

# Advances in Statistical Compact Modeling

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Dr. Volker Glöckel,  
MunEDA



Dr. Elmar Gondro,  
Infineon Technologies



## Overview

- Statistical Compact Modeling and use model
- Method is available as GUI based MunEDA WiCkeD App “Statistical Fit”
- Data normalization for fitting
- Excluding mismatch effects by calculation from PCM measurements
- Helpful input data consistency checks
- Validation of results

## Task and implementation idea

### Target User Group

- CAD Modeling & Characterization engineers

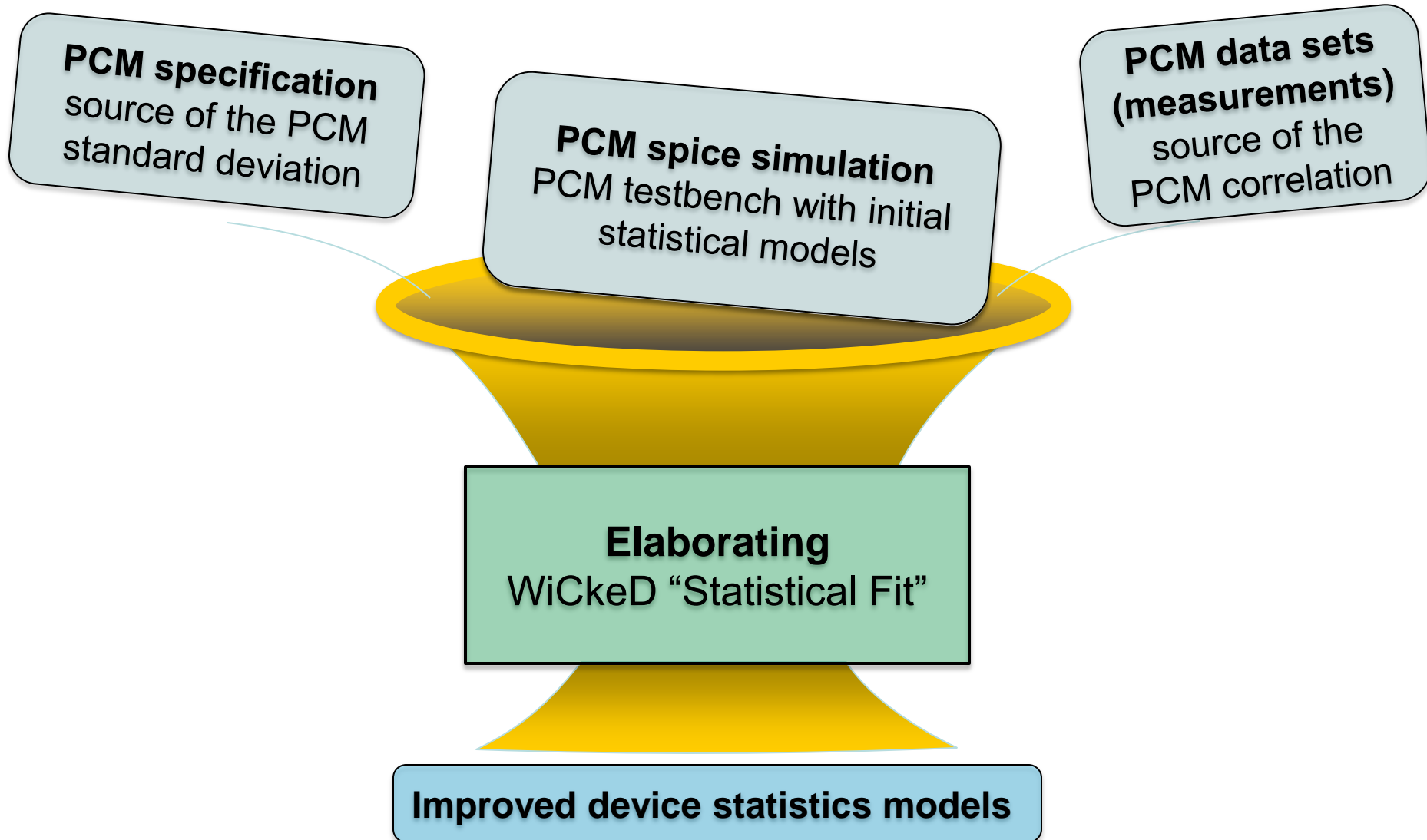
### Task

- Improve consistency of statistical spice models according to variations and correlations in order to ...
- Improve the variation of analog device behavior in statistical (Monte Carlo) simulations

