

Variation-Aware *TED* – Based Approach for Nano-CMOS RTL Leakage Optimization

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Abstract—As technology scales down to nanometer regime the process variations have profound effect on circuit characteristics. Meeting timing and power constraints under such process variations in *nano-CMOS* circuit design is increasingly difficult. This causes a shifting from worst-case based analysis and optimization to statistical or probability based analysis and optimization at every level of circuit abstraction. This paper presents a *TED* (Taylor Expansion Diagram) based *multi* – T_{ox} techniques during high-level synthesis (*HLS*). A variation-aware simultaneous scheduling and resource binding algorithm is proposed which maximizes the power yield under timing yield and performance constraint. For this purpose, a *multi* – T_{ox} library is characterized under process variation. The delay and power distribution of different functional units are exhaustively studied. The proposed variation-aware algorithm uses those components for generating low power *RTL* under a given timing yield and performance constraint. The experimental results show significant improvement as high as 95% on leakage power yield under given constraints.

I. INTRODUCTION

As *CMOS* technology continues to scale down to achieve higher performance and higher level of integration, power consumption and process variation pose new and difficult challenges for integrated circuit designers. The scaling down of technology has resulted in significant deviations from the nominal values of transistor parameters, such as channel length, threshold voltage, and gate-oxide thickness. For example, variation in gate length increases from 35% in a 130 *nm* technology to almost 60% in a 65 *nm* technology [1], resulting in the large variation in leakage and performance of the designed circuit.

A lot of work on low-power high-level synthesis (*HLS*) can be found in the literature. Most of these works [2], [3], [4] have considered dynamic power reduction without considering process variation. But, if the variations are not estimated properly and use existing worst-case analysis, leakage power may exceed the power limit of the design which degrades circuit performance. In [1], authors analyzed the *multi* – $V_{th}/V_{dd}/T_{ox}$ design space with consideration of process variation at the gate level. A recent work on parametric yield-driven *HLS* work can be found at [5]. Here, authors present impact of process variations on the *multi* – V_{dd}/V_{th} techniques at the behavioral level. But, power reduction based

on *multi* – V_{dd}/V_{th} requires extra power supply voltages and is not applicable in performance-critical circuit design. Thus, variation-aware low-power exploration for behavioral synthesis still needs to be investigated further.

The gate-oxide leakage current (I_{ox}) in *CMOS* can be described as:

$$I_{ox} \propto \left(\frac{V_{dd}}{T_{ox}}\right)^2 \exp\left(-\gamma\frac{T_{ox}}{V_{dd}}\right) \quad (1)$$

where γ is an experimentally derived factor. So, I_{ox} is proportional to the square of supply voltage and inversely proportional to the square of T_{ox} (gate-oxide thickness). Reducing supply voltage will increase the delay of the circuit and hence would affect the performance of the design. On the other hand, increase in the gate-oxide thickness leads to increase in propagation delay. So, multiple gate-oxide thickness can serve as a leakage power and delay trade-off. In [6], authors have used *dual* – T_{ox} based *CMOS* technology to minimize the leakage current during behavioral synthesis. However, they did not consider delay variations of the functional unit and their *RTL* generation is not optimal.

The paper presents a variation-aware leakage power optimization work in behavioral synthesis using *multi* – T_{ox} assignment. In this work, we have used *TEDs* (Taylor Expansion Diagrams) representation for high-level design description [7], [8] to generate the optimal *RTL* at the end of synthesis process. This representation is useful for modeling and supporting equivalence verification of designs specified at the behavioral level. *TED* is a canonical, graph based representation, similar to *BDDs* (binary decision diagrams) [9] and *BMDs* (binary moment diagrams) [10]. In contrast to *BDDs* and *BMDs*, *TED* is based on a non-binary decomposition principle, modeled along the Taylors series expansion. *TED* is capable of capturing an entire class of structural solutions, rather than a single *DFG* (data flow graph). By using decomposition, *TED* can be converted into a structural representation, *DFG*, optimized for a particular design objective. After obtaining *DFG*, we do statistical timing and power analysis to determine delay and power distribution through *DFG*. For this purpose, we explore the impact of process variation on delay and leakage power. A variation-aware resource library is constructed where all the library units

are characterized based on their delay and power distribution at different T_{ox} . A variation-aware simultaneous scheduling and resource binding algorithm is presented which takes time constraint as a performance (or delay) trade-off factor and offers user to maximum leakage power yield. The algorithm schedules nodes of DFG at the appropriate control steps and simultaneously binds them to the best available resources while considering resource constraint so as to achieve the desire performance with maximum leakage power yield. **The contributions of the paper can be summarized as,**

- To best of our knowledge so far, this is the first work to use TED techniques during behavioral synthesis in presence of both delay and leakage power variation.
- The HLS flow for variation-aware leakage power optimization in $multi - T_{ox}$ is proposed.
- Consideration of both resource and time constraints to provide user an optimal RTL by taking account of process variations.

