, the disentanglement of the incomplete item aggregation stage is cultivated by utilizing a addition tree, in which the quantity of fractional items is decreased by a variable of 2 at each phase of the tree. This addition tree is generally not executed utilizing precise multi-bit adders because of the long inertness. In any case, the proposed estimated Fig. 2. Images for (a) an OR entryway, (b) a full addition or a half addition and (c) a Truncated addition cell addition is reasonable for carrying out a addition tree, since it is less intricate than an ordinary addition and has a lot more limited basic way delay. Careful quick

multipliers frequently incorporate a Wallace or Dadda tree utilizing full adders (FAs) and half adders (HAs); blowers are additionally used in the Wallace or Dadda tree to additionally lessen the basic way with an expansion in circuit region. These plans require an appropriate determination of various circuit modules; for instance, 4:2 blowers, FAs and HAs are usually utilized in a Wallace tree and a wise association of these modules should be considered in a tree plan. This builds the plan intricacy, particularly when multipliers of various sizes are thought of; the proposed configuration is straightforward for different multiplier sizes.

III. ERROR REDUCTION

The inexact addition produces two signals: the surmised total S and the blunder E ; the utilization of the error signal is viewed as close to lessen the error of the multiplier. As (7) is material to the amount of each and every surmised addition in the tree, a blunder decrease circuit is applied to the last increase result rather than to the result of every addition. Two stages are needed to decrease errors: i) blunder gathering and ii) blunder recuperation by the expansion of the collected blunders to the addition tree yield utilizing a CPA. In the blunder amassing step, error signals are aggregated to a solitary error vector, which is then added to the result vector of the fractional

item collection tree. Two inexact blunder gathering strategies are proposed, yielding the Truncated multiplier 1 (AM1) and surmised multiplier 2 (AM2). Fig. 2 shows the images for an OR entryway, a full addition and half addition cell and an inexact addition cell utilized in the blunder amassing tree.

A. Error Accumulation for Approximate Multiplier 1

As displayed in Fig. 1, each surmised addition A_i produces a total vector S_i and a error vector E_i , where $i = 1, 2, \dots, 7$. In the event that the error signals are added utilizing precise adders, the gathered blunder can completely remunerate the erroneous item; but to lessen intricacy, an inexact blunder collection is presented.

Consider the perception that the error vector of each estimated addition will in general have more 0's than 1's. Thusly, the likelihood that the error vectors have a blunder bit '1' at a similar position, is tiny. Henceforth, an OR entryway is utilized to Truncated process the amount of the errors for a solitary cycle. In the event that m blunder vectors (signified by E_1, E_2, \dots, E_m) must be amassed, then, at that point, the amount of these vectors is acquired as $E_i = E_{1i} \text{ OR } E_{2i} \text{ OR } \dots \text{ OR } E_{mi}$. To lessen blunders, an aggregated error vector is added to the addition tree yield utilizing an ordinary CPA (for example a convey lookahead addition).

Notwithstanding, just a few (for example k) MSBs of the error signals are utilized to remunerate the results to additionally lessen the general intricacy. The quantity of MSBs is chosen by the degree that blunders should be redressed. For instance in a 8×8 addition tree, there are a sum of 7 blunder vectors, produced by the 7 estimated adders in the tree. Nonetheless, not every one of the pieces in the 7 vectors should be added, on the grounds that the MSBs of certain vectors are less huge than the most un-critical pieces of the k MSBs. In the case of Fig. 1, 5 MSBs (for example the (11 – 14)th pieces, no blunder is produced at the fifteenth piece position) are considered for error recuperation and in this manner, 4 error vectors are thought of (i.e., the error vectors E3, E4, E6 and E7). The error vectors of the other three adders are less huge than the eleventh piece, so they are not thought of. The gathered error E is acquired utilizing (8); then, at that point, the end-product is found by adding E to S utilizing a quick precise CPA. The error collection conspire is displayed in Fig. 3. As no error is created essentially critical two pieces of each inexact addition A_i ($i = 1, 2, \dots, 7$), the most un-huge two pieces of every blunder vector E_i are not aggregated.