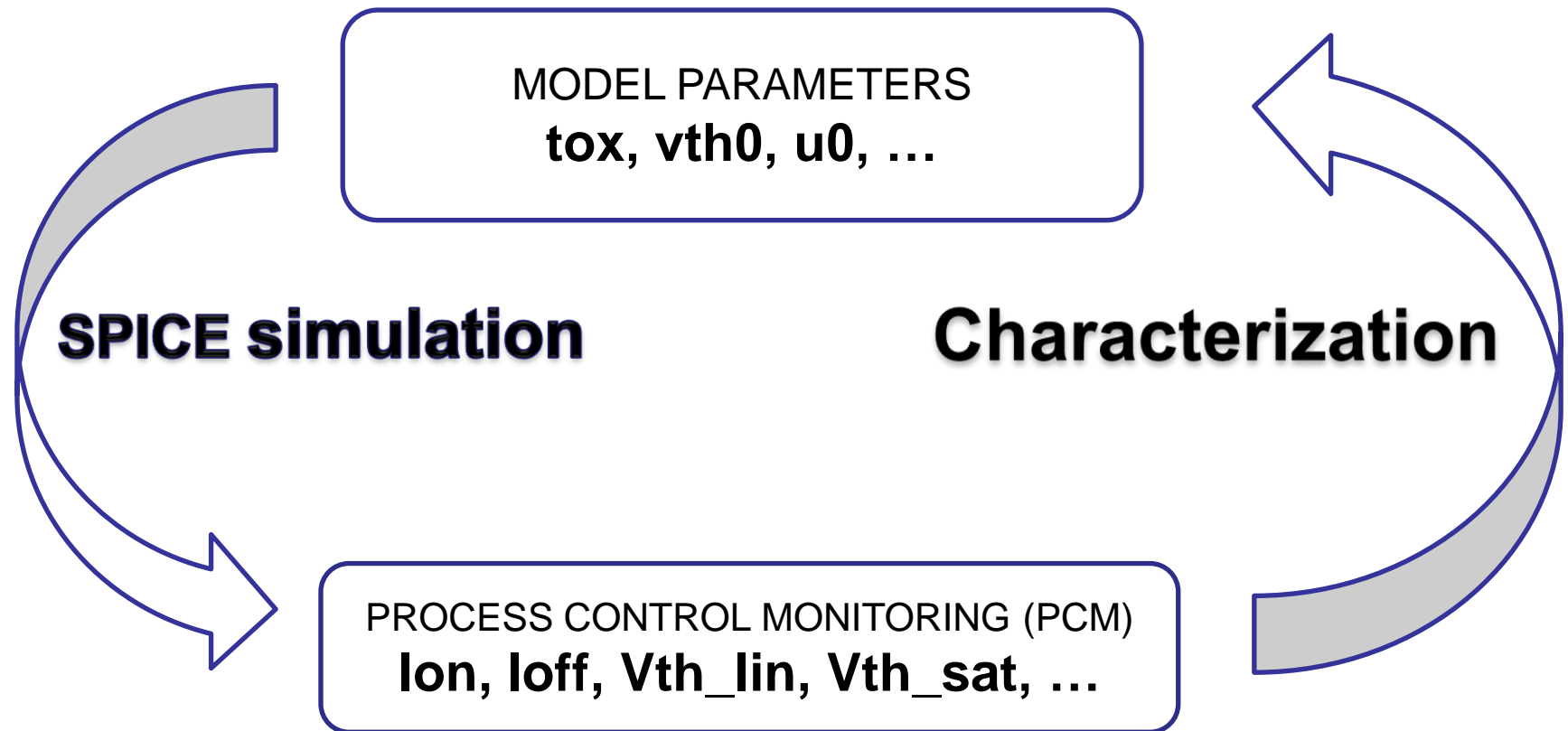
### Implementation idea

- Introducing correlations with statistical spice model card parameters
- The standard deviation is taken from PCM specification
- The correlations between PCMs can be extracted from PCM measurements

## General use model



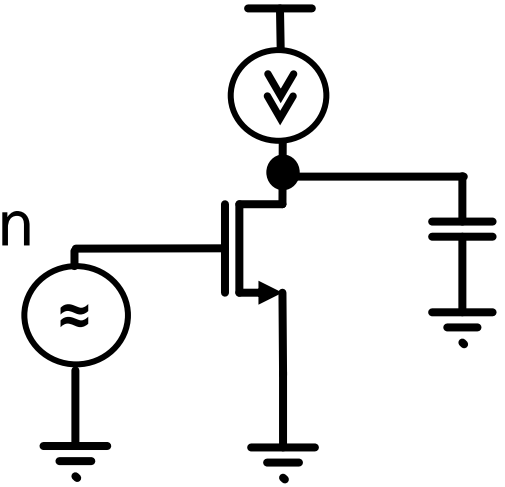
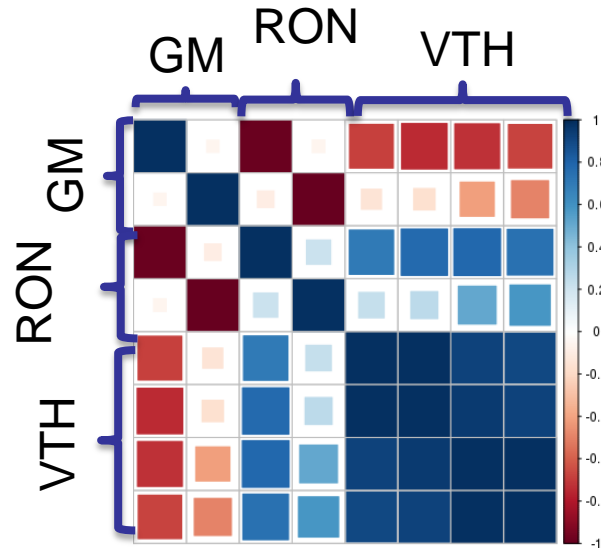
## Relation between model parameters and PCM re-simulation



Correlations between model parameters ( $v_{th0}$ ,  $u_0$ ,  $t_{ox}$ , ...) are necessary to reproduce correct variance and correlation in process MC for analog circuits

## Why is the correlation between PCMs so important for analog design?

- Common source single stage amplifier
- Low frequency gain:  $A_v = G_m \cdot R_{on}$
- There is strong negative correlation between  $G_m$  and  $R_{on}$ :



- The MC model must reproduce this correlation in order to correctly simulate the variance of the gain

## Typical problem with modeling of the correlation of PCMs

### ➤ Three typical model parameters for process MC:

- tox – oxide thickness
- u0 – mobility
- vth0 – threshold voltage

$$I_{ds} \approx u0/tox \cdot (V_{gs} - vth0 - V_{ds}/2)$$

$$GM \approx u0/tox \quad VTH\_LIN \approx vth0$$

### ➤ Consider two PCMs from the same device:

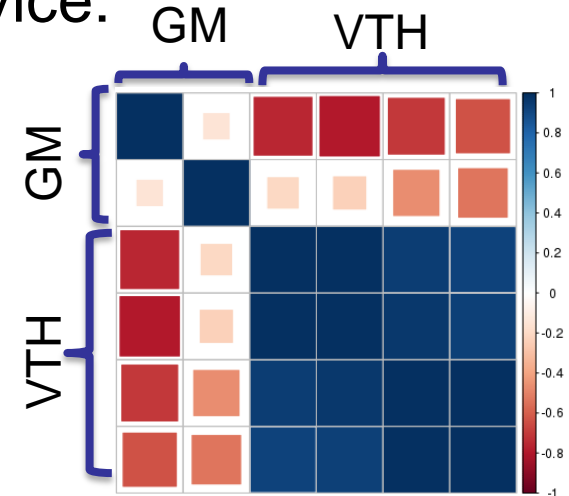
- GM – maximum gm
- VTH\_LIN – extrapolated Vth

### ➤ The variance of Gm:

- $\approx 60\%$  dependence on u0
- $\approx 40\%$  dependence on tox

### ➤ The variance of VTH\_LIN:

- $\approx 100\%$  dependence on vth0



**tox, u0 and vth0 must be correlated to see a correlation of GM and VTH\_LIN in simulation**



## Linear dependence of PCM variation on Model parameter variations

- Assumption of a linear dependency between PCM and model parameters

$$\Delta P = S \cdot \Delta M$$

- $\Delta P$  – variation of PCM
- $S$  – sensitivity of PCMs w.r.t. model parameters  $M$  (by simulation)
- $\Delta M$  – variation of model parameters

$$\text{cov}(P) = S \cdot \text{cov}(M) \cdot S^T$$

- Least-Squares fit:

$$\min_X \left\| \underbrace{\text{cov}(P)}_{\substack{\uparrow \\ \text{measured}}} - \underbrace{S \cdot X \cdot S^T}_{\substack{\uparrow \\ \text{simulated}}} \right\|^2$$

