

**ROBUSTIZING CIRCUIT OPTIMIZATION  
USING HUBER FUNCTIONS**

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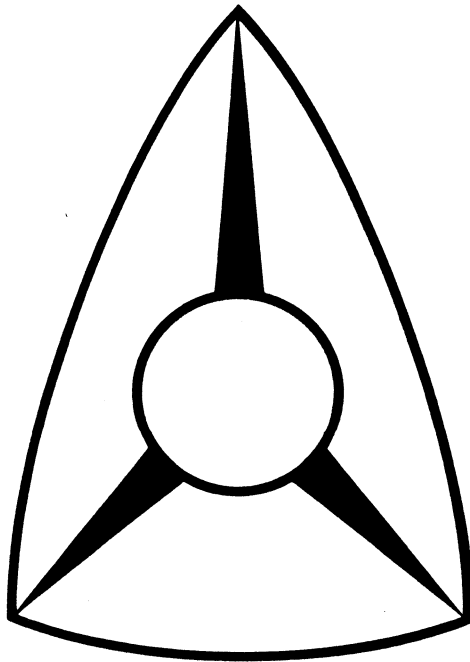
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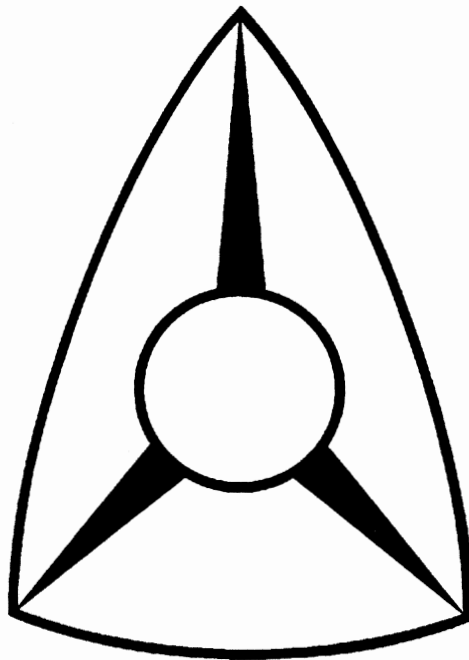
Optimization Systems Associates Inc.  
P.O. Box 8083, Dundas, Ontario  
Canada L9H 5E7



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## **Abstract**

We introduce a novel approach to "robustizing" circuit optimization using Huber functions: both two-sided and one-sided. We compare Huber optimization with  $\ell_1$ ,  $\ell_2$  and minimax methods in the presence of faults, large and small measurement errors, bad starting points and statistical uncertainties. We demonstrate FET statistical modeling, multiplexer optimization, analog fault location and data fitting.



## **Introduction**

designers are concerned with the robustness of numerical optimization techniques

engineering data is frequently contaminated by model/measurement/statistical errors

the classical least-squares method is vulnerable to gross errors: a few wild data points can alter the least-squares solution significantly

the  $\ell_1$  method is robust against gross errors, but it can be undesirably biased when the data contains many small variations

the Huber optimization is more robust than  $\ell_2$  w.r.t. large errors, and smoother, less biased than  $\ell_1$

the benefits of this novel approach are demonstrated by FET statistical modeling, analog fault location and data fitting

the one-sided Huber function can be more effective and efficient than minimax in overcoming a bad starting point for large-scale optimization problems



## Huber Functions

$$\rho_k(f) = \begin{cases} f^2/2 & \text{if } |f| \leq k \\ k|f| - k^2/2 & \text{if } |f| > k \end{cases}$$

where  $k$  is a positive constant

the Huber function  $\rho_k$  is a hybrid of the least-squares ( $\ell_2$ ) (when  $|f| \leq k$ ) and the  $\ell_1$  (when  $|f| > k$ )

if  $k$  is sufficiently large, the Huber function becomes the  $\ell_2$  function; if  $k$  approaches zero,  $\rho_k$  approaches the  $\ell_1$  function

## Huber Optimization

$$\underset{x}{\text{minimize}} \quad F(x) \triangleq \sum_{j=1}^m \rho_k(f_j(x))$$

