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THE SYNTHESIS OF MINIMUM SENSITIVITY NETWORKS

By

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January 21, 1963

Technical Report Number Six

Engineering Division

Case Institute of Technology

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The synthesis of networks with minimum sensitivity to element tolerances is studied from a computer viewpoint. The theory of equivalent networks is used to generate a sequence of networks whose transfer functions are identical to that of a given network but whose elements differ from one network to the next by an incremental amount. In the limit, differential equations result whose solution at any value of the independent variable give the elements of an equivalent network. Similarly, differential equations for the sensitivity of the transfer function to changes in each of the elements are derived. The differential equations in both cases are linear homogeneous with the elements of the transformation matrix as the forcing functions. With the aid of the exponential solution to the matrix differential equation, digital computer solution even for complex networks is very straightforward and rapid. As a measure of the optimality of the network, the sum of the magnitudes of the sensitivities is chosen as a performance criterion. The method of steepest descent applied to this criterion leads to a simple choice of the transformation parameters which is easily implemented on the digital computer, thereby allowing efficient synthesis of networks with minimum sensitivity to element tolerances.
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1. Introduction:

The advent of thin film and integrated circuit techniques has changed some of the criteria by which networks are evaluated. In particular, the new techniques place less emphasis on the number of elements in a network but require designs which are fairly insensitive to changes in the element values, due to the difficulty of maintaining tolerances at this stage in the development of the thin film techniques. Several authors have recently considered the design of insensitive networks\(^1-4\) from various viewpoints. In this paper, it is shown that the theory of equivalent networks is very useful in the synthesis of networks insensitive to large element tolerances.

Cauer in 1929 showed that by means of a congruence transformation, one physically realizable network could be generated from another in such a way that specified driving point and/or transfer functions were held invariant.\(^5,6\) This approach to network synthesis has been discussed by many authors and with the exception of the minimum-inductance transformation in filter theory has not realized its apparent potentialities.\(^7-10\)

Useful results from this theory can be derived by using the concept of continuously equivalent network theory together with
a digital computer for implementation. In this paper, this approach is used.

Given a network with the desired transfer function (synthesized by any of the known schemes), the theory of equivalent networks is used to generate another network with the same transfer function but with elements differing from those of the original network by an incremental amount. In the limit, differential equations result whose solution at any value of the independent variable give the elements of an equivalent network. In this way, it is possible to generate many networks equivalent to a given network but having widely differing element values while insuring at all times that no elements become negative. In addition, differential equations for the sensitivities of the transfer function to changes in element values are derived and an efficient computational algorithm derived which allows rapid realization of minimal sensitive networks on a computer. In an example 30 element network, computing times of several minutes were found.

2. Equivalent Network Theory:

Consider a network with \( n \) independent node-pairs. Such a network is described by a set of \( n \) equations of the form

\[
I = Y_o E
\]
where \( Y \) is the \( nxn \) admittance matrix of the network, \( E \) is the \( nxl \) column matrix of node voltages and \( I \) is the \( nxl \) column matrix of source currents. If this network is imbedded in a \( 2n \)-port network of ideal transformers, there results another \( n \)-port network and the variables of the original and new network are related by a linear transformation:

\[
E' = AE' \\
I' = A^tI
\]

where \( A \) is an \( nxn \) nonsingular constant matrix, \( t \) denotes transpose, and \( E' \) and \( I' \) are the new voltage and current variables. The last equation follows from the losslessness of the transformer network. The new variables are related by the equation

\[
I' = Y'E'
\]

where

\[
Y' = A^tY_A
\]

From the method of construction it is clear that the resulting admittance matrix is physically realizable. The usefulness stems from the possibility of interpreting Eq. 2.3 as the equilibrium equations of an equivalent transformerless network. In this case, not all elements will be positive in general and this fact has limited the applications of the theory. By proper choice of the transformation matrix \( A \), certain driving point and/or transfer functions can be held invariant. For
example, if the $k^{th}$ row of $A$ is the $k^{th}$ unit vector (all zeros except for the $k^{th}$ entry which is unity), the driving point impedance at the $k^{th}$ terminal pair is invariant to the transformation. If two rows are so chosen, the driving point impedance at each port is invariant and also the transfer impedance between the two ports. Thus it is possible to maintain desired transfer functions invariant to the transformation.$^9$

