An improved common subexpression elimination method for reducing logic operators in FIR filter implementations without increasing logic depth

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ABSTRACT

It is well known that common subexpression elimination techniques minimize the two main cost metrics namely logic operators and logic depths in realizing finite impulse response (FIR) filters. Two classes of common subexpressions occur in the canonic signed digit representation of filter coefficients, called the horizontal and the vertical subexpressions. Previous works have not addressed the trade-offs in using these two types of subexpressions on the logic depth and the number of logic operators of coefficient multipliers. In this paper, we analyze the impact of the horizontal and the vertical common subexpression elimination techniques on reducing the logic depth and number of logic operators in FIR filters. Further, we present an algorithm to optimize the common subexpression elimination that produces FIR filters with fewer numbers of logic operators when compared with other common subexpression elimination algorithms in literature. The design examples show that the average reduction of logic operators achieved using our method over the weight-2 horizontal common subexpression elimination method which produced the best trade-off between logic operators and logic depth (contention resolution algorithm, CRA-2 [F. Xu, C.-H. Chang, C.-C. Jong, Contention resolution algorithm for common subexpression elimination in digital filter design, IEEE Trans. Circuit Syst. II 52(10) (2005) 695–700 (October)]) is 15%. This reduction of logic operators is achieved without any increase in the logic depth. When compared with the recently proposed multiple adder graph (MAG) algorithm [Jeong-Ho Han, In-Cheol Park, FIR filter synthesis considering multiple adder graphs for a coefficient, IEEE Trans. Comput.-Aid. Design Integr. Circuit Syst. 27(5) (2008) 958–962 (May)], the average reduction of logic operators obtained using our method is 5% and the reduction of logic depth is 25%.

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1. Introduction

FIR filters find extensive use in mobile communication systems for the functions such as channelization, channel equalization, matched filtering and pulse shaping due to its absolute stability and linear phase property. The filters employed in mobile systems must be realized with low complexity and minimum delay. Although programmable filters based on digital signal processor cores are available, they are not very efficient as they consume more power and operate at low speed. Hence dedicated FIR filter architectures have received great deal of attention in the last decade.

The key computation in FIR filters is coefficient multiplications, which is implemented using shifts and adds, out of which the addition operation dominates the complexity because shifts are less complex and hence they can be hardwired. The number of adders (logic operators) used to compute the sum of the partial product terms obtained when the inputs signal is multiplied by the coefficients and the critical path lengths (logic depths, which is equal to the number of adder-steps) of the multiplication operation are the two metrics that determine the complexity of FIR filters. Hence, the methods that minimize the complexity of multiplication in FIR filters focus on reducing the number of logic operators (LOs) and logic depth (LD) used to implement the multipliers. Multiple Constant Multiplications (MCM) is a transformation closely related to the widely used substitution of multiplications with constants by shifts and additions [1]. While the latter considers multiplication of only one constant at a time, the MCM considers multiplication of one variable with multiple constants. Common subexpression elimination (CSE) tackles the MCM problem by eliminating redundant computations in multiplier blocks (MBs) using the most common bit patterns called common subexpressions (CSs) that exist in the canonical signed digit (CSD) representation of coefficients [2–6]. In [2], an algorithm based on a coefficient subexpression graph for the identification and elimination of two-nonzero bit subexpressions (2-bit CSs) was proposed. A method to eliminate the most commonly occurring 2-bit CSs was proposed in [3]. As an
additional criterion in the subexpression identification process, an estimation of a latch count improvement was also used in [3]. A modification of the method in [2] for identifying and eliminating the best subexpressions to maximize the optimization impact is proposed in [4]. In [5], a nonrecursive signed CSE (NR-SCSE) algorithm has been proposed as a modification of the technique in [3] that minimizes the logic depth into the digital structure. The main idea in [6] is reordering computations and identifying common computations that maximize computation sharing between different multipliers. However, the method in [6] offers only a slight improvement in reduction of adders (11%) over the CSE method [3]. Moreover, this method results in an increase in delay, corresponding to the delay of one adder-step on average. Instead of exploring optimizations over the original filter coefficients, differential coefficients were considered in [7,25], where differences between absolute values of filter coefficients were employed to reduce the dynamic range of computation. However the DCM suffers from overheads since it requires extra adders to compute the sum of the stored partial product of previous computation in order to compensate the effect of differential coefficients. In [8], the idea of using differential coefficient was applied to the multiplierless implementation of digital filters. In this work, a graph-based approach was developed to explore the low-complexity solutions for DCM. However complexity reduction achieved in [8] is usually smaller than the amount of reduction achieved by CSE approaches. A computation reduction technique called computation sharing differential coefficient (CSDC) method, which combines the strength of an augmented differential coefficient approach and subexpression sharing has been proposed in [9]. The augmented differential coefficient approach expands the design space by employing both differences and sums of filter coefficients through algorithmic equivalence. However the method in [9] has additional overheads since it requires extra adders to compensate the effect of differential coefficients if coefficient differences are used or extra subtractors if the sums of coefficients are used. A CSE algorithm that considers both the redundancy among the CSD coefficients and the LD in the MB was proposed in [10]. The reductions of LOs and LDs achieved using this method over the method in [4] is minimal. A contention resolution algorithm for weight-two horizontal subexpressions (CRA-2), based on an ingenious graph synthesis approach has been developed for the common subexpression elimination of the multiplication block of digital filter structures in [11]. CRA-2 saves 1–3% more logic operators than NR-SCSE [5]. In our recent work [12], we have proposed two techniques for optimizing the CSE methods. These techniques are based on the extension of conventional 2-bit CSs in [2–6] to form three-nonzero bit and four-nonzero bit super-subexpressions (SSSs) by exploiting identical shifts between a 2-bit bit CS and a third nonzero bit, or between two 2-bit CSs. These SSSs eliminate redundant computations of two-nonzero bit CSs and hence reduce the number of adders. However it must be noted that the formation of 3-bit and 4-bit SSSs is based on the occurrence of 2-bit CSs with identical shifts between them. Therefore, the main limitation of the method in [12] is its dependence on the statistical distribution of shifts between the 2-bit CSs in the CSD representations of FIR filter coefficients. It has been shown in [12] that the number of SSSs grows linearly with the wordlength and hence this technique is more advantageous only when the coefficient wordlength is relatively larger. Note that the routing complexity of the method in [12] is higher than that of the 2-bit CSE techniques in [2–6] as the former method has more number of subexpressions. However, using system-in-package (SiP) solutions which have higher integration capacity than conventional system-on-ship (SOC) solutions, the size and routing complexity can be significantly reduced [22]. The first two limitations of [12] still pause hardware reduction constraints.

