Indirect Test of M-S Circuits Using Multiple Specification Band Guarding

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Abstract—Testing analog and mixed-signal circuits is a costly task due to the required test time targets and high end technical resources. Indirect testing methods partially address these issues providing an efficient solution using easy to measure CUT information that correlates with circuit performances. In this work, a multiple specification band guarding technique is proposed as a method to achieve a test target of misclassified circuits. The acceptance/rejection test regions are encoded using octrees in the measurement space, where the band guarding factors precisely tune the test decision boundary according to the required test yield targets. The generated octree data structure serves to cluster the forthcoming circuits in the production testing phase by solely relying on indirect measurements. The combined use of octree based encoding and multiple specification band guarding makes the testing procedure fast, efficient and highly tunable. The proposed band guarding methodology has been applied to test a band-pass Butterworth filter under parametric variations. Promising simulation results are reported showing remarkable improvements when the multiple specification band guarding criterion is used.

Index Terms—Band Guarding, Multiple Specification, Mixed-Signal Testing, Alternate Test, Indirect Measurements, Indirect Measurements Selection, Test Escapes, Test Yield Loss, Octrees, Quadtrees, Classifiers, Butterworth Filter.

I. INTRODUCTION

ESTING analog and mixed-signal circuits is a challenging task due to the limitations of current analog automatic test equipment and the partial availability of systematic procedures for mixed-signal testing [1]. This fact causes a significant increase of the incurred costs in the final product. Built-in techniques are considered a promising solution to cope with the ever increasing analog ATE costs and the drawback of externalizing internal analog nets. Area overhead and impact on CUT signals are, in some applications, justifiable and cost effective. Indirect testing methodologies are widely used in contrast to the classic specification based testing and have been presented as a suitable solution [2]-[6]. Alternate testing techniques require a mapping between the specification space and the measurement space in order to allow the test decision to be performed. Machine learning techniques [7], [8] and regression techniques [3], [9] have been used with successful results to this purpose, as well as using octrees to represent

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the pass/fail regions [10], [11]. Test costs are heavily raised up due to the existence of decision errors, i.e. test escapes and test yield loss, so lowering these analog metrics usually becomes the goal of any testing strategy [1], [7].

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Indirect testing strategies require the selection of a set of easy to measure parameters to be used as indirect measures. To that purpose, many options exist, some of them entirely relying on designer's expertise and experience. Some authors have proposed the use of the sensitivity matrix between circuit's functional specifications and indirect measurements with the goal of maximizing its rank [12]. This allows the avoidance of redundant information. Statistical methods have been also proposed, most of them relying on correlations and regressions techniques between the set of functional specifications and the set of indirect measurements [13]. For instance, in [6], [14], the authors use the *Brownian distance correlation* together with a greedy algorithm in order to select a meaningful subset of measures adequate for analog/RF circuits testing.

This paper is an extension of the work presented at the SMACD conference in 2015 [15]. Here, the work has been extended to consider multiple specification band guarding factors while keeping the goal of optimizing and/or reaching certain analog test metrics targets. The use of multiple band guarding factors over multiple circuit specifications is the major improvement with respect to the conference paper but other enhancements have been included such as the assessment of the criterion used to select the indirect measurements as well the evaluation of the required computation times. The proposed indirect test method comprises two phases, namely, the training phase and the testing phase [10], [15]. In the training phase, a representative amount of circuit samples is generated and the acceptance/rejection regions are encoded using octrees in the indirect measurement space. In the former phase, the proposed multiple specification band guarding strategy is applied by adequately tuning a set of factors which modify the test specifications and therefore the number of misclassified circuits. The latter phase corresponds to the actual production testing of the incoming ICs using the previously generated octree data structure. Octree encoded pass/fail regions facilitate an efficient and fast circuit clustering due to the inherent sparsity of such data structures.

The rest of the paper is organized as follows. Section II focuses on the details of testing analog and mixed-signal circuits using octree tessellation in the indirect measurement space. The training and testing phases are described as well as a short introduction to octree encoding is given. Section III

presents the proposed multiple specification band guarding methodology and how it can be used to achieve certain test targets. Section IV presents a band-pass Butterworth filter which serves as the test vehicle to evaluate the proposed multiple specification band guarding methodology. Section V reports the simulation results resulting from the application of the proposed band guarding methodology. The efficiency of the testing procedure in terms of test escapes and test yield loss metrics is studied as well as in terms of computation time. Also, the noise impact in the indirect measurements is also explored. Finally, Section VI summarizes the work and concludes the paper.

