



Dynamic Power Estimation for Deep Submicron Circuits with Process Variation

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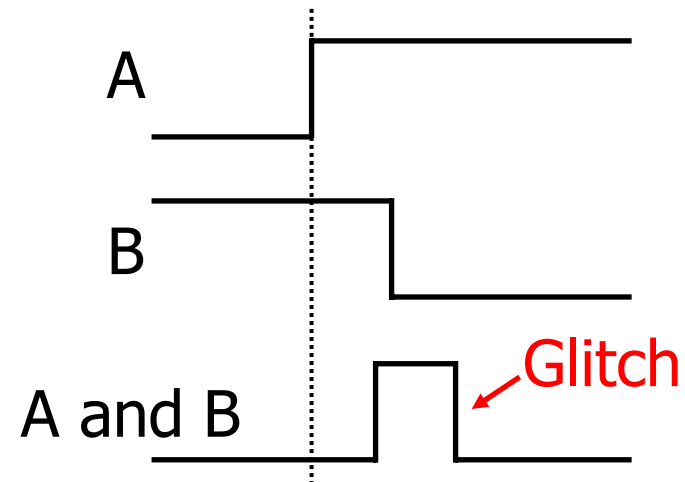
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Outline

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 - Power estimation for a switch segment
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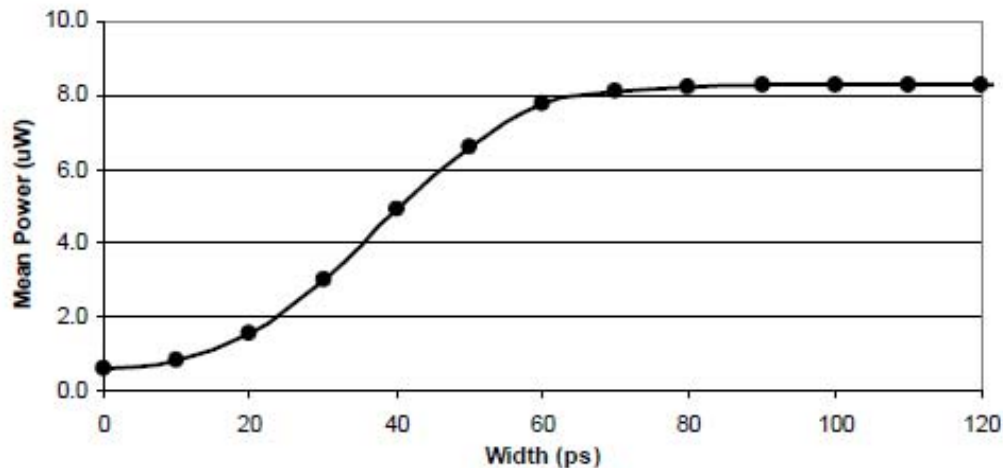
Glitches

- Signal transitions consists of:
 - Functional transitions
 - Spurious transitions (Glitches)

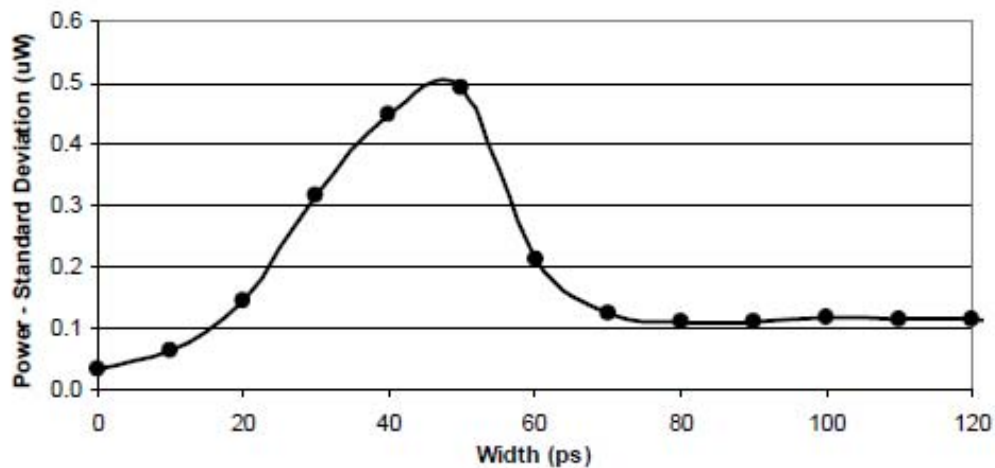


- Glitch transitions can be 4-5X of functional transitions in some data-flow intensive designs [Raghunathan, TCAD'99]
- They can be 20% of total power in FPGAs [Li, TCAD'05]

