



**Integrated Systems Engineering**  
Development, Modeling, and Optimization of Microelectronic  
Processes, Devices, Circuits, and Systems

# Automatic BSIM3/4 Model Parameter Extraction with Penalty Function

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# Outline

- Introduction
- Extraction Approach
- Objective Function
- Penalty Functions
- Auto-Selection of Binning Parameters
- Experimental Results
- Summary

# Introduction

- Difficulties of Traditional Extraction Approach
  - Poor convergence
  - Complicated extraction strategy
  - Need human interactions
  - Time consuming
- Solution: Extraction with optimization techniques

# Extraction Approach

- Reliable optimization algorithm
  - Quasi-newton
- Objective function with penalties
- Sequence of extraction steps
  - Initial guess
  - Full optimization
  - Auto-selection of binning parameters procedure

# Objective Function

$$F(X) = \sum_{\text{devices}} \sum_{\text{curves}} \left\{ \left( \frac{Id_{meas} - Id_{sim}}{Id_{meas}} \right)^2 + \Psi(P) \right\}$$

## Penalty Function

$$\Psi(P) = \begin{cases} 0 & P \in \text{feasible space} \\ \sum_i \psi(P_i) & P \notin \text{feasible space} \end{cases}$$

$P_i$  – parameter or intrinsic variable of the model

Penalty function guarantees physically correct parameters.

# Example of Penalty Function

Mobility model for mobMod=2 (BSIM3)

$$\mu_{eff} = \frac{\mu_0}{1 + P_\mu}$$

$$P_\mu = \left( U_A + U_C \cdot Vbs_{eff} \right) \left( \frac{Vgs_{eff}}{Tox} \right) + U_B \cdot \left( \frac{Vgs_{eff}}{Tox} \right)^2$$

$P_\mu$  - Function of more than 100 model parameters

BSIM3 Constraint

$$P_\mu > -0.8$$

Penalty Function

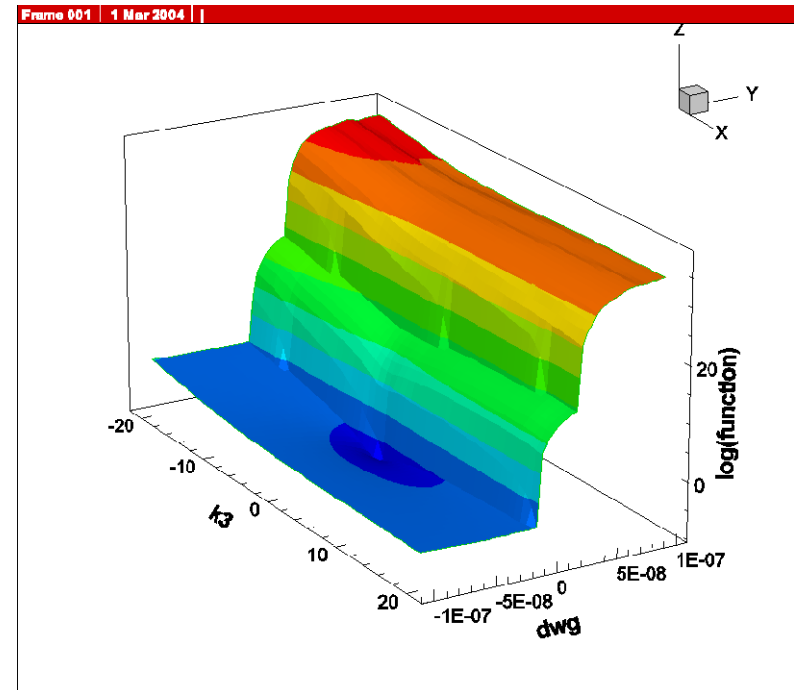
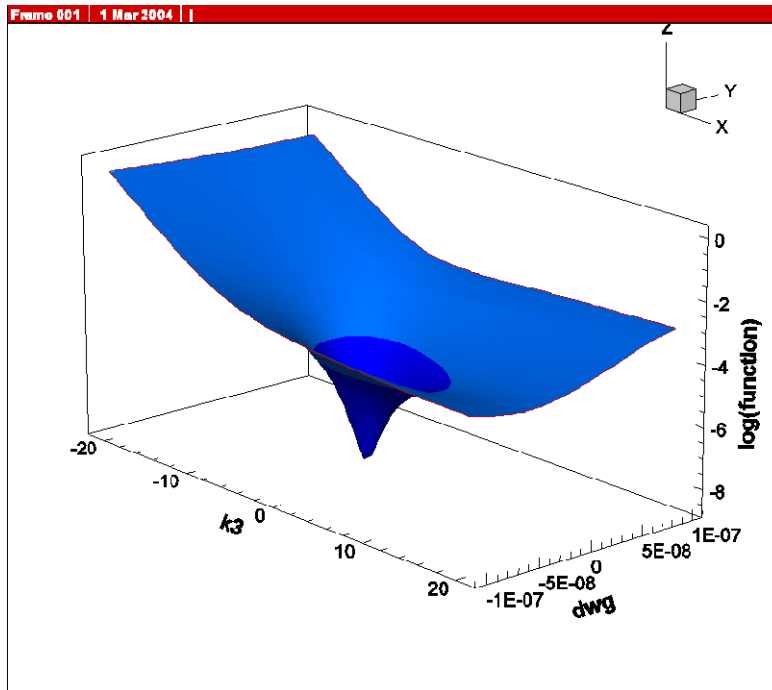
$$\psi(P_\mu) = \begin{cases} 0 & P_\mu > -0.8 \\ (0.8 + P_\mu)^2 & P_\mu \leq -0.8 \end{cases}$$