II. TEST OF MIXED-SIGNAL CIRCUITS IN THE INDIRECT MEASUREMENT SPACE

The procedure of testing analog and mixed-signal circuits can be formalized as the classification of a candidate circuit into pass/fail clusters. Such classification is usually performed in the *specifications space* or *performance space* in which circuit specifications/performances are directly measured. This procedure is referred as the well known *specification based test* as opposed to the *indirect test* or *alternate test*. Indirect testing aims to battle against the drawbacks of specification based test such as the need for high end technical resources or the difficulties in validating every single test specification. The indirect test procedure used in this work is formed by two phases, namely, the training phase and the testing phase. The former phase is in charge of generating the circuit data and encode them using an octree data structure which will be used in the latter phase to test the freshly fabricated ICs.

A. Training Phase

In order to perform the encoding of the acceptance/rejection regions in the measurement space, a certain amount of representative circuit data need to be generated [10]. This can be achieved by circuit simulation, model/macromodel evaluation or simply by using available data on the production run from already tested ICs. Of course, accelerating techniques to obtain representative border circuits can also be used, such as stratified sampling or statistical blockade [16], [17]. Then, when the circuit data are generated, they are digitally encoded by means of a classifier. This is, in general, considerably challenging since highly nonlinear boundaries are difficult to be represented by means of state of the art classifiers (neural networks, SVMs,...) or regression models (linear, MARS,...)

On the contrary, octree data structures present the advantage of being able to represent arbitrary *n*-dimensional regions as long as the training algorithm encounters circuit data points to generate the octree. Also, it has the capability of controlling the resolution by simply limiting the maximum depth level. Such structures can be represented digitally and are easy to evaluate since only a few comparisons are required to traverse the tree. Next subsection focuses on how to use an octree

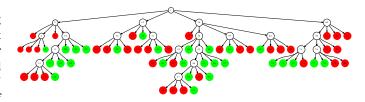


Fig. 1. Graph representing the quadtree depicted in Fig. 2. The obtained quadtree is 6 levels deep and has 85 nodes. Green/red leaf nodes correspond to pass/fail circuits.

data structure to encode the pass/fail regions in the indirect measurement space.

B. Octree Encoding

An octree data structure of dimension n is a tree in which each node has exactly 2^n children or none [19]. Octrees usually represent a geometric partition of an n-dimensional space. One of their main applications is in the field of 3D computer graphics, so every node has $2^3 = 8$ children, what gives its name (8-tree). For the case of two dimensional applications, they also receive the name of quadtrees (4-tree). The concept can be easily generalized to n-dimensional spaces where sometimes are referred as 2^n -trees or simply n-dimensional octrees. Fig. 1 shows an example quadtree data structure encoding the plane tessellation depicted in Fig. 2. The graph has 85 nodes and is 6 levels deep, what implies the length of the side of the smallest octree cell is 1/64th of the the initial cell at level zero.

In order to illustrate the process of encoding the acceptance/rejection regions in the measurements space, consider the set of data points shown in Fig. 2. The data shown represent circuits passing the test in green and circuits failing the test in red. Initially, the theoretical boundary separating them is unknown, so the octree encoding algorithm is solely based on the green/red data points. The first step is to consider an initial cell containing all the circuits and tessellate it into 4 equal regions by halving each dimension. Then, the following algorithm is applied to each of the resulting children. If it does exclusively contain single cluster data points, the square is tagged accordingly to that class and no further partitioning is performed. Otherwise, the square is marked as decision pending (white parent nodes in Fig. 1) and the procedure continues until all the generated squares only contain equal class data. After a tessellation, it may happen that some of the generated cells do not contain any data point. These cells are left untagged until the algorithm finishes and then treated separately. The way to proceed is to mark them accordingly to the majority of the clusters assigned to their neighboring octree cells [11].

C. Testing Phase

Trees, and in particular k-ary trees, are popular data structures in computer science for which efficient algorithms have been developed such as tree traversal, tree insertion and tree

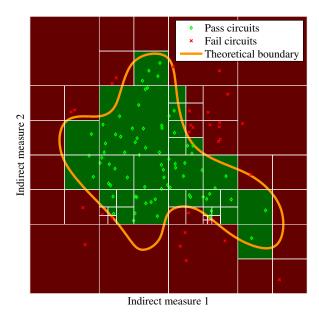


Fig. 2. Example of a 2-dimensional space encoded using quadtrees. The octree encoding algorithm solely relies on the green/red data points to approximate the acceptance/rejection test regions.

pruning. A k-ary tree can be stored as an array of pointers to memory locations what greatly facilitates its traversal from the root node to the leafs. The most common way of implementing a k-ary tree is by defining a node structure data type. The node structure has k members declared as pointers to a type of itself, what creates a recursive and efficient implementation of the tree which allows a fast traversal procedure in a similar way CPU indirect addressing works.