Dynamic power under process variation



Mean of the dynamic power of a logic gate vs. input glitch width



Standard deviation of the dynamic power of a logic gate vs. input glitch width

5X difference
(e.g., between width 50ps and 100ps)

Related work

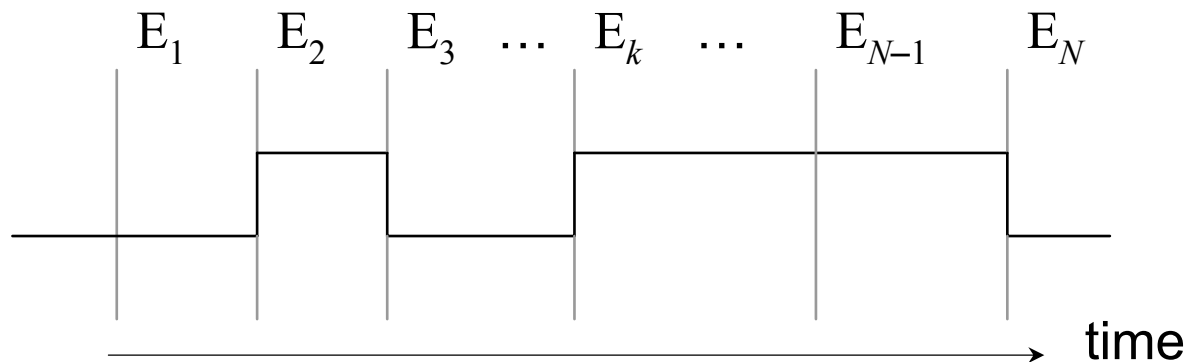
- Circuit power under process variation mainly focused on leakage power
 - [Mukhopadhyay, ISLPED'03], [Rao, TVLSI'04], [Chang, DAC'05], etc.
- Two previous works on dynamic power modeling with process variation
 - [Pilli & Sapatnekar, ISCAS'97]: Pioneer work; didn't estimate variation of dynamic power
 - [Alexander & Agrawal, ISVLSI'09]: only minimum and maximum bounds for delays are considered; no consideration of partial swing transitions

Definition: Transition Event

- A *transition event* at a particular node n in the circuit describes a possible transition at that node.
- No distinguishing between $0 \rightarrow 1$ and $1 \rightarrow 0$ transitions.
- Represented by three values (p, t, σ_t) as follows:
 - p is the probability for the transition to occur.
 - t is the mean of the time of the transition.
 - σ_t is the standard deviation of the transition time.
- For a transition event (0.2, 0.42 ns, 0.21 ns)
 - there is a likelihood of 20% that a transition occurs at a time around 0.42 ns, with a standard deviation of 0.21 ns.

Transition Waveforms

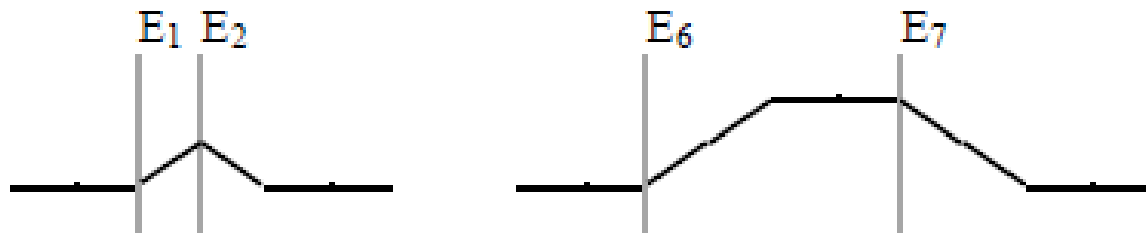
- The set of all transition events for a node are collectively represented by a *transition waveform*
 - $\{(p_1, t_1, \sigma_{t1}), (p_2, t_2, \sigma_{t2}), \dots, (p_N, t_N, \sigma_{tN})\}$.
 - N transition events: E_1, E_2, \dots, E_N
 - The events E_i of (p_i, t_i, σ_{ti}) are sorted according to the increasing order of t_i



In this example, events E_2, E_3 and E_N correspond to actual transitions, while events E_1 and E_{N-1} do not.