The Bull-Horrocks algorithm (BHA) [13] used a graph representation of the MB for reducing the number of LOs. Two methods that further reduce the number of LOs have been presented in [14], called the Bull-Horrocks Modified (BHM) algorithm and the n-dimensional Reduced Adder Graph (RAGn) algorithm. As the partial sums generated in multiplication are added in a serial manner in [13,14,24], these algorithms produce multipliers with large LDs, which increases the delay of the multiplier substantially. Even though the graph representation-based MB implementation reduces the number of LOs compared to CSE methods in [2–6], the LDs of the resulting coefficient multipliers are considerably larger. A new GD algorithm was proposed in [23] to optimize for minimum LOs for the MCM problem. The method in [23] resulted in longer LD, which in turn would increase the delay of the filter. Moreover, [23] is restricted to a maximum of 200 taps and its applicability for filters longer than 200 taps is not known (as per the details available on spiral.net). A multiple adder graph (MAG) based filter synthesis method has been recently proposed in [26]. While the previous graph-dependence algorithms [13,14,23,24] considered only one coefficient at a time and did not take into account the effect on the rest of the coefficients when synthesizing the coefficient, the MAG algorithm minimizes the adder cost by considering the effect on the remaining coefficients. A method for designing multiplier blocks with low LD was proposed in [27].

In general, the CSE methods utilize two types of CSs—the horizontal CSs (HCSs) that exist within each coefficient and the vertical CSs (VCSs) that exist across the adjacent coefficients. These techniques are called the horizontal common subexpression elimination (HCSE) and the vertical common subexpression elimination (VCSE), respectively. It has been shown in [15] that the VCSE offers better reduction of adders than the HCSE in realizing FIR filters. In our work [16], we have shown that the HCSs and the VCSs can be combined to produce better reduction of adders than the method in [15]. A new CSE method for implementing FIR filters using HCs and VCs has been proposed in [17]. The authors claim that the method in [17] reduces the average area by 6.4% and 3.8% over the methods in [8,9], respectively. The LD reductions achieved using [17] over [15,16] are 17.6% and 3.2%, respectively. However, the methods [15–17] only consider the implementation of the symmetric first half coefficient set of the FIR filter. These methods assume that the symmetric second half coefficient sets can be implemented by sharing the output of their symmetric first half coefficients. We denote the coefficients h(0) to h(N/2−1) of an N-tap FIR filter as symmetric first half coefficients and h(N/2) to h(N−1) as symmetric second half coefficients. We noted that the use of VCSs imposes constraints in implementing the symmetric second half of the coefficients. Considerable numbers of additional LOs are needed to realize the symmetric second half as the coefficient symmetry cannot be completely exploited when VCSs are used in CSE. The LO requirements shown in [15–17] do not account this overhead and therefore the hardware reductions claimed using these methods are incorrect. To the best of our knowledge, the constraints in utilizing the symmetry of FIR filter coefficients while employing VCSs have not been addressed in literature. This is because all the CSE-based FIR filter implementation methods in literature discuss the implementation of symmetric first half coefficients only. These methods assumed that since FIR filter coefficients are symmetric, the symmetric second half part can be implemented from the first half coefficients without using any additional LOs. But this is not true when VCSs are used. Note that the constraints discussed in this paper are not applicable to antisymmetric filters.
In this paper, we analyze the impact of HCSE and VCSE in exploiting the symmetry of FIR filter coefficients. Further, we present an optimization algorithm to reduce the number of LOs and LD in FIR filters. We show that our algorithm produces the best reduction of LOs when compared to the best known CSE algorithms in literature without increasing the LD of the coefficient multiplier. The rest of the paper is organized as follows. In Section 2, we present a complexity analysis of the filters realized using conventional CSE methods. Our CSE optimization technique is presented in Section 3. In Section 4, several design examples and comparisons are provided. Section 5 provides our conclusions.