II. POWER, LEAKAGE, DELAY, AND YIELD TRADE-OFFS AT RTL

In this section, we mainly discuss some preliminaries on variation-aware high-level synthesis (HLS), and present the motivation of our work.

A. Timing and Power Yield in HLS

HLS is a process of translating a behavior description into a register level structural description. Scheduling and resource binding are key steps during the synthesis process. The scheduler divides the set of arithmetic and logical operation in the DFG into groups so that the operations in the same group can be executed concurrently, while taking into consideration possible trade-offs between total execution cost and hardware cost. The binding process selects resources from the library, which involves trade-offs according to different features like delay, area, power, and leakage. The resource library contains different functional units with different characteristics such as delay, leakage, etc. Traditional HLS algorithms consider worst-case latency of each functional unit during scheduling and binding. However, as the magnitude of process variation grows rapidly, worse-case based analysis and optimization are no longer acceptable since they introduce too much pessimism in the design. This in turn creates problem for designers to meet the requirement. Instead, statistical description and analysis of functional units are introduced to tackle the timing problem [11], [12], [13], [14], [15], [16], [18].

In presence of process variation, the delay of the functional unit is no longer a fixed value, but spreads into wide distributions. In a statistical timing view, the distribution can be described by a probability density function (PDF). Timing yield is defined as the probability that a functional unit can finish execution in a given time period. Alternatively, it is the cumulative probability under a given T_{clk} in PDF . An concept of timing yield is shown in *Fig. 1*

Given the clock time T_{clk} , the overall timing yield of the entire DFG is the probability that the entire design can finish

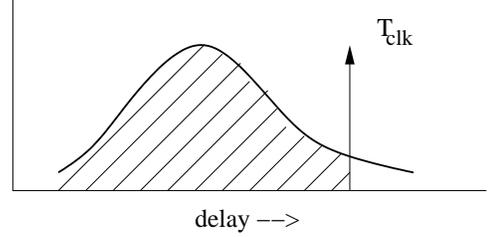


Fig. 1. Timing yield of an adder with clock period T_{clk}

execution within T_{clk} , and can be defined as

$$Yields = P(t_1 \leq T_{clk}, t_2 \leq T_{clk}, \dots, t_n \leq T_{clk}) \quad (2)$$

where $P()$ is the probability function, t_1, t_2, \dots, t_n are the execution time for the control steps 1, 2, \dots , n , respectively.

Dynamic power in $CMOS$ circuit is relatively immune to process variation and it affects the mean value of the total power consumption. Thus, in our work we have applied statistical analysis to the leakage power. The total leakage power consumption of a DFG can be calculated by adding leakage power of all functional units present in the DFG . Given a power limit P_L , the power yield of the DFG ($YieldP$) can be defined as the probability that of total power consumption of DFG ($PDFG$) is less than or equal to P_L , and can be defined as,

$$YieldP = P(PDFG \leq P_L) \quad (3)$$

In variation-aware HLS , a metric called parametric yield is introduced in [5]. The parametric yield is defined as the probability of the synthesized hardware meeting a specified constraint $Yield = P(Y \leq Y_{max})$, where Y can be delay and power.