Switching Segments

- A *switching segment* is the duration from the beginning of a transition to just before the beginning of the next transition. A switching segment corresponds to exactly one transition.



Switching segment for partial-swing transition (E_1, E_2) and full-swing transition (E_6, E_7).

Algorithm: propagation through a wire segment or a gate

Source: $E_S (p_S, t_S, \sigma_{tS})$
 Target: $E_T (p_T, t_T, \sigma_{tT})$

Input: $E_I (p_I, t_I, \sigma_{tI})$
 Output: $E_O (p_O, t_O, \sigma_{tO})$

$$p_T = p_S$$

$$t_T = t_S + d$$

$$s_{tT} = \sqrt{s_{tS}^2 + s_d^2}$$

$$t_O = t_I + d$$

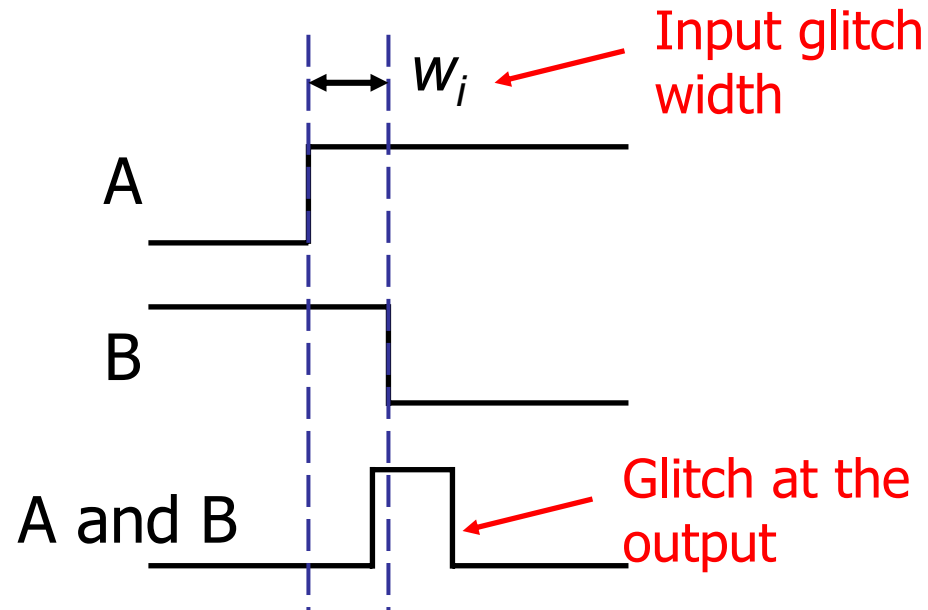
$$s_{tO} = \sqrt{s_{tI}^2 + s_d^2}$$

$$p_O = p_I \mp P \frac{\text{踐} f}{\text{顏} x_i}$$

Wire segment

Gate

Dynamic power for an input glitch width



- $DPower(w_i)$ is a random variable due to
 - variation in the load capacitance
 - variation in the gate delay from different inputs to the output

Gate power characterization

- Gate Characterization under
 - different input glitch widths (20 different widths, from 0ps to 190ps)
 - different load capacitances
- Build look-up tables
 - Look up and interpolate the mean $\mu_{\mathcal{C}}(w_i)$ and standard deviation $\sigma_{\mathcal{C}}(w_i)$ of dynamic power for a given w_i and a given load capacitance.

Compute probability for w_i

Discretize the width into w_0, w_1, \dots, w_L .

The probability that $w = w_i$ is:

$$\Pr\{w = w_i\} = \frac{1}{s_w \sqrt{2\pi}} \exp - \frac{(w_i - m_w)^2}{2s_w^2}$$

Normalize this probability as follows:

$$PN_i = \frac{\Pr\{w = w_i\}}{\sum_j \Pr\{w = w_j\}}$$

Compute the mean of the dynamic power for a switch segment

For each w_i , we can look-up and interpolate the corresponding mean power $\mu_C(w_i)$ and standard deviation $\sigma_C(w_i)$

$$m_C(w_i) = E \left[\overset{\text{鸽}}{\underset{\cdot}{D}}Power(w) \mid w = w_i \right]$$

$$s_C(w_i) = \sqrt{E \left[\overset{\text{鸽}}{\underset{\cdot}{D}}Power(w) - m_C(w_i) \right]^2 \mid w = w_i}$$