## Covariance matrix of PCM

- The variances of target PCMs are defined by specification:
  - $\sigma(P)$  are defined from technology , e.g.  $(USL-LSL)/9$
  
- The correlation of PCMs are measured by the FAB:
  - $corr(P)$  are calculated from wafer
  
- Covariance matrix of PCM are constructed from variances and correlations of PCMs:

$$cov(P) = \sigma(P) \cdot corr(P) \cdot \sigma(P)$$

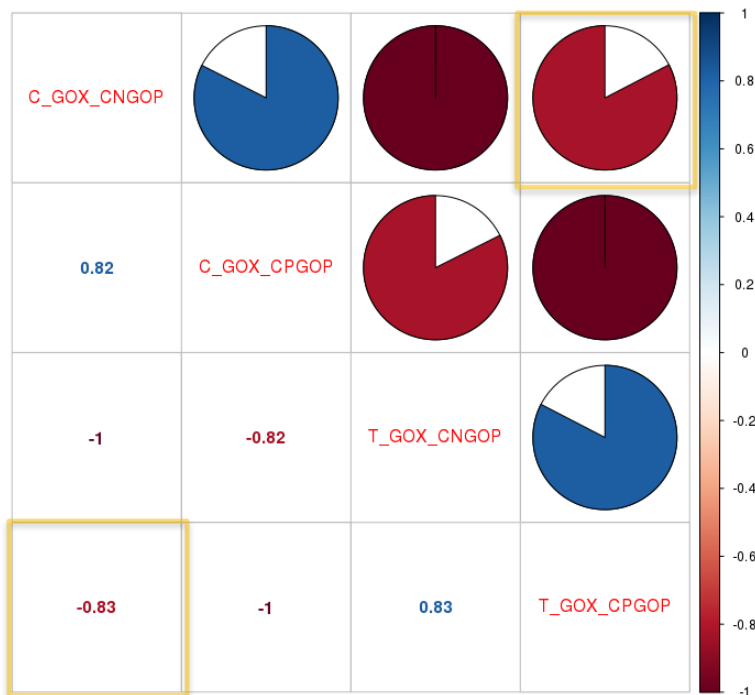
# Capacitor example

➤ E.g. two different types of capacitors:

– Modelled with two parameters and one correlation factor

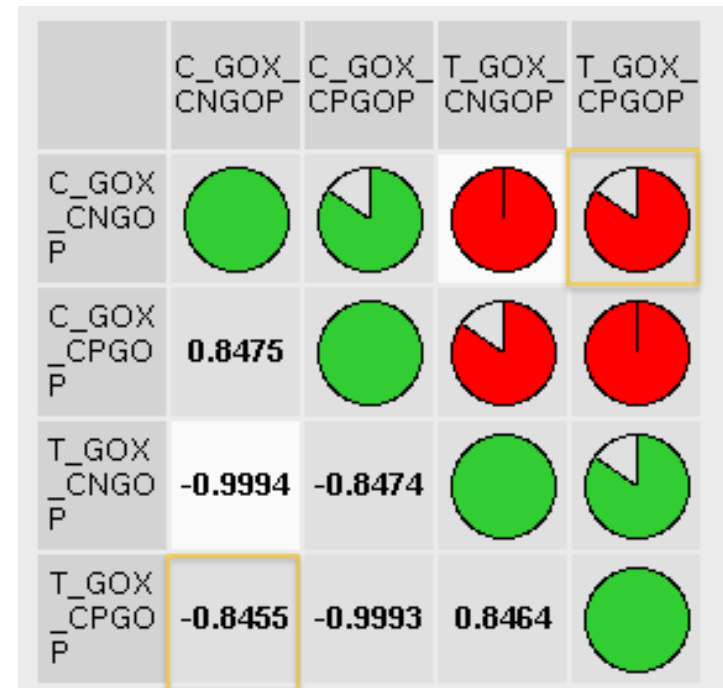
- toxn
  - toxp
- } cc\_toxn\_toxp

Correlation matrix of PCM measurements



**cc(T\_GOX\_CPGOP,C\_GOX\_CNGOP)=-0.83**

Simulation by Monte-Carlo



**cc(T\_GOX\_CPGOP,C\_GOX\_CNGOP)=-0.85**

## MunEDA WiCkeD App “Statistical Fit”

- GUI tool to create Statistical Models considering PCM correlations for analog transistor behavior that reproduce long-term process variation
- MunEDA WiCkeD is applied to the PCM netlists
  - Simulation setup with initial variation model of spice models
  - Selection of statistical model card parameters to consider
- Additional inputs of “Statistical Fit”
  - PCM measurements (data sets) for correlations target
  - PCM specification to determine the standard deviation of the statistical model card parameters as well
- Results of “Statistical Fit”
  - Set of model parameter sigmas (global MC parameters) and
  - correlations between them
- Benefit: More realistic sigmas / correlations of MC parameters
  - Safer verification
  - Less pessimism in design → better utilization of process technology

## Screenshot of the WiCkeD App “Statistical Fit”

Selection of file with target sigmas

Selection of strategy

Selection of PCM data file or Correlation file

Mapping PCM name to measurement

Start Analysis (Node 2) <1>

PCM targets file (PCM LSL TARGET USL [WEIGHT])

/PCM\_targets.dat Browse

Method

- Full Correlation Matrix
- Uncorrelated Model Parameters
- Uncorrelated Model Parameters ignore PCM correlation
- Block Correlation Structure
- Arbitrary Correlation Structure

Use PCM Samples or Correlation file

- Samples file /PCM\_samples.csv Browse
- Correlation file PCM\_correlations.dat Browse

PCM to performance mapping (PCM "performance name")

Use mapping file /PCM\_to\_sim.dat Browse

Cancel Execute

## Nonlinear PCM with respect to the statistical model parameters

- Assumption of a linear dependency between PCM and model parameters

$$\Delta P = S \cdot \Delta M$$

$\Delta P$  variation of PCM

$S$  sensitivity of PCMs w.r.t. model parameters  $M$  (by simulation)

$\Delta M$  variation of model parameters

- Sample PCMs are not linear; transform them to alternative:
- COX – measurements of capacitance  $\Rightarrow$  TOX – oxide thickness
  - RON – measurements of on-resistance  $\Rightarrow$  GON – on-conductance
  - ILEAK – leakage current  $\Rightarrow$  log(ILEAK) – logarithm of ILEAK