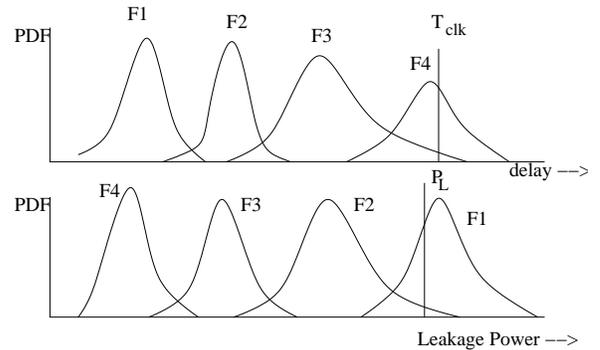


Fig. 2. Delay and power distributions of the functional unit

Fig. 2 shows an example to compare yield-driven approach and worse-case deterministic approach. Four functional units $F1, F2, F3$, and $F4$ have the same functional description. However, T_{ox} (gate-oxide thickness) of these functional units increases from $F1$ to $F4$. The leakage power and delay distribution of these units are shown in *Fig. 2*. The power limit P_L and clock cycle time T_{clk} are also shown in *Fig. 2*. The T_{ox} of $F1, F2, F3$, and $F4$ follows $T_{ox}(F1) < T_{ox}(F2) < T_{ox}(F3) < T_{ox}(F4)$. So, mean leakage power follows $\mu(F4) < \mu(F3) < \mu(F2) < \mu(F1)$ and delay follows up as $\mu(F1) < \mu(F2) < \mu(F3) < \mu(F4)$. In worse-case deterministic

TABLE I
LIBRARY WITH DIFFERENT GATE-OXIDE THICKNESS

Functional unit	$T_{ox} = 1.4nm$			$T_{ox} = 1.7nm$		
	I_{ox} (μA)	T_{pd} (ns)		I_{ox} (μA)	T_{pd} (ns)	
		yield 100%	yield 90%		yield 100%	yield 90%
Adder	2.155	11.68	10.94	0.2725	14.52	13.12
Subtractor	11.99	11.46	10.16	3.185	14.59	13.15
Multiplier	53.81	15.55	15.44	6.701	17.29	16.45
Comparator	3.30	0.2304	0.2266	0.123	0.2396	0.2307
Register	3.465	0.7934	0.7599	0.2025	0.7973	0.7534
Multiplexer	3.181	0.3763	0.3732	1.827	0.3997	0.383

approach $F4$ will be chosen under leakage power constraint as it has lowest leakage power consumption. But, from statistical point of view $F4$ has low timing yield and may cause timing violation. Similarly, from the performance constraint point of view $F1$ will be chosen as it satisfies timing constraint. But, $F1$ has larger leakage power and may lead to higher power dissipation. However, if we consider both power and performance constraint simultaneously, $F2$ and $F3$ can be chosen. Selection of $F3$ results in slightly loss in timing yield but satisfies power yield, while $F2$ results in slightly loss in power yield but satisfy timing yield. So, selection of $F2$ and $F3$ introduces a concept of tradeoff in between timing and power yield. Thus, a yield-driven statistical approach is needed which selects the functional units so that one parameter yield can be maximized under other parametric yield constraint.

B. Library setup for yield-driven multi – T_{ox} optimization

Leakage power is inversely proportional to the gate-oxide thickness (T_{ox}). The reduction in T_{ox} results increase in leakage power and decrease in delay. In the present work, we have created libraries with different T_{ox} components. For this purpose, we first characterized a library of 16 – bit components, such as adders, subtractors, multipliers, comparators, multiplexers, and registers following the structural description from [19]. We performed our library simulation using different gate-oxide thicknesses. The Berkeley Predictive Technology Model (*BPTM*) of 45nm technology node is used in this work, with base values of $T_{ox} = 1.4nm$, $V_{dd} = 0.7V$ and $V_{th} = 0.22V$. The nominal power supply is $V_{dd} = 0.7V$. The effect of varying oxide thickness was incorporated by varying the parameter t_{oxe} in the SPICE model deck directly. It may be noted that the length of the device is proportionately changed to maintain a constant (L/T_{ox}) ratio in order to minimize the impact of higher oxide thickness on device performance and to maintain the per width gate capacitance constant as per fabrication requirements [17]. The *PMOS* transistors are sized appropriately to ensure proper functionality of the building blocks. We have exhaustively evaluated the process variation effects through detailed 10000 Monte Carlo simulations to capture the effects. The primary goal of this analysis is to assess the extent of leakage I_{ox} and power variation as a result of process variations in gate oxide thickness T_{ox} . The distribution of these parameters is assumed to be Gaussian with the variance of 10%. *Table I* shows the statistical variation of the delay corresponding to oxide thickness 1.4nm and 1.7nm respectively.