2. Complexity analysis of CSE methods

A 6-tap FIR filter designed using Parks–McClellan algorithm is used to analyze the CSE methods. The passband and stopband edges of the filter are 0.2π and 0.25π, respectively. The 16-bit CSD representations of the coefficients are shown in Fig. 1. The numbers in the first row in Fig. 1 represent the number of bitwise right shifts and n represents −1.

2.1. The HCSE algorithm

The HCSE uses the HCSs, [10 1], [10 ¯1], [10 01] and [10 01 ¯], and their negated versions present in the CSD representation of coefficients to eliminate redundant multiplications. Hartley [3] showed that the use of the two most commonly occurring HCSs, [10 1] and [10 1 ¯], would reduce the routing complexity of the filter circuit when compared with the HCSE using other HCSs such as [10 01] and [10 01 ¯]. Therefore, we use Hartley’s HCSs [10 1] and [10 1 ¯] in our illustration. If \( h_1 \) is the input signal and \( 2^{-j} \) represents shift right by \( j \), the HCSs, [10 1] and [10 1 ¯], shown inside the solid rectangles in Fig. 1 are given by \( x_2 \) and \( x_3 \) respectively:

\[
x_2 = x_1 + 2^{-2}x_1 \quad \text{and} \quad x_3 = x_1 - 2^{-2}x_1
\]

Fig. 2 shows the filter implementation using the HCSE method. The numerals adjacent to the data paths in Fig. 2 represent the number of bitwise right shifts. There are two types of adders in the filter structure—structural adders (SAs) that compute the sum of convolved signals (shown between each delay stage in Fig. 2), and MB adders (MBAs) which compute the sum of partial products formed in coefficient multiplication. For a given filter length, the number of SAs is fixed (equal to the number of distinct delay stages). The focus of CSE is to reduce the number of MBAs since they dominate the hardware cost. If \( N_b \) represents the number of nonzero bits in the symmetric half coefficient set of an FIR filter of length \( N \), the total number of MBAs, \( T_{mba} \), needed to realize the filter using direct method (direct method is the implementation using shifts and adds and without using CSE techniques) is

\[
T_{mba} = N_b - \lfloor N/2 \rfloor
\]

In the CSD coefficients in Fig. 1, \( N_b \) is 18 and \( N \) is 6. Thus 15 MBAs are required to realize the filter using direct method. In the HCSE method, since all the nonzero bits forming an HCS exist within the coefficient, its symmetric counterpart can be easily implemented using delays and SAs, i.e., no additional MBAs are required for the symmetric part. Note that the coefficients \( h(3) \)–\( h(5) \) are symmetric with respect to \( h(0) \)–\( h(2) \) and hence their outputs can be shared as shown in Fig. 2 using the symbol ‘@’. Thus, only 11 MBAs (\( A_1 \)–\( A_{11} \)) are needed for the HCSE implementation in Fig. 2, which is a reduction of 26% over the direct method. The LDs of the filter circuit are identical (3 adder-steps) in direct method and CSE.

