

Unified Statistical Modeling for Circuit Simulation

Colin C. McAndrew
Patrick G. Drennan

WCM-MSM2002 April 12, 2002

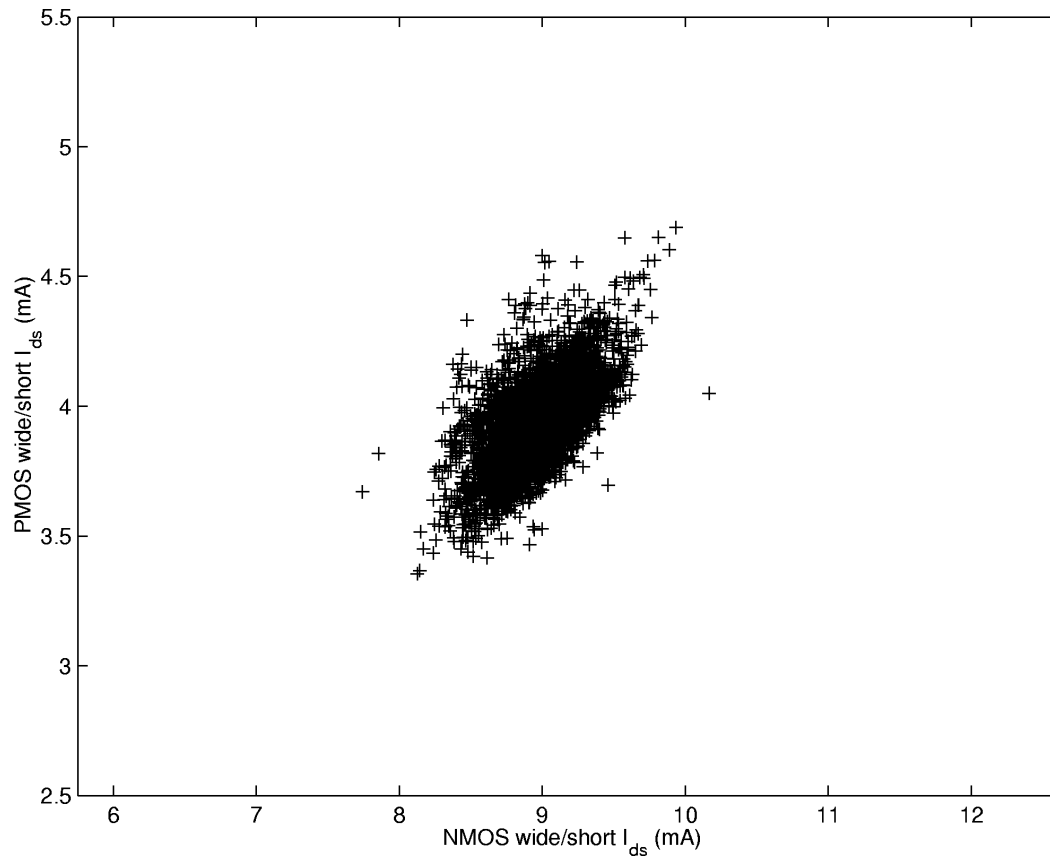
Statistical Modeling

- IC design flows can embody different types of statistical design practices
- The types of statistical simulation techniques used dictates the types of statistical models that are required
 - Corner simulations require case files
 - MC simulations need distributional models
 - Mismatch analysis needs mismatch models
 - Sensitivity analysis needs physical models

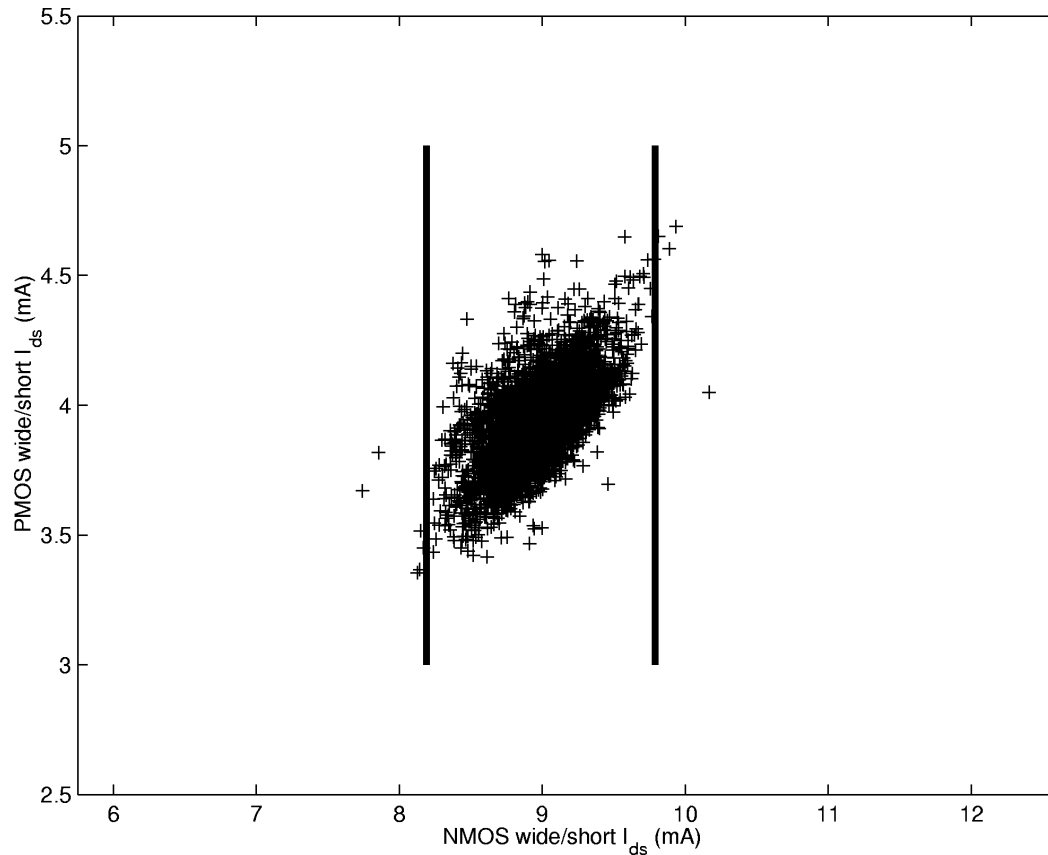
Goal of (Statistical) Modeling

- To accurately represent the (mean and standard deviation of) electrical performances E_i of devices (circuits)
- The goal is **NOT** to accurately represent the statistics of “parameters”
 - Effect of Δ_L differs between models
 - Different techniques to measure Δ_L give different values

