Statistics in Monte Carlo Analysis



PSpice Application Note | V2



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1 Introduction

The Monte Carlo analysis calculates the circuit response to changes in part values by randomly varying all of the model parameters for which a tolerance is specified. This provides statistical data on the impact of a device parameter's variance. Monte Carlo analysis is frequently used to predict yields on production runs of a circuit. With Monte Carlo analysis, model parameters are given tolerances, and multiple analyses (DC, AC, or transient) are run using these tolerances.

You can display data derived from Monte Carlo waveform families as histograms (using Performance Analysis).



Fig. 1: Histogram of Max(I(Meter))

The MONTE CARLO SUMMARY depended on the collating function e.g. the greatest difference from the nominal run (YMAX) is reported in PSpice output file.

SORTED DEVIATIONS OF I(V_Meter) TEMPERATURE = 27.000 DEG C **** MONTE CARLO SUMMARY Mean Deviation = 28.5690E-09 2.4445E-06 Sigma RUN MAX DEVIATION FROM NOMINAL 5.4934E-06 (2.25 sigma) higher at P = Pass 28 3.6 (139.92% of Nominal) 5.4736E-06 (2.24 sigma) higher at P = Pass 91 4.8 (128.52% of Nominal)

Fig. 2: MONTE CARLO SUMMARY in PSpice Output File

But how can we understand the statistical data?



2 Demo Project

The following example shows how the performance of a pressure sensor circuit with a pressuredependent resistor bridge is affected by manufacturing tolerances, using Monte Carlo analysis to explore these effects.

A zero volt voltage source is commonly used as a current meter in PSpice. A zero ohm resistor is not allowed.



In the above circuit, the model with the following definition

.model Rmc RES R=1 DEV=2% LOT=8%

is assigned to resistors R1, R2, R3 and R_Pressure while R4, R5 and R6 are normal resistors without tolerance.

Note

- Both tolerances DEV and LOT are available for each model parameter. Device tolerances, denoted as DEV within the model statement, cause components that share the same model to vary independently of one another. Lot tolerances, denoted by LOT in the model statement, cause components that share a common model to vary together.
- If you want to apply only a DEV tolerance to a usual R, L or C device, you can edit the TOLERANCE attribute on the part.
- To add both LOT and DEV tolerance to R, L and C devices, you must use the breakout symbols Rbreak, Lbreak and Cbreak in place of the usual R, L and C parts.



2.1 Simulation Settings and Simulation Results

The simulation profile could be a DC Sweep analysis as follows:

General Analysis Config	uration Files	Options	Data Collection	Probe Wind	ow	
Analysis Type: DC Sweep Options: Primary Sweep Secondary Sweep Monte Carlo/Worst Case Parametric Sweep Temperature (Sweep) Save Bias Point Load Bias Point	Sweep Va O Voltage Curren Global Model Tempe Sweep Ty Linear Logarit Value I	riable e source t source parameter parameter rature pe hmic ist3,6	Name: Model type Model nan Parameter	e: ne: r name: Start Va End Va Increme	P alue: lue: ent:	0 5 0.1

In Monte Carlo/Worst Case option, set I(Meter) as the Output variable, use Gaussian as distribution. If you left Random number seed blank, the default value 17533 will be used for the random number generator.

General	Analysis	Configu	uration Files	Options	Data Collection	Probe Window		
Analysis Typ DC Sweep Options:	e:	•	Monte C Monte C Monte Ca	Carlo ase/Sensitivity arlo Ontions –	Output Var	PSpice AA supp iable: I(Meter)	oort in legacy	
Primary Sweep Secondary Sweep Monte Carlo/Worst Case Parametric Sweep		Case	Number o Use Distri Random Save Dat	of runs: ibution: number seed: a From:	100 Gaussian 👻 All 👻]] [1.32767]	Distributions	
 Temperature (Sweep) Save Bias Point Load Bias Point 			Worst-cas Vary Devi Limit devi	se/Sensitivity (ice that have ices to type(s) lata from each	Dptions both DEV and sensitivity run	ILOT	✓ tolerances	



Note

• It is possible to set different distribution for different components but this needs to be done at model level. The distributions defined in the .model statement override the setting in the simulation setting options for MC analysis.