# Objective Function Surface Plot, BSIM3

## k3-dwg Surface Plot

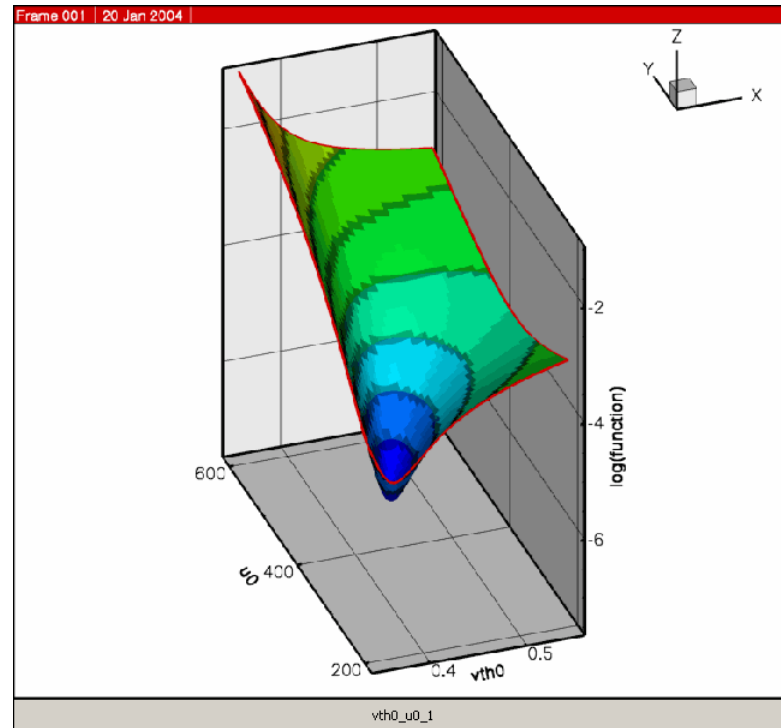
No penalties

With Penalties



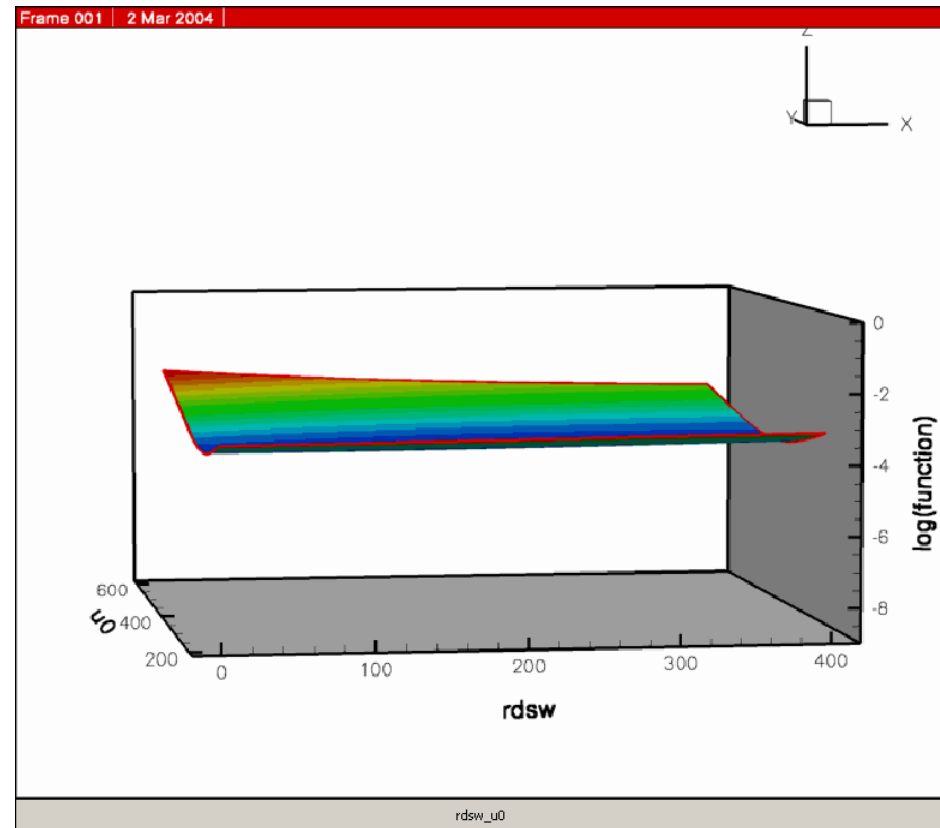