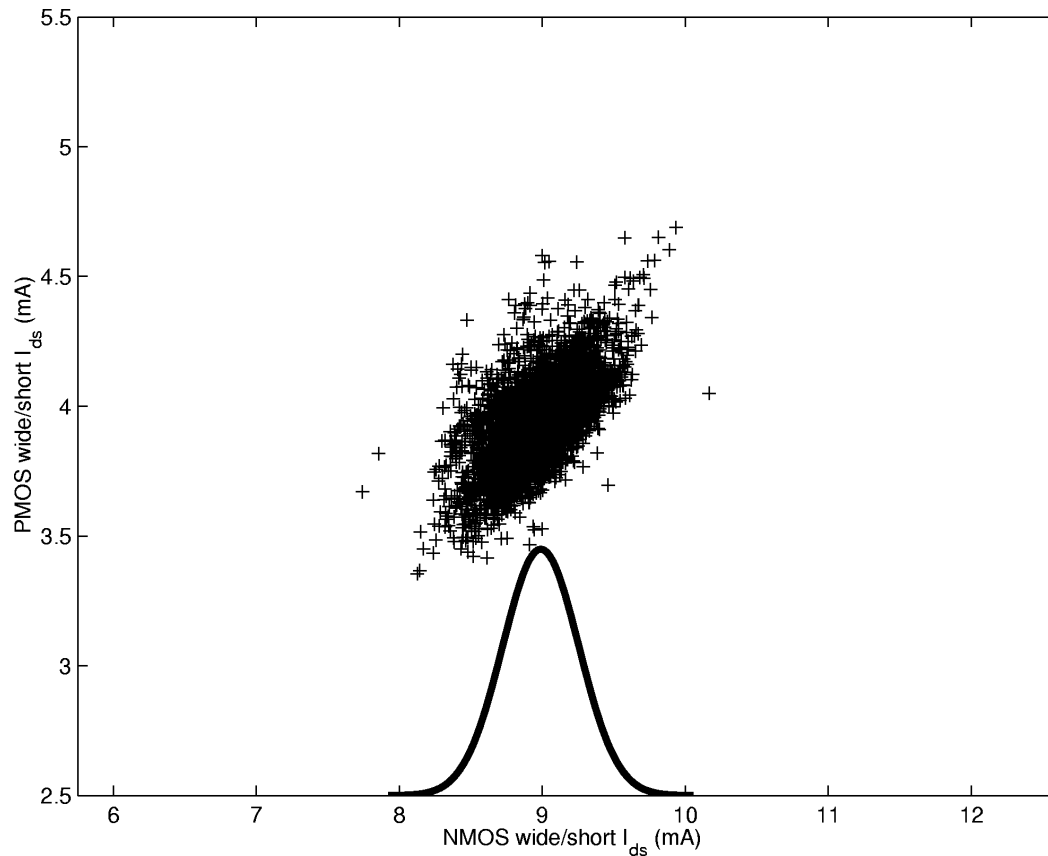
Goal: Model Fab Variations



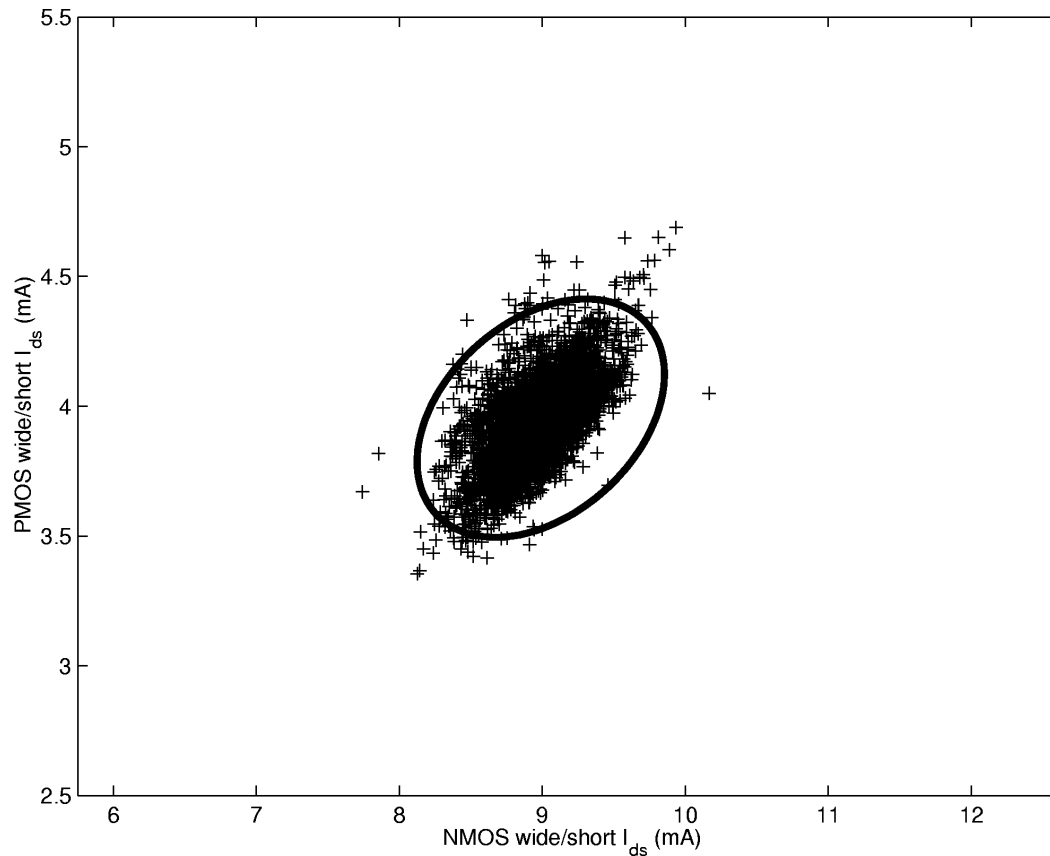
Goal: Model Fab Limits



Goal: Model Fab Distributions



Goal: Model Fab Correlations



Physical Modeling Basis

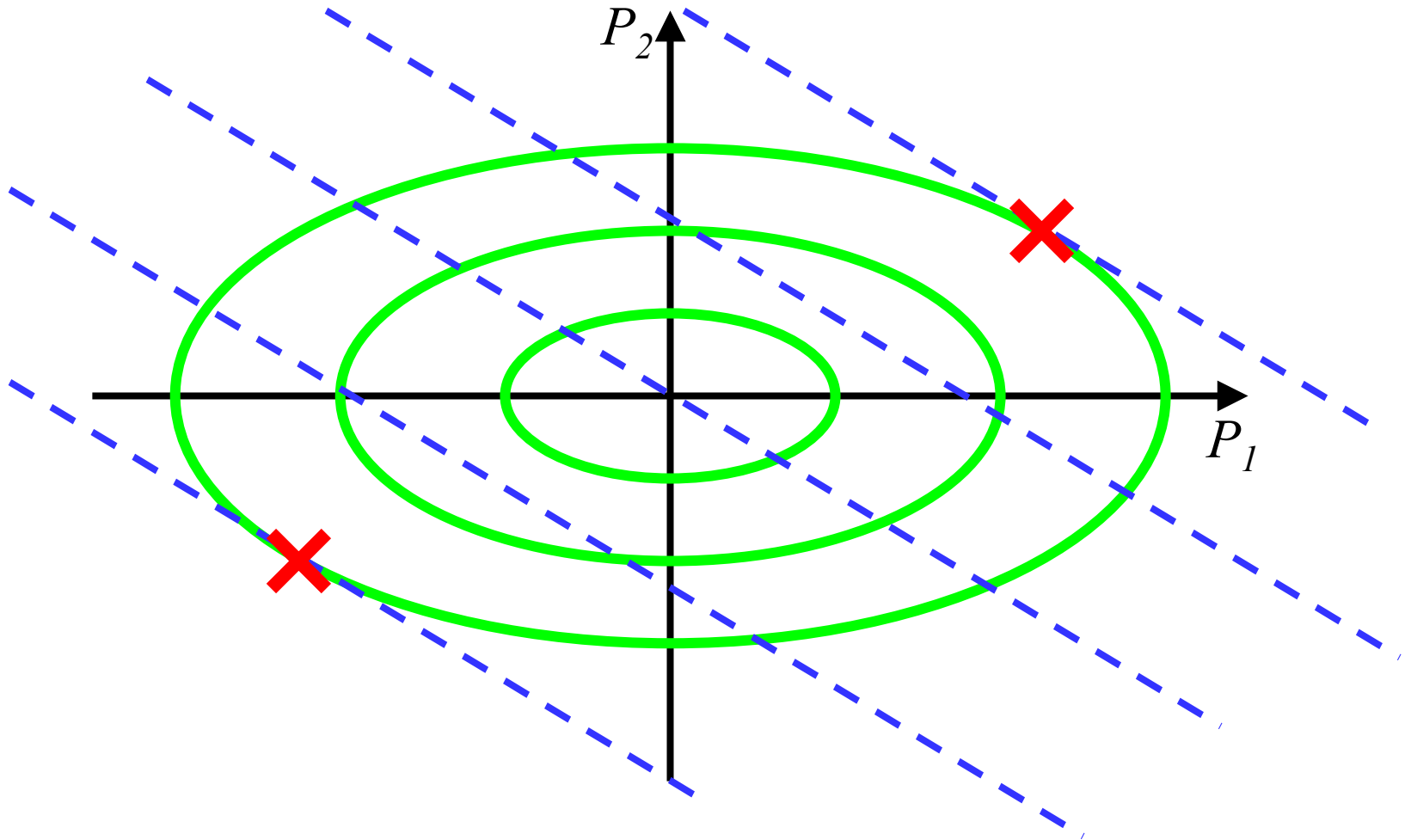
- Variations in physical, independent process parameters P_k cause variations in electrical performances E_i
 - E.g. T_{OX} , Δ_L , ρ_{SH} , N_B , V_{FB} , x_J
 - V_{TH} , β are **NOT** basic process parameters
- SPICE model parameters are characterized as functions of P_k
 - And geometry g if necessary
- Allows effective and efficient modeling of correlations with **uncorrelated** P_k

Statistical Modeling in a Nutshell

$$\delta E_i = \sum_k \frac{\partial E_i}{\partial P_k} \delta P_k$$
$$\delta P_j = \pm n \frac{\sigma_{P_j}^2 \left(\partial E_i / \partial P_j \right)}{\sqrt{\sum_k \sigma_{P_k}^2 \left(\partial E_i / \partial P_k \right)^2}}$$

$$\sigma_{E_i}^2 = \sum_k \left(\frac{\partial E_i}{\partial P_k} \right)^2 \sigma_{P_k}^2$$