For example:

.model Rmc1 RES R=1 DEV/UNIFORM=2% LOT/UNIFORM=8%
.model Rmc2 RES R=1 DEV/GAUSSIAN=2% LOT/GAUSSIAN=8%

 If you click MC Load/Save, the Load/Save Monte Carlo Parameter File dialog box comes up. In the text box that is enabled, specify the name and the location of the file in which the parameter data is to be saved. After running the Monte Carlo analysis, the model parameter values used for various simulation runs will be stored in the .mcp file. Of course, you can reuse these parameter values by using the Load parameter values from file option.

Load/Save Monte Carlo Parameter File	х
Load Parameter values from file:	
Browse.	
Save Parameter values in file:	
MC_parameter.mcp Browse.	
OK Cancel	

Click **More Settings** to choose a collating function e.g. YMAX. With the collating function you can define on which criteria the results shall be sorted in the output File You can check the **List model parameter values in the output file for each run** if you like.

Monte Carlo/Worst-Case Output File Options	х
Collating Function OK The collating function is performed on an output variable (for example, V(1)). The result is listed in the output (.OUT) file only. Cancel Find the greatest difference from the nominal run (YMA) Cancel Threshold Value	





After the simulation the curves of I(Meter) could be as follows:

Note

With many runs (100 runs in our case), the waveform display can look more like a band than a set of individual waveforms. This can be useful for seeing the typical spread for a particular output variable. As the number of runs increases, the spread more closely approximates the actual worst-case limits for the circuit.

Click the **Performance Analysis** button, add trace Max(I(Meter)) to display histograms.



A MONTE CARLO SUMMARY is reported according to the output variable in simulation set up and the used collating function. The deviations are printed in order of size along with their run number.

FlowCAD

The Quantity of the bars in the bar chart (histogram) can be set under **Tools > Options > Number of Histogram Divisions.**

****	SORTED	DEVIATIONS	OF I(V	_Meter)	TEMPER	ATURE	-	27.000	DEG C	2
		MONTE	CARLO	SUMMARY						
*****	*******	**********	*****	******	*****	*****	****	*****	****	•*•
Mean De Sigma	viation = =	28.5690E 2.4445E	09 06							
RUN		MAX	DEVIAT	ION FROM	NOMINA	L				
Pass	28	5.493 (13	84E-06 89.92%	(<mark>2.25</mark> s: of Nomina	igma) al)	higher	at	P =	3.6	
Pass	91	5.473 (12	86E-06 28.52%	(<mark>2.24</mark> s: of Nomina	igma) al)	higher	at	P =	4.8	
Pass	64	4.798 (7	8E-06 6.184%	(<mark>1.96</mark> s: of Nomin	igma) nal)	lower	at :	P =	5	
Pass	42	4.595	50E-06 8.4763	(<mark>1.88</mark> s: E+15% of	igma) Nomina	lower l)	at i	P =	0	

Fig. 3: MONTE CARLO SUMMARY with original Simulation Settings

Note

When you use the YMAX collating function, the output file also lists mean deviation and sigma values. These are based on the changes in the output variable from nominal at every sweep point in every Monte Carlo run.

2.2 MONTE CARLO SUMMARY

In the above diagram, the first **RUN** is **Pass 28**, because the deviation from nominal is the biggest in all of the Monte Carlo runs.

The maximal deviation is **5.4934E-06 (2.25 sigma)**, because of 2.4445E-06*2.25≈5.500E-06.

Higher at P=3.6 means I(Meter) at the 28th run is bigger than at the nominal run while the sweep parameter P=3.6.

Note

There could be some rounding errors while calculating. The exact value is not very important here, the point is to understand the methodology within Monte Carlo analysis.

If you are interested which parameter values are used at pass 28, you can get the information in PSpice output file (the parameter values are also saved in MC_parameter.mcp).



Now go back to **Simulation Settings** window and set the sweep variable P=3.6 and run the simulation again.

General Analy	sis (Configur	ation Files	Options	Data Collection	n Probe Wi	ndow	
Analysis Type: DC Sweep Options: Primary Sweep Secondary Sw Monte Carlo/M	o eep /orst Ca:	▼ Se	Sweep Va Voltag Currer Global Model Tempe	ariable —— e source nt source parameter parameter erature	Name: Model ty Model na Paramete	oe: ame: er name:	P	v
Parametric Sw Temperature (Save Bias Poir Load Bias Poir	eep Sweep) nt nt	•	Sweep Ty C Linear C Logari Value	thmic D List 3.6	Decade v	Start End Incre	:Value: Value: ement:	0 5 0.1

The new simulation results could be as follows:



FlowCAD

Just marking **I(Meter) > Edit > Copy**, then open the Texteditor and paste the data, save the file as e.g. my_data.txt.