The testing phase corresponds to the actual classification of the freshly fabricated circuits into pass/fail clusters using the former octree data structure. Consider the octree depicted in Fig. 2 whose graph representation is the one shown in Fig. 1. Let (M_1, M_2) be a pair of circuit indirect measurements which are assumed to be within the initial square. The first step is to check to which quadrant they belong to (SW, NW, SE or NE) by performing one comparison per coordinate. This decision brings the candidate circuit to a new bound since a deeper octree level is achieved. If the current node is a tagged leaf, the point is mapped to that cluster and the evaluation terminates. If not, the evaluation algorithm repeats the decision operation through out the graph until a tagged leaf node is found and the circuit is clustered as pass or fail [11], [20]. The algorithm listed in Fig. 3 implements the octree evaluation procedure in pseudocode.

III. MULTIPLE SPECIFICATION BAND GUARDING IN THE INDIRECT MEASUREMENT SPACE

Test specification band guarding can be used to improve test outcomes depending on the established test escapes and test yield loss trade offs or targets. The underlying idea is to create a series of band guards along the test decision boundaries. The actual purpose of these guards is to displace the test limits with

```
1: function EVALMEASUREMENTS(N, M)
       Precond: N is an octree node data structure
2:
       Precond: M is a vector of indirect measurements
3:
       if N.label \neq NULL then
4.
          return N.label
5:
6:
       else
          idchild \leftarrow compare(M, N.center)
7:
          N \leftarrow N.child[idchild]
8:
          return EvalMeasurements(N, M)
9:
       end if
10:
11: end function
```

Fig. 3. Recursive implementation of the testing procedure using octrees in the indirect measurement space. Octree evaluation is efficient since only one comparison per coordinate and level is required.

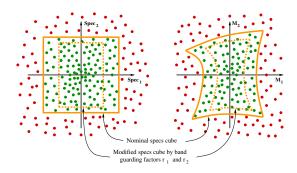


Fig. 4. Example of multiple specification band guarding in the specifications space and the resulting non linear mapping in the indirect measurement space. The applied shrink helps to reduce test escapes metric.

the objective of bypassing the mispredictions or uncertainties carried out by the actual test procedure, therefore achieving less misclassified circuits [15].

In this work, the proposed band guarding procedure is performed in the indirect measurement space using a set of factors r_1, r_2, \ldots, r_p , which are in charge of shrinking or enlarging the test specification cube and so the test decision boundary in the indirect measurement space. From now on, these factors will be referred as *band guarding factors*. Each of the factors is responsible of modifying one or more test specifications as sketched in Fig. 4. The use of multiple band guarding factors is the major novelty of this work with respect to the work presented in [15].

If all the band guarding factors are less than 1, it is clear that the acceptance region will be shrunk, what immediately translates into tighter test specifications. This implies that the probability of classifying an actual fail circuit into the pass cluster will be reduced, i.e. the test escape metric will be reduced. On the contrary, if all the band guarding factors are greater than 1, the acceptance region will be enlarged, what implies the relaxation of tests specifications and therefore the probability of classifying an actual pass circuit into the fail cluster gets reduced, i.e. the test yield loss metric will be reduced. Of course, according to the desired trade off between the test efficiency metrics or the needs for tightening or relaxing any of the circuit performances, a whole space of possibilities is available for the test engineer. It is worth to

TABLE I
BAND-PASS BUTTERWORTH FILTER DESIGN SPECIFICATIONS

Specification	Symbol	Value	Units
Center frequency	f_0	1.0	MHz
Bandwidth (−3 dB)	BW	1.0	MHz
Quality factor	Q	1.0	
Pass-band gain	G_{BP}	0.0	dB

note that the case $r_1 = r_2 = \cdots = r_p = 1$ results in a test scenario with no band guarding at all.

As a proof of concept, this work deals with two band guarding factors which are in charge of tuning the test specifications of a continuous time filter. The following section describes the case study and explains in which manner the test specifications are varied according to each of the factors.

IV. CASE STUDY: BAND-PASS BUTTERWORTH FILTER

A. Filter Design and Test Specifications

In order to demonstrate the viability of the proposed multiple specification band guarding technique, it has been applied to test a 6th order band-pass Butterworth filter in the indirect measurement space [21]. The schematic of the CUT is shown in Fig. 6. Filter's topology corresponds to an active implementation of a ladder filter in which inductors have been substituted by their active gyrator counterparts using operational transconductance amplifiers (OTA) and capacitors have been implemented taking advantage of technology's MiM capacitor capabilities. The OTA circuit is a current mirror operational transconductance amplifier designed using simulated annealing optimization. The Butterworth filter has been designed in an industrial 65 nm N-well bulk CMOS technology from ST-Microelectronics [22]. Filter design components are listed in Table III.