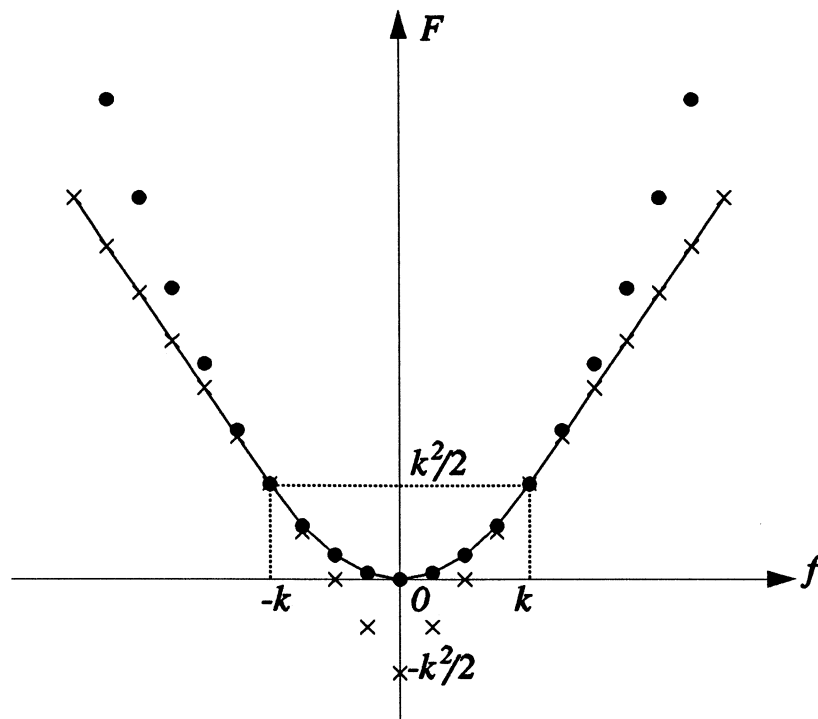
where

$x = [x_1 \ x_2 \ \dots \ x_n]^T$  is the set of variables

$f_j \ j = 1, 2, \dots, m$ , are the error functions



## Huber Function as a Hybrid $\ell_1 / \ell_2$



the Huber,  $\ell_1$  and  $\ell_2$  objective functions in the one-dimensional case

the strikes represent the discrete points on the  $\ell_1$  curve

the dots represent the discrete points on the  $\ell_2$  curve

the continuous curve indicates the Huber objective function



## **Comparison in Data Fitting**

approximating  $\sqrt{t}$  by a rational function for  $0 \leq t \leq 1$

large errors deliberately introduced at 5 of the sample points  
and small variations added to the remaining data

the  $\ell_1$ ,  $\ell_2$  and Huber approximations are compared

$\ell_2$  solution suffers significantly from the presence of gross  
errors

$\ell_1$  solution is dictated by a subset of residual functions which  
have zero values at the solution and all the nonzero residuals  
are treated as large errors

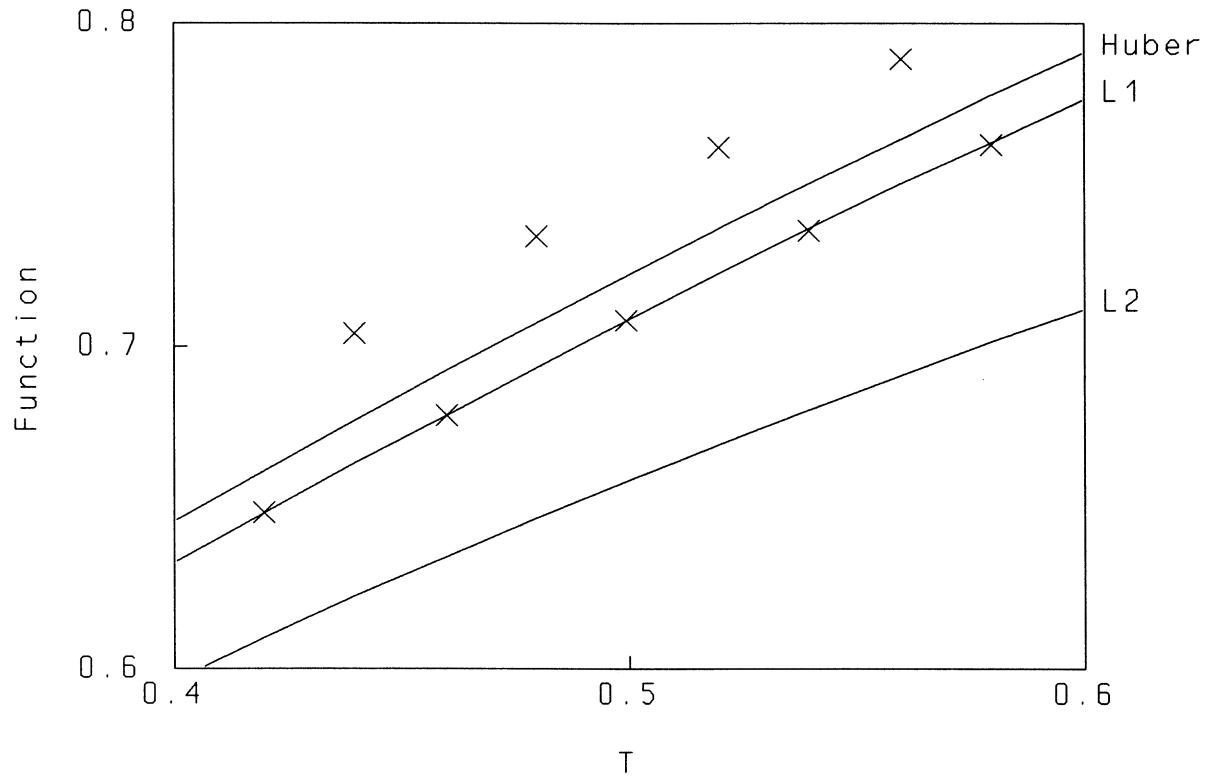
the Huber solution provides a flexible combination of the  
robustness of the  $\ell_1$  and the unbiasedness of the  $\ell_2$

