

**PROPOSED ALGORITHMS FOR THE AGGRESSIVE
SPACE MAPPING TECHNIQUE**

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SOS-98-41-V

November 1998

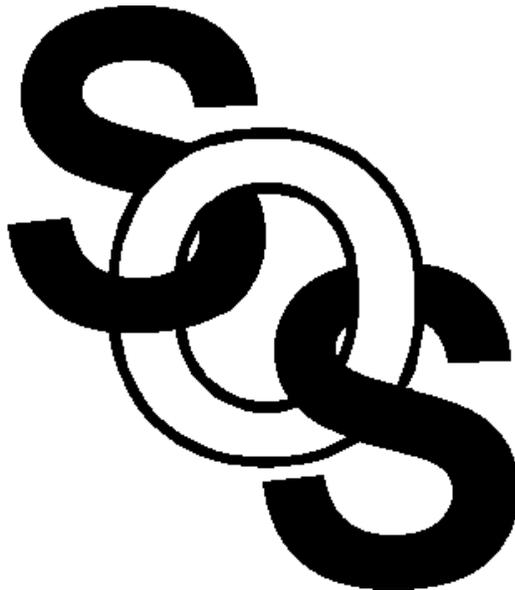
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PROPOSED ALGORITHMS FOR THE AGGRESSIVE SPACE MAPPING TECHNIQUE

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presented at

Supervisory Committee Meeting, Hamilton, November 1998



Basic Concepts

it is assumed that the circuit under consideration can be simulated using two models: a fine model and a coarse model

the fine model is accurate but is computationally intensive

\mathbf{x}_{em} is the vector of fine model parameters

the coarse model is a fast model but less accurate than the fine model

\mathbf{x}_{os} is the vector of coarse model parameters

space mapping aims at avoiding the computationally intensive direct optimization of the fine model

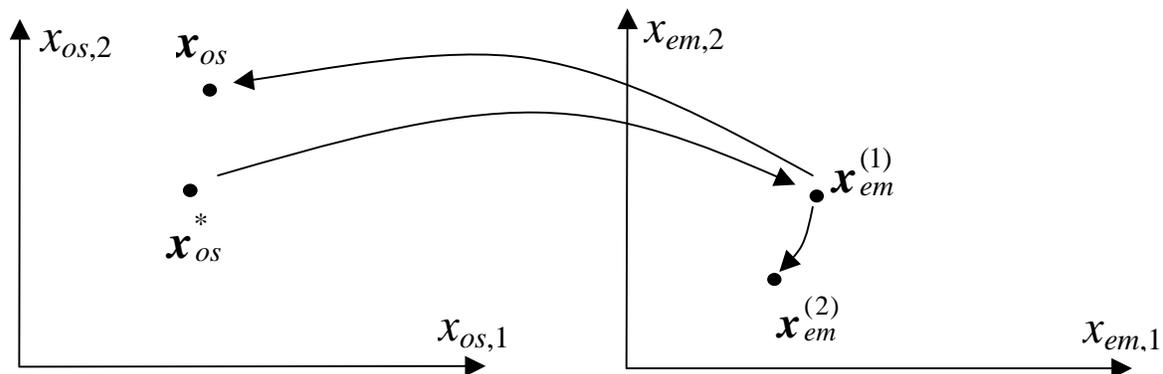


The Aggressive Space Mapping (ASM) Algorithm

the initial fine model design is the optimal coarse model design

$$\mathbf{x}_{os}^*$$

parameter extraction is used to predict the step to be taken in the fine model space



in the i th iteration the new iterate is given by

$$\mathbf{x}_{em}^{(i+1)} = \mathbf{x}_{em}^{(i)} + \mathbf{h}^{(i)}$$

and $\mathbf{h}^{(i)}$ is obtained by solving

$$\mathbf{B}^{(i)} \mathbf{h}^{(i)} = -\mathbf{f}(\mathbf{x}_{em}^{(i)})$$

where

$$\mathbf{f} = \mathbf{P}(\mathbf{x}_{em}^{(i)}) - \mathbf{x}_{os}^*$$

the nonuniqueness of the parameter extraction problem can lead to divergence or oscillation of the algorithm



The Trust Region Aggressive Space Mapping (TRASM) Algorithm

this algorithm integrates a trust region methodology with the aggressive space mapping technique

certain success criterion must be satisfied at each iteration so as to accept the predicted step

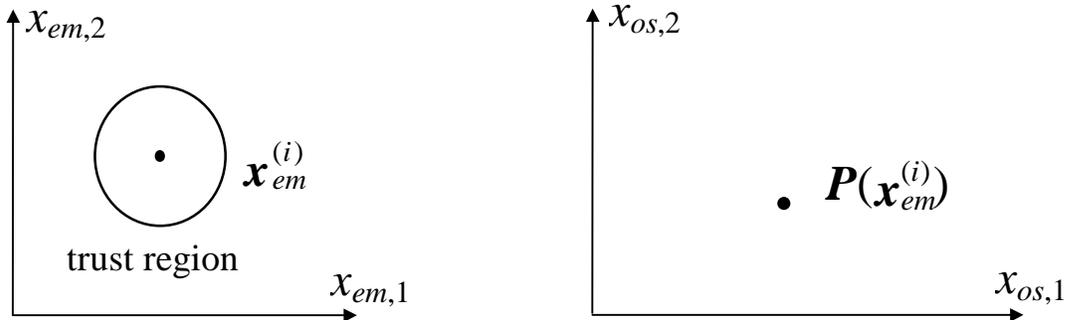
a recursive multi-point parameter extraction procedure was introduced in the context of this algorithm

all available fine model simulations are utilized in order to improve the uniqueness of the parameter extraction step