2.2. The VCSE algorithm

The VCSE methods [15–17] utilize the VCSs that occur across the adjacent coefficients to tackle the MCM. The VCSs, [11] and [11 ¯], that exist across the coefficients, shown inside the dotted rectangles in Fig. 1 by \( x_4 \) and \( x_5 \) respectively:

\[
x_4 = x_1 + x_1[-1] \quad \text{and} \quad x_5 = x_1 - x_1[-1]
\]
where \(x_t[-k]\) represents \(x_t\) delayed by \(k\) units. With these VCSs, the filter output using VCSE is

\[
\begin{align*}
-2^{-6}x_4 + 2^{-6}x_1 - 2^{-8}x_5 &+ 2^{-10}x_4 + 2^{-12}x_4 + 2^{-14}x_5 - 2^{-16}x_4 \\
-2^{-4}x_t[-1] + 2^{-2}x_t[-2] &- 2^{-5}x_t[-2] + 2^{-9}x_t[-2] - 2^{-15}x_t[-2] \\
+2^{-2}x_t[-4] &- 2^{-4}x_t[-4] + 2^{-8}x_t[-4] + 2^{-10}x_t[-4] + 2^{-12}x_t[-4] \\
-2^{-4}x_t[-4] &+ 2^{-8}x_t[-4] + 2^{-10}x_t[-4] + 2^{-12}x_t[-4] \\
-2^{-14}x_t[-4] &- 2^{-16}x_t[-4] + 2^{-8}x_t[-5] \\
\end{align*}
\]

(4)

Fig. 3 shows the VCSE realization of the filter. Since the bits form VCSs occur across the coefficients, the symmetry of VCSs cannot be utilized when the bits are of opposite signs. Hence in VCSE, additional MBAs are required to obtain the symmetric part of the coefficients when more than one VCSs with bits of opposite signs exist.

Consider the VCSs across the coefficients \(h(0)\) and \(h(1)\) in Fig. 1:

\[
\begin{align*}
2^{-2}x_4 + 2^{-6}x_1 - 2^{-8}x_5 &+ 2^{-10}x_4 + 2^{-12}x_4 + 2^{-14}x_5 \\
-2^{-16}x_4 &- 2^{-4}x_t[-1] \\
\end{align*}
\]  

(5)

Its symmetric VCS part across the coefficients \(h(4)\) and \(h(5)\) is

\[
\begin{align*}
2^{-2}x_4[-4] &- 2^{-4}x_t[-4] + 2^{-8}x_5[-4] + 2^{-10}x_4[-4] + 2^{-12}x_4[-4] \\
-2^{-14}x_5[-4] &- 2^{-16}x_4[-4] + 2^{-6}x_t[-5] \\
\end{align*}
\]  

(6)

Note that (6) cannot be directly obtained from (5) by simple delay operation since the signs and delays of certain terms of (6) are different from that of (5). Therefore, (6) needs to be obtained from (5) using (7) and (8) as given below:

\[
\begin{align*}
2^{-2}x_4 + 2^{-10}x_4 + 2^{-12}x_4 &- 2^{-16}x_4[-4] + 2^{-10}x_4[-4] \\
+2^{-12}x_4[-4] &- 2^{-16}x_4[-4] \\
-2^{-8}x_5 &+ 2^{-14}x_5[-4] \rightarrow 2^{-8}x_5[-4] - 2^{-14}x_5[-4] \\
\end{align*}
\]  

(7)

(8)

where \('[4]\) represents 4 units delay and \(\rightarrow\) represents negation. The adders, \(A_3\), \(A_4\) and \(A_5\) compute (7) and \(A_6\) computes (8) as shown in Fig. 3. The outputs of \(A_5\) and \(A_6\) corresponding to the left-hand side of (7) and (8) are utilized by \(A_{12}\) and \(A_{13}\), respectively, to obtain the right-hand side of these expressions and hence extra adders are not required in this case. However, the term \(2^{-8}x_1\) in (6) and \(2^{-8}x_1[-4]\) in (6) require two additional MBAs, \(A_7\) and \(A_{12}\). (But, the term \(2^{-4}x_t[-1]\) in (5) does not require an MBA since no other terms that has an identical delay and same is the case with \(2^{-8}x_1[-5]\) in (6). Thus these terms can be realized using SAs, \(A_2\) and \(A_4\), respectively.) Due to this constraint in exploiting the symmetry, the VCSE implementation requires more MBAs (13 MBAs in this case) than the HCSE despite the fact that the number of VCSs (16 VCSs as in Fig. 1) is more than the number of HCSs (12 HCSs as in Fig. 1). Furthermore, the LD in VCSE implementation (5 adder-steps) is larger than the HCSE (3 adder-steps). Hence the VCSE method results in increased LOS and LDs when compared with HCSE. It must be noted that the CSE methods in [15–17] which employ VCSs do not account the overheads in LOS and LDs in implementing the symmetric second half coefficients. Therefore, the reductions claimed by these methods are incorrect.