The theory becomes more useful if we imagine a transformation from a given network to one whose elements differ by only a small amount, and then pass to the limit. Let

$$A = U + B \Delta x$$ \hfill (2.5)

$$|b_{ij}| \leq 1$$

where the elements of $B$ are bounded, and $x$ is an independent variable. Consider the admittance matrix of the given network as a function of the independent variable $x$, $Y(x)$ so that the given admittance matrix is $Y(0) = Y_0$. Then applying this transformation yields a new matrix $Y(x + \Delta x)$ given by

$$Y(x + \Delta x) = Y(x) + \left[B^tY(x) + Y(x)B\right]\Delta x + B^tY(x)B(\Delta x)^2$$ \hfill (2.6)

In the limit as $\Delta x$ approaches zero, there results the matrix
differential equation

\[
\frac{dY}{dx} = B^T y + XB
\]

with initial conditions

\[ Y(0) = Y_0 \quad 2.7 \]

For any choice of \( B(x) \), the solution to this set of differential equations at any value of \( x \) gives a realizable admittance matrix. In order to maintain some transfer functions invariant, note from Eq. 2.5 that any row of \( A \) which is a unit vector implies that the same row of \( B \) must be identically zero. That is, equivalence at a port is assured if the row of \( B \) corresponding to that port is zero. The problem of negative elements which plagues the usual formulation of the equivalent network theory problem is easily surmounted here because the elements of the admittance matrix change continuously as a function of the independent variable \( x \) and the transformation matrix \( B \) is also chosen continuously as a function of \( x \), thereby allowing choices for \( B \) which keep elements positive or zero.

The set of equations 2.7 really consist of \( 3n^2 \) equations since the \( nxn \) matrices contain \( n^2 \) entries and the admittance matrix is really the sum of a conductance matrix, a capacitance matrix, and an inductance matrix. Since the node-to-datum set of equations for an \( RLC \) network have symmetrical admittance
matrices, only $3(n^2 + n)/2$ of the equations are different. The equations can be put into a more useful form for study and numerical solution by considering the relation between the admittance matrix and the elements of the network. The entry in the $i^{th}$ row and $j^{th}$ column ($i \neq j$) of the admittance matrix of a node to datum set of variables is the negative of the admittance connected between the $i^{th}$ and $j^{th}$ nodes, and each entry on the diagonal is merely the sum of all admittance attached to the corresponding node. Hence it is possible to derive the elements of the network from the admittance matrix (this assumes no mutual coupling between elements of the network).

Hence it is possible to convert the differential equations into another set whose variables are the elements of the network. For example, consider the RC network in Fig. 1 having two independent mode-pairs. The matrix differential equation is

$$\frac{dY}{dx} = B^T Y + YB \quad \text{2.8}$$

where

$$Y = \begin{bmatrix} g_1 + g_2 & -g_1 \\ -g_1 & g_1 + g_3 \end{bmatrix} + S \begin{bmatrix} c_1 + c_2 & -c_1 \\ -c_1 & c_1 + c_5 \end{bmatrix}$$

and

$$B = \begin{bmatrix} b_1 & b_2 \\ b_3 & b_4 \end{bmatrix} \quad \text{2.10}$$
This is the set of equations

\[
g_1' + g_2' = 2(b_1g_1 + b_2g_2 - b_3g_1) \\
-g_1' = -b_1g_1 + b_3g_1 + b_5 g_3 + b_2g_1 + b_2g_2 - b_4g_1 \\
g_1' + g_3' = 2(b_4g_1 + b_4g_3 - b_2g_1) \\
c_1' + c_2' = 2(b_1c_1 + b_2c_2 - b_3c_1) \\
-c_1' = -b_1c_1 + b_3c_1 + b_3c_3 + b_2c_1 + b_2c_2 - b_4c_1 \\
c_1' + c_3' = 2(b_4c_1 + b_4c_3 - b_2c_1) \tag{2.11}
\]

Here the prime denotes differentiation with respect to \( x \) and it is important to note that each \( b_i \) is in general an arbitrary (but bounded) function of \( x \).

Note that this set can be put into the convenient form of a first order matrix differential equation:

\[
\frac{dG}{dx} = M_1G 
\]

\[
\frac{dC}{dx} = M_1C \tag{2.12}
\]

where

\[
G^t = \begin{bmatrix} g_1 & g_2 & g_3 \end{bmatrix} 
\]

\[
C^t = \begin{bmatrix} c_1 & c_2 & c_3 \end{bmatrix} \tag{2.13}
\]

and
In general, the differential equations for the elements of the network will be of the form of Eqs. 2.12 where the matrix \( M \) has elements which are linear combinations of the \( b_i \) and are arbitrary. Some of the \( b_1 \) will be fixed by the requirement of invariance of a transfer function or the like. The remaining \( b_i \) are free and can be chosen to yield some optimal network. Note that given \( B(x) \), Eqs. 2.12 are linear homogeneous differential equations, and this greatly simplifies the computational problem.