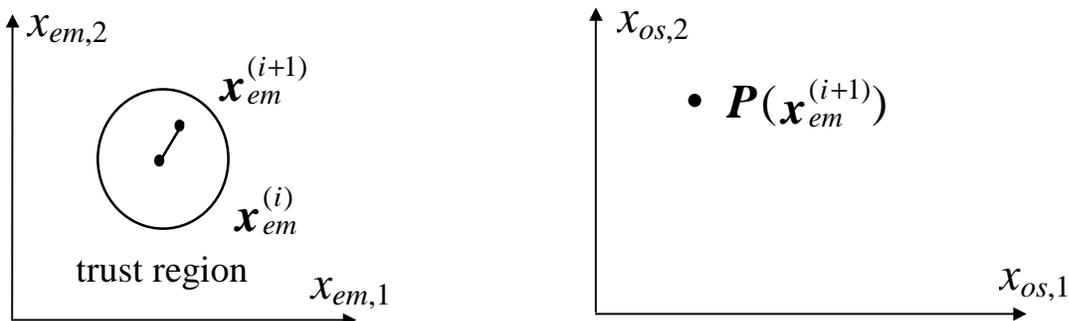
the available information about the mapping between the two spaces is integrated in this extraction procedure



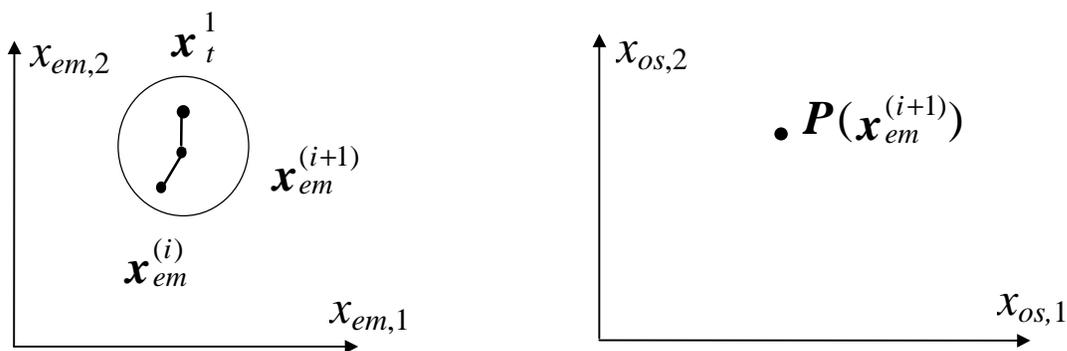
Illustration of the TRASM Algorithm



the current state at the i th iteration



initial parameter extraction at the
suggested point

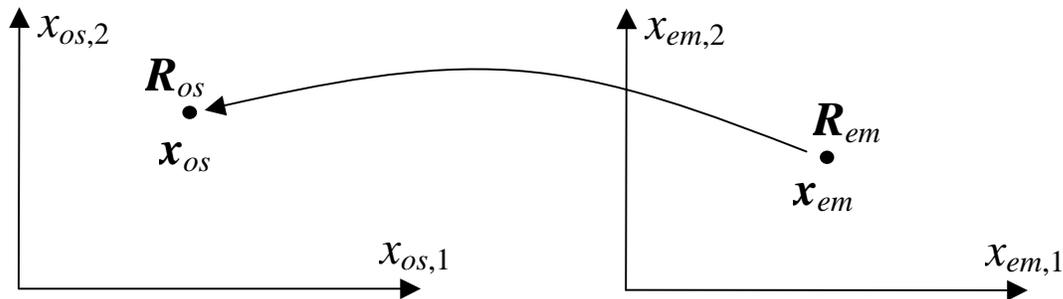


multi-point extraction is applied



Parameter Extraction

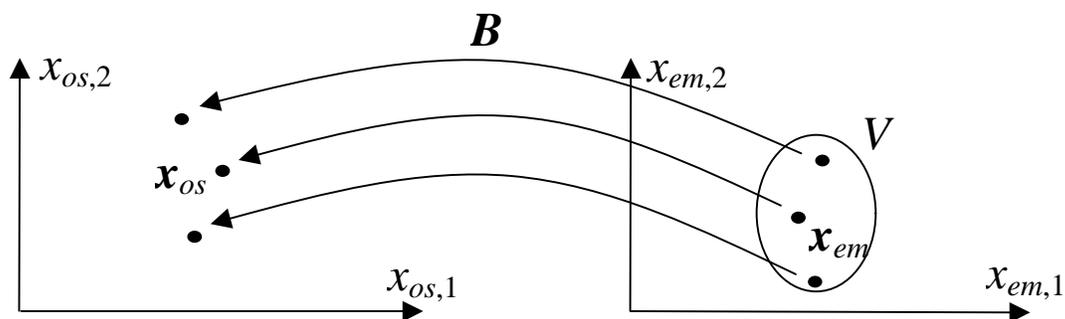
single point parameter extraction aims at matching the responses of both models at a single point



it can be formulated as an optimization problem;

$$\underset{\mathbf{x}_{os}}{\text{minimize}} \left\| \mathbf{R}_{em}(\mathbf{x}_{em}) - \mathbf{R}_{os}(\mathbf{x}_{os}) \right\|$$

multi-point parameter extraction aims at matching the responses of both models at a number of points



the extracted parameters should satisfy

$$\mathbf{R}_{os}(\mathbf{x}_{os} + \mathbf{B}(\mathbf{x} - \mathbf{x}_{em})) = \mathbf{R}_{em}(\mathbf{x})$$

simultaneously for a set of points $\mathbf{x} \in V$; the set of points used for the multi-point parameter extraction where $\mathbf{x}_{em} \in V$