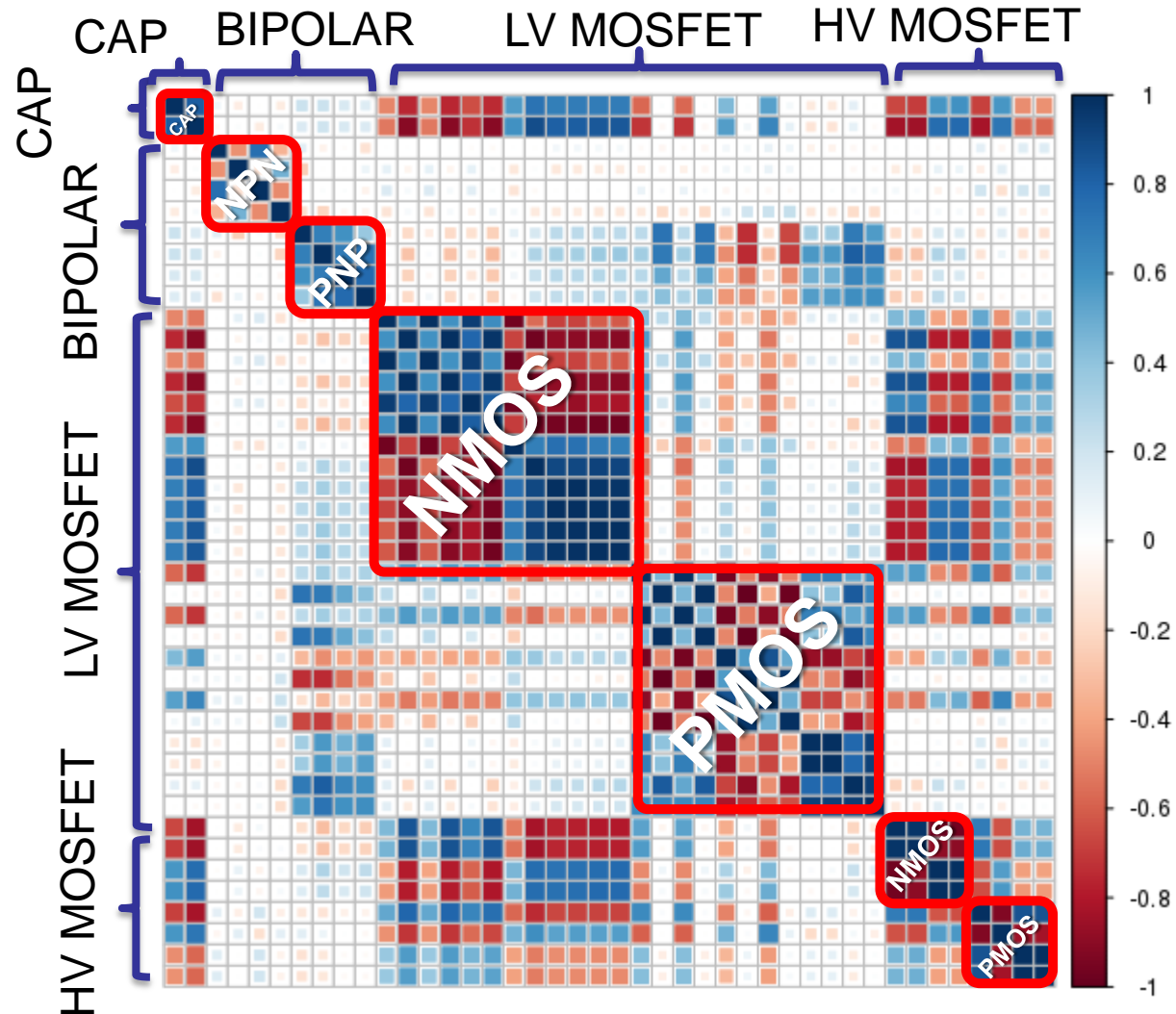
- In order to include second order nonlinearities into the dependence of PCMs on parameter variation, consider, “quadratic BPV” \*

\* I. Stevanovic, C.C.McAndrew, „Quadratic Backward Propagation of Variance for Nonlinear Statistical Circuit Modeling“, IEEE on CAD, V.28, No.9, 2009

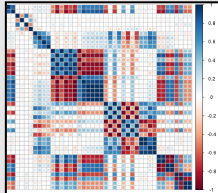

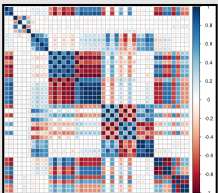
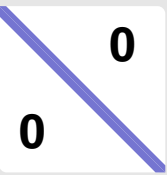

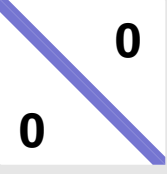
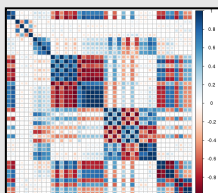
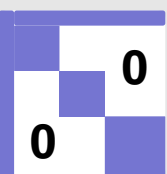
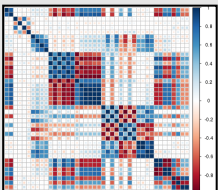
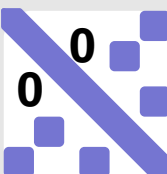
## Sample structure of correlation matrix of PCMs (Measurement)

Typical correlation matrix of PCMs (Long-term data – 3 years of meas)

- DEVICES
- CAPACITOR
  - Different types
- BIPOLAR
  - NPN
  - PNP
- Low Voltage MOSFET
  - NMOS (diff geom)
  - PMOS (diff geom)
- High Voltage MOSFET
  - NMOS (diff geom)
  - PMOS (diff geom)



# Strategies for Statistical Compact Modeling

Method	Purpose	Matrix of PCMs	Matrix of Model Parameters
<b>Full Correlation Matrix</b>	Mathematical approach without physical distinguishes of model parameters	Full Covariance 	Full Covariance 
<b>Uncorrelated Model Parameters</b>	Pay attention on correlation of PCMs, but produce uncorrelated model parameters matrix	Full Covariance 	Only diagonal elements 
<b>Uncorrelated Model Parameters ignore PCM correlation</b>	Unknown correlation between PCMs, preliminary estimation of model parameters' sigmas	Only diagonal elements 	Only diagonal elements 
<b>Block Correlation Structure</b>	Known correlation structure of model parameters, take into account correlation between PCMs	Full Covariance 	Block-wise model parameters 
<b>Arbitrary Correlation Structure</b>	Know correlation structure of model parameters, but not block wise	Full Covariance 	Arbitrary structure 