In the general \( \mathbb{R}LC \) case, Eq. 2.12 would be modified by the addition of an equation for \( L \).

3. **Sensitivity Equations:**

Consider the network function \( T \) (a driving point or transfer function). In general, \( T \) is a function of frequency and can be considered a function of the elements in the network. We are interested in determining how much the value of the transfer function \( T \) at a frequency changes when the value of one of the elements changes. To this end, define the sensitivity of the transfer function \( T \) to changes in the \( k \)th element.
by the usual relation: 

\[ S_k = \frac{e_k}{T} \frac{dT}{\partial e_k} \quad 3.1 \]

If now the given network is subjected to an equivalence transformation as described in the previous section, the elements vary continuously with \( x \) but the transfer function, being invariant, remains constant. As the elements change values, the sensitivity of \( T \) to changes in the elements also changes. Consider the derivative of sensitivity with respect to the independent variable \( x \):

\[ \frac{dS_k}{dx} = b \sum_{i=1} e_i \frac{\delta S_k}{\delta e_i} \frac{de_i}{dx} \quad 3.2 \]

where the summation is taken over all the elements of the network, and \( e_i \) denotes the \( i \)th element (R, L, or C).

By straightforward differentiation and noting that second partial derivatives can be taken in either order, it is easy to verify that

\[ \frac{1}{e_k} \frac{\delta S_k}{\delta e_i} = \frac{1}{e_i} \frac{\delta S_i}{\delta e_k} \quad 3.3 \]

Using Eq. 3.3 in Eq. 3.2 yields

\[ \frac{dS_k}{dx} = b \sum_{i=1} e_k \frac{\delta S_i}{\delta e_k} \frac{de_i}{dx} \quad 3.4 \]
But the transfer function \( T \) is a function of the elements \( e_i \) and is independent of \( x \). Hence

\[
\frac{dT}{dx} = \frac{b}{\sum_{i=1}^{b} \frac{de_i}{dx}} = 0 \quad 3.5
\]

If Eq. 3.1 is rearranged as

\[
\frac{dT}{de_1} = \frac{TS_i}{e_1} \quad 3.6
\]

Eq. 3.5 yields

\[
T \frac{b}{\sum_{i=1}^{b} \frac{S_i de_i}{dx}} = 0 \quad 3.7
\]

Dividing Eq. 3.7 by \( T \) and taking the partial derivative with respect to \( e_k \) yields

\[
\frac{b}{\sum_{i=1}^{b} \left\{ \frac{e_i e_k}{e_i} \frac{dS_i}{de_k} + \frac{S_i}{e_i} \frac{de_i}{de_k} - \frac{e_i S_i}{e_i} \frac{de_i}{de_k} \right\}} \quad 3.8
\]

Rearranging Eq. 3.8 supplies the factor needed for Eq. 3.4 and noting that \( \frac{de_i}{de_k} \) is zero for \( i \neq k \) and unity for \( i = k \), yields

\[
\frac{b}{\sum_{i=1}^{b} \frac{1}{e_i} \frac{de_i}{e_k} \frac{dS_i}{dx}} = \frac{b}{\sum_{i=1}^{b} \left\{ -\frac{S_i}{e_i} \frac{de_i}{de_k} \right\}} + \frac{S_k de_k}{e_k^2} \quad 3.9
\]
Substituting Eq. 3.9 into Eq. 3.4 yields a differential equation for the sensitivity:

\[
\frac{dS_k}{dx} = \frac{S_k}{e_k} \frac{de_k}{dx} + \sum_{i=1}^{b} \frac{e_k S_i}{e_i} \frac{\partial}{\partial e_k} \left( \frac{de_i}{dx} \right)
\] 3.10

This equation simplifies considerably if an additional variable \( q_k \), related to \( S_k \) by

\[
S_k = q_k e_k
\] 3.11

is defined.