the Huber solution is equivalent to an  $\ell_2$  solution with the  
gross errors reduced to the threshold value  $k$





## $\ell_1$ , $\ell_2$ and Huber Data Fitting



$\ell_1$ ,  $\ell_2$  and Huber solutions for data fitting in the presence of large and small errors: an enlarged view



## **Huber Estimator for Statistical Modeling**

use the Huber function as an automated robust estimator for FET statistical modeling

model parameters are extracted from the measurements of 80 FETs using HarPE™ and then postprocessed to estimate the parameter statistics

to estimate the mean we define the error functions as

$$f_j(\bar{\phi}) = \bar{\phi} - \phi^j, \quad j = 1, 2, \dots, N$$

where  $\phi^j$  is the extracted parameter value for the  $j$ th device and  $N$  is the total number of devices

choose threshold  $k$  for the Huber function according to the normal spread of the parameter values (e.g.,  $k = 0.25$  for  $\tau$ )

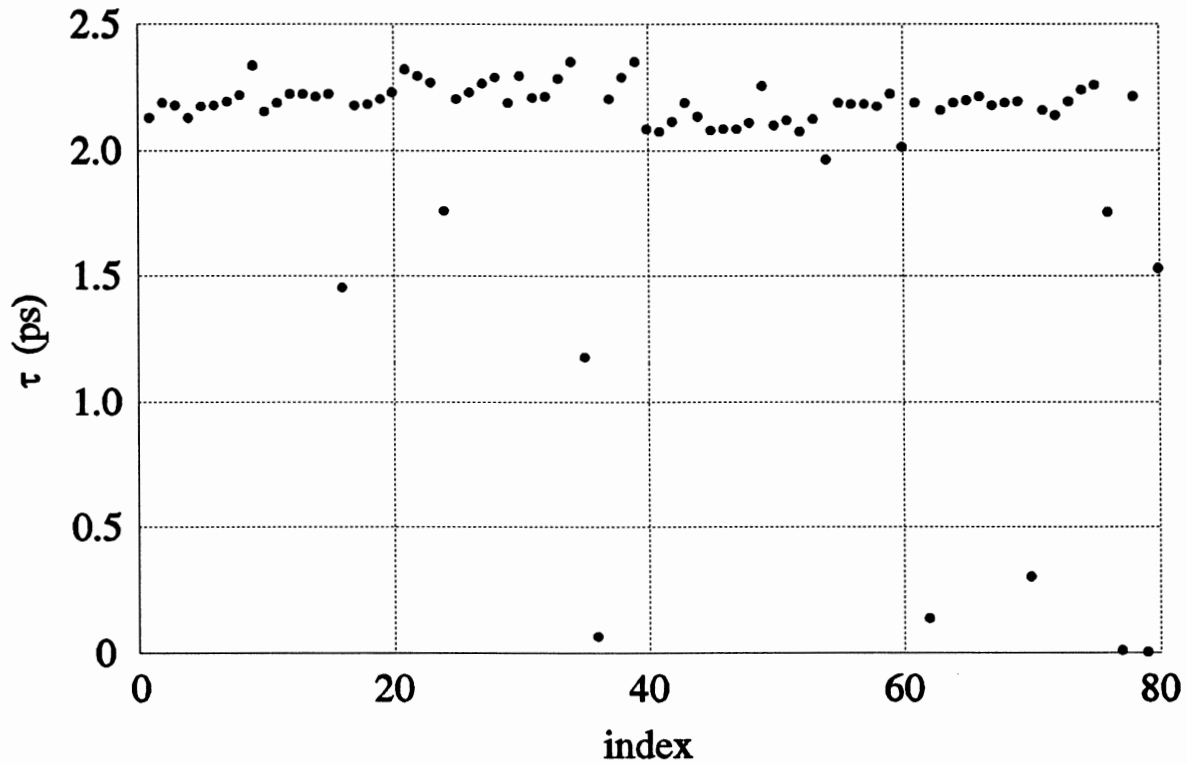
to estimate the variance we define the error functions as