We have examined the reduction of LOS (MBAs) for FIR filters of different lengths (N), 8 bits to 24 bits wordlengths and frequency response specifications (passband and the stopband frequencies, \(\omega_p\) and \(\omega_s\), respectively). We noted that VCSE offered better reduction of LOS than the HCSE only when the coefficient wordlength is 8 bits. For wordlengths larger than 8 bits, the HCSE produced filters with fewer numbers of LOS than the VCSE. The LDs of the filters realized using VCSE are larger when compared with HCSE in most of the cases. In most practical filter applications, the frequency response of the filter will deteriorate considerably if the coefficients are coded using 8 bits. Therefore, the VCSE offers no advantage over the HCSE in practical FIR filter implementations if the proper VCSs are not chosen by carefully examining their signs. In next section, we present an optimization algorithm that efficiently combines HCSE and VCSE to minimize the number of LOS without increasing the LDs in FIR filters.

### 3. Proposed CSE optimization method

The core of our algorithm is to extract the maximum number of most frequently occurring common subexpressions. The HCSs, \([101], [101], [1001], [10001]\) and their negated versions, are used in our method since they are the most commonly occurring subexpressions. Among all the possible VCSs, we only use \([11], [101]\) and their negated versions, since the signs of nonzero bits in these VCSs are identical (we designate these two VCSs as ‘compatible VCSs’). Therefore, the use of these compatible VCSs facilitates better utilization of coefficient symmetry. Note that other HCSs such as \([10001]\) and \([100001]\) and VCSs such as \([1001]\) and \([1001]\) also exist in the CSD representation of coefficients. However, their frequency of occurrence is relatively smaller when compared to the HCSs and VCSs we have chosen. It has been shown in [3] that the use of large number of CSs with low frequency would have adverse effect on the routing complexity of the filter circuit.

Our algorithm first scans the coefficients to determine the frequency of HCSs and VCSs. For any coefficient, the CSs (HCSs or VCSs) with highest frequency are selected with priority given to HCSs first. If two or more HCSs occur common to different coefficients and if they are having identical shifts between them, then they are known as identical-shift HCSs (IS-HCSs). Each coefficient is compared with all the other coefficients for IS-HCSs. If more than one common IS-HCSs occur between a coefficient pair, the IS-HCSs can be grouped together to further eliminate redundant computations. Our optimization procedure is explained below.
3.1. CSE optimization procedure

The steps of our CSE optimization are as follows.

Step 1: Let \( C_{ij} \) represent the correlation index (CI) of the coefficient pair \( h(i) \) and \( h(j) \), and \( L \) is the number of filter taps.

**Definition ([Correlation Index CI]).** The correlation index of a coefficient pair is defined as the number of IS-HCSs obtained after HCSE algorithm. Thus, the CI of a coefficient pair is given by the coefficient pair is defined as the number of IS-HCSs obtained after HCSE algorithm.

\[
C_{ij} = C(h(i), h(j)), \quad 0 \leq i < n, 0 \leq j < n
\]

Determine the CIs of all the coefficient pairs and form the correlation matrix, \( C[h_{ij}] \) given by (9):

\[
C[h_{ij}] = \begin{bmatrix}
C_{01} & C_{02} & C_{03} & \ldots & C_{0L} \\
C_{12} & C_{13} & \ldots & C_{1L} \\
C_{23} & C_{24} & \ldots & C_{2L} \\
\vdots & \vdots & \ddots & \vdots \\
C_{L-1L} & \end{bmatrix}
\]

Step 2: The correlation matrix \( C[h_{ij}] \) is scanned row wise and the coefficient pair corresponding to the largest CI is grouped together to extract the IS-HCSs of each row. It may be noted that while selecting the best coefficient pairs, matching at one level must take into account how a particular match influences matching at the next level. This is done as follows. Set \( i_0 = 1 \) initially.

(i) Compute the largest CI, \( C_{\text{max}} \) of the \( i_0 \)th row from \( C[j_{0}] = \max_{j \in \{1,2,\ldots,L\}} C[i_{0},j] \), where \( j_{0} \) corresponds to the column in which the largest CI lies.

(ii) Check all the CIs in the \( j_{0} \) column to find whether any other CI greater than \( C_{\text{max}} \) exists. If no such CI exists, choose \( C_{\text{max}} \) as the largest CI of the \( i_0 \)th row and group the corresponding pair \( [h(i_0), h(j_{0})] \). Otherwise, choose the second largest CI of the \( i_0 \)th row as the largest CI and obtain the IS-HCS from respective coefficient pair.

Step 3: Let the largest CI obtained in previous step be \( C_{\text{max,h}} \). Replace all the elements of corresponding rows and columns by zero to exclude the coefficient pair chosen above from further search.