Filter design specifications are listed in Table I. As can be appreciated, the Butterworth band-pass filter has been designed with a center frequency of 1 MHz and a bandwidth of 1 MHz. The Q-factor resulting for such combination is Q = 1, what ensures a wide flat band in the pass-band. A series of test specifications have been imposed over the filter as listed in Table II. The test limits have been defined using boxes in the magnitude Bode plot as Fig. 5 depicts in blue dashed lines. As can be observed, test specifications regard to both bandwidth and gain, what makes the testing of the filter a non trivial task. The filter has been simulated using HSPICE and the 65 nm technology models from ST-Microelectronics (statcrolles corner). Monte Carlo simulations based on the aforementioned statistical corner have been used to characterize the transconductance parameters of the OTA, namely, the transconductance gain g_m and the output resistance r_o , therefore facilitating the generation of the 10⁴ circuits to be used in the training phase. Fig. 5 shows some Monte Carlo simulations when the statistical corner of the process design kit is used.

TABLE II
BUTTERWORTH FILTER TEST LIMITS

Specification	Test Spec	Units
Minimum pass-band bandwidth	745.6	kHz
Maximum stop-band bandwidth	11.0	MHz
Maximum pass-band gain	1.2	dB
Minimum pass-band gain	-3.0	dB
Minimum stop-band attenuation	-59.0	dB

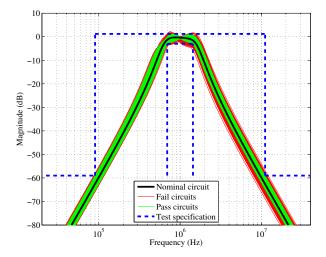


Fig. 5. Magnitude Bode diagrams obtained by Monte Carlo simulations using technology statistical corner (statcrolles). The test limits listed in Table II are also drawn in blue dashed line.

The band guarding factors used in this work are referred to r_1 and r_2 . The way the band guarding factors control the test limits is as follows. Factor r_1 contributes to the maximum and minimum allowable gains in the pass-band of the filter. If $r_1 > 1$ means that the two upper gain lines move away from the 0 dB line. On the contrary, factor r_2 contributes to the bandwidth of the filter. If $r_2 > 1$, the test specification related to filter's bandwidth is relaxed, what implies the band-pass corner frequencies get closer to f_0 .

B. Indirect Measurements Selection

The presented 6th order band-pass Butterworth filter has been excited with a multitone stimulus. The input signal is composed of 3 in-phase tones corresponding to filter's center frequency (1 MHz), an octave higher (2 MHz) and an octave lower (500 kHz), what yields to a periodic signal of period 2 μ s. Such stimulus has been proven to be effective for analog and mixed-signal testing purposes since it concentrates in a single stimulus the excitation of three single tones, making the

TABLE III
BUTTERWORTH FILTER COMPONENTS VALUES

Comp	Value	Units	Comp	Value	Units
C_1	2.12	pF	C_{L1}	74.94	pF
C_2	4.24	pF	C_{L2}	37.47	pF
C_3	6.37	pF	C_{L3}	24.98	pF
g_m	79.24	μ A/V	R_L	50.00	$k\Omega$

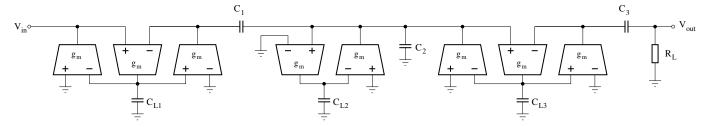


Fig. 6. Schematic of the 6th order band-pass Butterworth filter used as a case study. The filter has been designed and simulated using an industrial 65 nm technology from ST-Microelectronics. Filter topology corresponds to a ladder filter implemented with operational transconductance amplifiers (OTA).

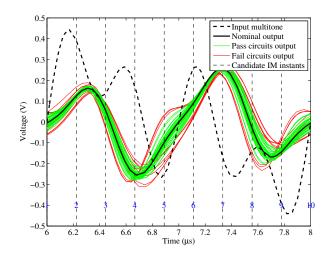


Fig. 7. Response of the Butterworth filter when it is excited with a multitone signal. The candidate indirect measurements correspond to equally spaced samples of filter's output.

whole process faster [11], [23]–[25]. Fig. 7 shows the applied multitone input signal and the corresponding output responses for a few Monte Carlo samples once the sinusoidal steady state has been reached. Green traces correspond to circuits fulfilling the test specifications listed in Table II while red traces correspond to circuits violating, at least, one of the test specifications.