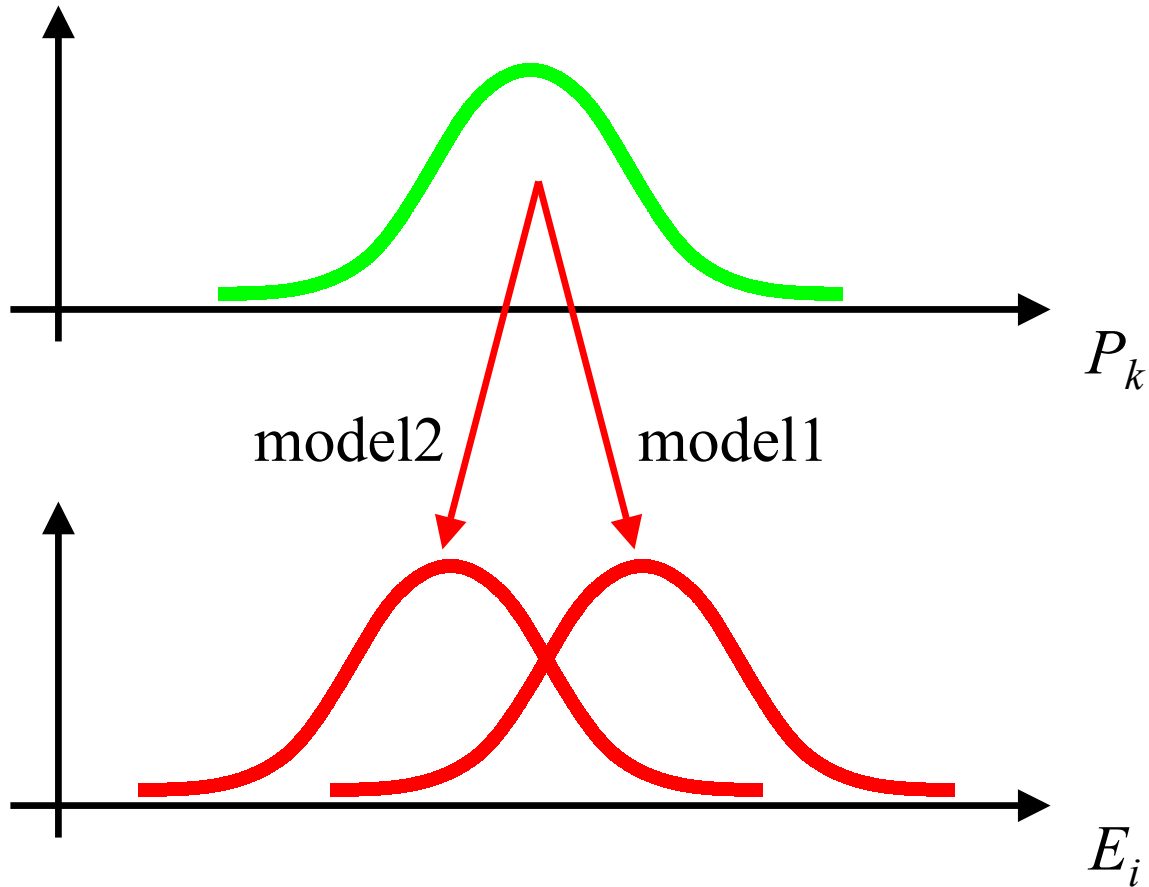
Constant E_i in P -Space



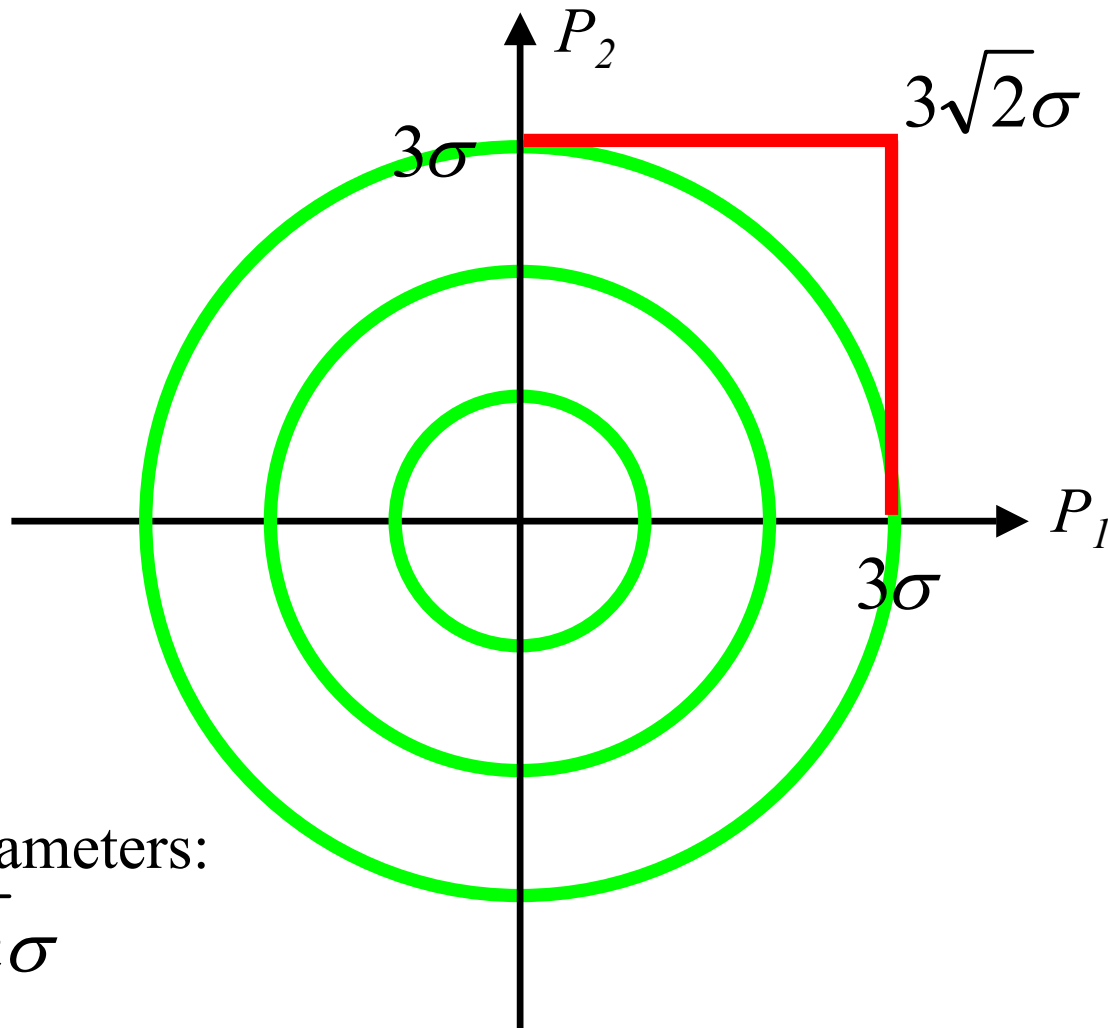
Backward Propagation of Variance

- Obtain statistics for E_i
- Calculate sensitivities from SPICE models
 - To process, not SPICE, parameters
- Solve linear equations to get σ_{P_k}
- Mathematical constraints
 - SPICE models must be reasonable
 - P must be observable in E
 - σ_E must be self-consistent