Step 4: If \( i_0 \leq L \), set \( i_0 = i_0 + 1 \) and go to step 2. Thus all the IS-HCSs are determined and redundant computations are eliminated.

Step 5: Eliminate the compatible VCSs [11] and [101].

3.2. Illustrative example

Our method can be illustrated using the example in Fig. 4, in which the CSD form of the filter coefficients are shown. The HCSs [101], [101] and [1001] and the VCSs [11] and [111], are indicated inside rectangles in Fig. 4.

Substituting the HCSs in Fig. 4, \( x_2 = [101] = 2, x_3 = [10 n] = 3 \) and \( x_4 = [100 n] = 4 \), and the VCSs, \( x_5 = [11] = 5 \), and \(-x_5 = [n n] = 3, \) we get Table 1.

The HCSs of 2 and 3 with a shift of 3 between them in \( h_0 \) and \( h_1 \) in Table 1 form the IS-HCS, \( x_6 = [2 0 0 0 0 3] \) as shown in Table 2.

From Table 2, the expression for filter output \( y_k \) is

\[
y_k = 2^{-1} x_6 + 2^{-10} x_3 + 2^{-14} x_5 - 2^{-1} x_6[1] + 2^{-3} x_5[1] + 2^{-3} x_4[2]
\]

(10)

The realization of (10) using our optimization is shown in Fig. 5.

The LD is 4 adder-steps and a total of 8 LOs are required for implementing the MB. For the direct implementation of the MB using the representation in Fig. 4, 13 LOs are required. Thus our CSE optimization offers 38% reduction of LOs compared to the direct implementation. The LD of the MB realized using our optimization method is one adder-step more than the direct implementation. In next section, we show that the LDs of filters realized using proposed method are comparable with the existing minimum LD CSE method.

4. Design examples

In this section, we present design examples of several FIR filters using proposed CSE optimization method. We also provide comparisons of the number of LOs and LDs needed to realize the filters using our method and the CSE methods in [3–6,9,11,14]. We use FIR filters designed using Parks-McClellan algorithm for different frequency response specifications (passband and stopband edges), filter lengths and coefficient wordlengths.

**Example 1:** In this example, we have compared the number of LOs and LDs generated by our algorithm with other algorithms for five benchmark filters FIR1 to FIR5. FIR1 has a passband frequency of 0.15\( \pi \) and stopband frequency of 0.25\( \pi \). For FIR2, the passband and stopband frequencies are 0.021\( \pi \) and 0.071\( \pi \), respectively. FIR3 is the high pass filter L1 from [19]. FIR3 has a stopband frequency of 0.37\( \pi \) and passband frequency of 0.55\( \pi \). FIR4 is a linear phase FIR filter employed in the filter bank channelizer of Digital Advanced Mobile Phone Systems (D-AMPS) receiver with passband and stopband frequencies of 0.6173\( \pi \) and 0.6276\( \pi \), respectively. FIR5 is the filter employed in the receivers for the Personal Digital Cellular (PDC) receiver. The passband and stopband frequencies of FIR5 are 0.6836\( \pi \) and 0.6973\( \pi \), respectively. The LOs and LDs obtained using these specifications for our method is compared with the BHM [14], NR-SCSE [5], Pasko [4] and Hartley [3]. Tables 3(A) and (B) show the comparison of the number of LOs and LDs. In Tables 3(A) and (B), \( N \) represents the filter length and \( W \) represents the coefficient wordlength. From Tables 3(A) and (B),
it is clear that our method produces the best reduction of LOs when compared to all other methods. The LDs achieved using our method is comparable with the CSE method in [5], which has the shortest LDs compared to other methods.

The graph-dependence-based BHM algorithm [14] produces the largest LDs since the partial sums generated in multiplication are added in a serial manner. Among the previous CSE methods, Hartley [3], Pasko [4] and NR-SCSE [5], the latter method [5] offers the best reduction both in terms of LOs and LDs. For the five benchmark filters FIR1 to FIR5, our method offers an average LO reduction of 10.2% over the second best method, i.e., the NR-SCSE [5]. The average LO reductions achieved using our method over the Hartley [3], Pasko [4], BHM [14], C1 algorithm [27] and Multiple Adder Graph Method (MAG) [26] are 22.4%, 16.1%, 12.9%, 6.7% and 4.4%, respectively. The LDs of proposed filters are shorter than other methods in most cases and in a few cases, they are comparable.

