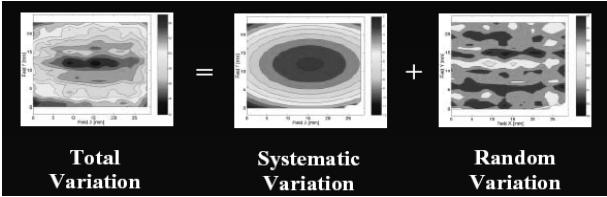


Statistical Timing Analysis

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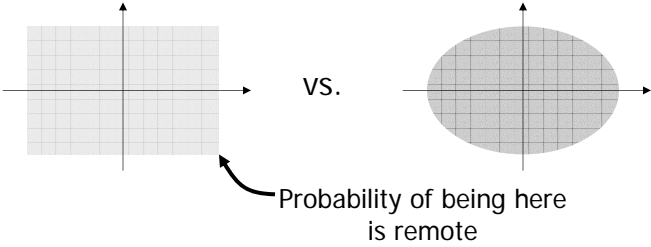


Corner-based vs. statistical analysis



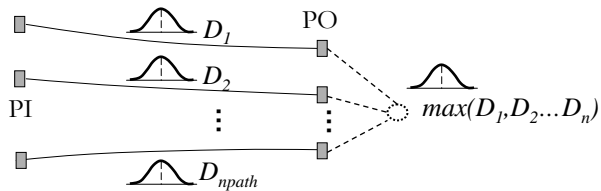
[Cain/Friedberg, UC Berkeley]

- Ellipse or square?



Statistical static timing analysis

- Intra-die variations in addition to inter-die variations
- Deterministic timing analysis \Rightarrow Statistical timing analysis
- Path-based analysis: find variability along a single path



- Block-based analysis: Find the distribution (PDF/CDF) of:

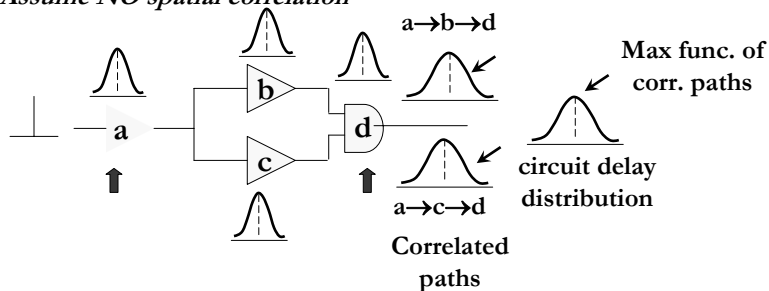
$$D_{max} = \max(D_1, D_2, \dots, D_{npaths})$$

D_i : distribution of i^{th} path delay

Difficulties in statistical timing analysis

- Path correlation due to reconvergent fanouts

Assume NO spatial correlation



- Spatial correlation makes the path correlation structure more complicated

Problem statement

- Find the PDF/CDF for circuit delay distribution:

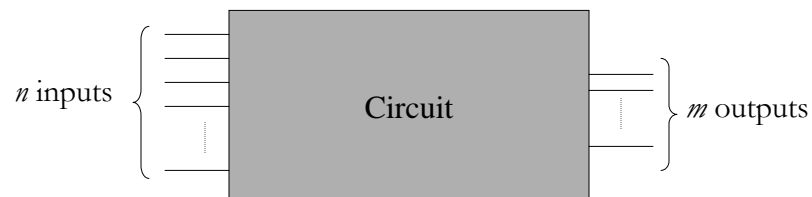
$$D_{max} = \max(D_1, D_2, \dots, D_{npaths})$$

where D_i : delay distribution of i^{th} path in the circuit

- Assume normal distributions on process parameter values
 - Why?
 - Is this reasonable? If not, what is?
- Parameter correlations
 - L_{eff} shows high spatial correlations
 - T_{ox}, N_d are largely uncorrelated