FPV Error in Modeling E



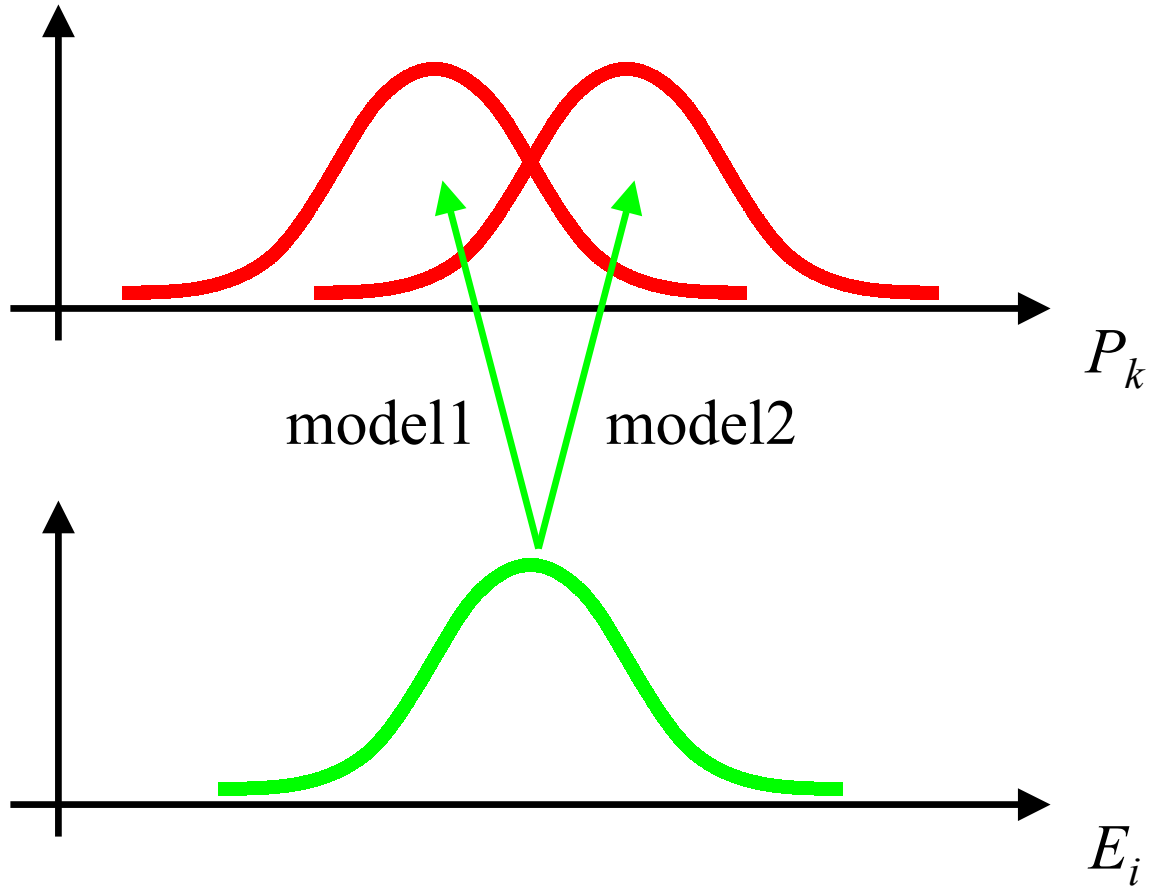
FPV Low Probability



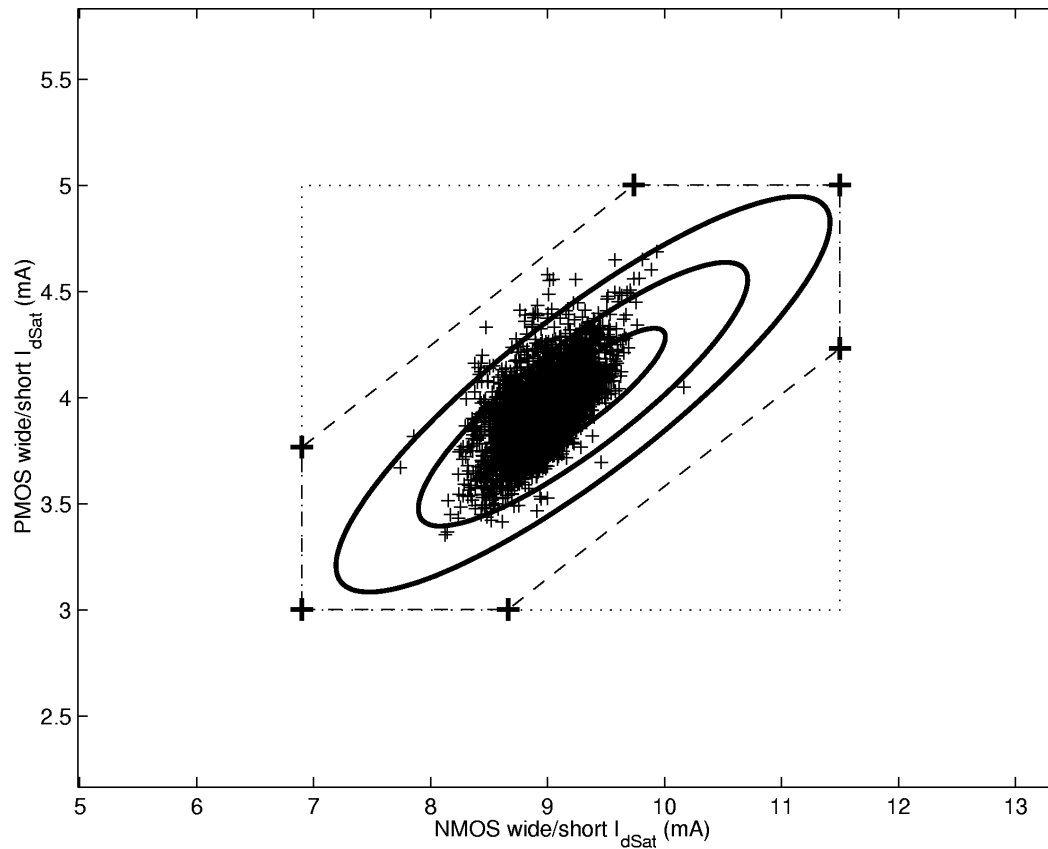
For n parameters:

$$3\sqrt{n}\sigma$$

BPV Process



Statistical Models



Statistics

- There are two sources of variation in semiconductor manufacturing
 - Global variation
 - Local variation
- The BPV analysis does **NOT** distinguish between them
 - Applies to both, target E_i may be different
 - **Identical** modeling/extraction formalism

Global and Local Variation

δP_G	δP_L
Global	Local (mismatch)
Inter-die	Intra-die
Correlated across die	Uncorrelated
Geometry independent	Geometry dependent