$$f_j(V_\phi) = V_\phi - (\phi^j - \bar{\phi})^2, \quad j = 1, 2, \dots, N$$

where  $V_\phi$  denotes the estimated variance



## Data Containing Wild Points



run chart of the extracted FET time-delay  $\tau$

a few abnormal values in the data due to faulty devices  
and/or gross measurement errors

manual deletion has been adopted to exclude abnormal data  
points in earlier work using  $\ell_2$  estimator



## Statistical Modeling Using Huber Estimator

### ESTIMATED STATISTICS OF SELECTED FET PARAMETERS

Parameter	$\bar{\phi}(\ell_2)$	$\bar{\phi}(H)$	$\bar{\phi}(\ell_2^*)$	$\sigma_{\phi}(\ell_2)$	$\sigma_{\phi}(H)$	$\sigma_{\phi}(\ell_2^*)$
$L_G(\text{nH})$	0.04387	0.03464	0.03429	94.6%	21.8%	17.4%
$G_{DS}(1/\text{K}\Omega)$	1.840	1.820	1.839	28.6%	6.3%	4.9%
$I_{DSS}(\text{mA})$	47.36	47.53	47.85	14.0%	12.7%	11.3%
$\tau(\text{ps})$	2.018	2.154	2.187	26.3%	5.8%	3.4%
$C_{10}(\text{pF})$	0.3618	0.3658	0.3696	8.2%	4.6%	3.5%
$K_1$	1.2328	1.231	1.233	15.5%	10.8%	8.7%

$\bar{\phi}$  denotes the mean and  $\sigma_{\phi}$  the standard deviation

H denotes Huber estimates

$\ell_2^*$  denotes  $\ell_2$  estimates after 11 abnormal data sets are manually excluded

$\ell_2$  and Huber estimates of the statistics for selected model parameters

Huber estimator does not require manual manipulation of the data and is more appropriate when there are data points which cannot be clearly classified as abnormal