In this work, the candidate set of indirect measurements are formed by several samples of filter's response to the aforementioned multitone input signal. The candidate set of samples are taken at evenly spaced time intervals as indicated with the dashed vertical lines in Fig. 7. Note that, since the sampling signal is periodic, a stroboscopic or sequential equivalent-time sampling technique may be carried out over more than one period in order to reduce the effective sampling rate. Such procedure yields to a set of 10 candidate indirect measurements from which a final subset needs to be selected. As intuition suggests, a good subset of indirect measurements has to satisfy two requirements. The measurements need to reflect circuit's performance variability and should not be redundant to avoid incurring in extra test costs.

Having a look at the Bode diagrams shown in Fig. 5 and the steady state transient responses depicted in Fig. 7, it is clear that the candidate measurements are correlated with circuit performances. Regarding the second condition, Kendall's Tau

TABLE IV
KENDALL'S TAU RANK CORRELATION COEFFICIENT BETWEEN
CANDIDATE INDIRECT MEASUREMENT PAIRS

	1	2	3	4	5	6	7	8	9	10
1	1.000									
2	0.688	1.000								
3	0.910	0.620	1.000							
4	0.835	0.576	0.830	1.000						
5	0.658	0.626	0.645	0.501	1.000					
6	0.766	0.712	0.755	0.633	0.812	1.000				
7	0.331	0.272	0.313	0.171	0.587	0.425	1.000			
8	0.705	0.706	0.689	0.550	0.916	0.865	0.520	1.000		
9	0.240	0.033	0.256	0.308	0.053	0.015	0.130	0.046	1.000	
10	0.981	0.689	0.912	0.835	0.658	0.768	0.331	0.706	0.240	1.000

rank correlation coefficient is used to quantify the correlation level among the final subset of indirect measurements [26]. Since the aim is to reduce redundant information, the pair presenting the lowest Kendall's Tau correlation is included in the target subset.

Table IV shows the absolute values of Kendall's Tau rank correlation coefficient among the 10 candidate indirect measurements under consideration. The pair of measurements presenting the lowest correlation is the one formed by measurements 9 and 6 with a correlation value of 0.015. Here forth, these measurements will define the indirect measurement space for the presented Butterworth filter and are the only measurements that need to be performed in the test application phase. Fig. 9 shows the indirect test space formed by these measurements.

It is worth to note that for the present case study two indirect measurements are enough to define the test decision boundary. If two measurements were not enough, Kendall's Tau rank correlation coefficient is also useful for such scenario [11]. The underlying idea is to add indirect measurements to the final subset while keeping the sum of all their Kendall's Tau the smallest possible. This is a general procedure based on a similar concept in order to make an appropriate selection when more than two measurements are needed.

C. Kendall's Tau Criterion Assessment

In order to verify that the criterion used to select a subset of indirect measurements yields to reasonable results, all the possible pairs (45) have been used for the training/testing phases with $r_1=r_2=1$. The training set size is composed of 10^4 pairs of indirect measurements while the testing set size

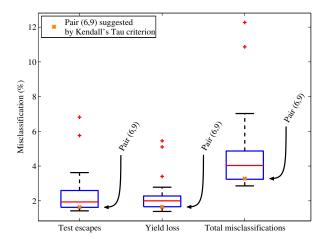


Fig. 8. Box plots showing the distribution of misclassified circuits for all the possible pairs of indirect measurements. The pair suggested by the Kendall's Tau criterion is within the first quartile of misclassifications.

is formed of 10^5 pairs. The results of this bruteforce study, are shown in Fig. 8 where the box plot of the distribution of the number of misclassified circuits are shown. As usual, box height represents the 25%-75% quartile spread while whisker's lengths relate to the interquartile range and/or extreme values within the distribution. As can be appreciated, the pair suggested using Kendall's Tau criterion reports misclassification levels which are within the first quartile of the population. More over, if the whole spread of total misclassification is taken as reference, the total misclassification level of the pair selected using Kendall's Tau criterion is among the 4.4% best values, fact that confirms Kendall's Tau rank correlation coefficient as a suitable indicator to select indirect measurements.

D. Training Phase

Taking into account the previously selected subset of indirect measurements, 10^4 Monte Carlo samples of the Butterworth filter have been simulated resulting in the samples shown in Fig. 9. The application of the octree encoding algorithm detailed in Section II results in the octree depicted in Fig. 10. As can observed the octree gets finer exactly where it is needed, i.e. in the surroundings of the test decision boundary. The resulting octree is 12 levels deep.

In order to better understand the benefits of octrees as classifiers, it is worth to take a look at the levels distribution of the octree cells forming the octree data structure. Fig. 11 shows such distribution for the octree depicted in Fig. 10. As can be stated, the vast majority of octree cells concentrate at levels 7, 8, and 9 despite the maximum level is 12. This fact implies a considerable advantage in terms of test application time since most of the circuits will be resolved without the need of traversing the tree down to its maximum level as will be shown in Section V where the simulation results are presented.