## Normalization for least square fit

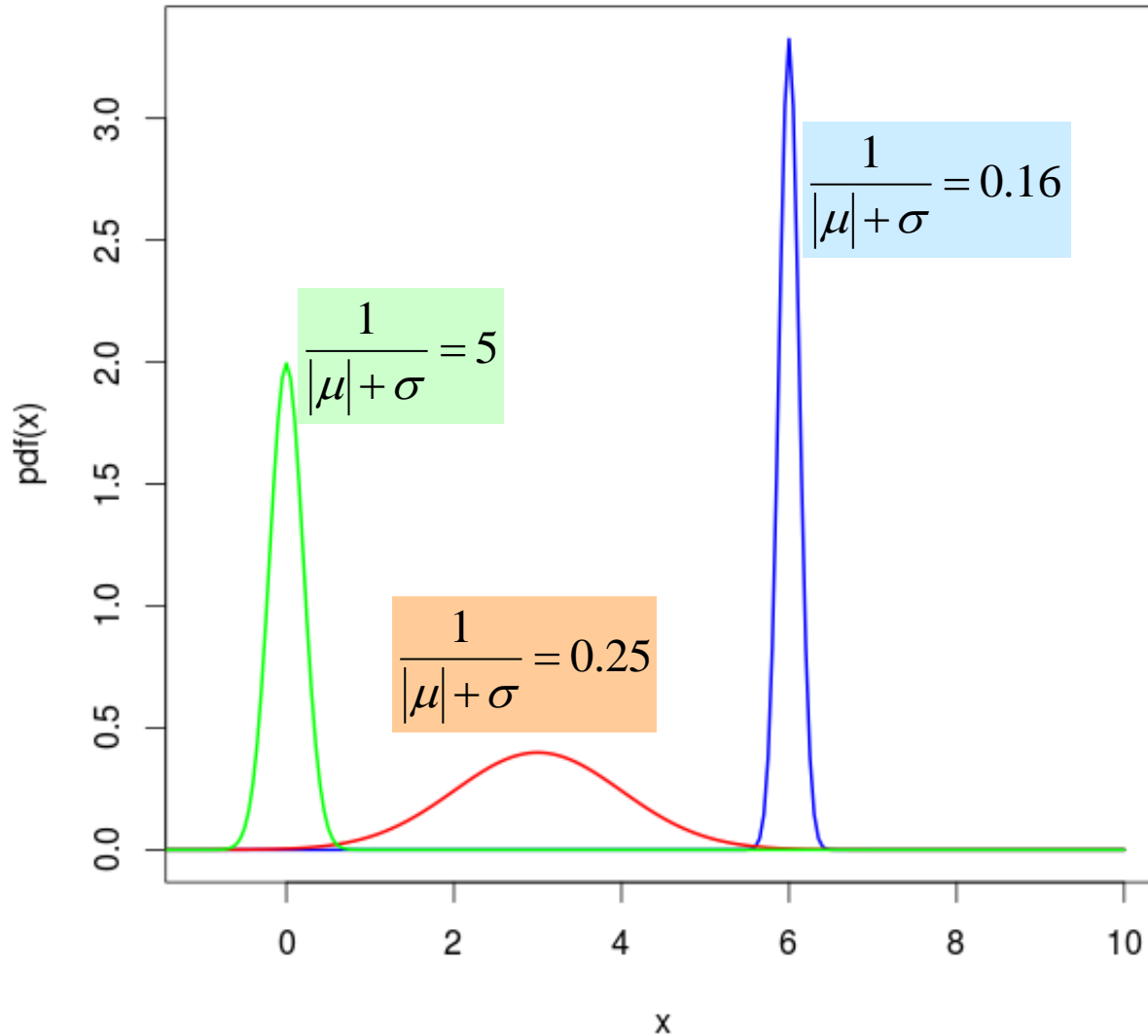
- PCMs have different orders of magnitude (kΩ, μA, ...).  
Normalization needed to avoid large values dominating.
- PCMs have different relative variation  $\sigma/\mu$ .  
Normalization needed to reduce weights of PCMs with  $\sigma/|\mu| \ll 1$

Normalization factor for every PCM: 
$$\frac{1}{\sigma} \cdot \frac{\sigma/|\mu|}{1 + \sigma/|\mu|} = \frac{1}{|\mu| + \sigma}$$

- i.e. instead of minimizing the difference in  $\sigma^2$ ,  
we minimize the difference in  $\sigma^2 / (|\mu| + \sigma)^2$ .

$$\min_X \| N \cdot (\text{cov}(P) - S \cdot X \cdot S^T) \cdot N \|^2 \quad \text{with } N = \text{diag}( 1/(|\mu| + \sigma) )$$

## Weight comparison of three example PCM



- PCM with smaller  $\sigma/\mu$  are weighted less
- Works for  $\mu=0$
- Works for  $\sigma=0$

## Removing of mismatch effects from the target PCM covariance matrix

➤ The PCM measurements contain both process and mismatch effects.  
But, the correlation shall be just generated for the process variation

➤ Mismatch can be removed via the covariance matrix

$$\text{cov}(P, \text{process}) = \text{cov}(P, \text{measured}) - \text{cov}(P, \text{mismatch})$$

➤ The mismatch covariance matrix is determined assuming a linear dependency between PCM and mismatch parameters

$$\Delta P = S_{\text{mm}} \cdot \Delta MM$$

$\Delta P$  variation of PCMs

$S_{\text{mm}}$  sensitivity of PCMs w.r.t. mismatch parameters (simulation)

$\Delta MM$  variation of mismatch spice model parameters

using

$$\text{cov}(P, \text{mismatch}) = S_{\text{mm}} \cdot \text{cov}(mm) \cdot S_{\text{mm}}^T$$

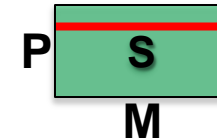
➤  $\text{cov}(mm)$  is the covariance matrix of the mismatch model card parameters that are  $mm \sim N(0, 1^2)$  distributed.

So,  $\text{cov}(mm)$  is typically an identity matrix.