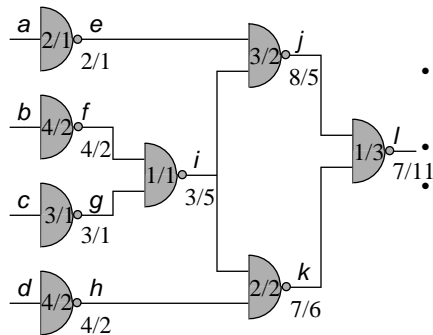
Basics of static timing analysis

- Main features
 - Fast
 - Input pattern-independent
 - Uses coarse delay models that are good for a first order approximation, and for optimization purposes



The critical path method (CPM)

- Often (incorrectly) called "PERT" in the literature



- Place PI's on a queue
- Process gate from head of queue
- Add gate to queue if all input delays known
- Complexity $O(E)$
- Can handle gate delay dependence on order of input arrivals

Interconnect and gate delay models

- Approximate a function by first-order Taylor expansion:

$$d = f(\vec{P}) \quad \vec{P} : \text{Set of normally distributed random variables}$$

$$d = \bar{d} + \sum_{\forall \text{ Parameter } p_i} \left[\frac{\partial f}{\partial p_i} \right]_{\bar{p}_i} \Delta p_i$$

Nominal Value of d
Sensitivity of f to p_i
 $\Delta p_i = p_i - \bar{p}_i$

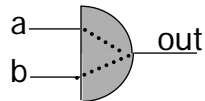
- d is a linear function of normally distributed random variables
- Sensitivities of gate/interconnect delay to the process parameters

Simple approaches

- Build in slacks to reduce variation
 - Bai *et al.*, DAC 02
- More exact methods
 - Perform statistical STA

Berkelaar's method

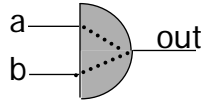
- Types of operations in STA
 - SUM: $T_{a \rightarrow \text{out}} = T_a + d_{a \rightarrow \text{out}}$; $T_{b \rightarrow \text{out}} = T_b + d_{b \rightarrow \text{out}}$
 - MAX: $T_{\text{out}} = \max(T_{a \rightarrow \text{out}}, T_{b \rightarrow \text{out}})$



- Gate delay modeled as a Gaussian
 - SUM is easy: sum of Gaussians = Gaussian
 - MAX of Gaussians is not a Gaussian
 - Approach: approximate max by a Gaussian

Tsukiyama's method

- No correlations in Berkelaar's method



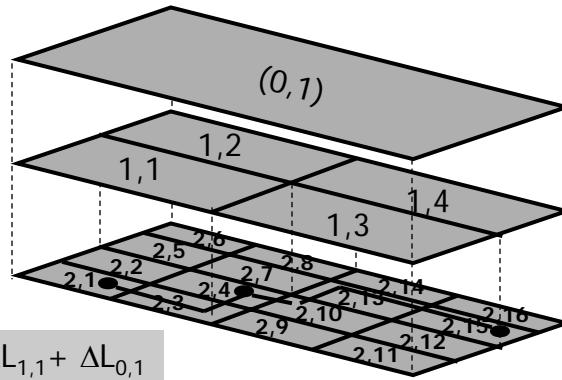
- Assume $d_{a \rightarrow out}, d_{b \rightarrow out}$ to be correlated Gaussians
- Basic operations:
 - SUM: still easy
 - MAX: use Clark's results from 1961 to find mean, variance of the max function

Incorporating spatial correlations

- In reality, there is a strong spatial correlation between process variables
- Orshansky's method (DAC02)
 - Path-based approach
 - Finds path pdf's, covariances, etc.
 - Uncorrelated distribution provides upper bounds on expected value
 - Uses results from probability theory to estimate pdf of the max function