# Objective Function Surface Plot, BSIM3

- vth0-u0 Surface Plot Animation



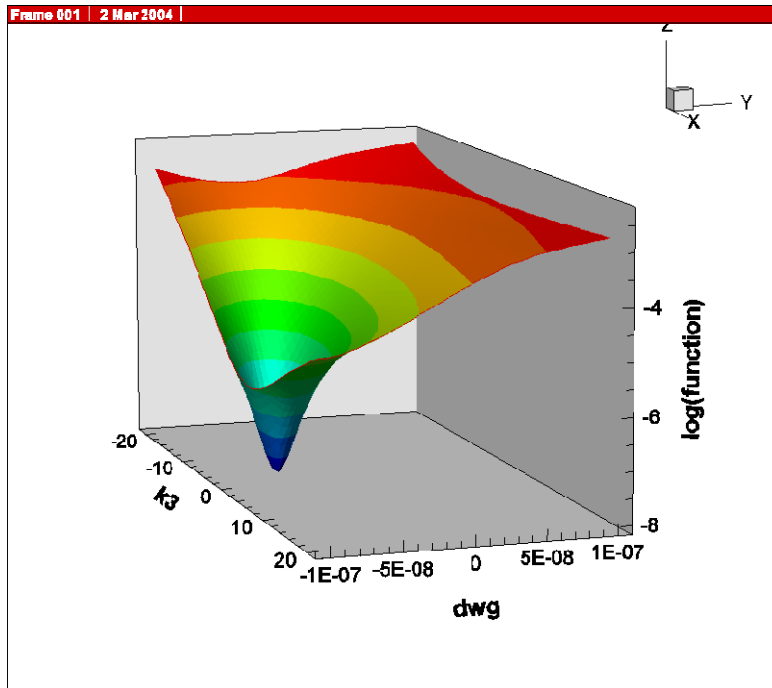
# Objective Function Surface Plot, BSIM3