B. Blunder Accumulation for Approximate Multiplier 2

The error gathering plan for AM2 is displayed in Fig. 4. To present the plan of AM2, a 8×8 multiplier with two data sources X and Y is thought of. For instance, consider the initial two incomplete item vectors $X_0Y_7, X_0Y_6, \dots, X_0Y_0$ and $X_1Y_7, X_1Y_6, \dots, X_1Y_0$ amassed by the primary inexact addition (A1 in Fig. 1), where X_i and Y_i are the i th least critical pieces of X and Y , individually.

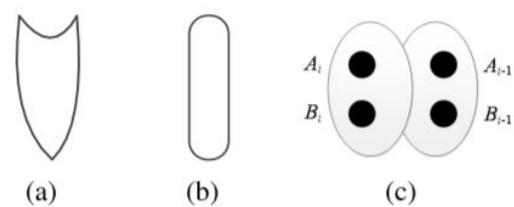


Fig. 2. Symbols for (a) an OR gate, (b) a full adder or a half adder and (c) an approximate adder cell

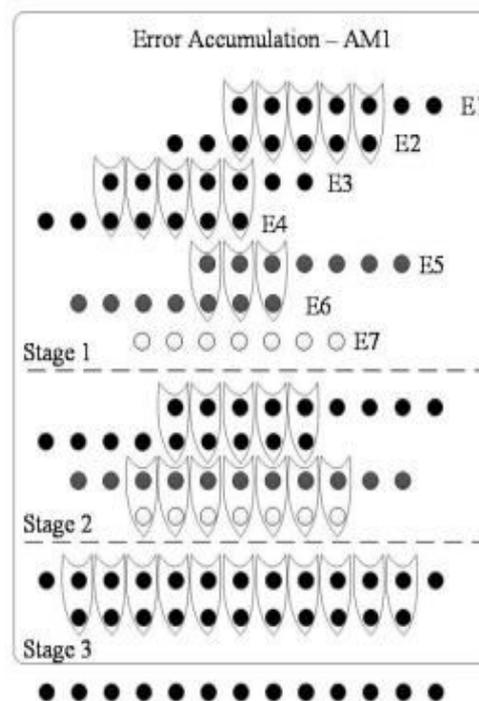


Fig. 3. Error accumulation tree for AM1.

Review from (6) for the estimated addition, the condition for $E_i = 1$ is $A_{i-1} = B_{i-1} = 1$ and $A_i = B_i$. (9) For the principal Truncated addition in the fractional item aggregation tree, its bits of feedbacks are $A = X_0Y_7, X_0Y_6, \dots, X_0Y_0$ and $B = X_1Y_7, X_1Y_6, \dots, X_1Y_0$. Subsequently, the i th least huge pieces for A and B are $A_i = X_0Y_i$ and $B_i = X_1Y_{i-1}$, separately.

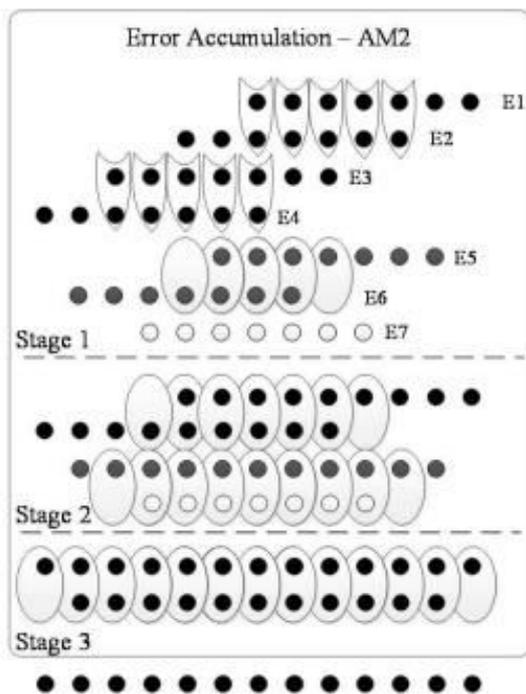


Fig. 4. Error accumulation tree for AM2.

