

Interval-Based Robust Statistical Techniques for Non-Negative Convex Functions, with Application to Timing Analysis of Computer Chips

**Michael Orshansky¹, Wei-Shen Wang¹,
Gang Xiang², Vladik Kreinovich²**

**¹Department of Electrical and Computer Engineering
University of Texas at Austin**

**² NASA Pan-American Center for Earth and Environmental Studies
University of Texas at El Paso**

Overview

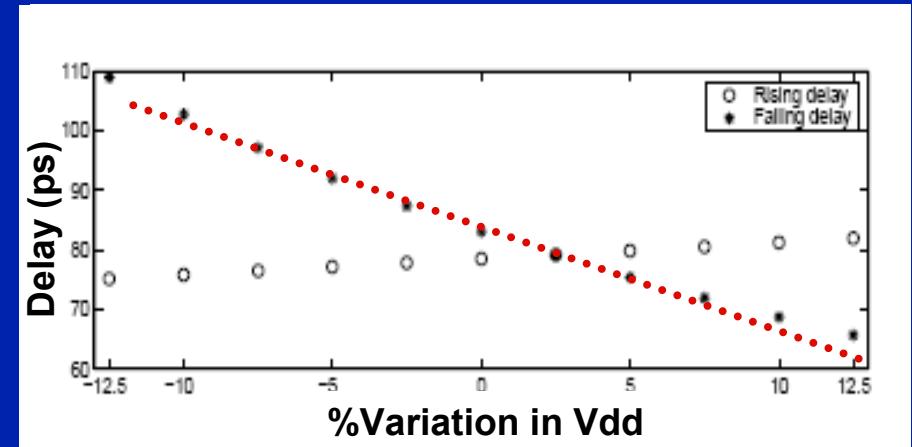
- Strategy of handling uncertainty: probabilistic interval analysis
- Timing analysis with partial probabilistic descriptions of uncertainty
 - ◆ Path delay computation
 - ◆ Circuit timing computation
- Experimental results
 - ◆ Robust timing estimates

Motivation

- Process and environmental uncertainty has great impact on timing performance of chip designs
 - ◆ Worst-case (interval) technique is conservative
 - ◆ Need to take into account probability
- Basic assumption of statistical timing analysis may not always be true: full probabilistic descriptions of parameters are not available
 - ◆ Only partial probabilistic information is available
- Need for strategy of handling partial probabilistic descriptions

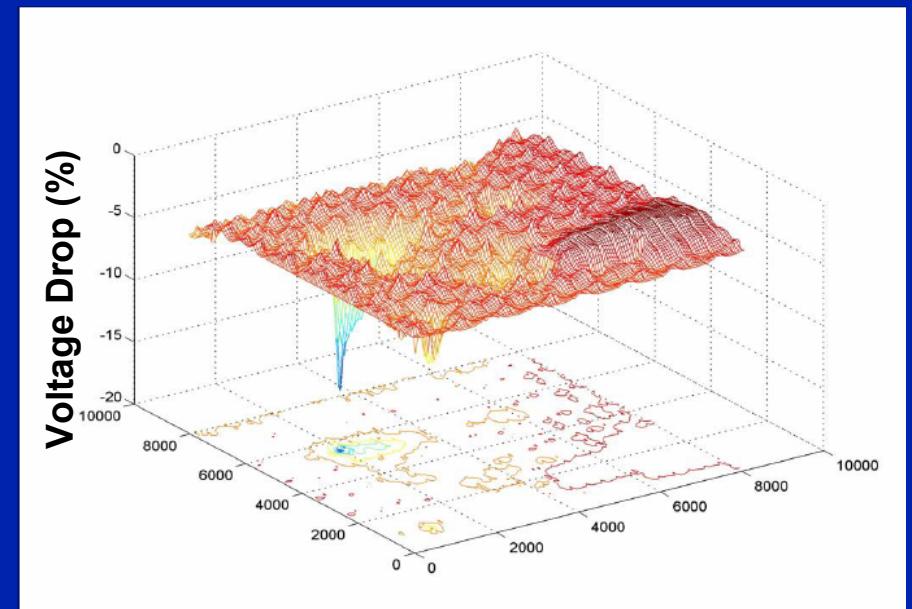
Limited Uncertainty Description

- Voltage drop affects gate delay
 - ◆ Difficult to fully characterize V_{dd} distribution
 - ◆ Use interval of V_{dd} to estimate worst/best-case delay