Table I indicates the delay values of the functional units under different performance yields. In order to obtain these values, we first generate the delay distribution for each functional unit. The delay for certain timing yield can be calculated by finding the area under the curve. For example, *Fig. 3* shows the PDF for the adder of *Table I*. Under 100% yield, the delay of the adder is 11.68 ns ($T_{ox} = 1.4 nm$) which corresponds to point *A* in *Fig. 3*. But, if we scarifly 10% yield, the delay becomes 10.94 ns which corresponds to point *B*. Similarly, points *C* and *D* represent the delay value of 11.09 ns and

10.98 ns correspond to 97% and 94% of timing yield.

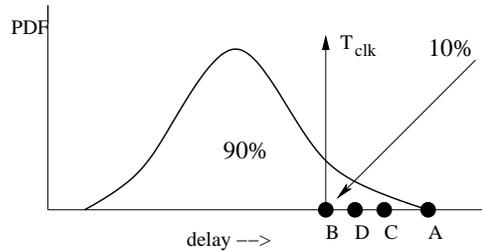


Fig. 3. The delay values of an adder under different timing yield

Similar to delay variation, we have generated the leakage power under different power yield for each functional unit which is not shown here due to space limitation. Once characterized, the next task is to create the *DFG* from the behavioral description of the given circuit. In order to create optimized *DFG*, in the present work we have used *TED* based approaches which is described in the next section.

III. CANONICAL *TED* FOR EFFICIENT HIGH-LEVEL REPRESENTATION

Taylor Expansion diagram [8] is a canonical, word-level data structure that offers an efficient way to represent computation in a compact, factored form. An Algebraic, multi-variable expression $f(x, y, ..)$, can be represented using Taylor series expansion, w.r.t. variable x as follows:

$$f(x, y, ..) = f(x = 0) + x f'(x = 0) + 1/2x^2 f''(x = 0) + .. \quad (4)$$

Where $f'(x)$, $f''(x)$, etc, are the successive derivatives of f w.r.t. x . The terms of the decomposition are then decomposed with respect to the remaining variables ($y, .., etc$), one variable at a time. A directed acyclic graph is used to store the resulting decomposition whose nodes represent the terms of the expansion. The detailed explanation of *TED* can be found in [7], [8].

A. *TED* – based RTL low-leakage optimization: A Finite Impulse Filter (*FIR*) Case Study

Since *FIR* (Finite-impulse response) filters are critical to most *DSP* application, an energy-aware filter design helps significantly in reducing the total power dissipation. The

polynomial corresponding to a 4 – tap FIR filter can be written as,

$$Y_n = a_0X_n + a_1X_{n-1} + a_2X_{n-2} + a_3X_{n-3} \quad (5)$$

TED corresponding to equation 5 is shown in Fig. 4. Given an optimized *TED*, the next task is to convert it to *DFG*, shown in Fig. 5. An *STA* on *DFG* is performed to generate the necessary timing information. Specifically, we need to calculate arrival time T_a , required time T_r , and slack $T_s = T_r - T_a$, for each node.

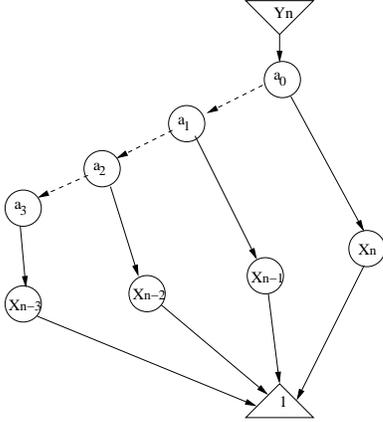


Fig. 4. *TED* for a 4 – tap FIR filter

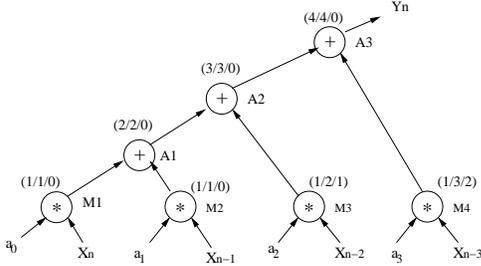


Fig. 5. *DFG* for the *TED* of Fig. 4

Definition 1: Arrival time T_a of a *DFG* node n is recursively defined as a sum of delay of node n and the maximum arrival time of its inputs:

$$T_a(n) = Delay(n) + \max(T_a(n_i)|_{n_i \in Input(n)}) \quad (6)$$

where $Delay(n)$ denotes the delay of the operation associated with node n , and $Input(n)$ is the set of input nodes to the node n .