The New Multi-Point Parameter Extraction Algorithm

we propose an algorithm that addresses the selection of the new points used for multi-point parameter extraction

the algorithm classifies the solutions of the parameter extraction problem based on the rank of the Jacobian matrix of the matched responses of the coarse model

a new point that aims at maximizing the improvement in the uniqueness of the parameter extraction step is obtained either by solving an eigenvalue problem or by solving a system of linear equations depending on the rank of the Jacobian matrix

the algorithm terminates when the extracted coarse model parameters in two consecutive iterations are close enough

the algorithm was applied successfully to a number of examples and the results obtained are encouraging



Illustrative Example: A 10:1 Impedance Transformer

the parameters of this problem are the two characteristic impedances of the two transmission lines

the coarse model is an ideal 10:1 impedance transformer

the fine model scales each of the impedances by 1.6

the responses of both models at 11 different frequencies in the band $0.5 \text{ GHz} \leq f \leq 1.5 \text{ GHz}$ are used to match the two models

three iterations are required to trust the extracted parameters

the set of points utilized is

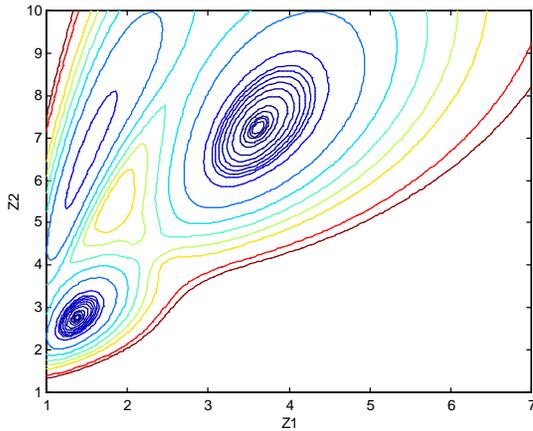
$$V = \left\{ \begin{bmatrix} 2.26277 \\ 4.52592 \end{bmatrix}, \begin{bmatrix} 1.49975 \\ 4.76634 \end{bmatrix}, \begin{bmatrix} 3.02024 \\ 4.26855 \end{bmatrix} \right\}$$

the extracted parameters in each iteration are

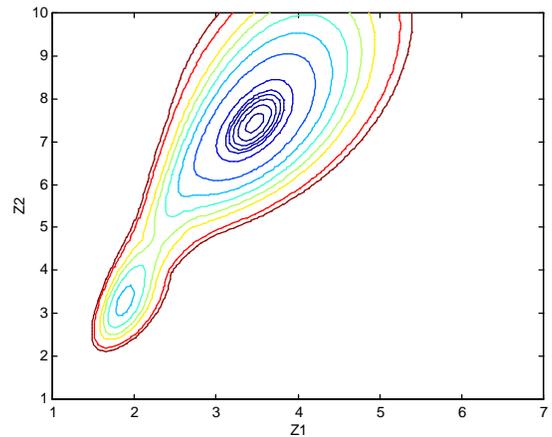
$$\mathbf{x}_{os}^{e,1} = \begin{bmatrix} 3.62043 \\ 7.24147 \end{bmatrix}, \mathbf{x}_{os}^{e,2} = \begin{bmatrix} 3.4716 \\ 7.43214 \end{bmatrix} \text{ and } \mathbf{x}_{os}^{e,3} = \begin{bmatrix} 3.60357 \\ 7.35052 \end{bmatrix}$$



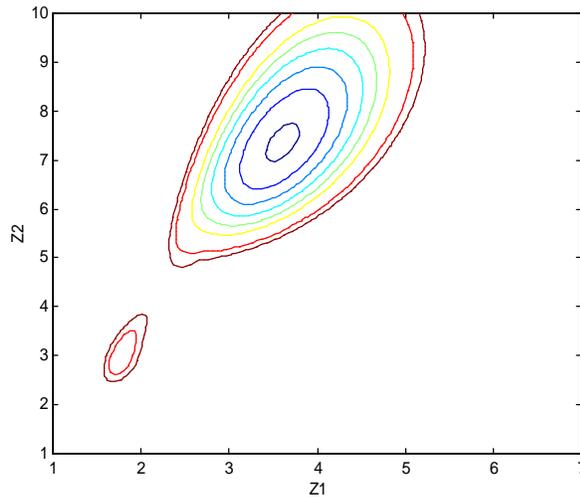
The Contours of the L_2 Objective Function for the 10:1 Transformer



single-point extraction



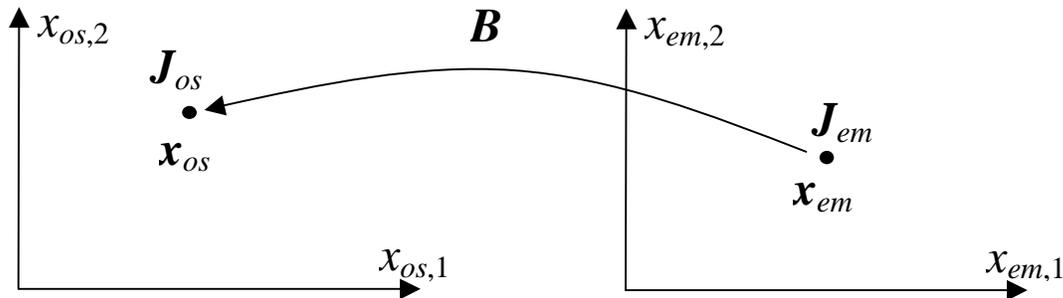
two-point extraction



three-point extraction



Space Mapping and Direct Optimization: Crossing the Gap



\mathbf{x}_{em} and \mathbf{x}_{os} are two corresponding points in the fine and coarse model parameter spaces, respectively