(Kouroussis '04)

- Mean and variance of voltage drop can be estimated
 - ◆ Use Monte Carlo sampling
 - ◆ Analytical technique also exists
 - ◆ Can compute moments – partial descriptions
 - ◆ Also can compute range



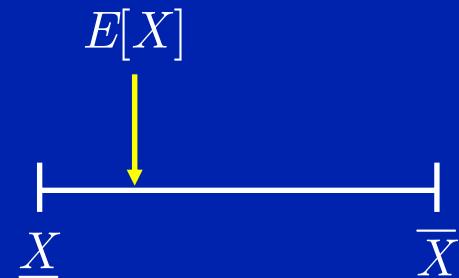
(Devgan)

Related Work

- Statistical Timing Analysis
 - ◆ Delay modeling: linear / nonlinear dependency on process variation
 - ◆ Process variability: Non-Gaussian Distribution (Zhan '05; Chang '05; Zhang '05)
- Interval / affine methods
 - ◆ Interval analysis (Moore '66)
 - ◆ Robust estimates using interval information
- Statistical delay computation based on affine arithmetic
 - ◆ Interconnect delay (Ma '04)
 - ◆ Need assumption about distributions within intervals

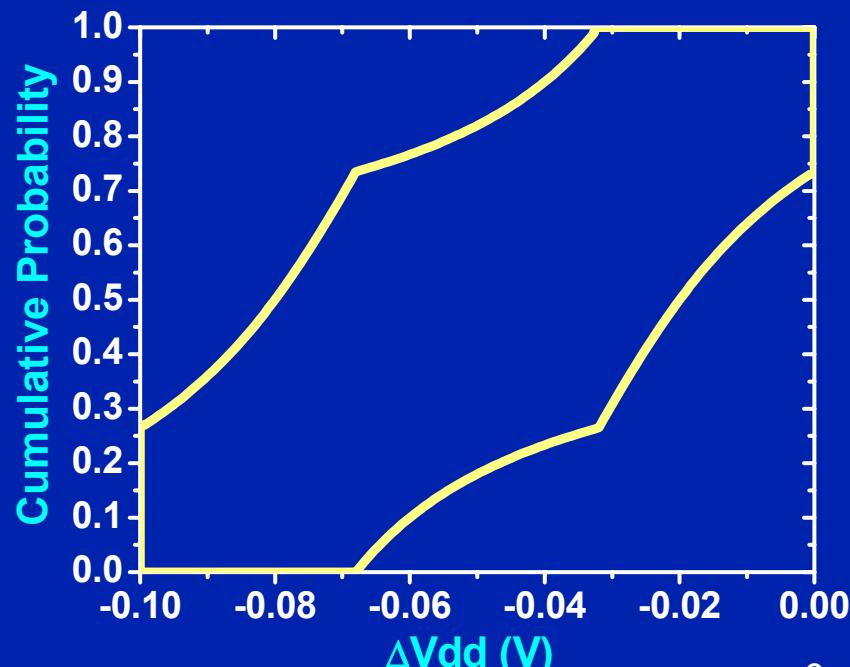
Handling Probabilistic Interval Uncertainty

- Probabilistic interval analysis
 - ◆ Use moments (mean, variance) and range to estimate probabilistic bounds
 - ◆ Include notion of probability
 - ◆ Distribution-free



- Probability box: representation of partially-specified variables

- ◆ A universal representation
 - ◆ Bounds for cumulative distribution functions
 - ◆ Knowledge of mean, variance, and interval permits constructing p-box