Example 2: In this example, our CSE optimization method is compared with CSE methods in [5,6,9,11,23] for the FIR filters with the passband and stopband frequencies of 0.2π and 0.22π, respectively. We have compared the LOs and LDs for different filter lengths of 20, 50, 80, 120, 200 and 400. The coefficient wordlengths considered are 12, 16, 20 and 24 bits. Tables 4(A)–(C) show the comparison of the LOs and Tables 5(A)–(C) show the LDs needed to implement the filters. Note that the results of [6] are not shown in the tables due to space constraints. However, the comparison with [6] is included in the figures. The results of [23] for filter length larger than 200 taps are indicated as 'NA' as [23] is restricted to a maximum of 200 taps (as per the details available on spiral.net).

![Fig. 5. FIR filter implementation using our CSE optimization.](image_url)
Table 4
Comparison of the number of LOs needed for realizing the FIR filter in Example 2.

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C. Filter length (N) | Multiple adder graph method [26] | C1 [27]
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Table 5
Comparison of the number of LDs needed for realizing the FIR filter in Example 2.

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C. Filter length (N) | Multiple adder graph method [26] | C1 [27]
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Fig. 6 shows the reductions of LOs achieved using our CSE method over other CSE methods when the filter length is 80 for wordlengths of 12, 16, 20 and 24 bits.

Our method offers an average LO reduction of 11.3% over [6], 15.1% over CRA-2 [11], 15.6% over NR-SCSE [5] and 19.1% over CSDC [9]. The method in [23] produces around 35% reductions of LOs than our method, but the LDs of [23] are larger than our method by 50–70%. The LO reductions achieved using our CSE method over other CSE methods when the filter length is 200 for wordlengths of 12, 16, 20 and 24 bits are shown in Fig. 7. Our method offers an average LO reduction of 13.1% over [6], 15.4% over CRA-2 [11], 19.5% over NR-SCSE [5] and 23.3% over CSDC [9].

For the filters in Example 2 (filter lengths of 20, 50, 80, 120, 200 and 400), the average reductions of LOs achieved using our method over [6] is 9.7%, CRA-2 [11] is 12%, NR-SCSE [5] is 13%, CSDC [9] is 18.5%, MAG method [26] is 5.6% and C1 [27] is 10.3%. From Tables 5(A)–(C), the LDs of our method are same as that of NR-SCSE [5] and CRA-2 [11]. The LDs of filters realized using [6] is one adder-step more than our method. Our method also offers similar LD reduction over MAG [26] and C1 [27]. The proposed method achieves 12% reduction of LOs compared to the best known minimum LOs method (CRA-2 [11]) for the same LD. When compared to [23], our method needs an average of 25% additional LOs, but our methods reduces the LDs by 50%. Moreover, [23] is
restricted to a maximum of 200 taps (as per the details available on spiral.net) whereas our method has no such filter length restrictions.

Fig. 8 shows the LO vs. LD characteristics of the 6 FIR filters in Example 2. It can be noted that the proposed method offers the best trade-off between LO and LD. The C1 algorithm [27] provides the second best trade-off up to LO = 160 (LD = 4), but its LD increases when LO increases further. The MAG method [26] needs only slightly more number of LOs than proposed method, but its LD values are very high.

**Example 3:** In this example, we consider FIR filters employed as channel filters in the channelizer of a wireless communication receiver. The channel filters of a receiver need to extract multiple narrowband signals (communication channels) from a wideband input signal. These filters must have a large number of taps due to the stringent adjacent channel attenuation specifications of wireless communications standards. We present examples of implementing channel filters using our method and provide comparisons with CSE techniques [5,6,9,11]. The channel filters employed in the filter bank channelizer of digital advanced mobile phone systems (D-AMPS) in [20] are considered. The sampling rate chosen is 34.02 MHz as in [20]. The channel filters extract 30 kHz D-AMPS channels from the input signal after down-sampling by a factor of 350. The passband and stopband edges are 30 and 30.5 kHz, respectively. The peak passband ripple is chosen as 0.1 dB. The filter stopband specifications are chosen as in the D-AMPS standard [20]. The length of the FIR filter \( N \) is determined using (11) [21]:

\[
N = -\frac{10 \log_{10} \beta_1 \beta_2 - 13}{14.6 f} + 1
\]

![LO vs. LD characteristic of the 6 FIR filters in Example 2.](image)

**Table 6**
Comparison of the number of LOs needed for realizing the channel filters in Example 3.

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where $q_1$ and $q_2$ are the passband and stopband ripples, respectively, and $\Delta f$ is the normalized width of the transition band. The comparison of LOs needed to implement the filters is shown in Tables 6(A) and (B). Filters of lengths 200, 460, 610, 940 and 1180 are chosen corresponding to peak stopband ripple (PSR) specifications of $-24$, $-48$, $-65$, $-85$ and $-96$ dB, respectively.