\mathbf{J}_{em} and \mathbf{J}_{os} are the Jacobians of the fine model responses and the coarse model responses at the points \mathbf{x}_{em} and \mathbf{x}_{os} , respectively

the mapping \mathbf{B} between the two spaces at the point \mathbf{x}_{em} is related to the matrices \mathbf{J}_{em} and \mathbf{J}_{os} by

$$\mathbf{J}_{em} = \mathbf{J}_{os} \mathbf{B}$$

it follows that the matrix \mathbf{B} can be approximated by

$$\mathbf{B} = \left(\mathbf{J}_{os}^T \mathbf{J}_{os} \right)^{-1} \mathbf{J}_{os}^T \mathbf{J}_{em}$$



Mapping-Based Modeling of Microwave Circuits

a fast and accurate model for a microwave circuit can be constructed using the mapping estimate

at any fine model point \mathbf{x}_{em}^* the corresponding coarse model point \mathbf{x}_{os} is obtained by applying single point parameter extraction

using an estimate of the Jacobian matrices \mathbf{J}_{em} and \mathbf{J}_{os} the corresponding mapping between the two spaces can be constructed

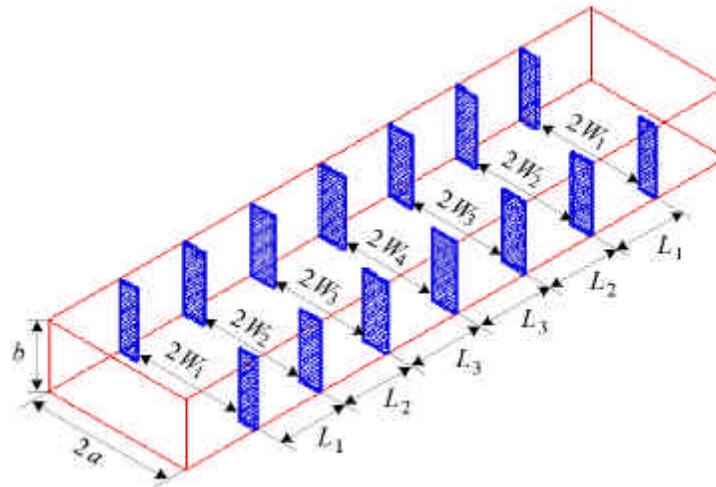
the mapping-based model is given by

$$\mathbf{R}_{em}(\mathbf{x}_{em}) \approx \mathbf{R}_{os}(\mathbf{x}_{os} + \mathbf{B}(\mathbf{x}_{em} - \mathbf{x}_{em}^*))$$

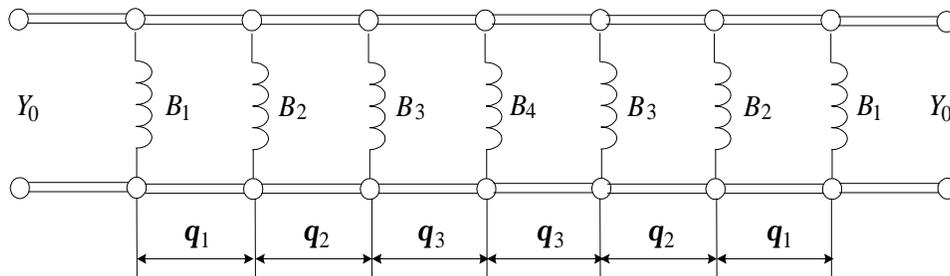
this modeling technique enables mapping-based modeling of microwave circuits without the need to carry out space mapping optimization



Example: Statistical Analysis of Six-Section H-Plane Waveguide Filter



fine model



coarse model

the designable parameters are W_1 , W_2 , W_3 , W_4 , L_1 , L_2 and L_3

the fine model is optimized and the mapping between the two spaces is established

the yield is estimated using the mapping-based model and using fine model simulations



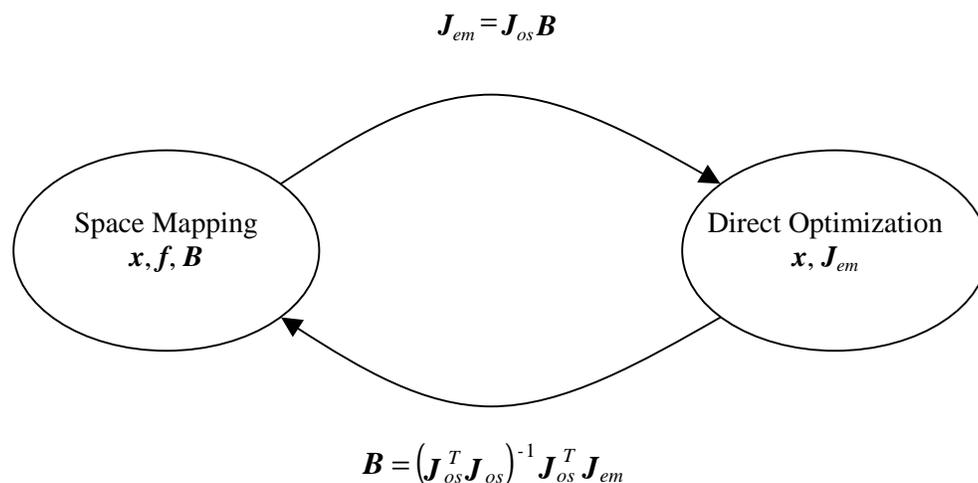
The Hybrid Aggressive Space Mapping Algorithm

space mapping assumes that the coarse model is sufficiently accurate

if the coarse model is severely different from the fine model space mapping is unlikely to converge

the final solution of the problem may not be the optimal fine model design due to the differences between the two models

this suggested algorithm defaults to the computationally efficient space mapping if convergence is smooth and defaults to direct optimization otherwise