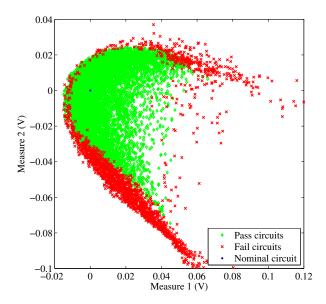


Fig. 9. Monte Carlo samples of a 6th order Butterworth band-pass filter in the indirect measurement space. Green/red circuits correspond to circuits fulfilling or not the test specifications detailed in Table II.

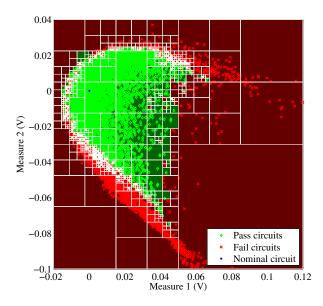


Fig. 10. Resulting octree after the application of the training phase to the set of 10^4 Butterworth circuit samples shown in Fig. 9. As can be observed, high octree levels concentrate in the test decision boundary.

V. SIMULATION RESULTS

A. Analog Test Metrics Evaluation

In order to evaluate the proposed testing procedure using octrees, a set of 10^5 Butterworth filters have been generated using Monte Carlo simulations and evaluated using the octrees generated in the training phase. For the case $r_1=r_2=1$, the results of the evaluation can be observed in Fig. 12. Correctly classified circuits are drawn using green and red colors, which respectively correspond to pass and fail circuits. Circuits that have not been correctly clustered are circled using dark green for pass circuits classified as fail and dark red for fail circuits

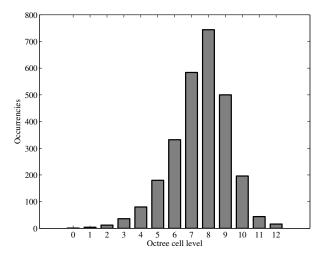


Fig. 11. Levels distribution for the octree shown in Fig. 10. Despite the maximum level is 12, the vast majority of octree cells correspond to levels 7, 8 and 9, what greatly facilitates its evaluation in the forthcoming testing phase.

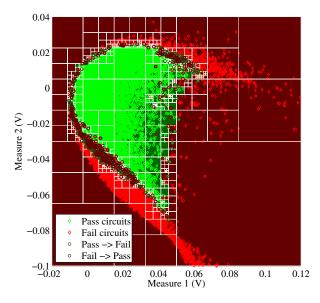


Fig. 12. Testing phase results using the octree shown in Fig. 10 with band guarding factors of $r_1=r_2=1$. Circled circuits correspond to circuits being misclassified by the octree, i.e. test escapes and test yield loss.

classified as pass. As can be appreciated, the misclassified circuits lie in the boundary of the test decision regions, which is the most prone area to misclassification.

Special attention needs the bar chart depicted in Fig. 13. It is showing the cumulative distribution of the required number of traversed levels in order to cluster every single circuit out of the 10^5 that have been generated. As can be observed, despite the maximum octree level is 12 (plus the root node), more than 95% or the circuits are classified by just descending down to level 7. Recalling the algorithm listed in Fig. 3, it turns that with only $2\times 7=14$ float number comparisons most of the circuits are clustered. This fact contrast with other classifier methodologies, such as support vector machines (SVM), which require the evaluation of multiple dot products

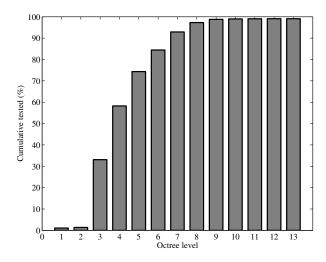


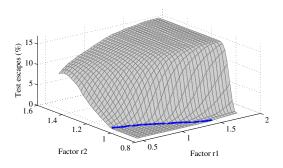
Fig. 13. Cumulative percentage of tested circuits as a function of the achieved octree level for the testing phase. As can be appreciated, more than 90% of circuits are tested in just 7 levels recursions.

plus a complex kernel function [11].

The training and testing procedures have been performed using different octrees with different levels of band guarding factors. The band guarding factor related to gain specifications, r_1 , has been varied from 0.5 to 1.78 in steps of 0.04. Similarly, the band guarding factor related to bandwidth specifications, r_2 , has been varied from 0.85 to 1.49 in steps of 0.02. Such ranges result in grid of size 33×33 , which allows to explore the $r_1 - r_2$ space. The resulting analog test metrics, i.e. test escapes and test yield loss, are shown in Fig. 14 as function of factors r_1 and r_2 . As mentioned before, the case $r_1 = r_2 = 1$ corresponds to the nominal case with no band guarding used. As can be appreciated, analog test metrics are drastically reduced down to zero as the band guarding factors vary along certain directions, while it is increased if opposite directions are taken.