Definition 2: Required time T_r of a node n is recursively defined as a difference between the minimum required time of its outputs and delay of node n :

$$T_r(n) = \min(T_r(n_o)|_{n_o \in output(n)}) - Delay(n) \quad (7)$$

Here $Output(n)$ is the set of output *DFG* nodes of node n

Definition 3: Slack time T_s of a *DFG* node n is defined as a difference between its required time T_r and the arrival time T_a .

$$T_s(n) = T_r(n) - T_a(n) \quad (8)$$

In Fig. 5, the arrival time T_a , the required time T_r , and the slack T_s of each node are denoted in the form of $[T_a/T_r/T_s]$. Here, we assume delay of each functional unit is 1 for simplicity. Based on the definition of slack, a critical node and critical path in *DFG* can be identified as follows,

Definition 4: A critical node in a *DFG* is a node which has a slack equal to 0. A critical path is a path which contains critical nodes only.

In Fig. 5, critical path 1 consists of 4 nodes ($M1, A1, A2, A3$) and critical path 2 consists of 4 nodes ($M2, A1, A2, A3$). In the next subsection, we present the generalized algorithm for variation-aware simultaneous scheduling-binding for general circuits.

B. An Algorithm for variation-aware Nano – CMOS RTL leakage optimization

In this section, we present a simultaneous scheduling and binding algorithm under resource constraint. The inputs to the algorithm are an unscheduled *DFG*, libraries with different resources made of transistors of different gate-oxide thickness, a delay trade-off factor T_d , and performance yield Y_d . The T_d is a user defined quantity which specifies the maximum allowed critical path delay of the target circuit. The present algorithm schedules the *DFG* in such a way that critical path delay is either less than or equal to T_d while improves the power yield under performance yield constraint Y_d .

The proposed algorithm performs an initial *STA* on the *DFG* to identify critical and non-critical nodes by calculating T_a , T_r , and T_s of each node. During this step, it uses delay value of 1 for each node. Once identified, it assigns T_{oxL} (low gate-oxide component) to critical nodes and T_{oxH} (high gate-oxide components) to non-critical nodes. After initial scheduling and binding, it calculates critical path delay and power yield. Now the algorithm searches iteratively on the *DFG* to reduce the leakage power or improves the power yield under Y_d and T_d constraints. Or in other words, it replaces nodes of T_{oxL} with T_{oxH} so that power yield can be maximized by satisfying Y_d and T_d . The pseudo code of the algorithm is presented in Algorithm 1.

Consider the *FIR* filter of Fig. 4 under the assumption that unlimited number of T_{oxL} and T_{oxH} components. The T_d and Y_d are assumed to 10 ns and 90% respectively. The delay of the library units under 100% and 90% timing yield is shown in Table II for both T_{oxL} and T_{oxH} components.

TABLE II
DELAY VALUES UNDER DIFFERENT TIMING YIELD

Functional Unit	delay (ns) (for T_{oxL})		delay (ns) (for T_{oxH})	
	yield 100%	yield 90%	yield 100%	yield 90%
Adder	2	1	3	2
Multiplier	4	3	5	4

Fig. 6 shows the *DFG* of equation 6 after initial scheduling and binding where critical nodes are bound to T_{oxL} and non-critical nodes to T_{oxH} . During this phase, the algorithm uses delay values of the functional units corresponding to 100% timing yield (or worse-case analysis). Once scheduled, the

