Chebyshev Affine Arithmetic Based Parametric Yield Prediction Under Limited Descriptions of Uncertainty

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Power Leakage

- Exponential rise of IC power dissipation
 - Device dimension scales down.
 - Threshold voltage shrinks.
- Great portion of total power consumption
 - May account for 50% of the total delay
 - Will be further aggravated
 - Significant impact on parametric yield

Parameter Variations

- Strong dependence of leakage on both process and environmental variations.
- Cause a large spread in leakage current.
- 2 kinds of variations are considered.
 - Process parameters
 - Leff, Vth, and Tox
 - Environmental parameters
 - Vdd, and T

Parametric Yield Prediction

- Parameter variations reduces the yield of designs.
- Yield prediction methods are required to model the dependency.
- Limited by fundamental features of IC design
 - Incomplete process characterization data.
 - Large uncertainty in statistic metrics.
 - Correlation between parameters.

Main Purposes

- Uncertainty representations
- Probability representations
- Consider both process variations and environme ntal uncertainty
- Consider correlation between parameters
- Provide reliable probability bounds for leakage c urrent

Chebyshev Affine Arithmetic

• Affine form:

$$\hat{x} = x_0 + x_1 \varepsilon_1 + x_2 \varepsilon_2 + \cdots + x_n \varepsilon_n$$

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$$\mathcal{E}_i \in [-1,1]$$
 and $E[\mathcal{E}_i] = 0$

- \mathcal{X}_0 , central value
- \mathcal{E}_i , noise symbols
- \mathcal{X}_i , partial deviations

Chebyshev Affine Operations

- Can be easily expanded
- 3 cases:

$$\hat{x} \pm \hat{y} = (x_0 + y_0) + (x_1 + y_1)\varepsilon_1 + \dots + (x_n + y_n)\varepsilon_n$$
$$\alpha \hat{x} = (\alpha x_0) + (\alpha x_1)\varepsilon_1 + \dots + (\alpha x_n)\varepsilon_n$$
$$\hat{x} \pm \zeta = (x_0 + \zeta) + x_1\varepsilon_1 + \dots + x_n\varepsilon_n$$

• Still in affine form

Non-affine Operations

•
$$z = f(\hat{x}, \hat{y}) = f^*(\varepsilon_1, \dots, \varepsilon_n)$$
, f^* is not affine

Approximations required

$$\hat{z} = f^a(\varepsilon_1, \cdots, \varepsilon_n) = z_0 + z_1\varepsilon_1 + \cdots + z_n\varepsilon_n + z_k\varepsilon_k$$

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$$\mathcal{Z}_k \mathcal{E}_k$$
 represents approximation error
 $e^*(\mathcal{E}_1, \dots, \mathcal{E}_n) = f^*(\mathcal{E}_1, \dots, \mathcal{E}_n) - f^a(\mathcal{E}_1, \dots, \mathcal{E}_n)$

• Returns an Affine form

Chebyshev Approximations

- in the form of affine combinations : $\alpha \hat{x} + \beta y + \zeta$
- Optimal: minimizes the maximum absolute error
- Geometric illustration $\hat{z} = \alpha \hat{x} + \zeta + \delta \varepsilon_k$



Leakage Model

- An empirical model
- Obtained from SPICE simulation
- Model the dependency on parameter variations
 - Leff: quadratic exponential dependency
 - Vth: exponential dependency
 - Tox: exponential dependency
 - Vdd: exponential dependency
 - T: supper linear dependency (approximated as expo nential)

Analytical Equations

- Mathematical representations of leakage model
 - Subthreshold leakage model

$$I_{sub} = I_{sub,nom} \cdot e^{a\Delta L^2 + b\Delta L + c\Delta V_{th} + d\Delta V_{dd} + e\Delta T}$$

- Gate leakage model

$$I_{gate} = I_{gate,nom} \cdot e^{h\Delta T_{ox} + k\Delta V dd}$$

- Total leakage is the summation

$$I_{total} = I_{sub} + I_{gate}$$

Parameter Decomposition

• Parameter variations further decomposed into two components.

$$\Delta P = \Delta P_{global} + \Delta P_{local}$$

- ΔP_{global} , the global (inter-chip) variations
- ΔP_{local} , the local (intra-chip) variations
- Assumed to be independent and normal
- Result in also normal distribution ΔP