Example: A Three-Section Waveguide Transformer

the design constraints are

$$vswr \leq 1.04 \text{ for } 5.7 \text{ GHz} \leq f \leq 7.2 \text{ GHz}$$

the designable parameters are the heights of the waveguide sections b_1 , b_2 and b_3 and the lengths of waveguide sections L_1 , L_2 and L_3

the fine model exploits HP HFSS through HP Empire3D

the coarse analytical model does not take into account the junction discontinuity effects

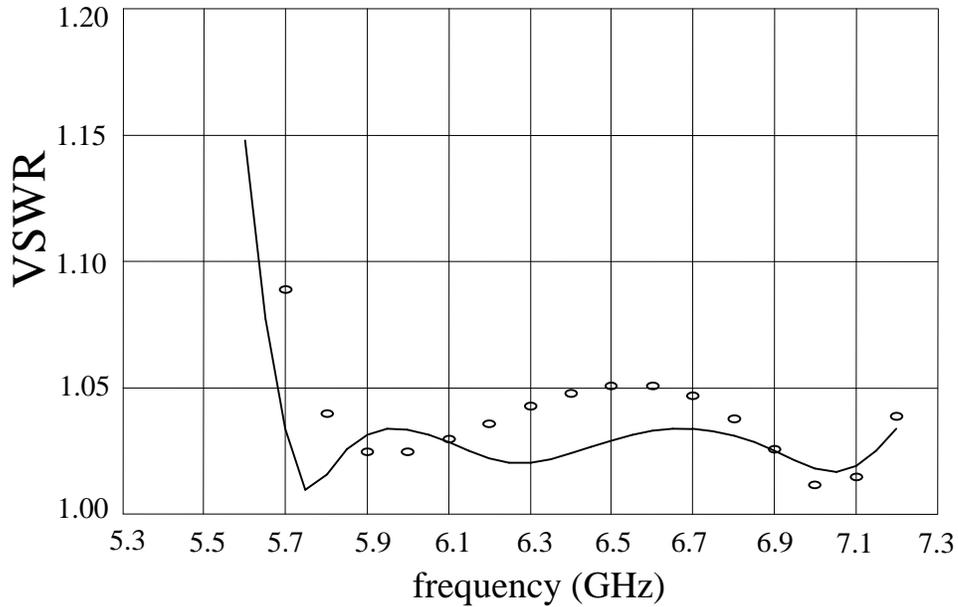
the first phase executed two iterations which required 4 fine model simulations

the second phase carries out only one iteration which required 2 fine model simulations

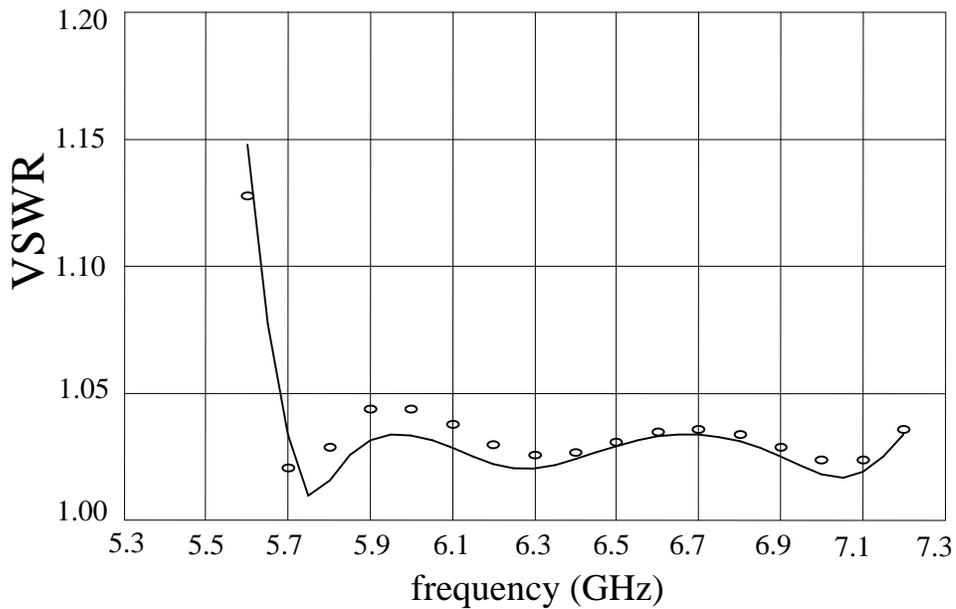
minimax optimization is then applied to the original problem starting from the second phase design