Agarwal's path-based method

- Variational model
 - Regions of variation



$$\begin{aligned}\Delta L_{g1} &= \Delta L_{2,1} + \Delta L_{1,1} + \Delta L_{0,1} \\ \Delta L_{g2} &= \Delta L_{2,4} + \Delta L_{1,1} + \Delta L_{0,1} \\ \Delta L_{g3} &= \Delta L_{2,15} + \Delta L_{1,4} + \Delta L_{0,1}\end{aligned}$$

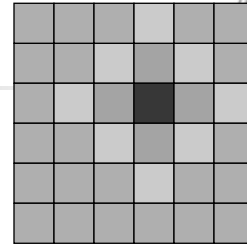
Agarwal's path-based method (contd.)

- For a small change
 - $\Delta D = S_L \Delta L + S_W \Delta W + \dots$
- Delay variations

$$\begin{aligned}\Delta D_{g1} &= K(\Delta L_{2,1} + \Delta L_{1,1} + \Delta L_{0,1}) \\ \Delta D_{g2} &= K(\Delta L_{2,4} + \Delta L_{1,1} + \Delta L_{0,1}) \\ \Delta D_{g3} &= K(\Delta L_{2,15} + \Delta L_{1,4} + \Delta L_{0,1})\end{aligned}$$

- Add up delay variations on each path using the sum operator to get path delay distributions

Chang's method



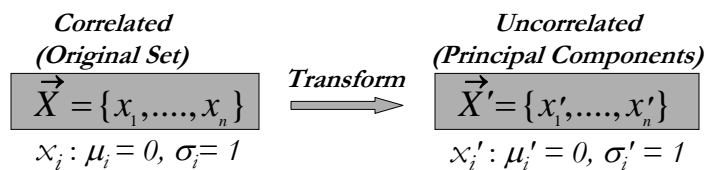
- Variational model
 - Chip area divided into n squares
 - Nearby squares are correlated
- Basic idea
 - Compute sensitivities of Delay to process parameters

$$\Delta D = \sum_k S_{Lk} \Delta L_k + \sum_k S_{Wk} \Delta W_k + \dots$$
 - Use principal component method to orthogonalize all variations

$$\Delta D = \sum S_i \Delta p_i$$
 - Calculating $\text{cov}(d_i, d_j)$ is easy!
 - Used to calculate approximation of max function
 - Sum function is easy, as before
 - PERT-like traversal: complexity = $n \times \text{STA complexity}$

Orthogonal transformation

- Principal Component Analysis (PCA):



- Element of original set can be expressed as linear function of the principal components (PC's)

$$x_i = \mu_i + \left(\sum_{j=1}^k \sqrt{\lambda_j} \alpha_{ij} x'_j \right) \cdot \sigma_i$$

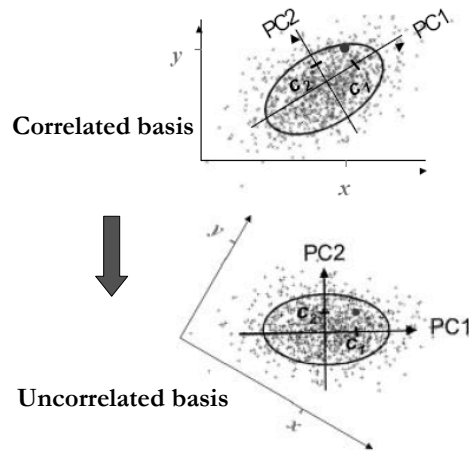
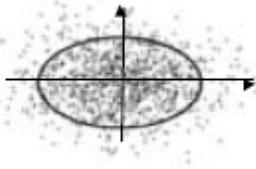
where λ_j : j^{th} eigenvalue of cov. matrix of \vec{X}

α_{ij} : j^{th} element of j^{th} eigenvector of cov. matrix of \vec{X}

Idea of orthogonal transformations

Set of equiprobable points for a 2D Gaussian = ellipse

Gaussian with a diagonal covariance matrix



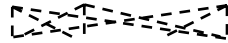
[Adapted from Tim Marks <http://cogsci.ucsd.edu/~desa/pca.pdf>]

Why?