- rdsu-u0 Surface Plot Animation

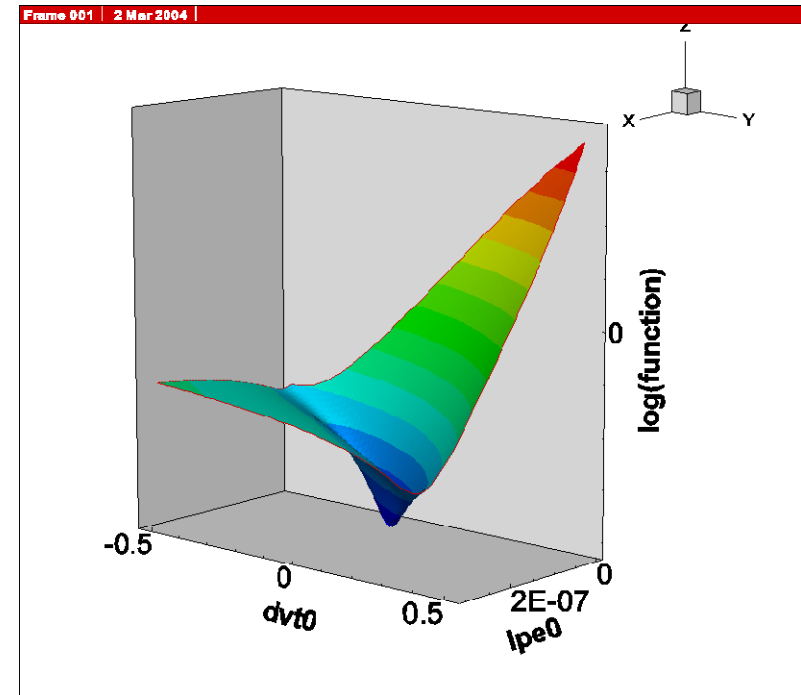


# Objective Function Surface Plot, BSIM4

## k3-dwg Surface Plot



## dvt0-lpe0 Surface Plot

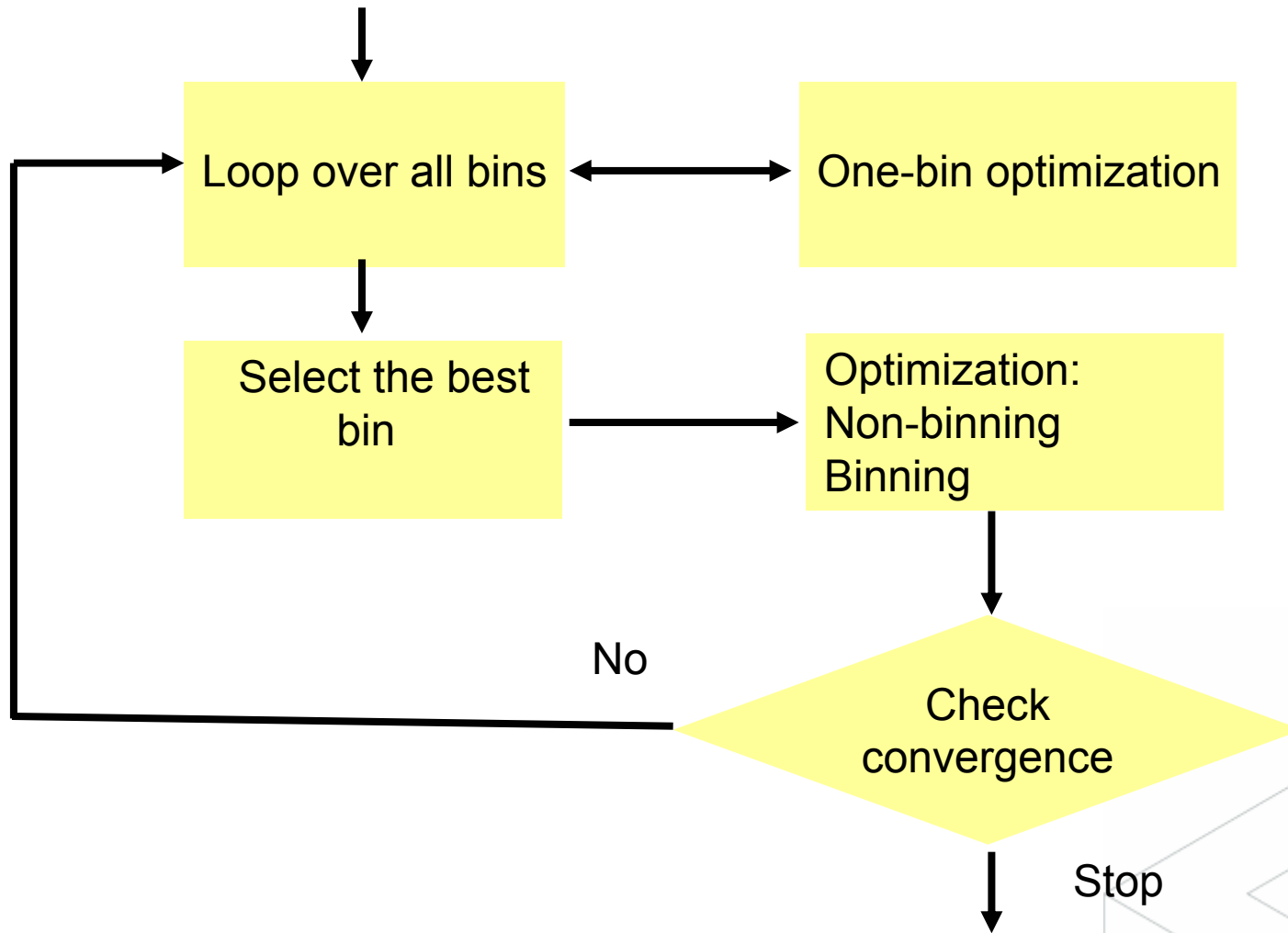