Operations of Probabilistic Interval Uncertainty in Timing Analysis

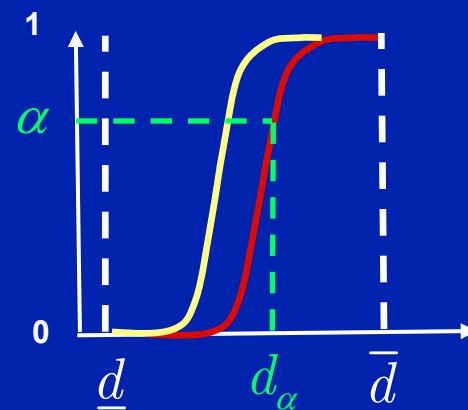
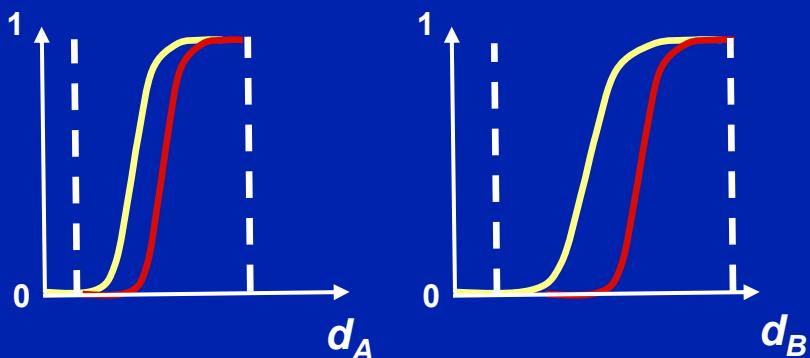
- Need to take into account correlation of gate delays

- Basic operations in timing analysis
 - Sum
 - Maximum

- Lower bound of cumulative probability (upper bound of delay) is a more important metric in timing analysis



d_A d_B



Timing Analysis with Partial Probabilistic Descriptions of Parameters

- Gate delay model
 - ◆ Consider global (die-to-die) and local (within-die) components

$$\begin{aligned}d_i &= \mu_i + \sum_{j=1}^n a_{i,j} \Delta x_{i,j} + \sum_{k=1}^m b_{i,k} \Delta y_{i,k} \\&= \mu_i + \sum_{j=1}^n a_{i,j} (\Delta x_{i,j,wd} + \Delta x_{j,dd}) + \sum_{k=1}^m b_{i,k} (\Delta y_{i,k,wd} + \Delta y_{k,dd})\end{aligned}$$

where Δx_i is normally distributed, and Δy_i is a *probabilistic interval variable* that mean and variance are available.

- Path delay

$$\begin{aligned}D^j &= \sum_{i \in P_j} (\mu_i + A_i^T X_{i,wd} + A_i^T X_{dd} + B_i^T Y_{i,wd} + B_i^T Y_{dd}) \\&= \underbrace{\sum_{i \in P_j} A_i^T X_{i,wd}}_{D_R^j} + \underbrace{\left(\sum_{i \in P_j} A_i^T \right) X_{dd}}_{D_{PI}^j} + \underbrace{\sum_{i \in P_j} B_i^T Y_{i,wd}}_{D_R^j} + \underbrace{\left(\sum_{i \in P_j} B_i^T \right) Y_{dd}}_{D_{PI}^j} + \sum_{i \in P_j} \mu_i\end{aligned}$$

Path delay due to Gaussian variables

Path delay due to probabilistic interval variables

Decomposition of Path Delay Function

- Computing distribution due to Gaussian variables (D_R^j) is straightforward
- Focus on path delay due to probabilistic interval uncertainty

$$D_{PI}^j = \sum_{i \in P_j} (\mu_i + u_i)$$

where $u_i = B_i^T Y_{i,wd} + B_i^T Y_{dd}$

- Range of delay variation

$$u_i \in \left[\sum_{k=1}^m b_{i,k} \underline{\Delta y_{i,k}}, \sum_{k=1}^m b_{i,k} \overline{\Delta y_{i,k}} \right]$$

- Mean and variance of path delay

$$E[D_{PI}^j] = \sum_{i \in P_j} \mu_i$$
$$Var\{D_{PI}^j\} = \sum_{i \in P_j} B_i^T \Sigma_{i,wd} B_i + \left(\sum_{i \in P_j} B_i^T \right) \Sigma_{dd} \left(\sum_{i \in P_j} B_i \right)$$