The overall (unconditional) mean dynamic power μ_{DP}

$$m_{DP} = E \left[\overset{\text{鸽}}{\underset{\cdot}{D}}Power(w) \right] = \sum_i P N_i \mp m_C(w_i)$$

Compute the standard deviation for a switch segment

The overall (unconditional) standard deviation σ_{DP} is

$$s_{DP} = \sqrt{E \left[\sum_i \left(DP_{Power}(w_i) - m_{DP} \right)^2 \right]}$$

$$E \left[\sum_i \left(DP_{Power}(w_i) - m_{DP} \right)^2 \right] = \sum_i \left(m_C(w_i) - m_{DP} \right)^2 + s_C(w_i)^2$$

$$s_{DP} = \sqrt{\sum_i PN_i \left[\sum_i \left(m_C(w_i) - m_{DP} \right)^2 + s_C(w_i)^2 \right]}$$

Compute dynamic power for a gate

For a gate with the transition waveform
 $\{(p_1, t_1, \sigma_{t1}), (p_2, t_2, \sigma_{t2}), \dots, (p_N, t_N, \sigma_{tN})\}$.

$$m_{DPgate} = \sum_{i=1}^{N-1} \sum_{j=i+1}^N Pr(i, j) \cdot m_{DP}(i, j)$$

$$s_{DPgate} = \sqrt{\sum_{i=1}^{N-1} \sum_{j=i+1}^N Pr(i, j) \cdot s_{DP}(i, j)^2}$$

$$Pr(i, j) = p_i \cdot p_j \cdot \prod_{k=i+1}^{j-1} (1 - p_k)$$

Dynamic power for the circuit

- Assume that the power for the gates are independent random variables of normal distribution, the total dynamic power is:

$$m_{DP\ circuit} = \sum_{\text{all gates}} m_{DP\ gate}$$

$$s_{DP\ circuit} = \sqrt{\sum_{\text{all gates}} s_{DP\ gate}^2}$$

Experiment Settings

- We used a 45nm Nangate FreePDK45 Generic Open Cell Library
- We introduced 20% random variation of normal distribution to four process parameters
 - W_g , L_{eff} , T_{OX} and na
 - Similar variation is added to the unit R and C for wires
- 10000 Monte Carlo samples, each with 1000 clock cycles

Dynamic power estimation vs. HSPICE-based Monte Carlo simulation

	Dynamic Power (μ W)				Differences (%)	
	HSPICE		Our Model			
	mean	σ	mean	σ	mean	σ
Circuit 1	2.649	0.2346	2.719	0.2479	-2.64	-5.67
Circuit 2	3.094	0.3816	2.961	0.3996	4.30	-4.72
Circuit 3	3.481	0.3865	3.514	0.4114	-0.95	-6.44
Circuit 4	3.186	0.3371	3.097	0.3539	2.79	-4.98
Circuit 5	4.491	0.4829	4.664	0.4970	-3.85	-2.92
Average of Absolute Differences					2.91	4.95

Runtime comparison

	HSPICE (hours)	Our Model (seconds)	Speed-up ($\times 10^6$)
Circuit 1	1692	2	3.0
Circuit 2	1824	2	3.3
Circuit 3	4326	2	7.8
Circuit 4	6120	2	11.0
Circuit 5	9776	2	17.6
Average			8.5

Results for ISCAS89 circuits

	Dynamic Power		Runtime (sec)
	Mean (mW)	σ (%)	
C1355	1.10	36.7	8
C1908	2.56	16.8	11
C2670	4.24	16.1	43
C3540	6.86	34.0	40
C432	0.85	20.5	4
C499	1.01	32.5	7
C5315	10.46	19.0	90
C6288	20.16	24.4	64
C7552	13.42	17.2	119
C880	1.02	27.4	8

Conclusions

- We presented a novel dynamic power estimation method for circuits with process variation
- Our estimation takes into account the difference between partial-swing and full-swing glitches
- We obtained high accuracy estimates, with average errors of 3% for the means and 5% for the standard deviations
- Our estimation is also very fast, more than six orders of magnitude comparing to SPICE-Monte Carlo
- Future work includes consideration of the correlation among the process parameters

Thank You