As a rule of thumb, higher costs are incurred if circuits not accomplishing the specified test limits are shipped to the customer, i.e. false positive circuits. For that reason, fixing a test escape target and tune the test strategy to achieve it becomes a common practice in the test industry. For instance, let us assume the test escapes ratio must be below 1000 ppm (0.1%). In order to better elucidate the concepts throughout the example, Fig. 15 shows the test escapes contour levels in the r_1-r_2 plane. The established test escape level of 1000 ppm has been highlighted using a thick blue line. This means that any combination of band guarding factors lying on this level will result in a test especifications with an associated test escape metric of 0.1%.

The counterpart metric to test escapes is the test yield loss. Test yield loss is defined as the ratio of circuits certainly accomplishing the test specifications but classified as fail by the test strategy, i.e. false negatives. In the used example, the test yield loss will serve to determine which is the optimum pair of (r_1, r_2) band guarding factors. Fig. 16 shows the contour levels of the yield loss metric in the $r_1 - r_2$ plane



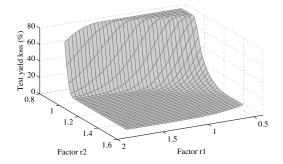
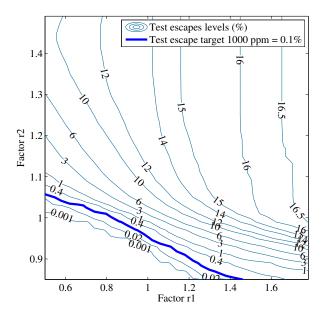
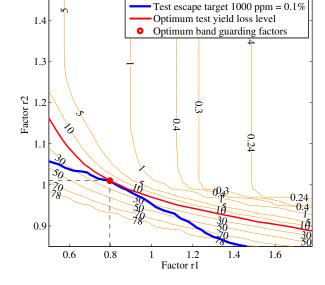


Fig. 14. Test escapes (left) and test yield loss (right) metrics as a function of the band guarding factors r_1 and r_2 evaluated using the set of 10^5 circuit samples in the testing phase. The test escapes target of 1000 ppm level is plotted in blue.





Test yield loss levels (%)

Fig. 15. Test escapes contour levels in the $r_1 - r_2$ plane. The test escapes target of 1000 ppm level is highlighted in blue. Any band guarding factor pair lying on the thick blue level guarantees a test escape of 1000 ppm.

Fig. 16. Test yield loss contour levels in the $r_1 - r_2$ plane together with the resulting 1000 ppm test escape level from Fig. 15. The optimum band guarding factors are selected by choosing the minimum test yield loss level.

together with the 0.1% test escape level elucidated from Fig. 15. The optimum band guarding factors would be the ones with the minimum test yield loss but still maintaining the test escapes target. It is clear that the contour level to choose is the one being tangent to the test escapes contour. Such yield loss contour level is highlighted with a thick red line in Fig. 16. The contour corresponds to a test yield loss of 16.77% and the needed band guarding factors to achieve the target are $r_1=0.797$ and $r_2=1.010$.

TABLE V
MULTIPLE SPECIFICATION BAND GUARDING RESULTS

TE (%) 2.05	0.10 ÷ 20.5
YL (%) 3.11 10	6.77×5.4
r_1 1.000 0.	.797
r ₂ 1.000 1.	.010

The results of the optimization process are listed in Table V. Initially, when no band guarding was used, the test escapes and yield loss were 2.05% and 3.11%, respectively. When using the proposed multiple specification band guarding method, the test escapes metric has been reduced by a factor of 20.5 at the expense of increasing the test yield loss metric by a factor of 5.4. In the presented example, test escapes has been fixed as a target, but a different approach can be carried out using importance weights. For instance, an objective function $f(r_1, r_2)$ can be formed as a linear combination (or whatever) of the test escapes and test yield loss functions as follows,

 $f(r_1, r_2) = w_{\text{TE}} \cdot \text{TE}(r_1, r_2) + w_{\text{YL}} \cdot \text{YL}(r_1, r_2)$. Where w_{TE} and w_{YL} are the weights assigned, for instance, proportionally to the incurred costs of false positives and false negatives, respectively. Such a function can be numerically optimized without even the need of exploring the whole $r_1 - r_2$ plane.