$$\sigma_P^2 = \sigma_{P_G}^2 + \sigma_{P_L}^2 (L, W, \dots)$$

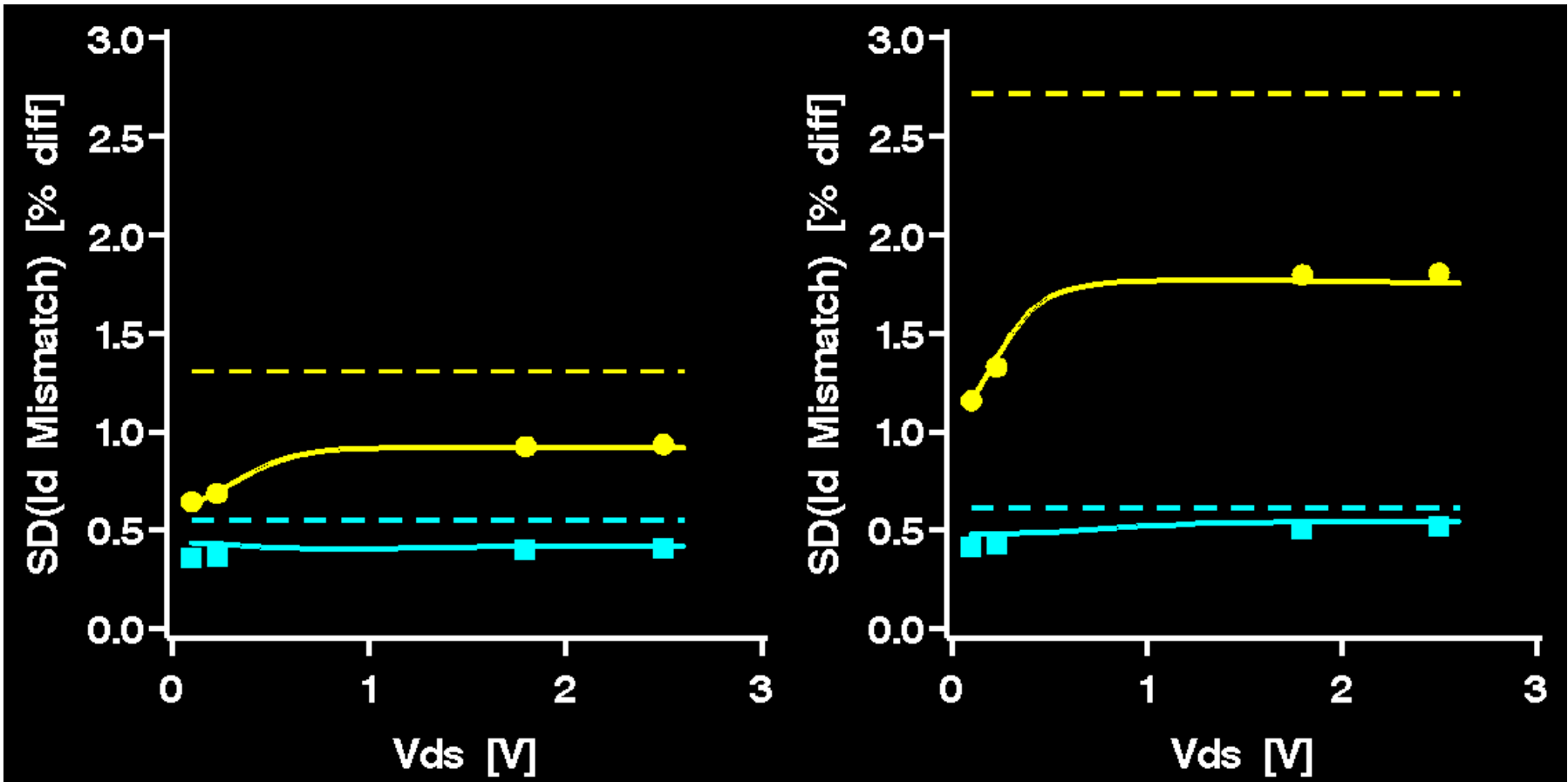
Common Inaccuracies

- V_{TH} and β are considered to be mismatch parameters for MOSFETs
 - Neither is a process parameter
 - T_{OX} variation is double counted
 - Misses effects like body effect mismatch
 - Measured in linear region, but analog circuits mostly operate in saturation
 - No guarantee I_d mismatch is modeled well
 - Inverse area fixation has precluded testing of wide/short and narrow/long devices
 - Some P_L vary as $1/L$ or $1/W$