## **Analog Diagnosis Using Huber Optimization**

penalty function approach

$$\underset{x}{\text{minimize}} \quad \sum_{j=1}^{n+K} \rho_k(f_j(x))$$

where

$$f_i(x) = \Delta x_i / x_i^0, \quad i = 1, 2, \dots, n$$

$$f_{n+i}(x) = \beta_i(V_i^c - V_i^m), \quad i = 1, 2, \dots, K$$

$\beta_i$                       appropriate multipliers for the  
penalty terms

$x \equiv [x_1 \ x_2 \ \dots \ x_n]^T$       vector of circuit parameters

$x^0$                       nominal values

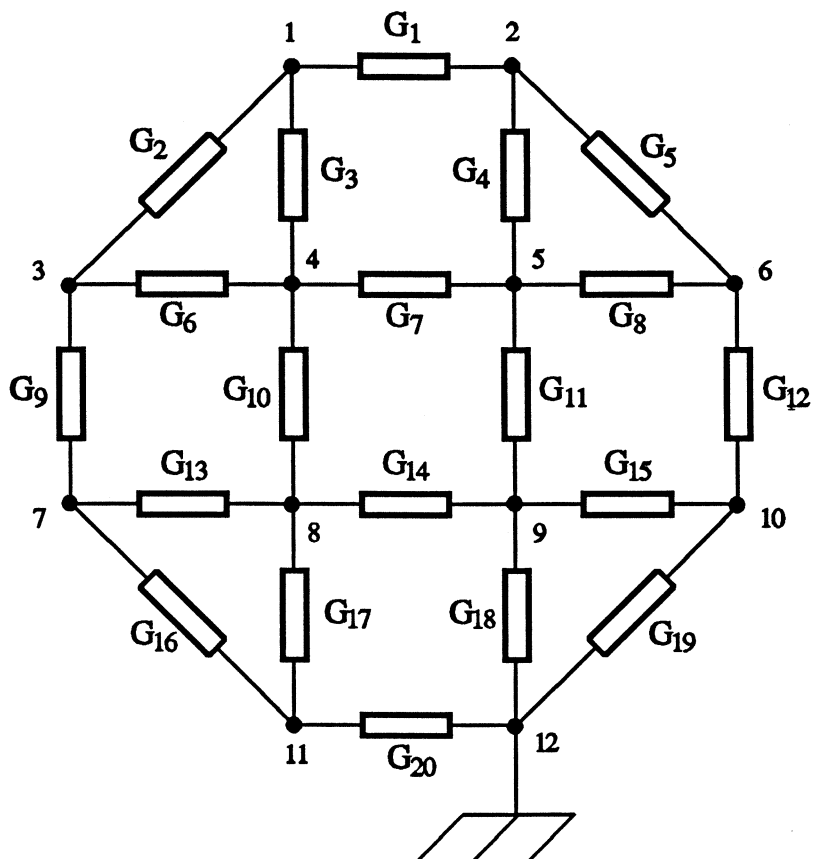
$\Delta x \equiv x - x^0$               deviations from the nominal

$V_1^m, \dots, V_K^m$                $K$  measurements

$V_1^c, \dots, V_K^c$               calculated circuit responses



## Analog Diagnosis Using Huber Optimization



resistive mesh circuit

only external nodes are available for excitation and measurements



## Huber Solution of Analog Diagnosis Problem

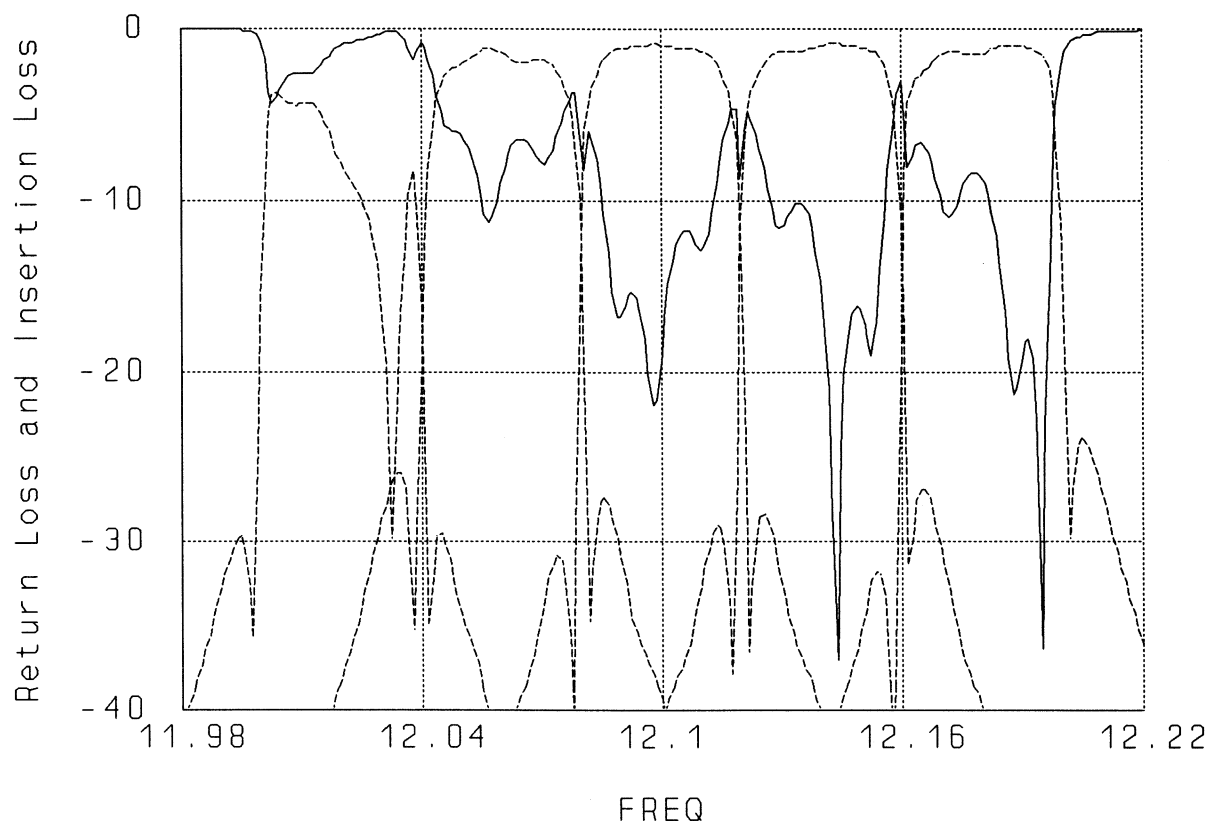
### FAULT LOCATION OF THE RESISTIVE MESH CIRCUIT