Improved Leakage Model

Subthreshold leakage model

$$I_{sub} = I_{sub,nom} \cdot e^{a\Delta L_l^2 + (2a\Delta Lg + b)\Delta L_l + c\Delta V_{th,l} + d\Delta V dd + \Delta T} \cdot e^{a\Delta L_g^2 + b\Delta L_g + c\Delta V_{th,g}}$$

• Gate leakage model

$$I_{gate} = I_{gate,nom} \cdot e^{h\Delta T_{ox,l} + k\Delta V dd} \cdot e^{h\Delta T_{ox,g}}$$

• They are correlated

Issues with New Technology Nodes

- Parameters are difficult to extract: uncertainty in probability distributions (70 nm and below)
- We use a set of CDFs consisting of a left and a right bound $\underline{F}(x) \le F(x) \le \overline{F}(x)$



Chebyshev vs. Discretized Method $\overline{F}(x)$ $\overline{F}(x)$ $\kappa_{\underline{F}(x)}$ $\underline{F}(x)$ X x Probabili $\overline{F}(x)$ $\overline{F}(x)$ \mathbf{K} <u>F(x)</u> K $\underline{F}(x)$ mulati

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x

PLPB Representation

- Linearization on CDF: Piece-wise Linear Probability Bounds (PLPB)
- Computation on Parameters' CDF functions



Piece-wise Linear CDF with its inverse



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Dependency Bounds of Z=X+Y

• Upper bound

$$\overline{F}_{X+Y}^{(-1)}(p) = \begin{pmatrix} \max[\overline{F}_X^{(-1)}(u) + \overline{F}_Y^{(-1)}(p-u)] & \text{if } p \neq 1 \\ \overline{F}_X^{(-1)}(1) + \overline{F}_Y^{(-1)}(1) & \text{if } p = 1 \end{cases}$$

• Lower bound

$$\underline{F}_{X+Y}^{(-1)}(p) = \begin{pmatrix} \min_{u \in [p,1]} [\underline{F}_{x}^{(-1)}(u) + \underline{F}_{Y}^{(-1)}(p-u+1)] & \text{if } p \neq 0 \\ \underline{F}_{X}^{(-1)}(0) + \underline{F}_{Y}^{(-1)}(0) & \text{if } p = 0 \end{cases}$$

Dataflow of Computing $\underline{F}_{X+Y}^{(-1)}(p)$



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Dependency bounds of X-Y

• Upper bound

$$\overline{F}_{X-Y}^{(-1)}(p) = \begin{pmatrix} \max[\overline{F}_X^{(-1)}(u) - \underline{F}_Y^{(-1)}(u-p+1)] & \text{if } p \neq 1 \\ \overline{F}_X^{(-1)}(1) - \underline{F}_Y^{(-1)}(0) & \text{if } p = 1 \end{cases}$$

• Lower bound

$$\underline{F}_{X-Y}^{(-1)}(p) = \begin{pmatrix} \min_{u \in [x,1]} [\overline{F}_X^{(-1)}(u) - \underline{F}_Y^{(-1)}(u-p)] & \text{if } p \neq 0 \\ \overline{F}_X^{(-1)}(0) - \underline{F}_Y^{(-1)}(1) & \text{if } p = 0 \end{cases}$$

Yield Prediction Procedure



Yield Prediction Procedure (continued)



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Experiment Environments

- 65nm Technology node PTM model.
 - Leff = 24.5nm
- Coefficients extracted by SPICE simulations.
- Parameter variations
 - Modeled as truncated Gaussian distributions.
 - Can be well handled if non-Gaussian.
 - Leff: 20% variation, Vth: 10% variation, Tox: 8% variation

Vdd: 10% variation, T: 10[°]C variation.

Comparison with MC simulation and inter val analysis: I_{sub}



Comparison with MC simulation and inter val analysis: I_{gate}



Comparison with MC simulation and inter val analysis: I_{total}



 Improvements:
 50% percentile 10.9%
 95% percentile 23.6%
 Mean value 21.7% (1.566->1.226)

Contours for inter-chip L Variation



 Shorter channel len gth causes more si gnificant variation of leakage current.

Conclusion

- Based on Chebyshev affine arithmetic
- Handle uncertainty of distributions
- Deal with correlations among variations
- Efficient and reliable yield prediction