Correlated

$$D_i = d_a^1 + d_a^2 + \dots + d_a^{N_i}$$

$$PI_i \begin{bmatrix} d_a^1 & d_a^2 & \dots & d_a^{N_i} \end{bmatrix} PO_i$$



$$PI_j \begin{bmatrix} d_b^1 & d_b^2 & \dots & d_b^{N_j} \end{bmatrix} PO_j$$

$$D_j = d_b^1 + d_b^2 + \dots + d_b^{N_j}$$



Orthogonalized

$$D_i = d_a^1 + d_a^2 + \dots + d_a^{N_i}$$

$$PI_i \begin{bmatrix} d_a^1 & d_a^2 & \dots & d_a^{N_i} \end{bmatrix} PO_i$$

$$PI_j \begin{bmatrix} d_b^1 & d_b^2 & \dots & d_b^{N_j} \end{bmatrix} PO_j$$

$$D_j = d_b^1 + d_b^2 + \dots + d_b^{N_j}$$

Experimental results

- Verified with Monte-Carlo simulation: Avg. error **0.2%** for mean, **0.9%** for s.t.d. and 0.2% for 98 percentile point
- Fast run-time: < 200s

Circuit Name	Error% w.r.t. Monte-Carlo		Monte-Carlo	Proposed Method	
	Mean	S.T.D.	CPU (s)	CPU(s)	PCA(s)
s38417	-0.1%	-0.7%	15295	130.32	0.15
s38584	-0.1%	-0.9%	19024	132.08	0.15
s35932	-0.3%	-3.6%	48087	182.31	0.15
s15850	-0.2%	-0.3%	9932	56.00	0.15
s13207	-0.1%	0.8%	5082	50.48	0.15
s9234	-0.1%	0.4%	2952	9.42	0.02
s5378	-0.3%	-0.5%	1531	5.27	0.02
s1196	-0.3%	-0.3%	378	0.41	0.01
s27	-0.1%	-0.3%	67	0.00	0.00

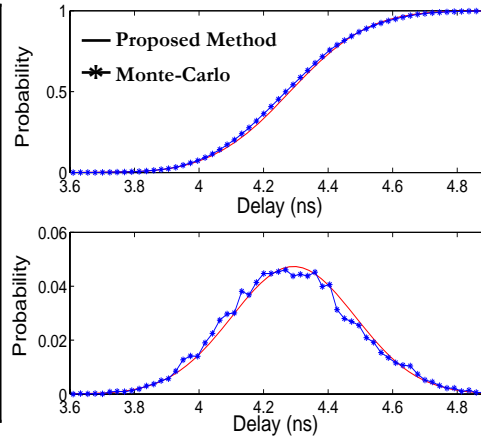
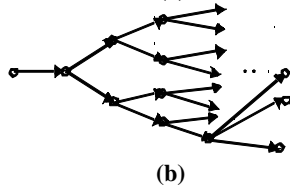
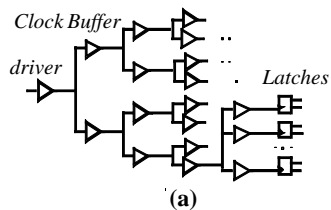


Fig 1: CDF and PDF Curves (ckt s38417)