Element	Nominal Value	Actual Value	Percentage Deviation		
			Actual	$\ell_1$	Huber
$G_1$	1.0	0.98	-2.0	0.00	-0.11
$G_2$	1.0	0.50	-50.0*	-48.89	-47.28
$G_3$	1.0	1.04	4.0	0.00	-2.46
$G_4$	1.0	0.97	-3.0	0.00	-1.18
$G_5$	1.0	0.95	-5.0	-2.70	-3.16
$G_6$	1.0	0.99	-1.0	0.00	-0.06
$G_7$	1.0	1.02	2.0	0.00	-0.19
$G_8$	1.0	1.05	5.0	0.00	-0.41
$G_9$	1.0	1.02	2.0	2.41	3.75
$G_{10}$	1.0	0.98	-2.0	0.00	0.39
$G_{11}$	1.0	1.04	4.0	0.00	-0.37
$G_{12}$	1.0	1.01	1.0	2.73	1.32
$G_{13}$	1.0	0.99	-1.0	0.00	-0.26
$G_{14}$	1.0	0.98	-2.0	0.00	-0.50
$G_{15}$	1.0	1.02	2.0	0.00	-0.05
$G_{16}$	1.0	0.96	-4.0	-3.36	-2.67
$G_{17}$	1.0	1.02	2.0	0.00	-0.61
$G_{18}$	1.0	0.50	-50.0*	-50.09	-47.33
$G_{19}$	1.0	0.98	-2.0	-1.41	-3.81
$G_{20}$	1.0	0.96	-4.0	-4.40	-4.72

\* Faults



## **Five-channel 12 GHz Waveguide Manifold Multiplexer**



responses before optimization

(——) common port return loss

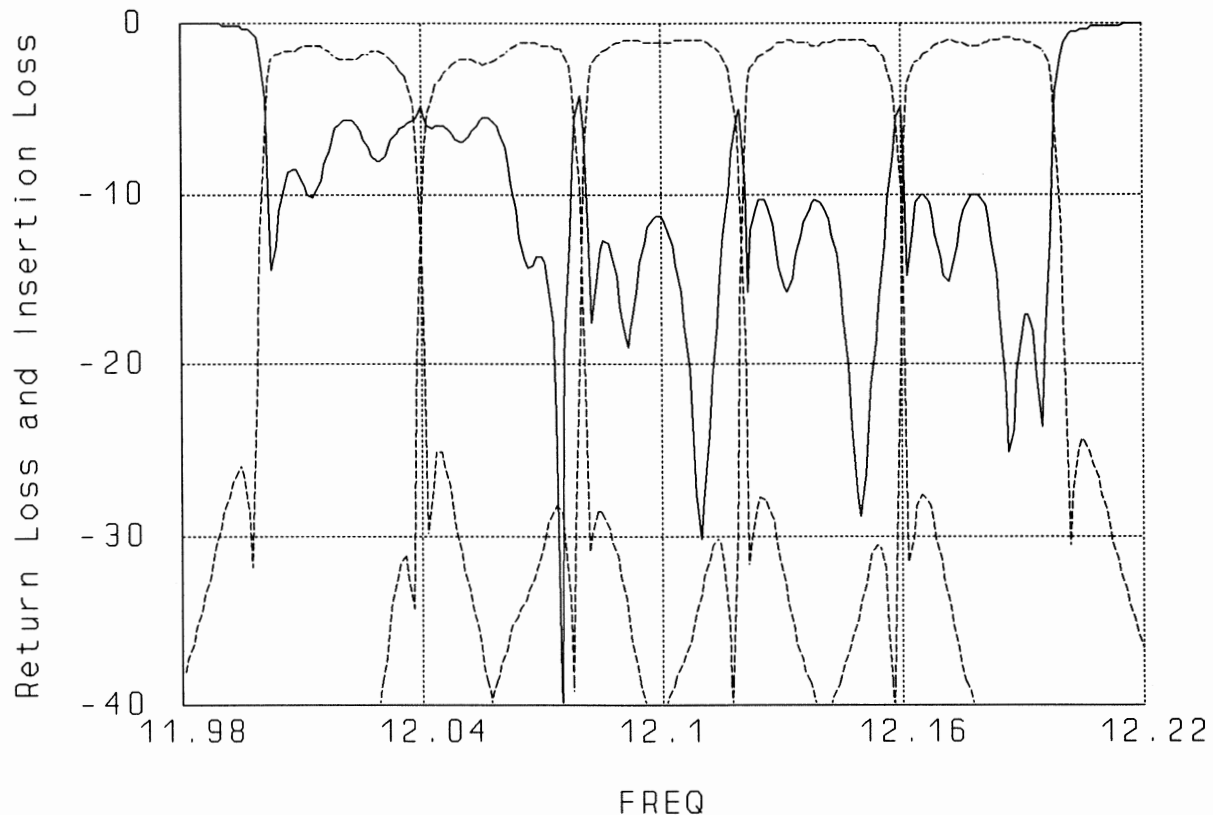
(-----) individual channel insertion losses