B. Noise Impact

It is clear that noisy indirect measurements may affect the test efficiency. The proposed multiple specification band guarding methodology is also capable of greatly improving

TABLE VI MULTIPLE SPECIFICATION BAND GUARDING RESULTS WHEN GAUSSIAN NOISE IS ADDED TO THE INDIRECT MEASUREMENTS

	Without Band	With Band	Factor
	Guarding	Guarding	ractor
TE (%)	3.21	0.50	÷ 6.4
YL (%)	7.68	20.18	× 2.6
r_1	1.000	0.989	
r_2	1.000	0.970	

one of the metrics at the expense of worsening the other in the presence of noise. In order to study such scenario, simulations using Gaussian noise in the measurements have been carried out. Table VI shows the analog test metrics under the influence of Gaussian noise with a σ spread of 2 mV. As expected, results are worsened with respect to the results shown in Table V, but the method is still able to provide some benefits by just choosing the right band guarding factors depending on the desired targets. Here, a test escapes target has been fixed to 5000 ppm (i.e. 0.5%). The best obtained yield loss metric to achieve the target is 20.18%, what means that the method has been able to reduce the test escapes by a factor of 6.4 while increasing the test yield loss metric by a factor of 2.6. Results get affected by the noise but the method is still usable under such conditions.

C. Performance Evaluation

In order to provide some performance metrics of the proposed multiple specification band guarding technique using octrees, the computation times of the whole process have been annotated. The machine used in the benchmarking is an 8 core Intel Pentium i7 processor at 3.4 GHz with 16 Gbytes of RAM. The employed electrical simulator is Synopsis HSPICE version 2010.03 with no multiprocessing capabilities, i.e. all the simulations have been carried out using a single processor core. The algorithms for octree encoding and evaluation have been ad-hoc implemented in ANSI C using a recursive data structure as mentioned in Section II, while the postprocessing has been done using MATLAB R2012a.

Table VII shows the computation times for the training phase. As can be appreciated, the most time consuming items are the 10^4 AC and transient Monte Carlo simulations to generate the training data sets. These two take about 1 min 17 s to be completed. The application of the multiple specification band guarding strategy using the presented 33×33 grid search for the band guarding factors takes about 1 min to complete. This means that, with a training set consisting of 10^4 circuits, an octree data structure encoding the acceptance/rejection regions is computed and stored to disk in just 54 ms. The total computation time devoted to the training phase is slightly over 2 min for the considered Butterworth filter case study.

Regarding the testing phase, Table VIII lists the required computation times. Again, the generation of the 10^5 circuits using Monte Carlo simulations is the bottle neck of the testing phase. The AC and transient simulations for the Butterworth

TABLE VII COMPUTATION TIME FOR TRAINING PHASE

Item	Time
10k circuits AC simulation	18.1 s
10k circuits TRAN simulation	59.3 s
33×33 grid octree encodings with 10k circuits	58.8 s
Total time for training phase	2 min 16 s

TABLE VIII
COMPUTATION TIME FOR TESTING PHASE

Item	Time
100k circuits AC simulation	3 min 13 s
100k circuits TRAN simulation	9 min 57 s
33×33 grid octree evaluations with 100k circuits	4 min 0 s
Total time for testing phase	17 min 10 s

filter take about 10 min 10 s. The evaluation of these 10^5 circuits in the previously encoded octrees for each of the (r_1,r_2) band guarding factors within the considered grid takes about 4 min to be completed. In other words, the evaluation of 10^5 circuit samples in a single octree including the storage of the results to disk can be done in just 220 ms what implies that a single circuit can be clustered in 2.2 μ s in average. These remarkable evaluation performances can be achieved because only one float number comparison per coordinate and level traversed is required. For the considered case study, the total computation time devoted to the testing phase is 17 min 10 s.

VI. CONCLUSIONS

A multiple specification band guarding technique for testing mixed-signal circuits in the indirect measurement space has been proposed. The method is based on a set of factors which are in charge of shrinking or enlarging the CUT specifications cube. An adequate tuning of the band guarding factors allows the optimization of any test target involving the number of misclassified circuits by the test methodology. The test method comprises two phases. The training phase encodes the indirect measurement using octrees to represent the acceptance/rejection test regions. In this phase, the multiple specification band guarding methodology tunes the CUT specifications. The production testing phase uses the octree data structure in order to efficiently test the forthcoming circuits taking advantage of the inherent octree sparsity.

The proposed multiple specification band guarding methodology has been applied to test a band-pass Butterworth filter with promising results. The band guarding factors have been set to act over the gain and bandwidth specifications of the CUT. The optimum selection of band guarding factors has allowed the achievement of the initially fixed test targets. Simulations conducted under the presence of noise have revealed the proposed band guarding methodology is time efficient and also usable under noisy indirect measurements.

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