Computation of Probabilistic Bounds for Path Delay

- Compute probabilistic bounds for a set of distributions satisfying constraints of range, mean, and variance
 - ◆ Optimization problem
- Analytical expressions for CDF bounds: a generalization of Chebyshev and Cantelli inequalities (Ferson et al, '02)

$$\boxed{\begin{array}{ll} P(X \leq x) = 0 & x < \mu + \sigma^2 / (\mu - \bar{X}) \\ P(X \leq x) \geq 1 - (m(1+y) - s^2 - m^2) / y & \mu + \sigma^2 / (\mu - \bar{X}) \leq x \\ & \text{and } x < \mu + \sigma^2 / (\mu - \underline{X}) \\ P(X \leq x) \geq 1 / (1 + \sigma^2 / (x - \mu)^2) & \mu + \sigma^2 / (\mu - \underline{X}) \leq x < \bar{X} \\ P(X \leq x) = 1 & \bar{X} \leq x \end{array}}$$

where $y = (x - \underline{X}) / (\bar{X} - \underline{X})$ $m = (\mu - \underline{X}) / (\bar{X} - \underline{X})$ and $s^2 = \sigma^2 / (\bar{X} - \underline{X})^2$

- Probabilistic bounds for total path delay ($D_R^j + D_{PI}^j$) can be computed by convolution

$$CDF(D^j) = CDF(D_{PI}^j) \otimes PDF(D_R^j)$$

Circuit Delay Computation

- Path-based circuit timing analysis

$$\begin{aligned} D_{\max} &= \max(D^1, \dots, D^N) \\ &= \max(D_R^1 + D_{PI}^1, \dots, D_R^N + D_{PI}^N) \\ &\leq \underbrace{\max(D_R^1, \dots, D_R^N)}_{D_{R\max}} + \underbrace{\max(D_{PI}^1, \dots, D_{PI}^N)}_{D_{PI\max}} \end{aligned}$$

- Delay due to Gaussian variables:** $D_{R\max} = \max(D_R^1, \dots, D_R^N)$
 - Can be computed by SSTA based on first-order delay model and normal assumption
- Delay due to interval uncertainty:** $D_{PI\max} = \max(D_{PI}^1, \dots, D_{PI}^N)$
 - Can be bounded if mean and variance are available
 - Monte Carlo technique is used

Robust Monte Carlo Technique

- Monte Carlo techniques are used in timing analysis (Hitchcock '82; Jyu '93)
 - ◆ Generates samples drawn from given distributions
- Challenge: parameters with unknown distributions
 - ◆ Heuristically generates samples of various distributions
 - ◆ Time consuming
- Robust Monte Carlo technique is needed
 - ◆ Generates samples following *specific distributions* that cause extreme values of target function
- Circuit delay due to probabilistic interval uncertainty is a non-negative convex function
 - ◆ Allows to devise a technique for robust Monte Carlo

Convexity of Maximum Path Delay

- Path delay is linear thus convex function of probabilistic interval variables (y_i)

$$D_{PI}^j = \sum_{i \in P_j} (\mu_i + B_i^T Y_{i,wd} + B_i^T Y_{dd})$$

- If $D_{PI}^1, \dots, D_{PI}^N$ are convex, their pointwise maximum is also convex

$$D_{PI\max}(Y) = \max(D_{PI}^1(Y), \dots, D_{PI}^N(Y))$$

- Gate delays are always positive: non-negativity of maximum path delay

Robust Monte Carlo Simulation

- Let $\{v_1, \dots, v_M\}$ be a set of *independent* random variables, where $v_i \in [\underline{v}_i, \bar{v}_i]$, and $E[v_i] = E_i$ for $i=1$ to M . Let $y = f(v_1, \dots, v_M)$ be a *non-negative convex* function of the random variable v_i , for $i=1$ to M .
- Among all possible *cdfs* of $\{v_1, \dots, v_M\}$ that correspond to the range and the mean, the k^{th} moment of the function, $E[y^k]$, achieves the maximum value when each random variable v_i follows the 2-point distribution:

$$P(v_i = \underline{v}_i) = \underline{p}_i$$
$$P(v_i = \bar{v}_i) = \bar{p}_i$$

where $\bar{p}_i = \frac{E_i - \underline{v}_i}{\bar{v}_i - \underline{v}_i}$ and $\underline{p}_i = \frac{\bar{v}_i - E_i}{\bar{v}_i - \underline{v}_i}$.