75 optimizable variables





## **Partial Minimax Optimization of the 5-Channel Multiplexer**



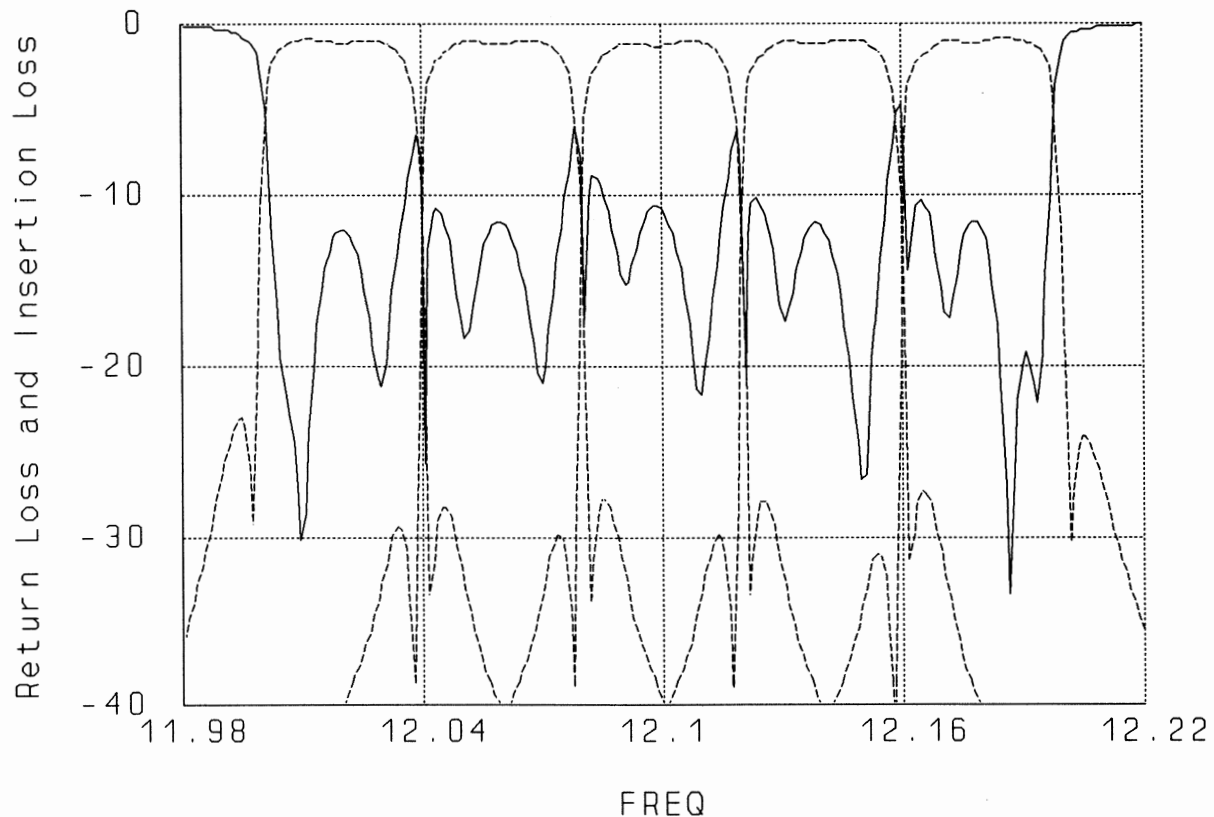
**multiplexer responses after minimax optimization with 10 variables: spacings and channel input transformer ratios**