Then

\[
\frac{dS_k}{dx} = q_k \frac{de_k}{dx} + e_k \frac{dq_k}{dx}
\] 3.12

With this change of variable, Eq. 3.10 simplifies to

\[
\frac{dq_k}{dx} = - \sum_{i=1}^{b} \frac{e_k}{q_i} \frac{\partial}{\partial e_k} \left( \frac{de_i}{dx} \right)
\] 3.13

In section 2, the differential equations for the elements were shown to be of the form

\[
\frac{dc}{dx} = MC
\]

\[
\frac{dc}{dx} = MC
\]

\[
\frac{dl}{dx} = ML
\] 3.14
Hence in Eq. 3.13, if $e_i$ is a conductance for example, $\frac{d e_i}{d x}$ is a function only of the conductances---there is no coupling in Eqs. 3.14 between the various element-kinds. Hence unless $e_k$ is also a conductance, $\frac{\partial}{\partial e_k} \left( \frac{d e_i}{d x} \right)$ will be zero. That is, in Eq. 3.13, if $e_k$ is a conductance, only those values of $i$ corresponding to the conductances of the network appear and if $e_k$ is a capacitance, only those values of $i$ corresponding to capacitances in the network will appear, etc. Hence the differential equations for the sensitivities will be of a form similar to Eqs. 3.14 above, in that the equations for one kind of element will not involve the other element-kinds.

To display these equations succinctly, define $S_{Gk}$, $S_{Ck}$, $S_{Lk}$ the sensitivity of the $k$th conductance, $k$th capacitance, and $k$th inductance respectively and similarly for $q_{Gk}$, $q_{Ck}$, $q_{Lk}$. Then Eq. 3.15 becomes

$$\frac{d q_{Gk}}{d x} = - \sum_{i=1}^{z} q_{ii} \frac{\partial}{\partial e_k} \left( \frac{d e_i}{d x} \right)$$

$$\frac{d q_{Ck}}{d x} = - \sum_{i=1}^{z} q_{ii} \frac{\partial}{\partial c_k} \left( \frac{d C_i}{d x} \right)$$

$$\frac{d q_{Lk}}{d x} = - \sum_{i=1}^{z} q_{ii} \frac{\partial}{\partial l_k} \left( \frac{d L_i}{d x} \right)$$

From Eq. 3.14, it is clear that
\[
\frac{dq_i}{dx} = u_i^T M \quad 3.16
\]

and

\[
\frac{3}{dg_k} \left( \frac{dg_i}{dx} \right) = u_i^T G u_j \quad 3.17
\]

where \( u_r \) is the \( r \)th unit vector. Hence the sensitivity equations for the conductances (the others are similar) simplify as follows:

\[
\frac{dq_k}{dx} = - \sum q_i u_i^T M u_j \quad 3.18
\]

\[
= - u_j^T M^T q_G \quad 3.19
\]

where

\[
q_G^T = \begin{bmatrix} q_{G1} \ q_{G2} \ \ldots \end{bmatrix} \quad 3.20
\]

The equations for all of the sensitivities in matrix form are then simply

\[
\frac{dQ_G}{dx} = - M^T q_G \quad 3.21
\]

The equations for the capacitance and inverse inductance sensitivities are identical in form:

\[
\frac{dQ_C}{dx} = - M^T q_C
\]
Thus, the equations for the sensitivities of the transmission are merely the adjoint equations. The actual sensitivities are then determined by Eq. 3.11. To summarize, six sets of first order linear homogeneous differential equations for the three element kinds (R, L, and C) and the sensitivities of the transfer function of the network have been found such that their solution—for each value of x—gives a set of elements of an equivalent network, one having the same transfer function—and also the sensitivity of that network to changes in each of its elements. The given network and its sensitivities serve as initial conditions for these differential equations. The particular transfer function being held invariant is determined by the rows of the matrix B which are set equal to zero. The matrix M is derived from B by the method discussed in section 2. It is easy to formalize the whole procedure by adopting standard numbering schemes etc. in order to give formulae for M in terms of B, but this adds nothing conceptually or even practically and therefore no space is devoted to this problem here.

There remains the problem of determining the network equivalent to the given network with minimum sensitivity. This is done in a straightforward manner by first defining a criterion.
At any frequency, sensitivity may be complex, but only the magnitude is of interest. Choose a criterion which is essentially the sum of the magnitudes-squared of the sensitivities of interest:

\[
\varphi = \frac{1}{2} \sum_k \left[ q_{k*} q_k + q_{k*} q_{k*} + q_{Lk*} q_{Lk*} \right]
\]

where * indicates conjugate and where the summation is taken over those k's of interest.

The method of steepest descent is most useful for minimizing \( \varphi \). Using Eq. 3.21 and 3.22, the derivative of \( \varphi \) may be written as

\[
\frac{d\varphi}{dx} = -Q_G^t M_s Q_G^* - Q_{C}^t M_s Q_{C}^* - Q_{L}^t M_s Q_{L}^*
\]

Here \( M_s \) is the symmetric part of \( M \) and the fact that \( M \) is real is used to derive Eq. 3.24. Eq. 3.24 is pure real as desired.