## Input data consistency checks and improved data cleaning

## 1. Detection of PCMs with a low/no sensitivity towards the process parameters

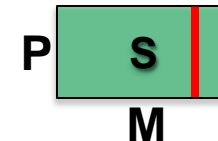
- There is no process parameter that can be used to model the variation of that PCMs
- Case can be detected in the sensitivity matrix as (close to) zero row(s)



- ➔ User to add the missing process parameters
- ➔ To continue, those PCMs are excluded from further calculations

## 2. Detection of process parameters with a low/no sensitivity towards the PCMs

- Without a sensitivity to PCMs, the statistics of those process parameters cannot be determined
- Case can be detected in the sensitivity matrix as (close to) zero column(s)



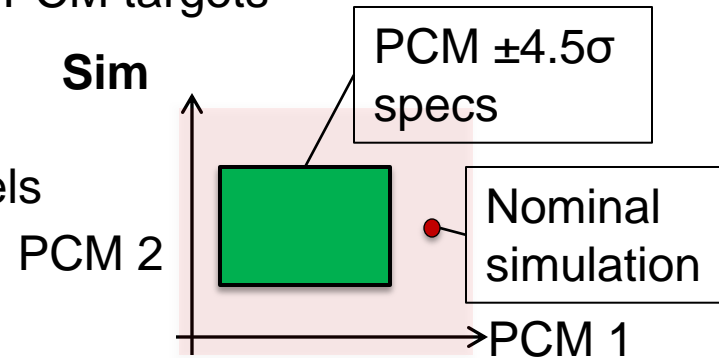
- ➔ User to pick the correct process parameter set
- ➔ To continue, such process parameters are excluded from further calculations

## Input data consistency checks and improved data cleaning

## 3. Detection of PCMs nominal simulation outside of the PCM specification

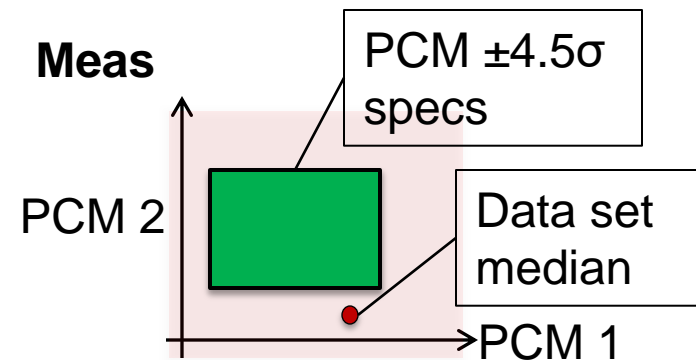
- The nominal PCM simulation does not fit to the PCM targets
- Can be detected comparing the PCM nominal simulation with the PCM targets

- ➔ Adjust the nominal simulation in the spice models
- ➔ To continue, those PCMs are excluded from the calculations



## 4. Detection of PCM measurement median outside of the PCM specification

- The PCM median of the measurements does not fit to the PCM specification
  - Can be detected comparing the median with the PCM targets
- ➔ Verify that the preprocessing steps of the PCM measurements are correct
  - ➔ Correct the PCM targets and specifications



## Input data consistency checks and improved data cleaning

### 5. Removing outliers from PCM measurements

- Values outside of median  $\pm 5.5\sigma$  are considered as outliers
  - PCM data sets have many missing values
    - Do not remove complete data sets → would reduce the number of data sets significantly
    - Instead mark outliers as missing value
- Simplifies preprocessing of PCM measurement data

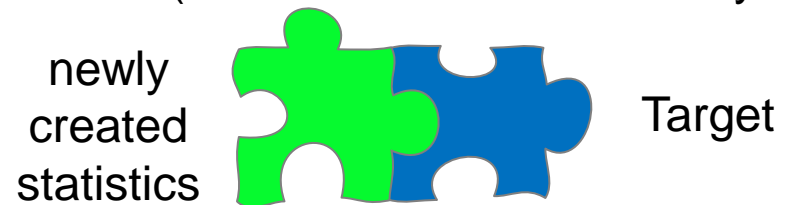
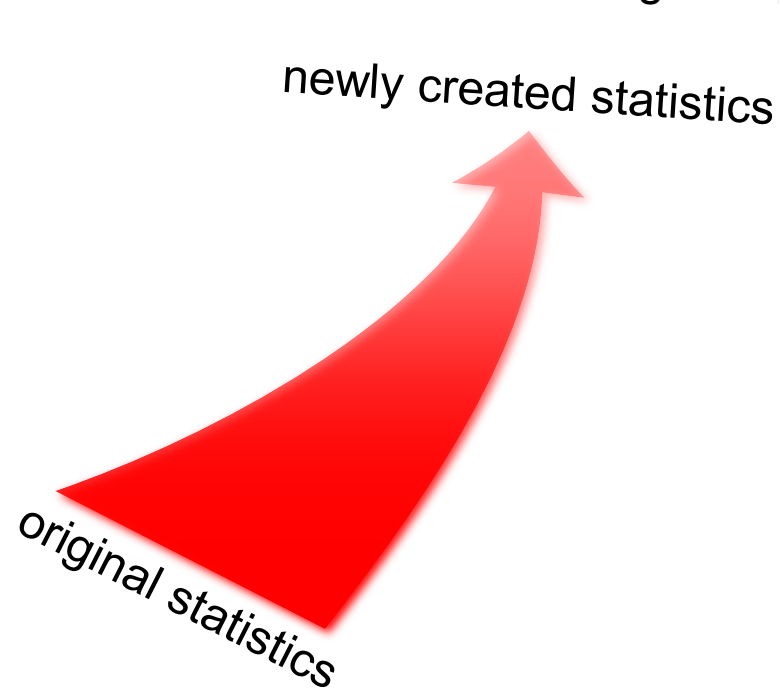
### 6. After determining the new statistics:

identify almost not varying process model parameters

- The calculated standard deviation of the process parameter will be very small
- Those model card parameter do not need to be varied in the spice models

## Monte Carlo based validation of determined statistics

- Allows comparisons of new model statistics to target and original statistics
- Three PCM covariance matrixes to compare
  - PCM target covariance
  - Covariance based on newly created statistics (result of Monte Carlo analysis)
  - Covariance based on original spice models (result of Monte Carlo analysis)



- New result displays
  - Correlation matrix
    - multi-column/row sort
    - rearranging columns/rows manually
  - Delta in correlation matrix
  - Comparison of standard deviation

## Comparison of the standard deviations

- NMOS and PMOS PCMs of an IC technology
- Standard deviation is matched very well

WiCkeD - Statistical Fit (Node 11) <1>

File Analysis Options Help

PCM sigma values | PCM sigma ranges | PCM correlations

PCM	PCM sigma target	PCM sigma simulated new
Vt_N_10/4_gm/5	22.22m	21.53m
G_N_10/4_gm/5	5.556u	5.635u
Vt_N_10/4_250n/0.1	22.22m	22.28m
Vt_N_10/4_25n/5	22.22m	22.2m
Vt_N_10/4_250n/5	22.22m	22.35m
Dibl_N_10/4_5	1.333m	114.9u
Slope_N_10/4_5	2m	378u
Ron_N_10/4_5/10u	55.56	8.426
Isat_N_10/4_5/5	44.44u	44.45u
Vt_P_10/3_gm/5	28.89m	21.54m
G_P_10/3_gm/5	6.667u	6.02u
Vt_P_10/3_133n/0.1	28.89m	22.41m
Vt_P_10/3_13.3n/5	28.89m	22.56m
Vt_P_10/3_133n/5	22.22m	22.49m
Dibl_P_10/3_5	1.778m	1.78m
Slope_P_10/3_5	1.556m	662.5u
Ron_P_10/3_5/10u	133.3	31.82
Isat_P_10/3_5/5	11.11u	28.33u

