



Jitter and Eye Estimation in High-Speed Channels

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Industry Advisory Board (IAB) November 2019

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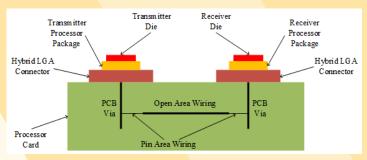
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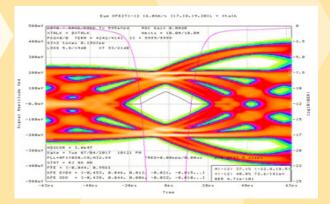
Georgia 1. Problem statement



- Evaluation of high-speed serial channels:
 - Estimation of jitter and eye-diagram
 - For channels with BER $\leq 10^{-12}$
- Traditional methods:
 - Transient eye analysis: Costly
 - Statistical methods: Only LTI systems
- Focus of this study: Data dependent jitter
 - Challenging to model
 - Caused by intersymbol interference (ISI)
- Machine learning:
 - Find the pattern between data and jitter with surrogate models.
 - Proposed approach:
 - Modified Polynomial Chaos theory





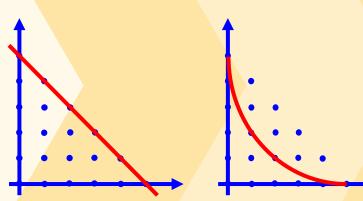


2. Polynomial Chaos (PC) theory 2.1. Formulation

- In uncertainty quantification:
 - For random variables $\lambda = [\lambda_1, \lambda_2, ..., \lambda_n]: f(\lambda) \approx \sum_{i=0}^{P} c_i \phi_i(\lambda)$
 - c_i = unknown coefficients, $\phi_i(\lambda)$ = known orthogonal polynomials
- $\phi_i(\lambda)$ is a product of 1-D orthogonal polynomials:
 - $\phi_i(\boldsymbol{\lambda}) = \prod_{j=1}^n \phi_{d_j}(\lambda_j)$
 - d_j shows indices of 1-D polynomials:
 - $\sum_{j=1}^{n} d_j \le m$, (linear constraint) m = order of expansion (2 or 3).
- Curse of dimensionality:
 - Length of expansion increases near exponentially $P + 1 = \frac{(m+n)!}{m!n!} = O(n^m)$
- Solution → Hyperbolic Polynomial Chaos (HPC) expansion:
 - $\sum_{j=1}^{n} d_j^u \le m^u$, 0 < u < 1, (hyperbolic constraint)
 - The selected terms have a higher impact on the output (Sparsity of effects).

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Georgia 2. PC theory 2.2. Training the model

- The system is evaluated at N points: $N \ge 2 * (P + 1)$
- Matrix form: $A\Gamma = E$,

$$\mathbf{A} = \begin{bmatrix} \phi_0(\boldsymbol{\lambda}^1) & \dots & \phi_P(\boldsymbol{\lambda}^1) \\ \vdots & \ddots & \vdots \\ \phi_0(\boldsymbol{\lambda}^N) & \dots & \phi_P(\boldsymbol{\lambda}^N) \end{bmatrix}, \mathbf{\Gamma} = \begin{bmatrix} c_0 \\ \vdots \\ c_P \end{bmatrix}, \mathbf{E} = \begin{bmatrix} f(\boldsymbol{\lambda}^1) \\ \vdots \\ f(\boldsymbol{\lambda}^N) \end{bmatrix}$$

- Solving with Ridge regression:
 - $\Gamma \approx (A^{\tau}A + B^{\tau}B)^{-1}A^{\tau}E$
 - Regularized with B:

•
$$B = \sqrt{\beta}$$
 I, except B(0,0) = 0

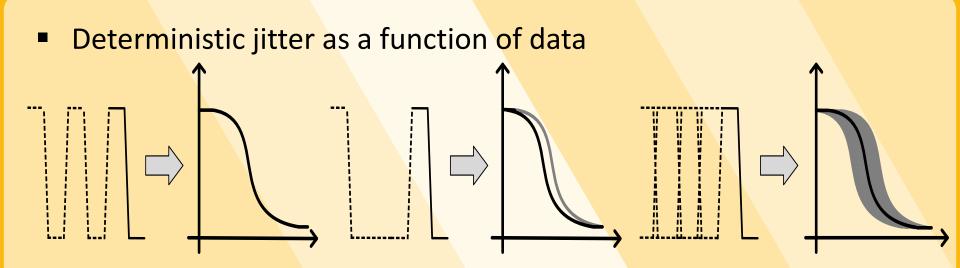
- Statistics directly from the coefficients:
 - Mean: $E(f(\lambda)) = c_0$
 - Variance: $Var(f(\lambda)) = \sigma^2 = \sum_{i=1}^{P} c_i^2$
- PDF: Sampling the surrogate model.