# Conclusions from the Surface Plots

In our implementation of objective function calculation

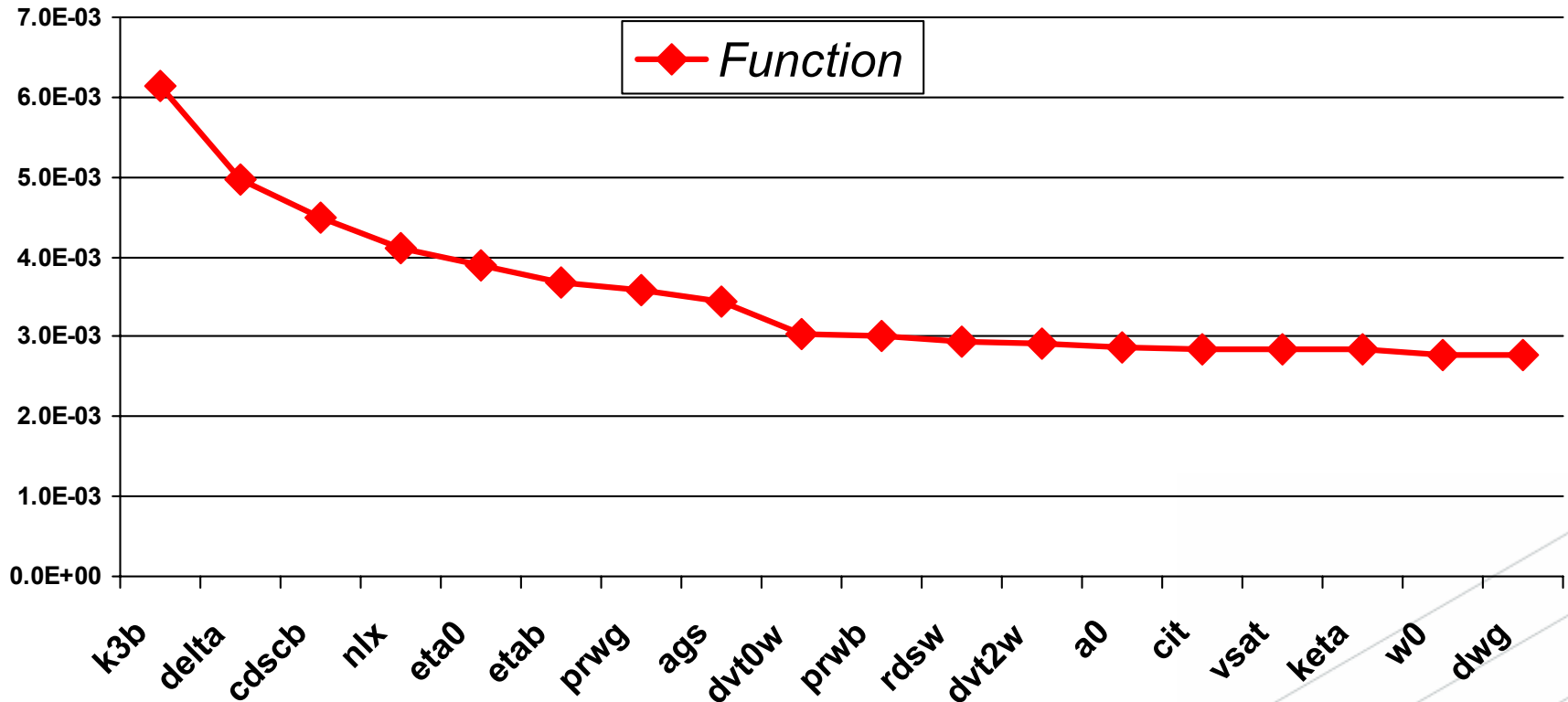
- Objective Function – smooth
- No multiextremum behavior
- Gradient algorithms of optimization – preferable
- Random search algorithms – not necessary