0.25 μm NMOS Example

$V_{bs}=0$

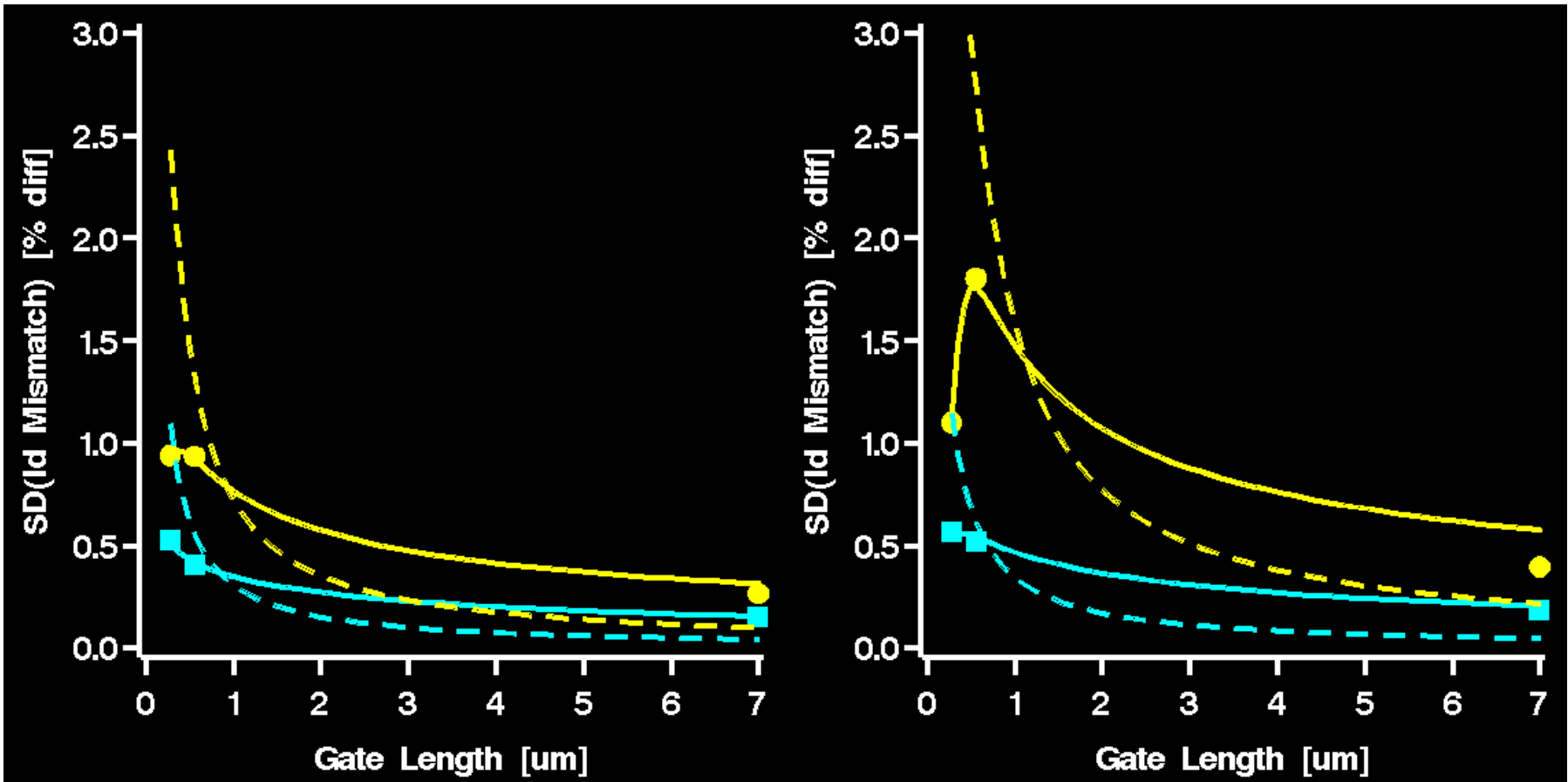
$V_{bs}=-2.5$



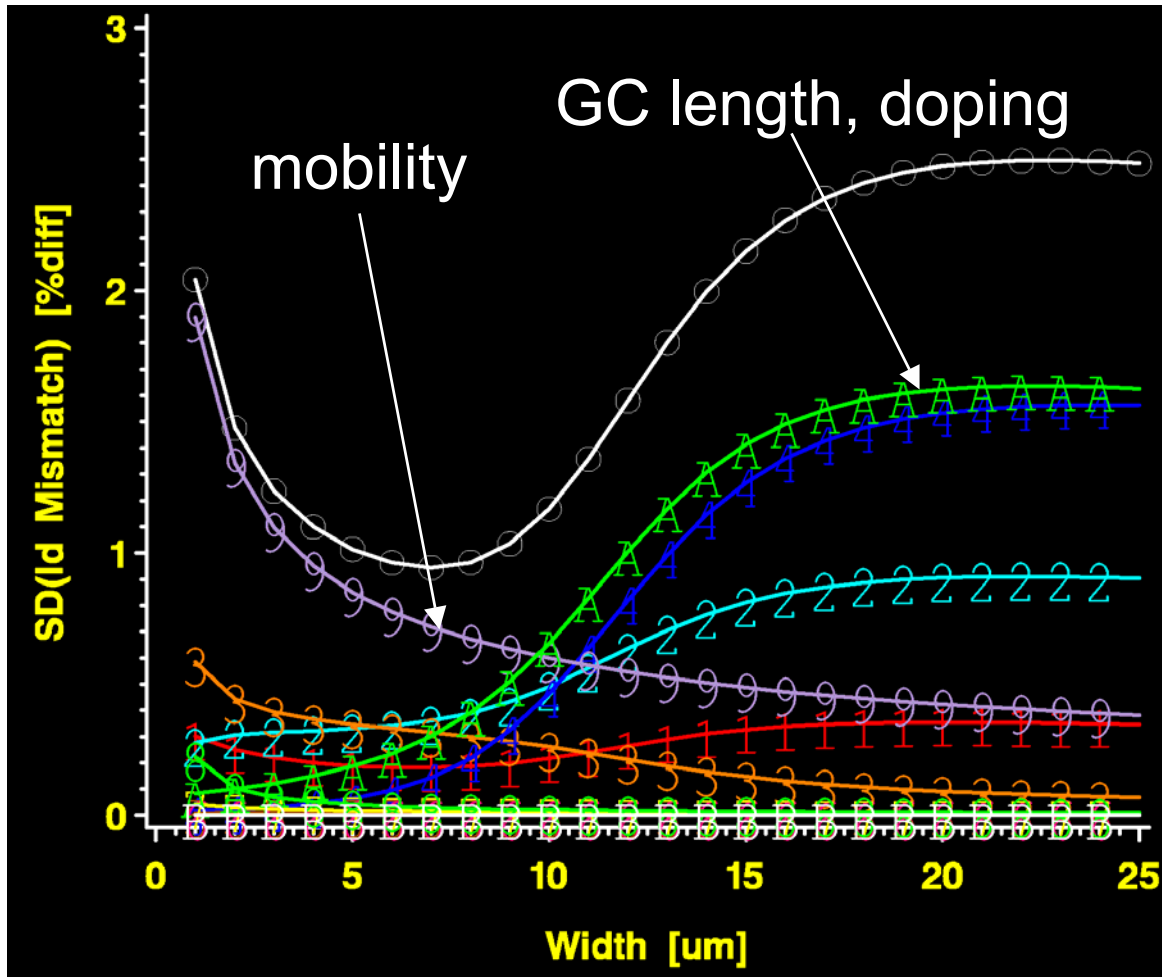
0.25 μm NMOS Example

$V_{bs}=0$

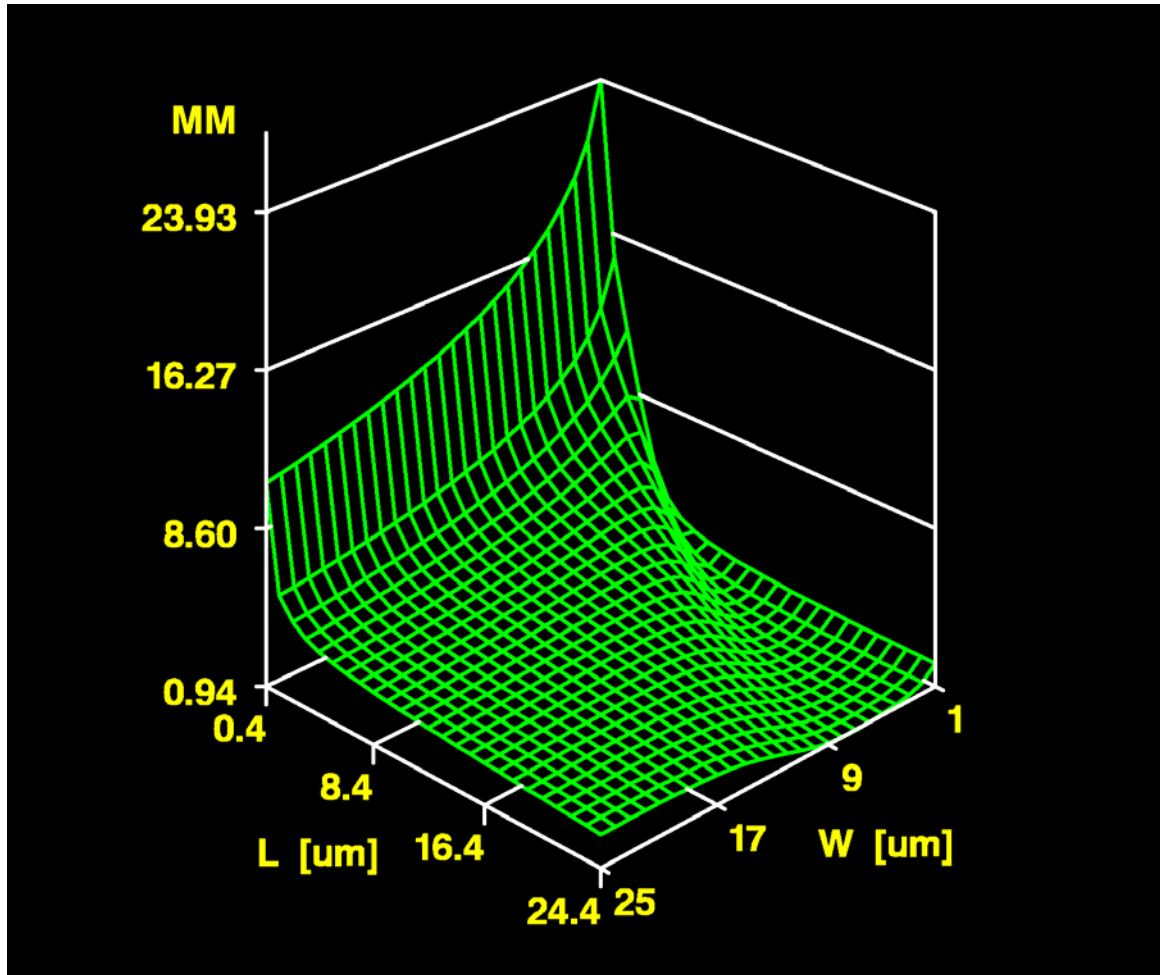
$V_{bs}=-2.5$



Mismatch Components



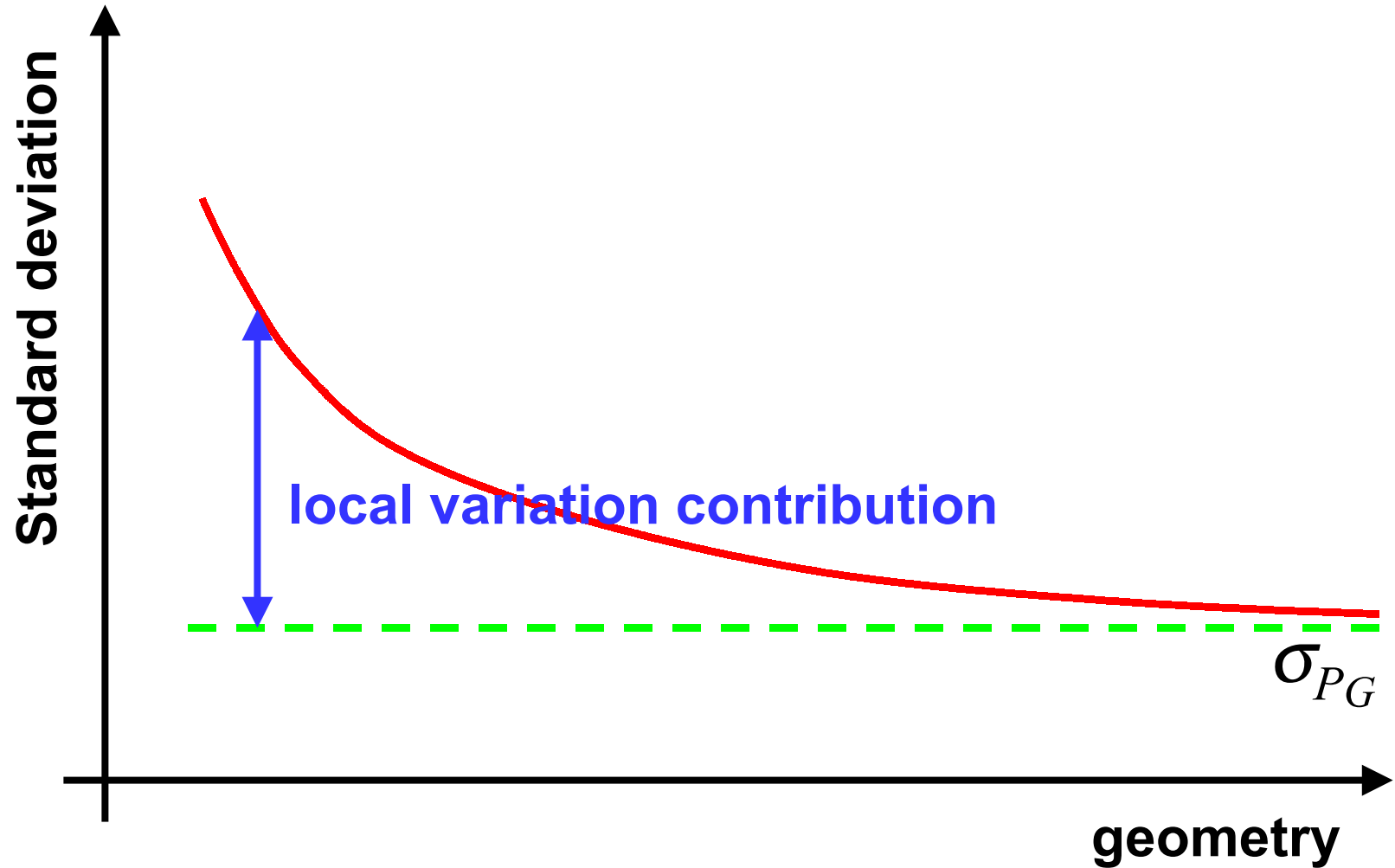
Current Mirror, $10\mu\text{A}$



Issues when σ_P is Assumed to be σ_{PG}

- ET dependence on geometry ignored
- Global statistical simulations do not account for the geometry variation of σ_P
- Local variation can be double counted
- Standard “corner” simulation methodology fails
 - Assumes devices are correlated
 - Becomes very pessimistic

The Wall



Issues with Correlation Viewpoint

- Physically incorrect
- Can miss need to handle >2 devices
- Can miss need for different geometries
- N devices require $N(N+1)/2$ correlations
- Can miss geometry dependence of σ_P
- Cannot handle σ_{P_G} being zero

$$\rho = \frac{\sigma_{P_G}^2}{\sqrt{(\sigma_{P_G}^2 + \sigma_{P_L}^2 (g_1))(\sigma_{P_G}^2 + \sigma_{P_L}^2 (g_2))}}$$

Summary

- **Statistical modeling should be based on physical process parameters**
- **Process parameters can always be formulated to be uncorrelated**
 - Yet still model correlated E_i accurately
- **This plus BPV process gives a unified methodology that is identical for global and local variation**
- **Substantially improves accuracy**