Algorithm 1 *NanoCMOS RTL Optimization for Yield, Power, and Time Tradeoff*

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1: Apply STA to DFG under resource constraint
2: Assume each node is assign to a delay of 1
3: Identified critical and non-critical nodes
4: for all critical nodes  $n_i$  do
5:   if  $FU_j(k, T_{oxL})$  is available for control step  $C[n_i]$  then
6:     Assign  $FU_j(k, T_{oxL})$  to node  $n_i$ 
7:   else
8:     Assign  $FU_j(k, T_{oxH})$  to node  $n_i$ 
9:   end if
10: end for
11: for all non-critical nodes  $n_i$  from root of the DFG do
12:   for all possible control steps (slack) of  $n_i$  do
13:     if  $FU_j(k, T_{oxH})$  is available for control step  $C[n_i]$  then
14:       schedule  $n_i$  in control step  $C[n_i]$ 
15:       Assign  $FU_j(k, T_{oxH})$  to node  $n_i$ 
16:       Update  $T_s$  for all the nodes connected to  $n_i$ 
17:     end if
18:   end for
19:   if  $n_i$  is not scheduled then
20:     for all possible control steps (slack) of  $n_i$  do
21:       if  $FU_j(k, T_{oxL})$  is available for control step  $C[n_i]$  then
22:         schedule  $n_i$  in control step  $C[n_i]$ 
23:         Assign  $FU_j(k, T_{oxL})$  to node  $n_i$ 
24:         Update  $T_s$  for all the nodes connected to  $n_i$ 
25:       end if
26:     end for
27:   end if
28: end for
29: Calculate  $T_a$ ,  $T_r$ , and  $T_s$  for all nodes
30: Calculate critical path delay  $T_{cp}$  and power yield  $Y_p$  of the DFG
31: Sort all critical nodes according to ascending order of leakage current
32: Calculate timing yield  $Y_t$  of the DFG
33: for all critical nodes  $n_i$  do
34:   if ( $T_d$  greater than  $T_{cp}$ ) and ( $Y_t$  greater than  $Y_d$ ) then
35:     if  $FU_j(k, T_{oxH})$  is available for control step  $C[n_i]$  then
36:       Assign  $FU_j(k, T_{oxH})$  to node  $n_i$ 
37:       Calculate  $Y_t$  and modified power yield  $Y_{pt}$  of DFG
38:       if ( $(Y_{pt} - Y_p)$  greater than 0) and ( $Y_t$  greater than  $Y_d$ ) then
39:         calculate  $T_{cp}$ 
40:         if  $T_{cp}$  less than or equal to  $T_d$  then
41:            $Y_p = Y_{pt}$ 
42:           continue to next critical node
43:         end if
44:       end if
45:     end if
46:     Assign  $FU_j(k, T_{oxL})$  to node  $n_i$ 
47:   end if
48: end for

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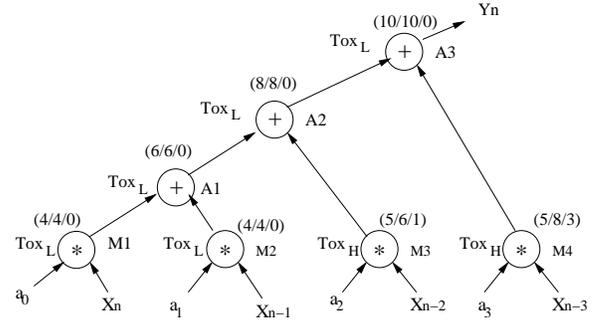


Fig. 6. The DFG of Fig. 4 after initial scheduling and binding

algorithm searches iteratively to bind T_{oxL} node with T_{oxH} components under Y_d and T_d constraints. It is clear from Fig. 6 that M1 can be replaced with T_{oxH} as under 90% yield delay of the multiplier is $4ns$ (see Table II). After replacement, timing yield of the DFG will be 90% which is equal to Y_d . Similarly, instead of M1 one can also replace A3 with T_{oxH} components. But, replacement of multiplier saves much more leakage power than an adder. The final scheduled DFG is shown in Fig. 7.

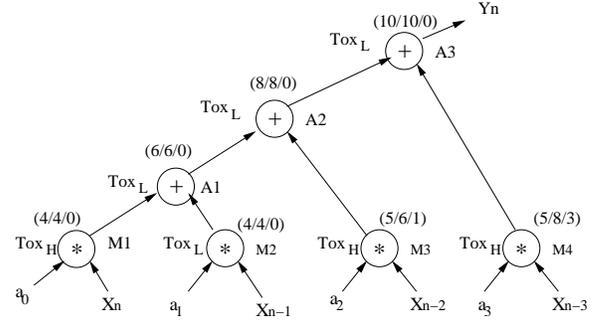


Fig. 7. Final DFG after variation-aware scheduling and binding

IV. EXPERIMENTAL RESULTS

In this section, we present the experimental results of our variation-aware leakage power yield improvement framework for HLS. We implement our variation-aware synthesis algorithm in C and perform the experiment on a set of HLS benchmarks [6]. The results show that our method can effectively improve the overall leakage power yield or minimize the leakage power dissipation under process variation.