- Furthermore, $E[y^k]$ achieves the minimum when $P(v_i = E_i) = 1$.
- This theorem and its corollary allow us to bound mean and variance of $y = f(v_1, \dots, v_M)$ using generated samples.

Algorithm of Fast Robust Monte Carlo Simulation

for $i = 1..N$

Generate samples for die-to-die components

for each gate

Generate samples for within-die components

Compute gate delay

end

Compute circuit delay, D_i .

end

Compute sample mean and variance,

$$\bar{D} = \sum_{i=1}^N D_i / N$$

$$s_D^2 = \sum_{i=1}^N (D_i - \bar{D})^2 / (N - 1)$$

With \bar{D} , s_D^2 , and range of circuit delay, use generalized Chebyshev inequality to compute lower bound for cdf.

Experimental Setup

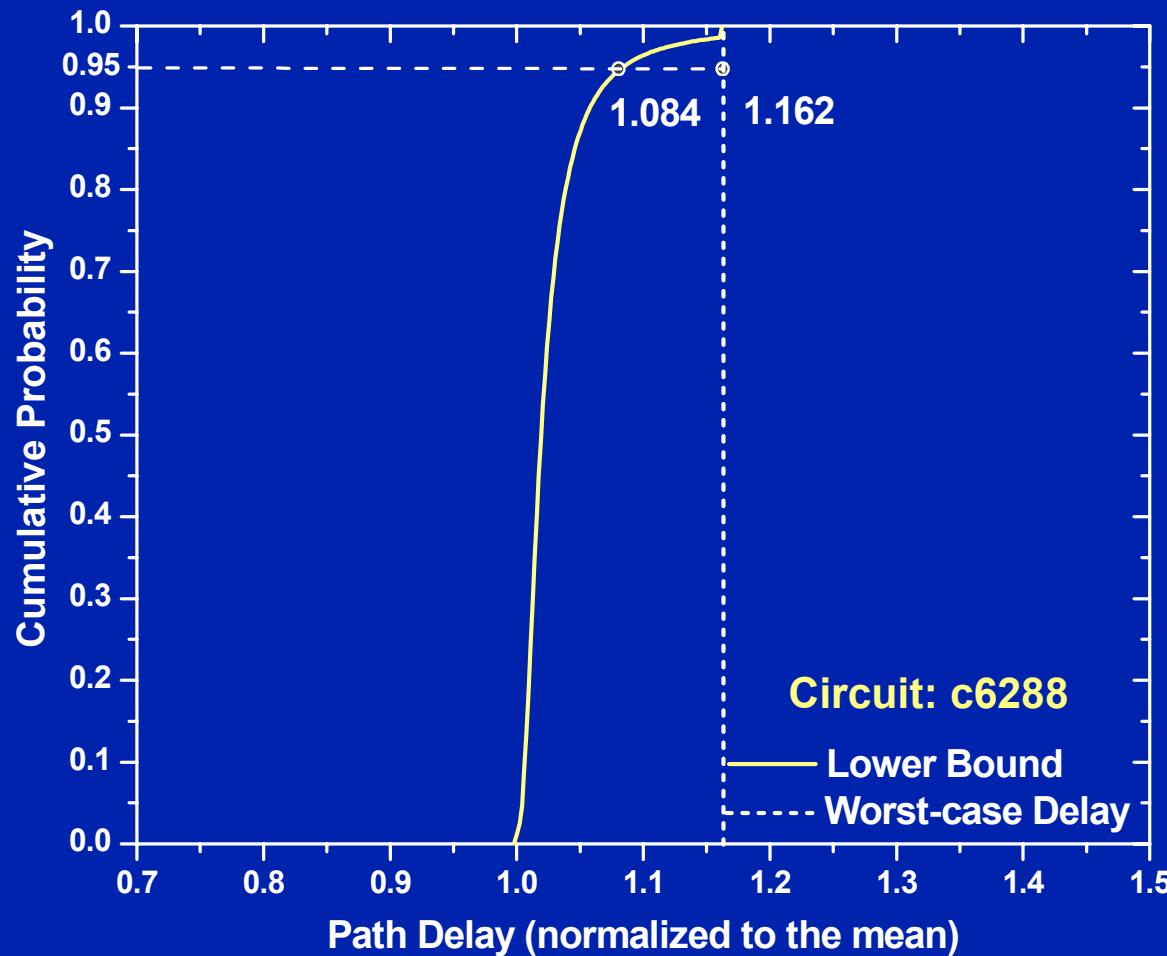
- Path and circuit timing analysis algorithms implemented in C++
- Cell library
 - ◆ Characterized using 130nm BPTM technology
- Process uncertainty