Spectre model file written to



# Target PCM correlations

➔ NMOS and PMOS PCMs of an IC technology

WiCkeD - Statistical Fit (Node 9) <1>

File Analysis Options Help

PCM sigma values PCM sigma ranges PCM correlations

Target PCM correlation (TARGET)

PCM	Vt_N 10/ 4_25 0n/5	Isat_N 1 0/4 5/5	G_N 10/4 gm/ 5	Ron_N 10 /4_5 /10u	Vt_P 10/ 3_13 3n/5	Isat_P 1 0/3 5/5	G_P 10/3 gm/ 5	Ron_P 10 /3_5 /10u	Vt_N 10/ 4_25 0n/0 .1	Vt_P 10/ 3_gm/ 75	Vt_P 10/ 3_13 3n/0 .1	Vt_P 10/ 3_13 .3n/ 5	Vt_N 10/ 4_gm/ 75	Vt_N 10/ 4_25 n/5	Slope e_P 10/3 _5	Slope e_N 10/4 _5	Dibl P_1 0/3 _5	Dibl N_1 0/4 _5
Vt_N_10/4_250n/5	+1.00	-0.91	-0.86	+0.76	-0.24	+0.54	-0.36	+0.55	+1.00	-0.28	-0.36	-0.33	+0.99	+1.00	-0.41	+0.79	-0.18	+0.13
Isat_N_10/4_5/5	-0.91	+1.00	+0.96	-0.94	+0.33	-0.72	+0.43	-0.73	-0.92	+0.36	+0.44	+0.41	-0.89	-0.91	+0.55	-0.78	+0.22	-0.15
G_N_10/4_gm/5	-0.86	+0.96	+1.00	-0.91	+0.33	-0.68	+0.43	-0.70	-0.87	+0.31	+0.40	+0.37	-0.82	-0.86	+0.51	-0.76	+0.22	-0.17
Ron_N_10/4_5/10u	+0.76	-0.94	-0.91	+1.00	-0.35	+0.76	-0.43	+0.78	+0.80	-0.38	-0.46	-0.42	+0.73	+0.80	-0.57	+0.69	-0.21	+0.11
Vt_P_10/3_133n/5	-0.24	+0.33	+0.33	-0.35	+1.00	-0.79	+0.09	-0.67	-0.35	+0.95	+1.00	+1.00	-0.23	-0.35	+0.46	-0.19	+0.01	-0.03
Isat_P_10/3_5/5	+0.54	-0.72	-0.68	+0.76	-0.79	+1.00	-0.38	+0.97	+0.54	-0.72	-0.80	-0.77	+0.53	+0.53	-0.68	+0.42	-0.13	+0.06
G_P_10/3_gm/5	-0.36	+0.43	+0.43	-0.43	+0.09	-0.38	+1.00	-0.44	-0.35	-0.18	+0.11	+0.06	-0.34	-0.35	+0.52	-0.24	+0.23	+0.04
Ron_P_10/3_5/10u	+0.55	-0.73	-0.70	+0.78	-0.67	+0.97	-0.44	+1.00	+0.54	-0.60	-0.67	-0.65	+0.53	+0.54	-0.65	+0.41	-0.13	+0.03
Vt_N_10/4_250n/0.1	+1.00	-0.92	-0.87	+0.80	-0.35	+0.54	-0.35	+0.54	+1.00	-0.28	-0.36	-0.33	+0.99	+1.00	-0.41	+0.78	-0.18	+0.19
Vt_P_10/3_gm/5	-0.28	+0.36	+0.31	-0.38	+0.95	-0.72	-0.18	-0.60	-0.28	+1.00	+0.95	+0.95	-0.28	-0.27	+0.37	-0.15	-0.03	-0.05
Vt_P_10/3_133n/0.1	-0.36	+0.44	+0.40	-0.46	+1.00	-0.80	+0.11	-0.67	-0.36	+0.95	+1.00	+1.00	-0.36	-0.36	+0.46	-0.20	+0.07	-0.03
Vt_P_10/3_13.3n/5	-0.33	+0.41	+0.37	-0.42	+1.00	-0.77	+0.06	-0.65	-0.33	+0.95	+1.00	+1.00	-0.33	-0.33	+0.41	-0.18	+0.00	-0.03
Vt_N_10/4_gm/5	+0.99	-0.89	-0.82	+0.73	-0.23	+0.53	-0.34	+0.53	+0.99	-0.28	-0.36	-0.33	+1.00	+0.99	-0.41	+0.79	-0.18	+0.14
Vt_N_10/4_25n/5	+1.00	-0.91	-0.86	+0.80	-0.35	+0.53	-0.35	+0.54	+1.00	-0.27	-0.36	-0.33	+0.99	+1.00	-0.41	+0.76	-0.18	+0.14
Slope_P_10/3_5	-0.41	+0.55	+0.51	-0.57	+0.46	-0.68	+0.52	-0.65	-0.41	+0.37	+0.46	+0.41	-0.41	-0.41	+1.00	-0.30	+0.07	-0.02
Slope_N_10/4_5	+0.79	-0.78	-0.78	+0.69	-0.19	+0.42	-0.24	+0.41	+0.78	-0.15	-0.20	-0.18	+0.79	+0.76	-0.30	+1.00	-0.17	-0.01
Dibl_P_10/3_5	-0.18	+0.22	+0.22	-0.21	+0.01	-0.13	+0.23	-0.13	-0.18	-0.03	+0.07	+0.00	-0.18	-0.18	+0.07	-0.17	+1.00	-0.03
Dibl_N_10/4_5	+0.13	-0.15	-0.17	+0.11	-0.03	+0.06	+0.04	+0.03	+0.19	-0.05	-0.03	-0.03	+0.14	+0.14	-0.02	-0.01	-0.03	+1.00

Spectre model file written to /spectre\_statistics.scs.