On the off chance that X_0 or X_1 is 0, there will be no error in this Truncated addition in light of the fact that either A or B is zero. In this manner, no blunder happens except if $X_0X_1 = 11$. At the point when $X_0X_1 = 11$, A_i and B_i are rearranged to Y_i and Y_{i-1} , individually. Then, at that point, to ascertain E_i , A_{i-1} , B_{i-1} , A_i and B_i are supplanted by Y_{i-1} , Y_{i-2} , Y_i and Y_{i-1} ,

separately. For E_i to be 1, $Y_iY_{i-1}Y_{i-2} = 011$ as indicated by (9). Hence, aerror possibly happens when the info has "011" as a piece succession. In light of this perception, the "distance" between two errors in a Truncated multiplier is no less than 3 pieces. In this manner, two adjoining estimated adders in the primary phase of the fractional item tree can't have blunders at a similar section, on the grounds that the errors in a lower inexact addition are those in the upper addition moved by 2 pieces when the two errors exist. The blunders in two adjoining inexact adders would then be able to be precisely amassed by OR entryways, e.g., an OR door can be utilized to aggregate the two pieces in the error vectors E_1 and E_2 in Fig. 1. Subsequent to applying the OR entryways to aggregate E_1 and E_2 just as E_3 and E_4 , the four error vectors are compacted into two. For E_5 , E_6 and E_7 , they are created from the surmised amount of the fractional items rather than the incomplete items. Hence, they can't be precisely collected by OR additions.

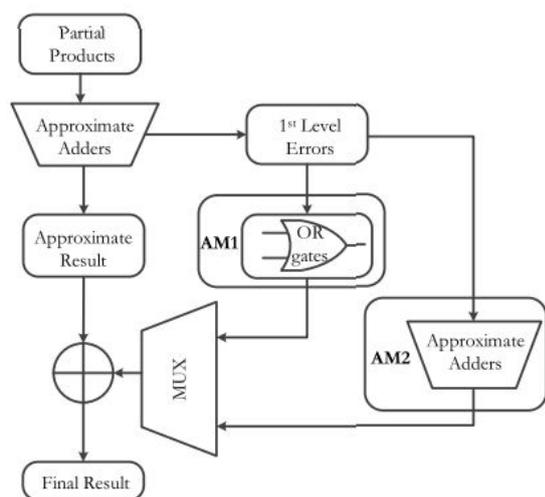


Fig. 5. Block graph of the proposed multipliers.

One more fascinating component of the proposed Truncated addition is as per the following. Accept $E_i = 1$ in (6), then, at that point, $A_{i-1} = B_{i-1} = 1$ and $A_i = B_i$. Since $A_{i-1} = B_{i-1} = 1$, i.e., $A_{i-1} \oplus B_{i-1} = 0$, it is not difficult to show that $E_{i-1} = 0$. Additionally, as $A_i = B_i$, i.e., $A_i B_i = 0$, then, at that point, $E_{i+1} = 0$. In this way, once there is a error in the slightest bit, its adjoining pieces are sans blunder, i.e., there are no continuous error bits in a single line. Along these lines, there is no convey proliferation way longer than two pieces when two error vectors are gathered, and the blunder vectors are precisely collected by the proposed Truncated addition. In view of the above examination, E_5 and E_6 are precisely aggregated by one estimated addition in the primary phase of the error amassing. After the main phase of blunder collection, three vectors are produced, and one more

two inexact adders are then used to gather these three vectors just as the error vector staying from the past stage (E_7). Recreation results (found in later areas) show that the changed error aggregation outflanks the OR-door blunder collection with minimal overhead on deferral and power. In the future, the proposed $n \times n$ inexact multiplier with k -MSB OR-door based blunder decrease is alluded to as a n/k AM1, while a $n \times n$ estimated multiplier with k -MSB surmised addition based error decrease is alluded to as a n/k AM2. The constructions of AM1 and AM2 are displayed in Fig. 5.