Fig. 9 shows the reductions of LOs achieved using our CSE method over other CSE methods when the filter length is 460 for wordlengths of 16, 20 and 24 bits. Our method offers an average LO reduction of 16.6% over [6], 17.6% over CRA-2 [11], 20.4% over NR-SCSE [5] and 22.9% over CSDC [9]. The LO reductions achieved using our CSE method over other CSE methods when the filter length is 940 for wordlengths of 16, 20 and 24 bits are shown in Fig. 10. Our method offers an average LO reduction of 17.6% over [6], 17.9% over CRA-2 [11], 20.2% over NR-SCSE [5] and 26.9% over CSDC [9]. As in the case of previous example, the LDs of filters realized using [6] is one adder-step more than our method. The LDs of filters realized using our method are same as that of [5,11]. We also compared with [26,27] and found that our method offers average LO reductions of 10% and 12.8% over [26,27], respectively. The LD reductions obtained using our algorithm over [26,27] were 25% and 15%, respectively. The detailed results are omitted here for brevity.

![Fig. 9. Percentage reduction of LOs achieved using our method over the methods in [5,6,9,11] for the 460-tap filter in Example 3.](image)

![Fig. 10. Percentage reduction of LOs achieved using our method over the methods in [5,6,9,11] for the 940-tap filter in Example 3.](image)
Our design examples show that in terms of number of LOs, proposed CSE method offers an average reduction of 13.7% over the best known minimum LO method [6]. It must be noted that the method in [6] requires one adder-step more than proposed method. The proposed method offers average LO reduction of 15% over the best known minimum LD method, CRA-2 [11]. The LDs of proposed method are similar to that of [11]. Thus our method offers the best tradeoff in terms of number of LOs and LDs when compared to other CSE methods in literature.

Example 4: In this example, our method is compared with the recently proposed MAG filter synthesis method [26]. We used the same filter specifications as that of filters 1–6 in [26] for comparison. The lengths of filters 1–6 are 60, 100, 101, 101, 60 and 60, respectively. All the filters have coefficient wordlength of 16 bits. The passband and stopband edge frequencies (Fs1 and Fp1) of filters 1 and 2 (both lowpass filters) are {0.1π, 0.14π} and {0.2π, 0.67π}, respectively. The passband and stopband edge frequencies (Fs1 and Fp1) of filters 3 and 4 (both highpass filters) are {0.3π, 0.42π} and {0.3π, 0.76π}, respectively. For the bandpass filter 5, edge frequencies are Fs1 = 0.2π, Fp1 = 0.3π, Fp2 = 0.8π and Fs2 = 0.9π. The bandpass filter 6 has edge frequencies are Fs1 = 0.2π, Fp1 = 0.45π, Fp2 = 0.65π and Fs2 = 0.9π. Table 7 shows the LOs and LDs for filters 1–6 realized using our method and that using [26]. The LO (adder cost) and LD (adder-step) values of filters realized using the multiple adder graph method are directly taken from Han and Park [26]. Our method offers average LO reduction of 1.8% and average LD reduction of 32.3% over [26].

Example 5: We present a comparison of our method with the C1 algorithm in [27]. The FIR filter specification chosen in this example is exactly same as that of the design example in [27]. Filter order is 24 and normalized passband and stopband edge frequencies are 0.25 and 0.3, respectively. Flooded 12-bit quantized coefficients are taken as in [27]. Table 8 shows the comparison of LOs and LDs for Example 5 for RAGn and BHM, applied once and twice, C1 [27] and proposed method. LO and LD values are directly taken from Table 1 in [27]. It can be noted that proposed method results in least LO and LD.

5. Conclusions

We have compared the reduction of logic operators (adders) and logic depths (critical path lengths) achieved using the horizontal and the vertical common subexpressions in realizing FIR filters. It has been noted that the common subexpression elimination technique employing horizontal common subexpressions offer better reductions in the number of logic operators as well as logic depths than their vertical common subexpressions counterpart in FIR filter implementations. Further, we have presented a method to optimize the horizontal and vertical common subexpression elimination techniques. Our method produced FIR filters with fewer numbers of logic operators and shorter logic depths when compared with other common subexpression elimination algorithms in literature. Our CSE optimization method offered an average reduction of 15% in terms of the number of logic operators over the best known common weight-two horizontal subexpression elimination method without any increase in logic depth. Our method reduces the number of structural adders in some cases at the cost of a slight increase in the number of delay elements. When compared with the recently proposed multiple adder graph (MAG) algorithm [26], the average reduction of logic operators obtained using our method is 5% and the reduction of logic depth is 25%.

References

[17] Y. Takahashi, M. Yokoyama, New cost-effective VLSI implementation of multiplexerless FIR filter using common subexpression elimination, in:
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