Example: A Three-Section Waveguide Transformer



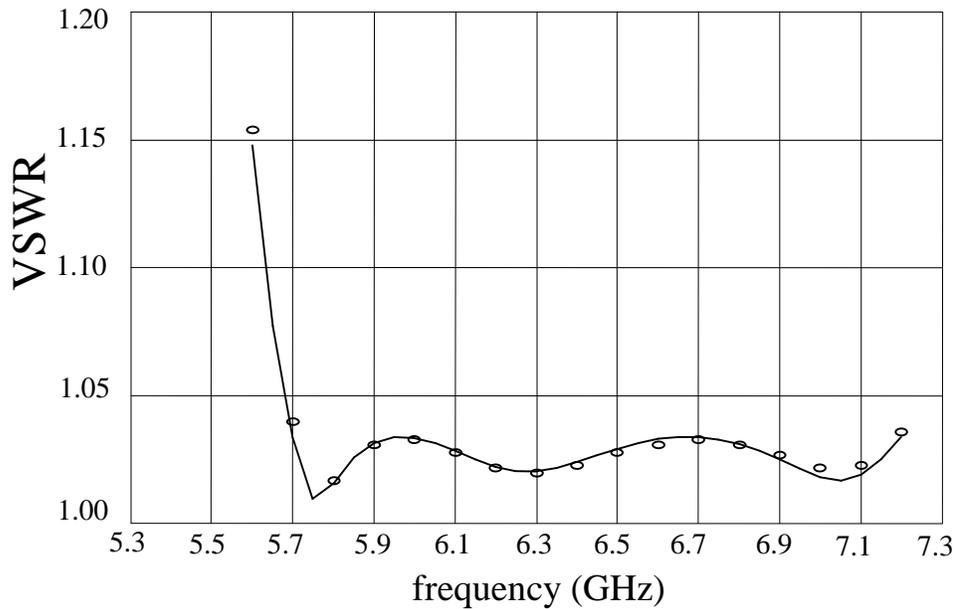
initial response of the fine model



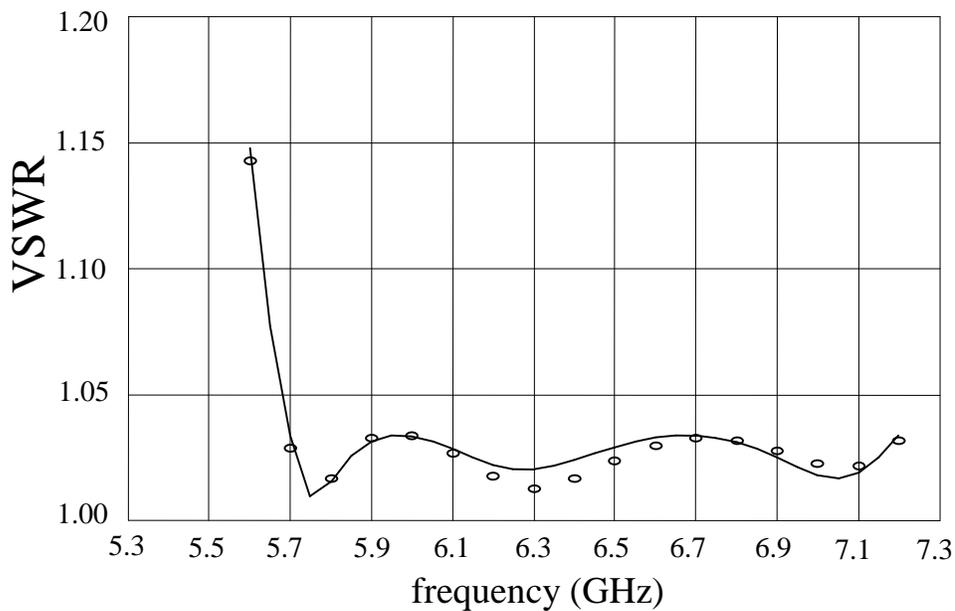
the first phase design



Example: A Three-Section Waveguide Transformer



the second phase design



the optimal fine model design



Example: The Six-Section H-plane Waveguide Transformer

the coarse model consists of lumped inductances and dispersive transmission line sections

a simplified version of a formula due to Marcuvitz is utilized in evaluating the inductances

the fine model exploits HP HFSS through HP Empire3D

the first phase executed 4 iterations requiring a total of 5 fine model simulations

the second phase did not carry out any successful iterations

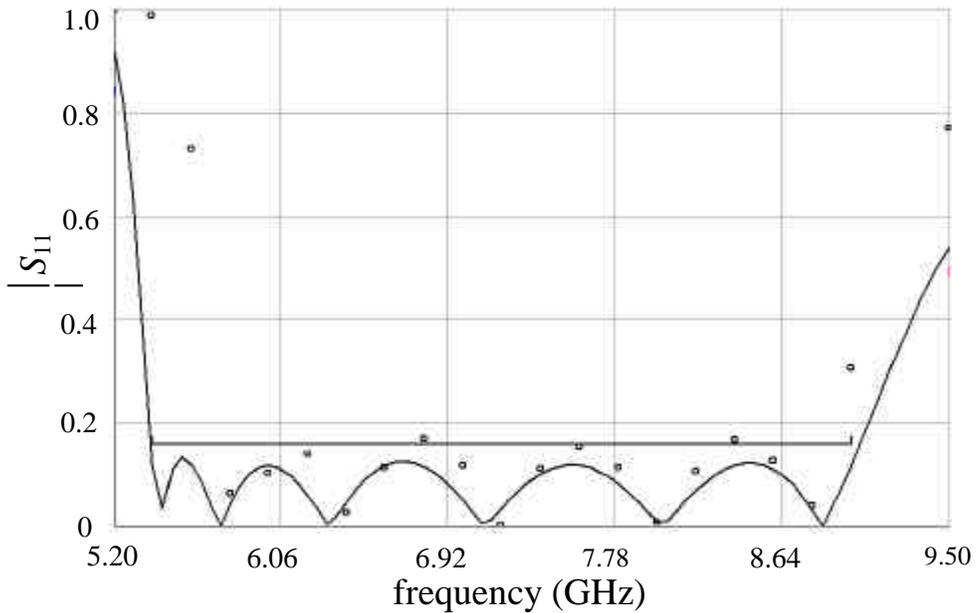
the optimal fine model design is obtained using minimax optimization