C. 16×16 Approximate Multipliers

In both AM1 and AM2, all the error vectors are packed to one blunder vector, which is then added back to the Truncated result of the halfway item tree. Contrasted with 8×8 plans, 16×16 multipliers produce more blunder vectors, and a lot of data would be disregarded if a similar error decrease techniques are utilized. That is, utilizing just one packed blunder vector doesn't make a decent gauge of the general error. In the altered plans, the blunder vectors created by the inexact adders are packed to two last error vectors. Take a 16×16 AM1 for instance, the eight blunder vectors created at the primary phase of the fractional item gathering tree are packed to one error vector, EV_1 , utilizing OR entryways. The leftover seven error

vectors from the second, third and fourth stages are packed to another blunder vector EV2. Then, at that point, both EV1 and EV2 are added back to the result of the halfway item at the fourth stage. Essentially, the proposed inexact adders are utilized in a 16x16 AM2 to pack the eight error vectors from the principal stage to one blunder vector and the leftover error vectors to another blunder vector. Truncation can likewise be applied to the proposed plans for huge info operands. Accordingly, 16 LSBs of the incomplete items are Truncated in 16x16 AM1 and AM2, coming about in Truncated AM1 (TAM1) and Truncated AM2 (TAM2).

4.Simulation Results

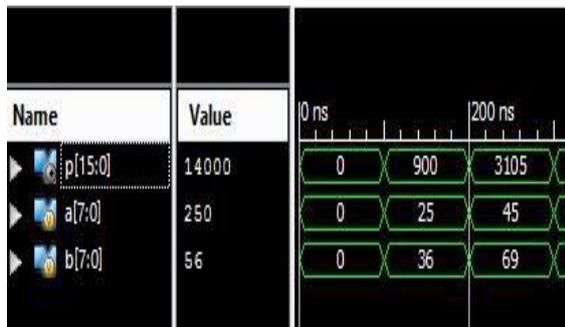


Figure 6: Simulation outcome

Device Utilization Summary (estimated values)			
Logic Utilization	Used	Available	Utilization
Number of Slice LUTs	4982	204000	2%
Number of fully used LUT-FF pairs	0	4982	0%
Number of bonded IOBs	131	600	21%

Figure 7: Design summary

The above result represents the synthesis implementation by using the Xilinx ISE software. From the above table, it is observed that only 4982 look up tables are used out of available 204000. It indicates

very less area (2%) was used for the proposed design.

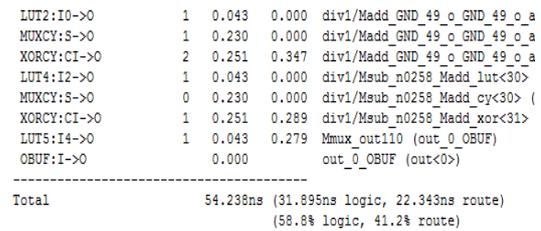


Figure 8: Time summary

Figure 8 represents the time consumed such as path delays by using the Xilinx ISE software. The consumed path delay is 54.238ns. Figure 9 represents the power consumed by using the Xilinx ISE software. The consumed power is 0.143uw.



Figure 9: Power summary

Table 1: Performance comparison

Parameter	Existing Method	Proposed Method
Time delay	59.110 ns	54.238 ns
Power	1.293uw	0.143 uw
LUTs	5277	4982

5. Conclusion

This paper proposes a high-performance and low-power approximate partial product accumulation tree for a multiplier using a newly designed approximate

adder. The proposed approximate adder ignores the carry propagation by generating both an approximate sum and an error signal. OR gate and approximate adder based error reduction schemes are utilized, yielding two different approximate 8×8 multiplier designs: AM1 and AM2. Moreover, modifications are made on the error reduction schemes for 16×16 multiplier designs, such that TAM1 and TAM2 are obtained by truncating 16 LSBs of the partial products. The proposed approximate multipliers have been shown to have a lower power dissipation than an exact Wallace multiplier optimized for speed. Functional analysis has shown that on a statistical basis, the proposed multipliers have very small error distances and thus, they achieve a high accuracy.

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