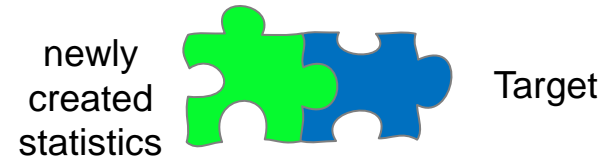
Relevant NMOS PCMs

Relevant PMOS PCMs

Further NMOS/PMOS PCMs

# Comparing the newly created model statistics with the target

➔ NMOS and PMOS PCMs of an IC technology



➔ Correlations match very well

Relevant NMOS PCMs

Relevant PMOS PCMs

Further NMOS/PMOS PCMs

WiCkeD - Statistical Fit (Node 9) <1>

File Analysis Options Help

PCM sigma values | PCM sigma ranges | PCM correlations

Difference between NEW and TARGET

PCM	Vt_N_10/4_250n/5	Isat_N_10/4_5/5	G_N_10/4_5gm/5	Ron_N_10/4_5/10u	Vt_P_10/3_133n/5	Isat_P_10/3_5/5	G_P_10/3_5gm/5	Ron_P_10/3_5/10u	Vt_N_10/4_250n/0.1	Vt_P_10/3_133n/0.1	Vt_P_10/3_13.3n/5	Vt_N_10/4_25n/5	Slope_P_10/3_5	Slope_N_10/4_5	Dibl_P_10/3_5	Dibl_N_10/4_5		
Vt_N_10/4_250n/5	+0.00	-0.00	-0.00	+0.01	-0.05	-0.02	-0.01	+0.01	+0.00	+0.09	+0.06	+0.08	-0.02	-0.00	-0.13	-0.21	-0.00	-0.02
Isat_N_10/4_5/5	-0.00	-0.00	-0.00	-0.00	+0.04	+0.02	-0.00	-0.01	+0.00	-0.11	-0.06	-0.08	+0.02	+0.01	+0.04	+0.17	+0.00	-0.00
G_N_10/4_5gm/5	-0.00	-0.00	-0.00	-0.00	+0.02	+0.02	-0.00	-0.01	+0.01	-0.08	-0.04	-0.06	-0.00	+0.02	+0.07	+0.18	+0.00	-0.00
Ron_N_10/4_5/10u	+0.01	-0.00	-0.00	-0.00	-0.03	-0.02	+0.01	+0.00	-0.03	+0.12	+0.06	+0.08	-0.01	-0.04	+0.05	-0.11	+0.00	+0.01
Vt_P_10/3_133n/5	-0.05	+0.04	+0.02	-0.03	-0.00	+0.00	-0.01	-0.00	+0.06	-0.01	-0.00	-0.00	-0.05	+0.06	-0.24	-0.06	+0.00	-0.00
Isat_P_10/3_5/5	-0.02	+0.02	+0.02	-0.02	+0.00	+0.00	-0.03	-0.00	-0.02	+0.08	+0.00	-0.01	-0.06	-0.04	+0.32	+0.12	-0.01	-0.05
G_P_10/3_5gm/5	-0.01	-0.00	-0.00	+0.01	-0.01	-0.03	+0.00	+0.02	-0.01	-0.04	-0.01	+0.02	+0.06	+0.03	-0.51	-0.36	-0.00	-0.02
Ron_P_10/3_5/10u	+0.01	-0.01	-0.01	+0.00	-0.00	-0.00	+0.02	+0.00	+0.01	+0.07	-0.01	-0.01	-0.02	-0.00	+0.27	+0.11	+0.00	+0.03
Vt_N_10/4_250n/0.1	+0.00	+0.00	+0.01	-0.03	+0.06	-0.02	-0.01	+0.01	+0.00	+0.08	+0.06	+0.08	-0.01	+0.00	-0.16	-0.24	-0.00	-0.02
Vt_P_10/3_133n/0.1	+0.09	-0.11	-0.08	+0.12	-0.01	+0.08	-0.04	+0.07	+0.08	+0.00	-0.01	-0.02	+0.04	+0.07	-0.07	+0.16	-0.03	-0.13
Vt_P_10/3_13.3n/5	+0.06	-0.06	-0.04	+0.06	-0.00	+0.00	-0.01	-0.01	+0.06	-0.01	+0.00	-0.00	+0.07	+0.06	-0.25	-0.07	+0.02	-0.00
Vt_N_10/4_25n/5	+0.08	-0.08	-0.06	+0.08	-0.00	-0.01	+0.02	-0.01	+0.08	-0.02	-0.00	+0.00	+0.09	+0.08	-0.26	-0.07	+0.01	+0.05
Vt_N_10/4_25n/5	-0.02	+0.02	-0.00	-0.01	-0.05	-0.06	+0.06	-0.02	-0.01	+0.04	+0.07	+0.09	+0.00	-0.00	-0.25	-0.38	+0.02	+0.17
Slope_P_10/3_5	-0.00	+0.01	+0.02	-0.04	+0.06	-0.04	+0.03	-0.00	+0.00	+0.07	+0.06	+0.08	-0.00	-0.00	-0.15	-0.26	+0.01	+0.03
Slope_N_10/4_5	-0.13	+0.04	+0.07	+0.05	-0.24	+0.32	-0.51	+0.27	-0.16	-0.07	-0.25	-0.26	-0.25	-0.15	+0.00	+0.21	-0.10	-0.64
Dibl_P_10/3_5	-0.21	+0.17	+0.18	-0.11	-0.06	+0.12	-0.36	+0.11	-0.24	+0.16	-0.07	-0.07	-0.38	-0.26	+0.21	+0.00	-0.06	-0.52
Dibl_N_10/4_5	-0.00	+0.00	+0.00	+0.00	+0.00	-0.01	-0.00	+0.00	-0.00	-0.03	+0.02	+0.01	+0.02	+0.01	-0.10	-0.06	+0.00	-0.00
Dibl_N_10/4_5	-0.02	-0.00	-0.00	+0.01	-0.00	-0.05	-0.02	+0.03	-0.02	-0.13	-0.00	+0.05	+0.17	+0.03	-0.64	-0.52	-0.00	+0.00

Done.

## Further advances in WiCkeD Statistical Fit

### ➤ Input data

- Further input file formats supported
- Comments in input files allowed
- Automatic matching of PCM names of data sets and of spice simulation

### ➤ Log

- Detailed log added

### ➤ While operating

- Progress bar added
- Improved status messages
- Extended error handling
- Improved handling of missing PCM values

### ➤ Export of results

- Export of sigma values and covariance matrixes to Excel format
- Spice include file with determined correlations

## Conclusion

- Ultimate goal of statistical modeling to reproduce the correlations between measurements of PCMs can be achieved
- Ready to use WiCkeD Statistical Fit solution to determine a correlated statistics
- Removing of mismatch effects from PCM measurements
- Support of different structures of correlation matrixes
- Several input data checks
- Monte Carlo based Validation of results



**WICKED**

Circuit Design Optimization

Thank You !