Parameter	Modeling	Magnitude of Uncertainty ($3\sigma / \mu$)	Magnitude of With-Die Variation
Effective Channel Length	Gaussian	20.0%	50%
Threshold Voltage	Probabilistic Interval	20.0%	50%
Oxide Thickness	Probabilistic Interval	20.0%	50%

- Environmental uncertainty

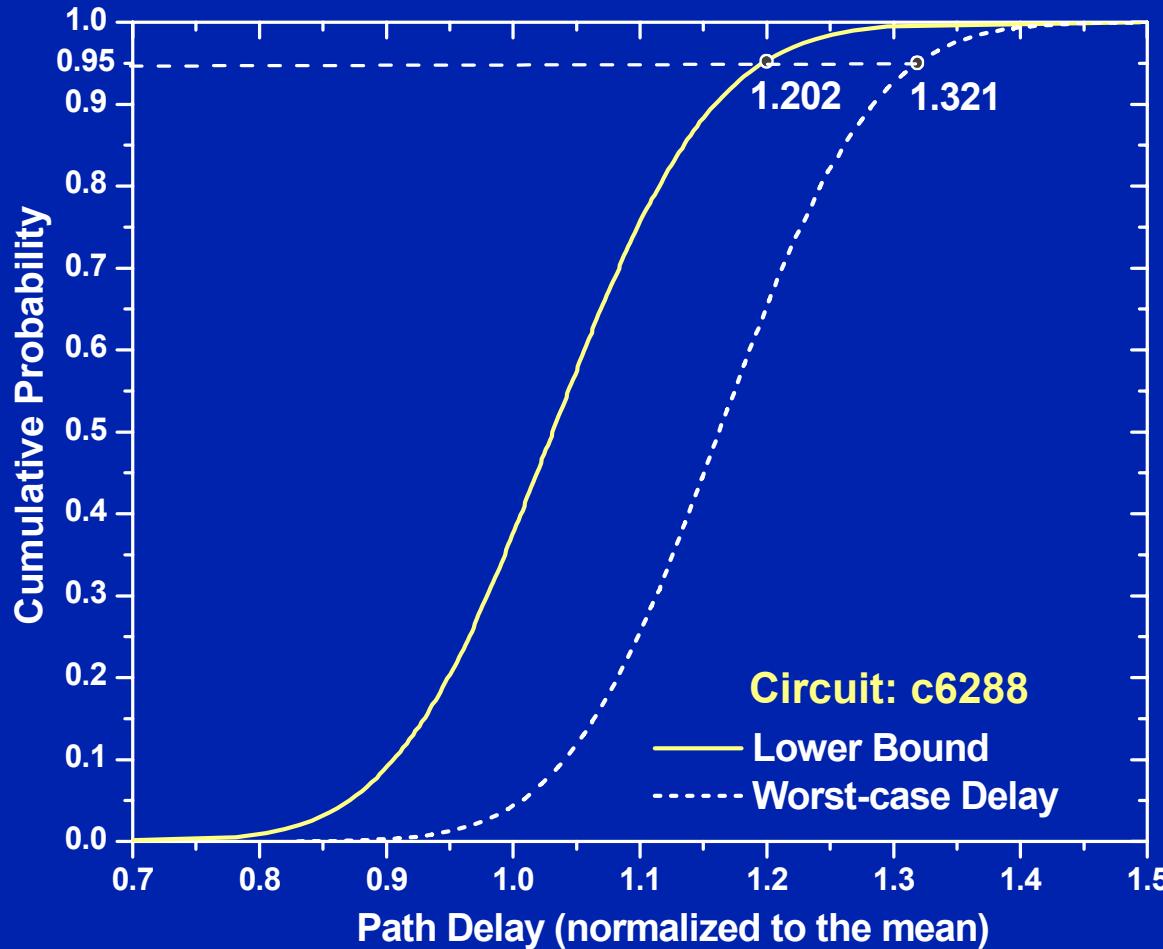
Parameter	Modeling	Magnitude of Uncertainty ($3\sigma / \mu$)	Magnitude of With-Die Variation
Power Supply Voltage	Probabilistic Interval	12.5%	100%