Variability in the clock network



Routed clock network with tree topology composed of driver gates

- Timing tree
 - Root – clock driver
 - Sink nodes – Latches
 - Edges – gate + interconnect
- For a Deterministic Timing Tree:

$$d_{min} = \min(d_{p,1}, d_{p,2}, \dots, d_{p,n})$$

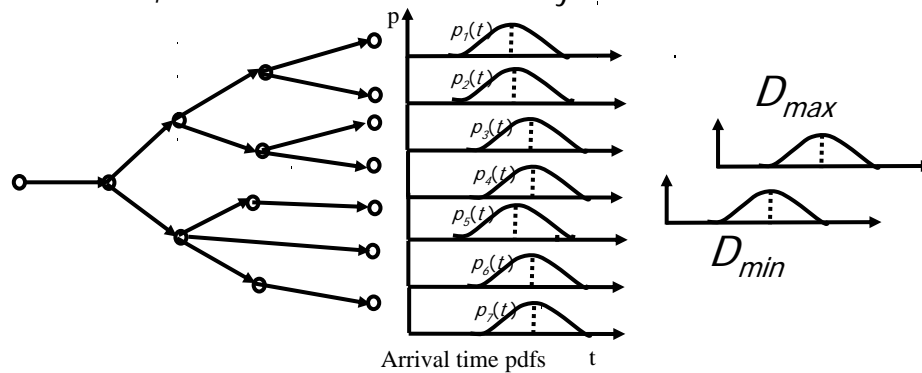
$$d_{max} = \max(d_{p,1}, d_{p,2}, \dots, d_{p,n})$$

$$skew = d_{max} - d_{min}$$
- Aim : Determine the skew distribution for a Probabilistic Timing Tree (PTT)

[Agarwal, ICCAD03]

Statistical clock skew analysis

- Propagate arrival time distributions to each sink node
- Difficulties: correlations between shared sections of the tree, between min and max delay

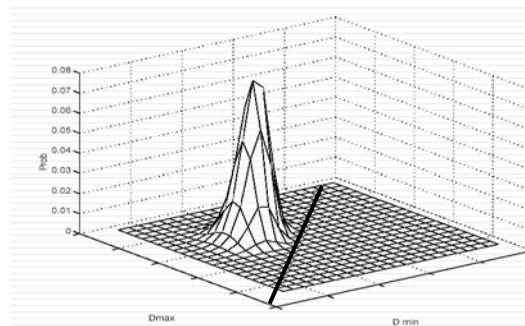


[Agarwal, ICCAD03]

Approach

- Key ideas
 - Compute the JPDF for D_{min} and D_{max}
 - Propagate bottom up
 - Details: see Agarwal *et al.*, ICCAD 2003

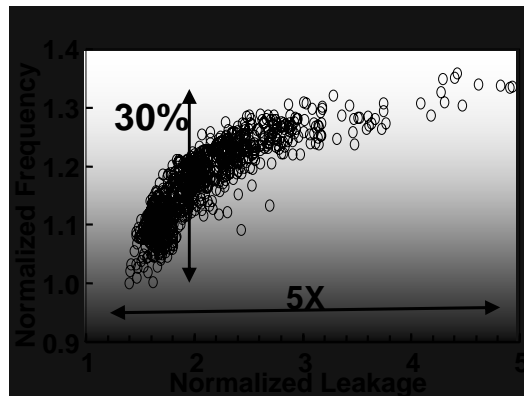
JPDF



[Agarwal, ICCAD03]

Statistical power analysis

- Leakage power
 - Most susceptible to variations
 - Components
 - Subthreshold leakage
 - Gate leakage
 - Problem amounts to summing up a set of correlated lognormals



What next?

- Statistical timing analysis methods will mature before statistical power analysis
 - The latter will be more “useful”
- Methodology issues
 - How does it all “come together” and work?
 - Statistical library characterization
 - Sources of uncertainty
 - Design uncertainty
 - Environmental uncertainty (Vdd/gnd levels, temperature, soft errors)
 - Process uncertainty
- Algorithmically, we still need solutions for
 - Nonlinear variations, non-Gaussian variations, multimodal distributions...



Conclusion



- Current approaches likely to be supplanted by statistical analysis
- Timing and power are both affected
- As a community, we now have an underlying algorithmic basis for this work
- Much more work to come...