**worst-case errors cannot be further reduced with only 10 variables by minimax**

**hardly improved upon the starting point**



## **Partial Huber Optimization of the 5-Channel Multiplexer**



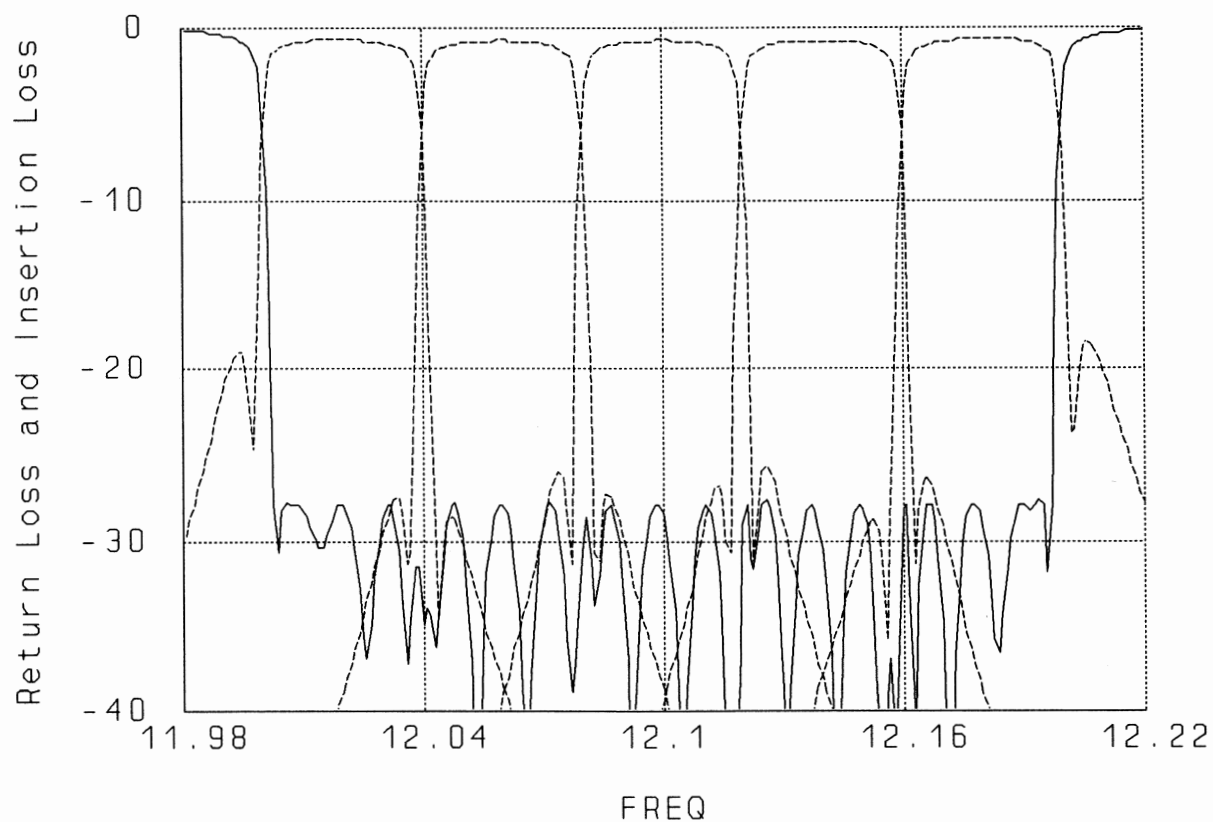
**multiplexer responses after one-sided Huber optimization  
with 10 variables**

**significantly better than minimax solution with the same 10  
variables**

**provides a good starting point for full-scale minimax  
optimization**



## **Full-scale Minimax Optimization of the 5-Channel Multiplexer**



**multiplexer responses after minimax optimization with the full set of 75 variables**



## **Conclusions**

we have introduced the concept and applications of the Huber method to circuit CAD

the novel Huber concept is consistent with practical engineering intuition

the Huber method will have a far-reaching and profound impact on modeling, design, design validation, fault diagnosis and statistical processing of circuits and devices

we have presented strong evidence in a number of application areas