Table III shows the comparison of variation-aware resource binding algorithm against traditional deterministic (worse-case) approach. Gate-oxide leakages current (I_{ox}) have been calculated under different performance yield (Y_d) and T_d constraints including worse-case delay based approach (i.e., when $Y_d = 100\%$). The results indicate significant reduction in I_{ox} when we scarify 10% of performance yield.

Fig. 8 shows the leakage power yield improvement against worst-case delay based approach under timing yield constraints 95% and 90% for different benchmark circuits when $T_d = 1.2ns$. Results indicate that the power yield improvement

TABLE III
EXPERIMENTAL RESULTS FOR THE PRESENT ALGORITHM

Circuits	resource cons	$T_d = 1.0 \text{ ns}$			$T_d = 1.2 \text{ ns}$		
		Y_d			Y_d		
		100%	95%	90%	100%	95%	90%
		I_{ox} (μA)					
ARF	1	886.82	865.13	854.72	853.47	847.19	839.46
	2	860.45	840.61	828.17	839.93	822.28	811.18
	3	835.13	810.41	798.10	810.68	796.81	780.11
	∞	790.61	767.53	752.76	774.15	749.17	730.48
BPF	1	667.56	652.51	645.64	654.15	643.91	635.70
	2	645.12	633.19	624.55	636.82	621.15	615.47
	3	631.65	621.97	607.17	624.17	607.76	599.12
	∞	606.63	594.17	586.28	595.31	582.49	578.97
FIR	1	462.85	451.54	445.63	449.72	438.57	431.98
	2	453.46	441.12	436.76	435.64	426.12	418.36
	3	446.72	434.65	423.31	427.15	415.91	408.01
	∞	414.58	401.40	390.87	398.56	391.32	383.83
EWF	1	486.51	478.45	470.91	471.23	462.64	456.23
	2	470.24	461.82	453.12	458.12	447.34	438.14
	3	462.87	452.19	440.52	443.91	432.89	421.63
	∞	441.31	430.52	418.57	413.24	401.44	388.17

depends on how much timing yield loss is affordable for the design.

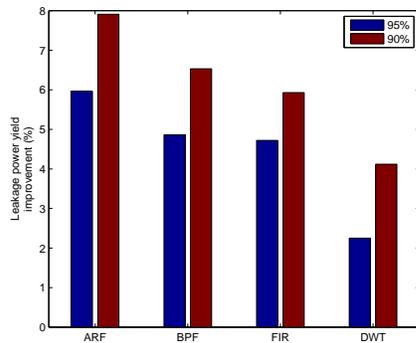


Fig. 8. Leakage power yield improvement against worst-case approach

Since there is no behavioural synthesis research dealing with gate-oxide leakage reduction for different performance yield, direct comparison is not possible. However, in view of power yield improvement we provide a broader comparative perceptive in Table IV with [5]. Δp indicates the power yield improvement against deterministic worst-case approach under different performance yield. From Table IV, it is clear that $multi - T_{ox}$ is an attractive approach for improving performance yield at the same time reducing gate leakage current for $nano - CMOS$ datapath circuits.

TABLE IV
A BROAD COMPARATIVE PERCEPTIVE WITH [5]

Circuits	$Y_d = 95\%$		$Y_d = 90\%$	
	[5]	This work	[5]	This work
	($multi - V_{th}/V_{dd}$)	($multi - T_{ox}$)	($multi - V_{th}/V_{dd}$)	($multi - T_{ox}$)
	Δp	Δp	Δp	Δp
EWF	3.9	4.2	4.0	4.4
FIR	4.8	4.6	5.4	5.8
BPF	4.5	4.7	6.6	6.4

V. CONCLUSIONS

In this paper we presented an effective way to the problem of variation-aware HLS. Here, we develop the timing and leakage power constraints based scheduling and resource binding algorithm for HLS. We have used TED based

approaches to generate the optimize DFG. The proposed algorithm maximizes the leakage power yield of the design circuit for a given performance constraint. The experimental results on several benchmark circuits show that performance yield can be maintained with increasing leakage power yield.

VI. ACKNOWLEDGMENTS

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