Results: Probabilistic Bounds for Path Delay due to Probabilistic Interval Uncertainty



- Proposed algorithm reduces worst-case delay by 6.7% at 95th percentile

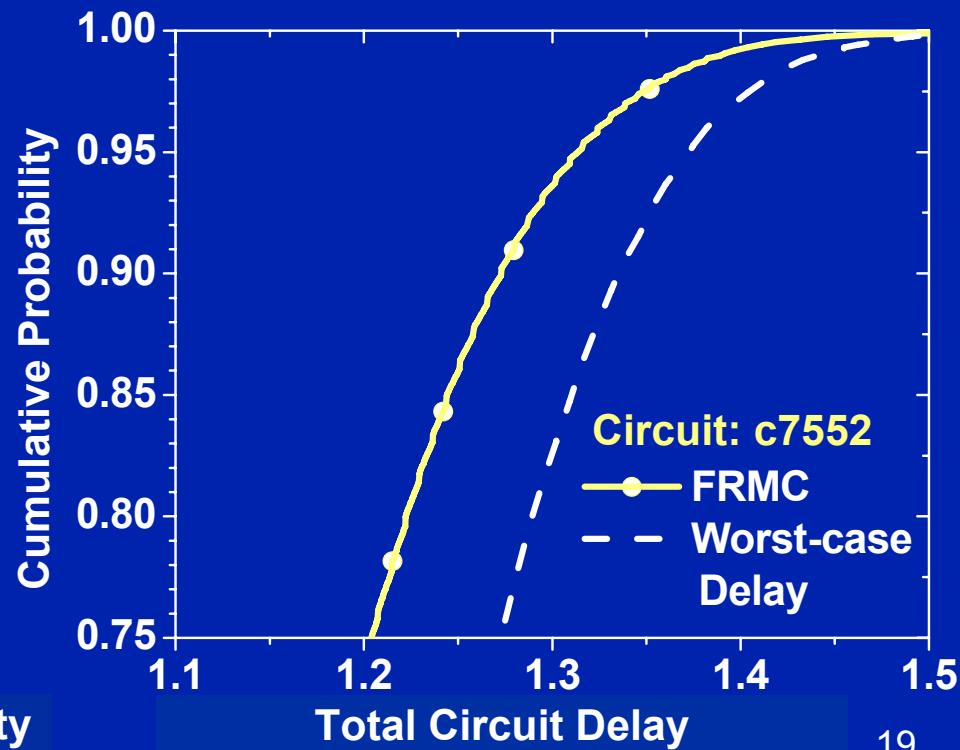
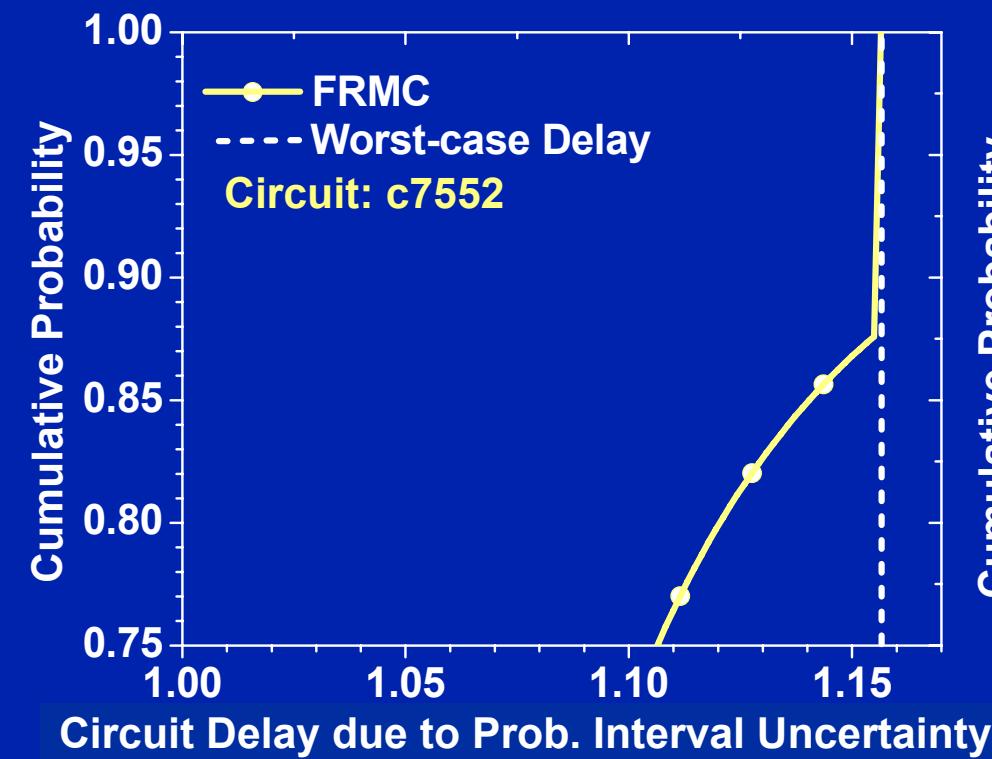
Results: Probabilistic Bounds for Total Path Delay