 $\mu + \sigma$

 $\mu - \sigma \mid \mu$



Georgia 3. Proposed approach 3.1. Intuition



- Rising/ falling edge is perturbed depending on the previous bits.
- Data and jitter are random variables → Uncertainty quantification.
- Idea behind the proposed approach:
 - Monte Carlo (MC): Traditional method for uncertainty quantification.
 - With numerous samples.
 - PC methods outperform MC.
 - Transient eye: Traditional method for eye analysis.
 - With numerous transient samples.

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Develop an eye analysis method with **PC models**.

3. Proposed approach

3.2. General setting

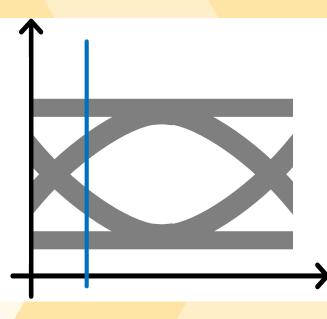
Input random variables:

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- Ideal value of the previous n bits with nontrivial ISI.
 - $\boldsymbol{\lambda} = [\boldsymbol{\lambda}_1, \boldsymbol{\lambda}_2, \dots \boldsymbol{\lambda}_n]$
- Output random variables:
 - Perturbation of rising/ falling edge.
 - Receiver voltage.
- A different model per transition is generated to increase accuracy.
 - $\lambda_{n-1} = 0, \lambda_n = 0 \rightarrow \text{Steady zero.}$
 - $\lambda_{n-1} = 0, \lambda_n = 1 \rightarrow \text{Rising edge}.$
 - $\lambda_{n-1} = 1, \lambda_n = 0 \rightarrow$ Falling edge.
 - $\lambda_{n-1} = 1, \lambda_n = 1 \rightarrow \text{Steady one.}$





3. Proposed approach 3.3. Two different type of analysis

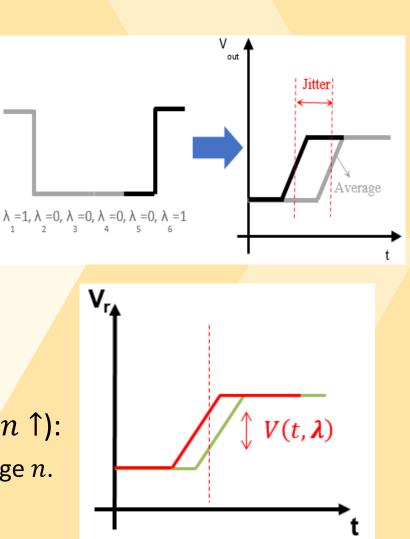
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Evaluating jitter directly:

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- Model perturbation of rising/ falling edges directly.
- Rising edge: $J_r(\lambda) \approx \sum_{i \in \alpha} c_r \phi_i(\lambda)$
- Falling edge: $J_f(\lambda) \approx \sum_{i \in \alpha} c_f \phi_i(\lambda)$
- Calculating the full eye diagram:
 - Model the receiver voltage over a unit interval.
 - $V_k(t,\lambda) \approx \sum_{i \in \alpha} c_{k_i}(t) \phi_i(\lambda)$
 - $0 \le t \le UI, 1 \le k \le 4$
- Modern channels have long delays $(n \uparrow)$:
 - We face curse of dimensionality for large n.
 - \rightarrow HPC expansion is used for n > 20







- For NRZ pulses: $\lambda_i \in \{-1, +1\}$
 - We do not need λ_i^2 , λ_i^3 , ... to model only 2 points.
- Only the relevant functions are selected from the PC sequence:

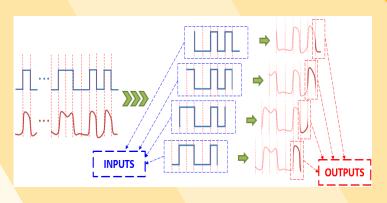
$$F(\lambda) = \sum_{j=0}^{p-1} c_j \phi_j(\lambda) =$$

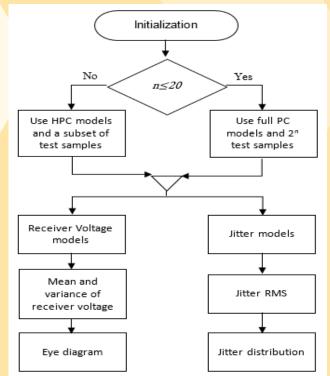
$$c_0 + \sum_{k=1}^n c_k \lambda^k + \sum_{r=1}^n \sum_{s=r+1}^n c_{rs} \lambda^r \lambda^s + \sum_{\alpha=1}^n \sum_{\beta=\alpha+1}^n \sum_{\gamma=\beta+1}^n \lambda^\alpha \lambda^\beta \lambda^\gamma + \cdots$$

- Interpretation:
 - Relationship of up to *m* bits at a time is considered.
 - m is maximum order of monomials in the PC expansion.
 - Distribution of $F(\lambda)$ is 2^n weighted impulse responses.
- For HPC expansion the hyperbolic constraint is applied to the above expansion.