the convergence of the TRASM algorithm is smooth for this problem

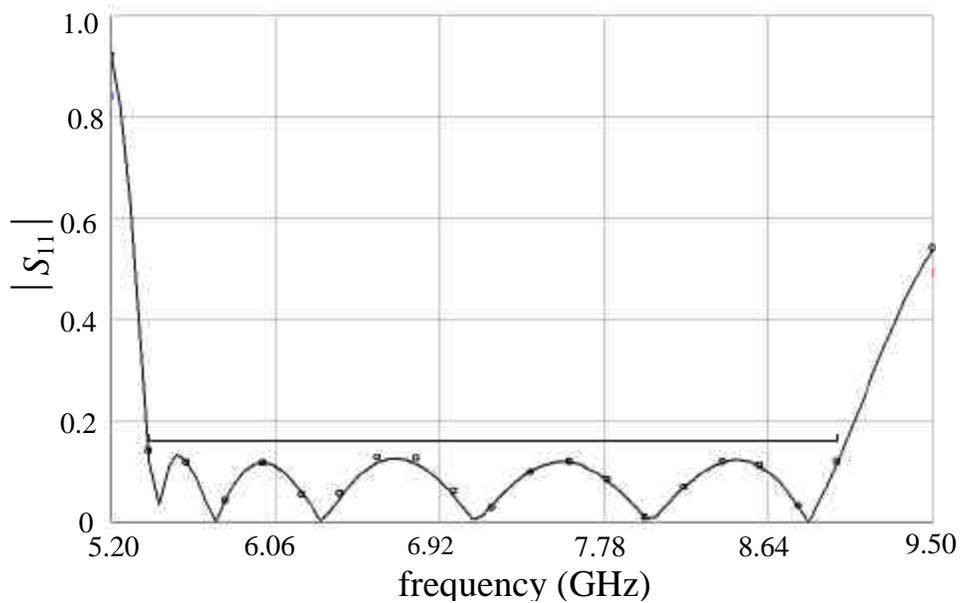
the fine model response at the end of the first phase is almost identical to the optimal fine model response



Example: The Six-Section H-plane Waveguide Transformer



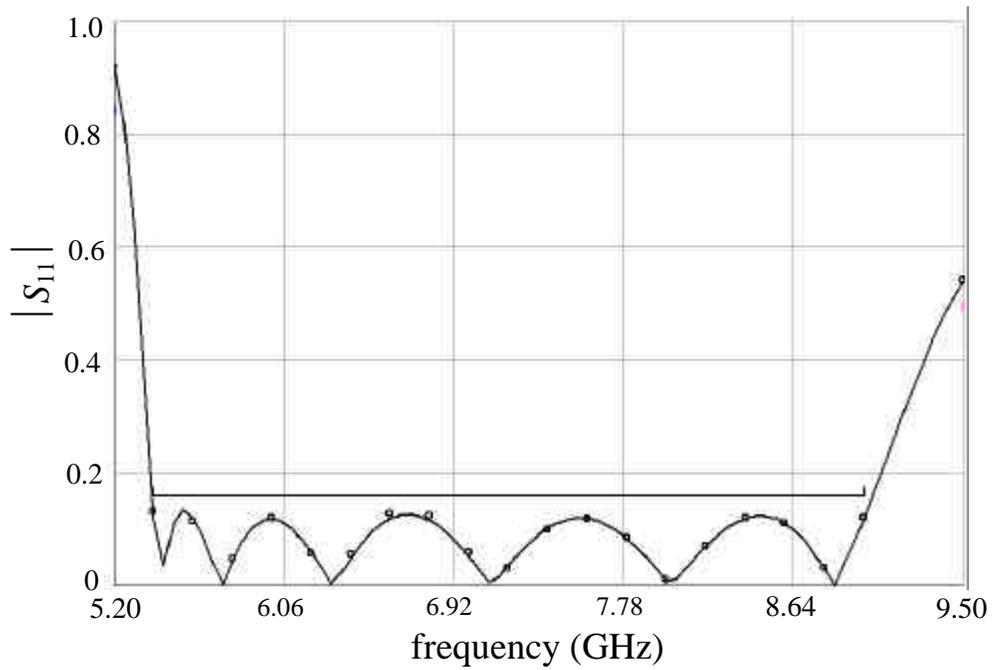
the initial fine model response



the second phase design



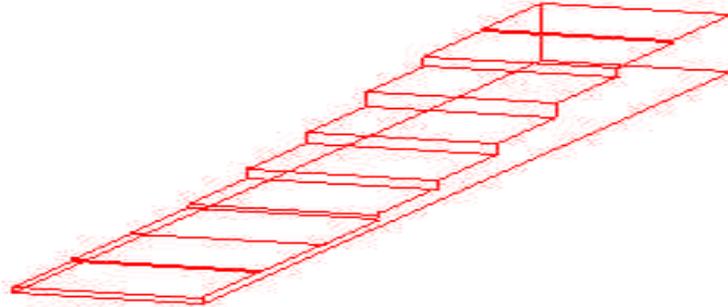
Example: The Six-Section H-plane Waveguide Transformer



the optimal fine model design



Example: The Seven-Section Waveguide Transformer



design specifications are

$$VSWR \leq 1.01 \text{ for } 1.08 \text{ GHz} \leq f \leq 1.8 \text{ GHz}$$

designable parameters for this problem are the height and length of each waveguide section

the fine model is simulated using HP HFSS through HP Empipe3D

the coarse model is an analytical model which neglects the junction discontinuity