- Worst-case delay for probabilistic interval uncertainty is added to path delay due to Gaussian variables
- Proposed algorithm reduces worst-case delay by 9.0% at 95th percentile

Results: Probabilistic Bounds for Circuit Delay

- FRMC gives a better bound than worst-case delay at lower than 87th percentile for circuit delay due to probabilistic interval variables.
- For total circuit delay, FRMC improves worst-case delay by 4% at 95th percentile.



Comparison of FRMC and Worst-case Delay

- For total circuit delay, FRMC improves worst-case delay, on average, by 4% at 95th percentile across ISCAS 85 benchmark.

Circuit	Number of Gates	Fast Robust Monte Carlo Simulation				Run Time (s)	Worst-case Delay for Probabilistic Interval Variables		
		90th Percentile		95th Percentile			90th Percentile	95th Percentile	
		Delay (ps)	Reduction	Delay (ps)	Reduction		Delay (ps)	Delay (ps)	
c880	456	2383	5.62%	2467	4.97%	12	2525	2596	
c1355	605	2264	4.59%	2335	4.26%	18	2373	2439	
c1908	975	2820	5.56%	2919	4.89%	26	2986	3069	
c2670	1544	3124	5.65%	3232	5.08%	38	3311	3405	
c3540	1787	4097	5.49%	4237	4.94%	52	4335	4457	
c6288	2448	17547	5.28%	18081	4.82%	87	18526	18996	
c5315	2600	3579	5.49%	3703	4.88%	79	3787	3893	
c7552	3874	3136	4.88%	3236	4.46%	114	3297	3387	

Conclusion and Future Work

- **New strategy of handling uncertainty**
 - ◆ **Permits handling parameters of incomplete information**
 - ◆ **Reduces over-conservatism of worst-case timing analysis that only uses interval information of parameters**
 - ◆ **Predicts probability of timing violation for circuits with error correction mechanism**
 - ◆ **Also compatible with SSTA based on first-order delay model and normal assumptions**
- **Future: experiments on real-life cases and circuits**