For steepest descent to the minimum, \( d\varphi/dx \) should be as negative as possible. Thus the free \( b_i \) which appear in \( M \) should be chosen to make this so. But multiplying out Eq. 3.24 shows that \( d\varphi/dx \) is linear in the \( b_i \) and since the elements of \( B \) were limited in magnitude to unity, \( d\varphi/dx \) is minimized if all of the \( b_i \)'s take on plus or minus one, depending
upon the sign of its coefficient in Eq. 3.24. In the next section, it is shown that this set of differential equations is extremely easy to solve on a digital computer. In this case, at each step of the solution, the machine merely chooses the proper b's to make the descent toward the minimum as swift as possible. The only constraint on the choice is that no element go negative at any step. Since at each step the elements change only slightly (in the numerical integration of a differential equation, small increments are always chosen), it is easy to modify the computation so that the elements stay positive (because the changes in any single element are linearly related to the b's by Eqs. 2.12).

The sensitivity equations give solutions at only a single frequency and hence if it is desired to determine sensitivities at more than one frequency, the equations must be solved at each frequency. This does not appreciably complicate the computing because the only change in the differential equations is in the initial conditions and most of the computation involves the matrix of coefficients M or $M^t$ which is independent of frequency.

4. Computer Synthesis:

It has been shown in the previous sections that equivalent networks may be generated from the solution of a set of linear first order differential equations. In this section,
an efficient computational procedure for the integration of such equations, leading to efficient machine synthesis of minimal sensitive networks is discussed.

Consider the vector differential equation

$$\frac{dV}{dx} = AV$$

where $V$ is an $n \times 1$ matrix and $A$ is an $n \times n$ matrix. It is well known that the solution may be written in terms of a matrix exponential:

$$V(x) = e^{Ax} V(0)$$

The matrix exponential is defined by the (absolutely and uniformly convergent) series

$$e^{Ax} = u + Ax + \frac{1}{2} A^2 x^2 + \frac{1}{6} A^3 x^3 + \ldots + \frac{1}{k!} A^k x^k + \ldots$$

For machine solution, we are interested in solutions at $x = 0$, $\Delta x$, $2\Delta x$, $3\Delta x$, $4\Delta x$, ..., where $\Delta x$ is small. In this case, the series above may be approximated by its first few terms because $x$ is small. Thus the computer must form $A$, $A^2$, and $A^3$ if four terms are used and then form the weighted sum. Such a calculation is well suited to computers, making the integra-
tion process straightforward.

In the case of the problem at hand, the one set of equations has matrix \( M \) and the other \( -M^t \). Hence the matrix exponentials which must be calculated are

\[ e^{Mx} \text{ and } e^{-M^tx} \]

From Eq. 4.3, it is clear that the same matrices, \( M, M^2, \) and \( M^3 \) can be used for both calculations, the only difference in the two computations being that the terms are added with proper signs.

The computation proceeds with the following steps. First, the values of the elements and sensitivities are used in Eq. 3.24 to determine the optimum \( b \)'s for the next step. Second, the matrices \( M, M^2, \) and \( M^3 \) are formed. Third, \( e^{Mx} \) and \( e^{-M^tx} \) are formed. Fourth, the new elements and sensitivities are calculated by multiplying the matrix exponentials by the present element values and sensitivities as in Eq. 4.2. Fifth, the elements are checked to determine if any has gone negative. If so, the step is repeated with the choice of \( b \)'s modified so that the element which went negative remains zero or positive. Then the procedure is repeated until a minimum is reached.

Although derived for RLC networks, the theory of equivalent networks is also valid for active networks with only
slight modifications. As an example of this procedure, a network containing two vacuum tubes and 30 RLC elements was chosen. The parameters of the tubes were fixed and it was desired to generate equivalent networks with the voltage transfer function held invariant so that the voltage transfer function was minimally sensitive to the gains of the tubes. Because the tube parameters were assumed fixed, the RLC formulation derived in this paper is applicable.

Four discrete frequencies were chosen and the linear combination of the sensitivities at these frequencies minimized. Using a Burroughs 220 computer, several minutes of computing time led to a 25% decrease in sensitivity of this network, and required only 6 steps. No attempt was made to make the program optimal and a more efficient program should decrease this time still further. In any case, the computing time for even complex networks does not limit the application of this approach.
REFERENCES


Fig. 1

Example RC Network
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