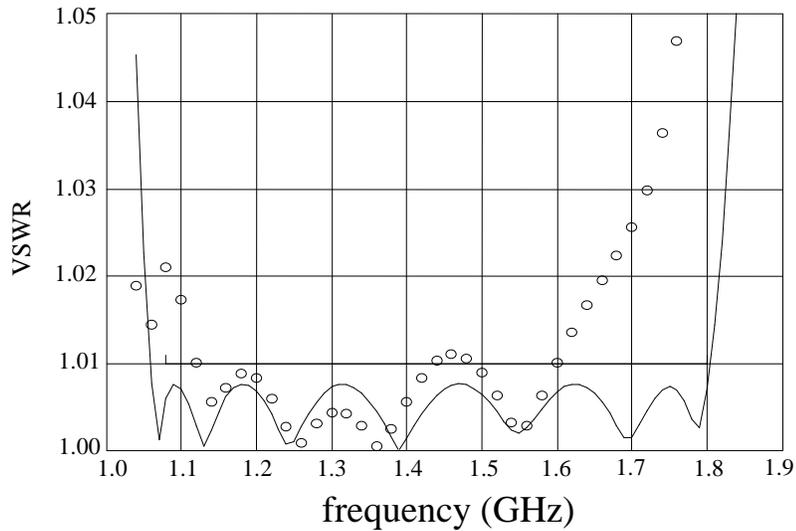
the first executed three successful iterations that required six fine model simulations

the second phase executed four iterations

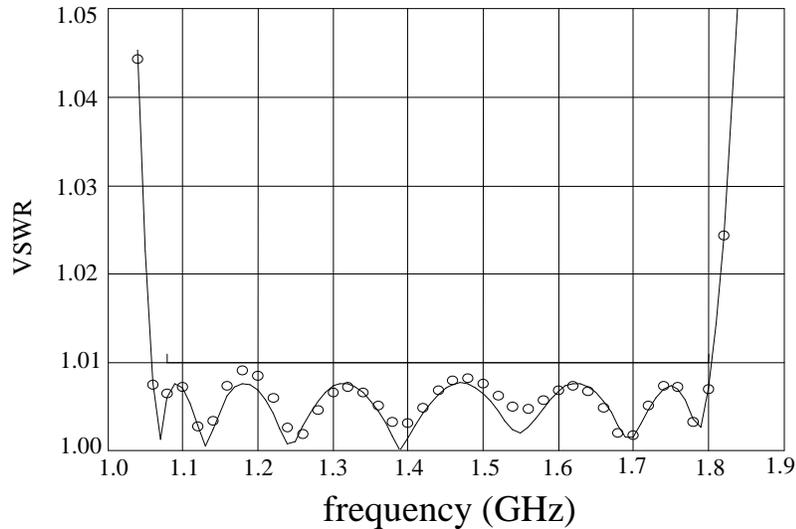
the convergence of this problem is also smooth as the fine model response obtained at the end of the first phase is similar to the optimal fine model response



Example: The Seven-Section Waveguide Transformer



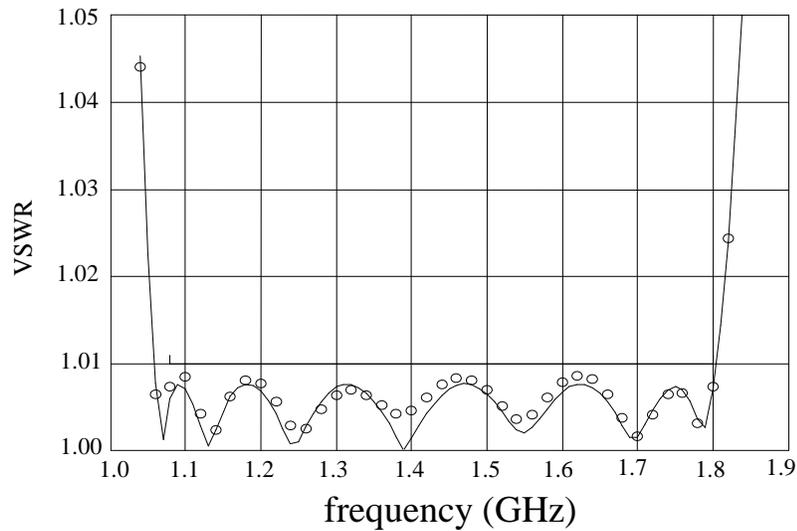
initial design of the fine model



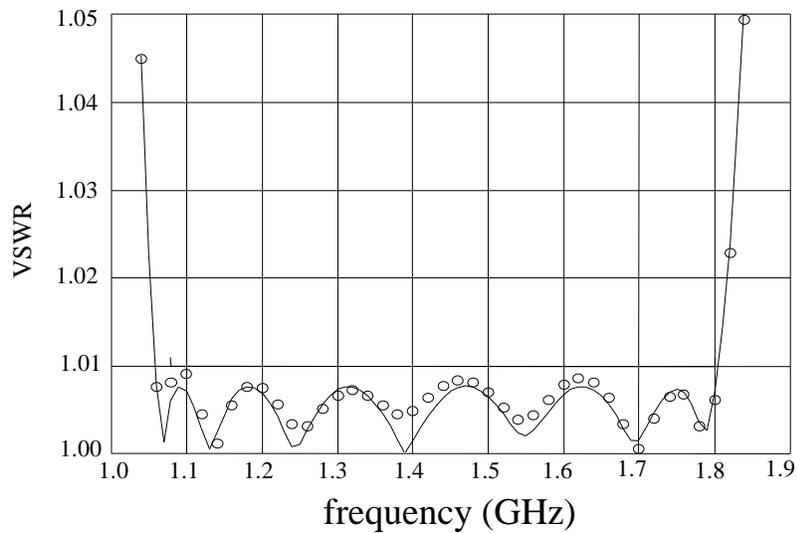
first phase design



Example: The Seven-Section Waveguide Transformer



second phase design



optimal fine model design