FlowCAD

In this file you will get I(Meter)

- at the nominal run is 1.3759909052169e-005
- at the 28th run is 1.92532643268351e-005

Note

In MC analysis, the first run is always the nominal run.

Therefore, the difference is:

1.92532643268351e-005 - 1.3759909052169e-005 ≈ 5.4934e-006

Please look at the MAX DEVIATION FROM NOMINAL at Pass 28 in MONTE CARLO SUMMARY section in PSpice output file (see Fig. 3), it is **5.4934E-06** and the same as our calculated value.

There is still **139.92% of Nominal** because 1.92532643268351e-005/1.3759909052169e-005 ≈ 139.92%.

2.3 Histogram and Statistical Data

Here we would like to reduce the number of runs to have less data for better calculation. But in fact, you may need much more runs to get more reasonable statistical data.

The sweep variable P remains with the value 3.6 and we just reduce the number of runs from 100 to 10.

General Analysis C	onfiguration Files	Options	Data Collection	Probe Window		
Analysis Type: DC Sweep Dptions: Primary Sweep Secondary Sweep Monte Carlo/Worst Cas Parametric Sweep Temperature (Sweep)	Monte C Monte C Monte C Number C Use Distr Random Save Dat Worst-ca	Carlo case/Sensitivity arlo Options of runs: ibution: number seed: ia From: se/Sensitivity (Enable Output Var I0 Gaussian All Dotions	PSpice AA supportiable: ((Meter)	Distributions	
 Save Bias Point Load Bias Point 	Vary Dev Limit dev	Vary Device that have both DEV and LOT voterances Limit devices to type(s) Save data from each sensitivity run MC Load Save More Settings				





After the simulation we can display the histogram for I(Meter).

Fig. 4: Histogram of I(Meter)

The associated statistics are as follows:

n samples = 10 10tl	h %ile = 1.23617e-005
n divisions = 10 medi	ian = 1.42536e-005
mean = 1.44607e-005 90tl	h %ile = 1.71466e-005
sigma = 1.7659e-006 maxi	imum = 1.81888e-005
minimum = 1.15008e-005 3*si	igma = 5.29769e-006

n samples = 10, because there are 10 runsn divisions = 10, because the Number of Histogram Divisions is 10



mean = 1.44607e-005



The values of I(Meter) at simulation runs are listed in the table below (my_data_10runs.txt).

Section Number	l(Meter)
1	1.3759909052169e-005
2	1.43037614179775e-005
3	1.39189451147104e-005
4	1.61043353728019e-005
5	1.32225886773085e-005
6	1.81887880899012e-005
7	1.44032792377402e-005
8	1.1500794244057e-005
9	1.50016412590048e-005
10	1.42034468808561e-005

The mean $\boldsymbol{\mu}$ is defined as

$$\mu = \frac{1}{n} \bullet \sum_{i=1}^{n} x_i$$

In this case

mean = (I(Meter)@1+I(Meter)@2+...+I(Meter)@10)/10 = (1.3759909052169e-005+...+1.42034468808561e-005)/10 ≈ 1.44607e-005

sigma = 1.7659e-006

The standard deviation σ is defined as

$$\sigma = \sqrt{E[(X - \mu)^2]}$$

Here

$$\sigma = \sqrt{\frac{1}{n-1} \sum_{i=1}^{n} (x_i - \mu)^2}$$

= $\sqrt{\frac{1}{10-1} [(I(Meter)@1 - \mu)^2 + ... + (I(Meter)@10 - \mu)2]}$

≈ 1.7659e-006

minimum = 1.15008e-005 at 8th run

maximum = 1.81888e-005 at 6th run

10th %ile = 1.23617e-005

10% goal function values (I(Meter)) are less than or equal 1.23617e-005, and 90% goal function values are greater than or equal 1.23617e-005.

$\tilde{x_p} = \frac{1}{2} (x_{n \bullet p} + x_{n \bullet p+1}), \quad \text{if n \bullet p integral,}$ $\tilde{x_p} = x \lceil n \bullet p \rceil, \quad \text{if n \bullet p not integral}$

In our case n=10, p=10% \rightarrow n•p=1, integral

$$\tilde{x_p} = \frac{1}{2}(x_1 + x_2) = \frac{1}{2}(I(Meter)@8 + I(Meter)@5)$$

\$\approx 1.23617e-005

Please see the file quantile.txt for the modified order of samples.