# Auto-Selection of Binning Parameters



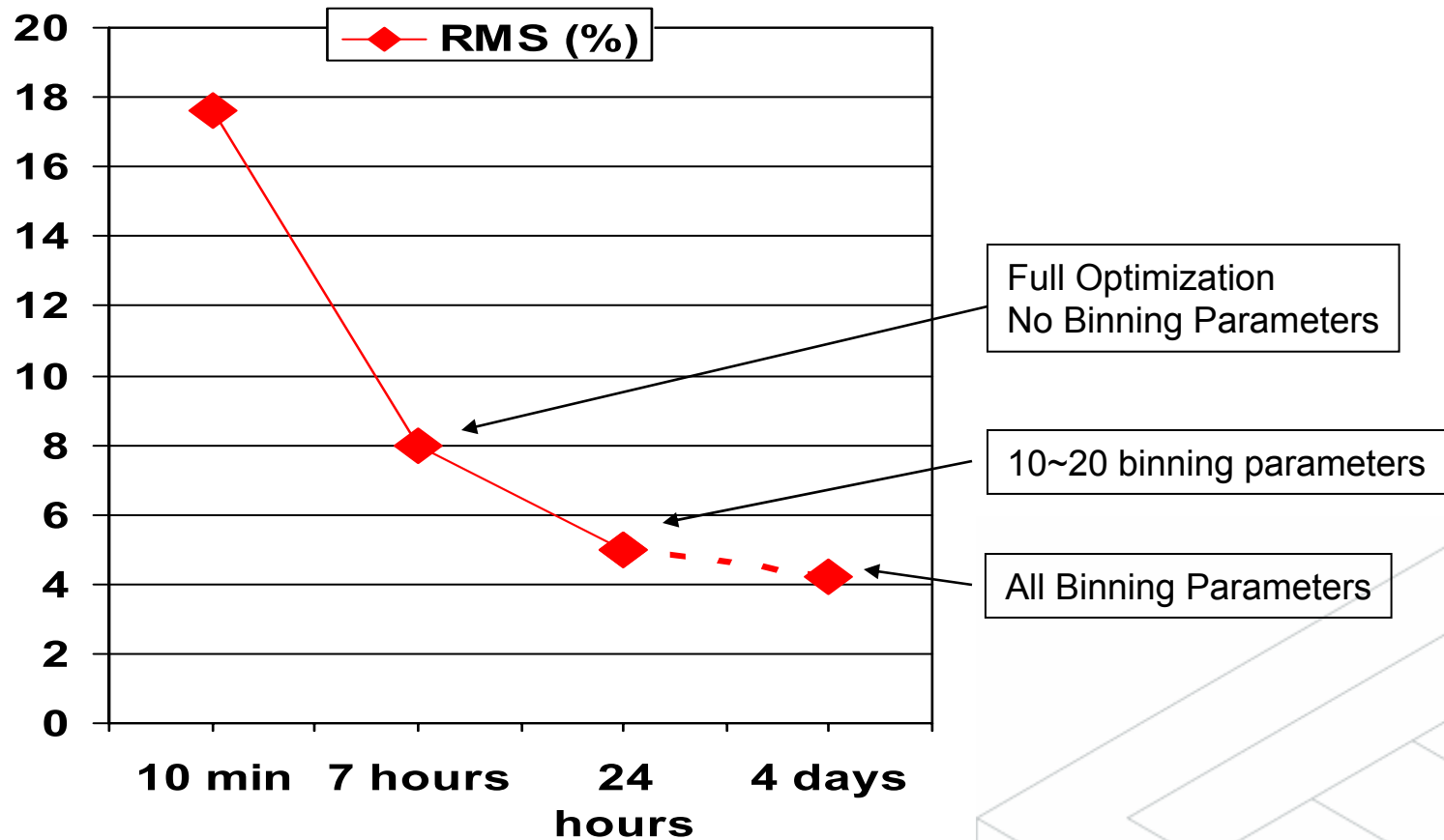
# Auto-Selection of Binning Parameters

## Example: Objective Function Evolution



# Auto-Selection of Binning Parameters

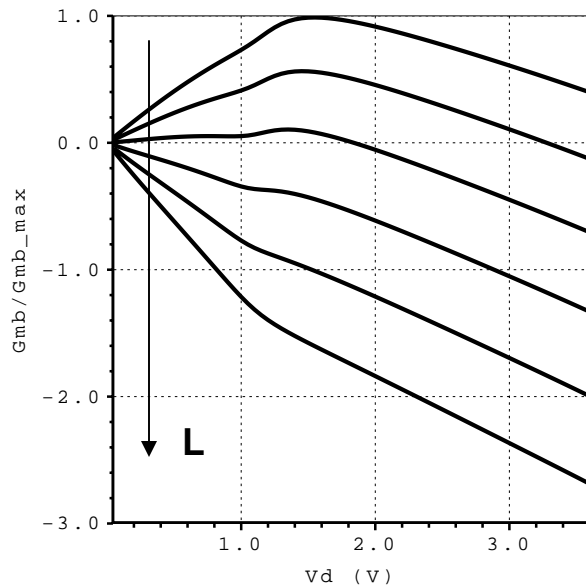
## RMS Error vs. Time Consumption, 1 CPU



# Experimental