Georgia 5. Implementation

- *N* training samples:
 - A short transient simulation.
 - A moving box of n UIs over the input.
 - A moving box of one UI over the output.
- Choose the suitable model:
 - $n \leq 20 \rightarrow \text{PC m} \text{odel}$
 - Otherwise \rightarrow HPC model
- Jitter models:
 - Trained directly with jitter values
 - Yields jitter RMS and distribution
- Receiver voltage models:
 - Trained with receiver voltage values
 - Yields mean and variance of the receiver voltage and the eye diagram







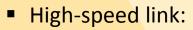
Georgia 6. Computation costs

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- Cost of transient simulation is significantly decreased →
 - Suitable for cases where cost of transient simulation is high.
- PC polynomials are known beforehand →
 - Reduce the cost by keeping the same sequence of random training bits.
- Total training cost:
 - = $O(\Psi P^2)$
 - Ψ : Number of time points in one unit interval.
 - P: Number of monomials in the PC expansion.
- Total testing cost:
 - $\bullet = O(\Psi \nu P)$
 - v: Number of samples evaluated for the eye diagram.

Georgia 7. Numerical example

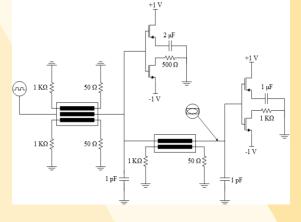


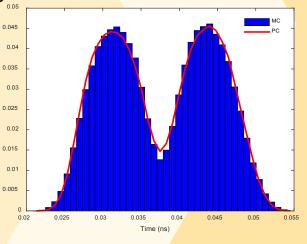


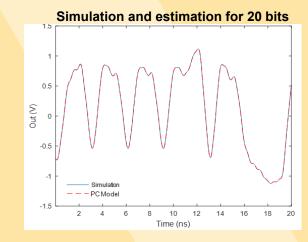
- Two sets of coupled lines
- Single ended signaling
- Nonlinear termination
- Speed: 1Gb/s
- Simulated in Ansys circuit simulator
- 20 bits considered for ISI
- Max order of polynomials = 3
- PC model
- Training: 60000 bits
- Testing: 1 million bits

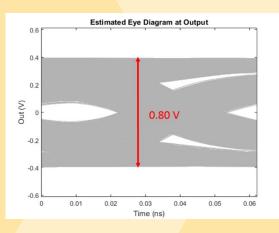
Computation times		Probability
	Time	LL.
60000 bits sim. with Ansys	663 s	
Training jitter and voltage models	3.93 s	
1 million bits estimation with PC	6.47 s	_
1 million bits sim. with Ansys	11055 s	

~16X speedup









RMS jitter values

	Low to high RMS jitter	High to l <mark>ow RMS jitter</mark>
Transient e <mark>ye analysis</mark>	<mark>8</mark> 7.2 ps	<mark>87.9 ps</mark>
Proposed PC approach	87.3 ps	87.8 ps

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Georgia 8. Conclusion

- Eye estimation:
 - Transient: Costly
 - Statistical: Inaccurate for non-LTI systems
- We developed modified PC and HPC surrogate models to:
 - Capture relation of data and jitter
 - From a single short transient simulation
 - Expands it to arbitrary input patterns
- Two types of analysis are provided:
 - Quick and direct jitter analysis
 - Full eye diagram analysis
- Proposed approach is applicable to non-LTI systems.
 - It can be integrated with estimation of other types of jitter.
- Future work focuses on quick estimation of the worst-case eye.



		2018	2019			2020			
		Q4	Q1	Q2	Q3	Q4	Q1	Q2	Q3
Done	1-Implementation of PC models								
Done	2-Direct Jitter distribution and analysis								
Done	3-Worst-case eye preliminary tests								
Done	4-Vectorizing PC models								
Progress	5-Worst-case eye implementation								
Done	6-Worst-case eye theory evaluation								
	7-Adding crosstalk to PC models								
	8-Improving efficiency of PC models								
	9-Further improvements on worst-case								
	10-Integrating with other types of jitter								
	11-Drafting final publications and thesis								

Light blue: Implementation and coding of established ideas. Dark blue: Research, study, and exploration of new ideas. Light yellow: Current time frame.

9. Schedule



Polynomial Chaos modes.
 Worst-case eye analysis.

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