📗 quantile.txt - Edito	or		<u>_ 🗆 ×</u>
<u>D</u> atei <u>B</u> earbeiten F <u>o</u> r	rmat <u>A</u> nsicht <u>?</u>		
Section Number	I(Meter)	×i	_
8 5 1 3 10 2 7 5 4 6	1.1500794244057e-005 1.32225886773085e-005 1.3759909052169e-005 1.39189451147104e-005 1.42034468808561e-005 1.43037614179775e-005 1.44032792377402e-005 1.50016412590048e-005 1.61043353728019e-005 1.81887880899012e-005	×1 ×2 ×3 ×4 ×5 ×6 ×7 ×8 ×9 ×10	

90th %ile = 1.71466e-005

90% goal function values (I(Meter)) are less than or equal 1.71466e-005, and 10% goal function values are greater than or equal 1.71466e-005.

$$\widetilde{x} = x_{\frac{n+1}{2}, \text{ if n odd}}$$
$$\widetilde{x} = \frac{1}{2} \left(x_{\frac{n}{2}} + x_{\frac{n}{2}+1} \right), \text{ if n even}$$

In this case, n = 10 and even

$$\tilde{x} = \frac{1}{2}(x_5 + x_6) = \frac{1}{2}(I(Meter)@10 + I(Meter)@2)$$

= 1.42536e-005

3*sigma = 5.29769e-006 simply because sigma = 1.7659e-006

FlowCAD



2.4 Difference of the Statistical Data Reported in Histogram and in Output File

The statistical data in histogram are for the samples measured using a function, e.g. Max(I(Meter)) in Fig. 1 or using a simulation output variable in **Add Traces** window, e.g. I(Meter) in Fig. 4. A histogram has nothing to do with the collating function.

The statistical data in output file are based on the output variable defined in Simulation Settings and the collating function. For example, the MONTE CARLO SUMMARY looks as follows:

Mean Sigma	Deviation	= =	778.7100E-09 1.7486E-06		
RUN			MAX DEVIATION FROM NOMIN	AL	
Pass	6		4.4289E-06 (2.53 sigma) (132.19% of Nominal)	higher at P =	3.6
Pass	4		2.3444E-06 (1.34 sigma) (117.04% of Nominal)	higher at P =	3.6
Pass	8		2.2591E-06 (1.29 sigma) (83.582% of Nominal)	lower at P =	3.6
Pass	9		1.2417E-06 (.71 sigma) (109.02% of Nominal)	higher at P =	3.6
Pass	7		643.3700E-09 (.37 sigma) (104.68% of Nominal)	higher at P =	3.6
Pass	2		543.8500E-09 (.31 sigma) (103.95% of Nominal)	higher at P =	3.6
Pass	5		537.3200E-09 (.31 sigma) (96.095% of Nominal)	lower at P =	3.6
Pass	10		443.5400E-09 (.25 sigma) (103.22% of Nominal)	higher at P =	3.6
Pass	3		159.0400E-09 (.09 sigma) (101.16% of Nominal)	higher at P =	3.6

MONTE CARLO SUMMARY

In this case there are 9 samples, not 10 samples. The sample value is the maximal deviation from nominal, not the value of a measurement or an available output variable.

Mean Deviation = 778.7100E-09

Here:

mean deviation =
$$\frac{1}{n} \sum_{i=1}^{n} x_i$$

 $\approx 778.71 \text{E-09}$



Please refer deviation.txt file which based on the data of MONTE CARLO SUMMARY. The values (x7 and x3) are negative, because they are lower than the nominal.

/ deviation.txt - Editor			<u> </u>	
<u>D</u> atei	<u>B</u> earbeiten F <u>o</u> rmat <u>A</u> nsicht <u>?</u>			
Pass	I(Meter):max deviation from nominal)	xi		
6 4 9 7 2 5 10 3	4.4289E-06 2.3444E-06 -2.2591E-06 1.2417E-06 643.3700E-09 543.8500E-09 -537.3200E-09 443.5400E-09 159.0400E-09	×9 ×8 ×6 ×5 ×4 ×3 ×2 ×1	T	

Sigma = 1.7486E-06

In our case

Sigma =
$$\sqrt{\frac{1}{n} \sum_{i=1}^{n} (x_i - Mean \, Deviation)^2}$$

≈ 1.7486E-06

3 Bibliography

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- [2] PSpice Reference Guide, Cadence
- [3] OrCAD Capture User Guide, Cadence
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