Results

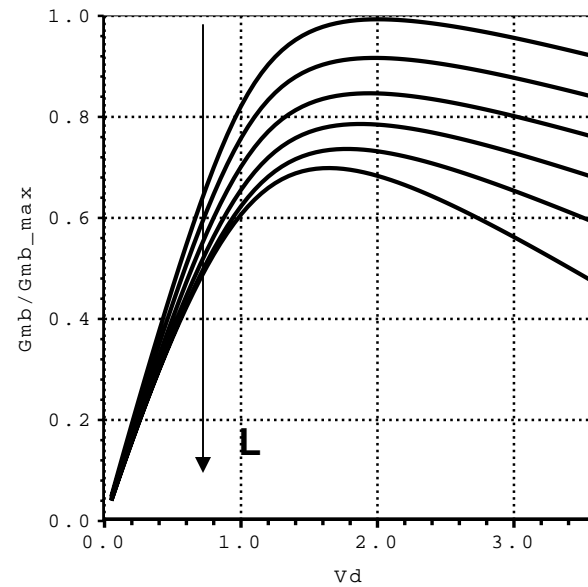
- BSIM3 model. HV 0.13 technology
- $L = 0.34, 0.32, 0.30, 0.28, 0.26, 0.24$

$G_{mb} < 0$ , old results



**No  $G_{mb}$  penalty**

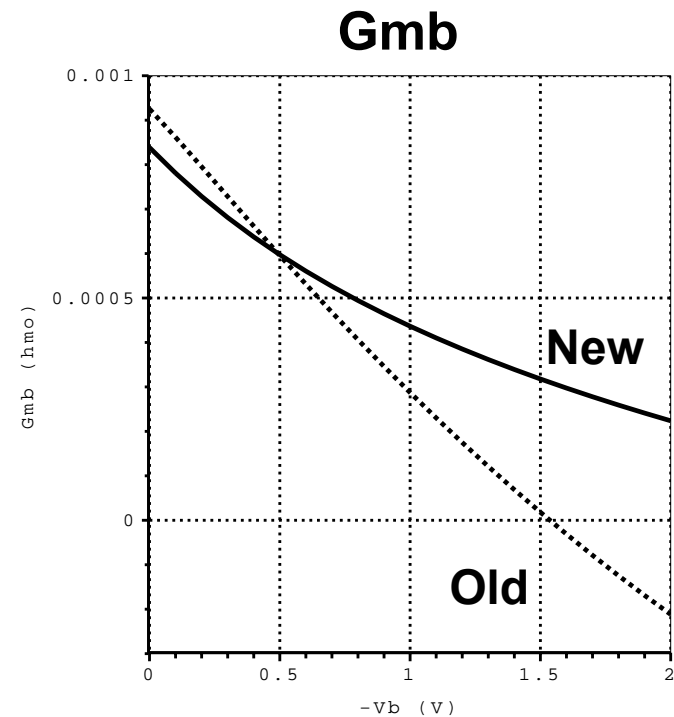
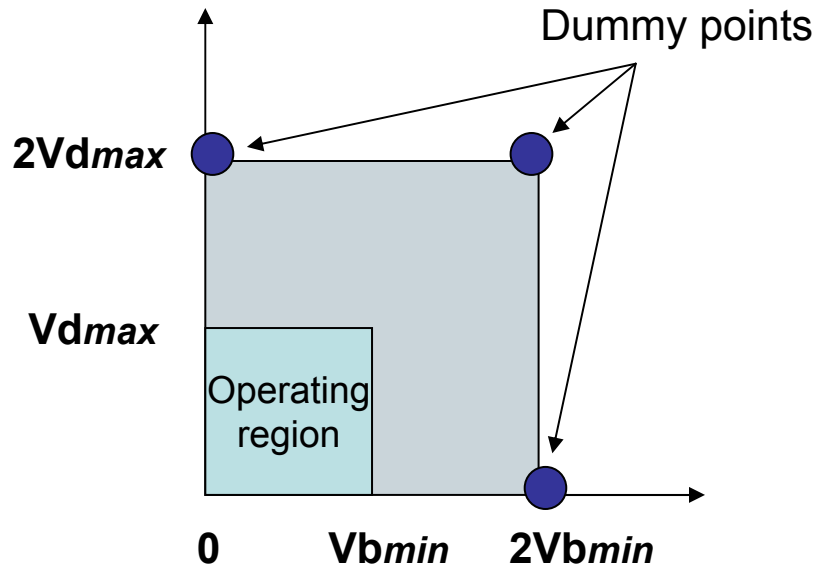
$G_{mb} > 0$ , new results



**$G_{mb}$  penalty turn on**

# Experimental Results

- BSIM4 model. 0.09 technology. Small device.



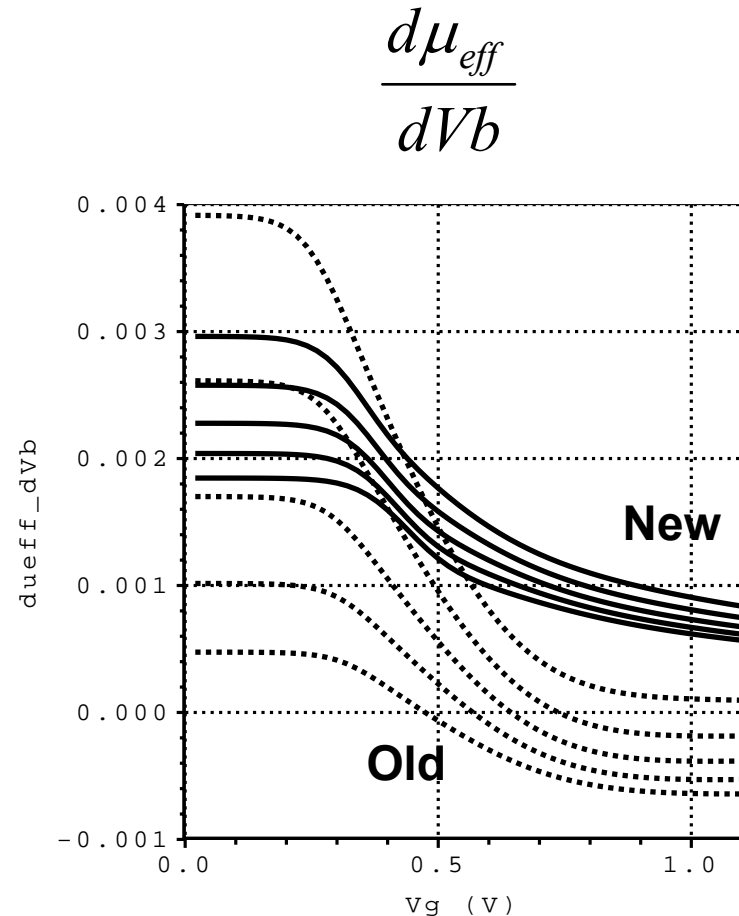
- Dummy Points
  - Do not optimize drain current
  - Keep under control the penalty functions
- $\Rightarrow$   $G_{mb}$  is positive within twice the range

# Experimental Results

BSIM4 model.

0.09 technology. Small device.

- No penalty -  $\frac{d\mu_{eff}}{dVb}$  may change sign
- Penalty for  $\frac{d\mu_{eff}}{dVb}$  is turned on -  $\frac{d\mu_{eff}}{dVb}$  always positive



# Summary

- Objective function with penalties demonstrates smooth behavior
- Gradient optimization algorithm is preferable
- Auto-selection of binning parameters helps to find the needed bins
- Model parameters within physical range is guaranteed by penalty functions
- No human interactions
- Reliable model card