Design of Microstrip Low-pass and Band-pass Filters using Artificial Neural Networks

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Abstract: Usually the design of microstrip filters is done using simulators and classical approximation methods as Butterworth and Chebyshev, these techniques takes a lot of time to run for designing filters. In this paper we develop a faster artificial neural network model for designing a microstrip low-pass and band-pass filters, when the input are the dimensions of filter, operating frequency, the features of substrate, and the output are the transmission and reflection coefficients. The database uses for training this model is generated by a linear simulator based on circuits model. Two filters designed by the developed model are a stepped impedance low-pass filter with a cut-off frequency of 1 GHz and a parallel coupled-line band-pass filter with fractional bandwidth of 25 % and a central frequency of 2.45 GHz. The proposed model has been shown to be as accurate as a linear simulator and much more efficient computationally in the design of microstrip low-pass and band-pass filters.

Keywords: artificial neural network model, microstrip low-pass and band-pass filters, parallel coupled-line, stepped impedance.

1. INTRODUCTION

Filters play an important role in many applications in radio frequency and microwave fields. They are used to select or confine signals within the assigned spectral limits. Microwave filters can be designed in various transmission line structures, such as waveguides, coaxial line and microstrip [1]. Microstrip line is one of the most popular types of planar transmission lines because it can be fabricated by photolithographic processes and is easily integrated with both passive and active microwave devices and offer a high range of frequency. The advantages of microstrip filters are their planar structure, low cost, insensitivity to manufacturing tolerances, reproducibility, a wide range of bandwidth and easy design process [2]. Development of simulation tools for microwave circuits remains a major challenge in microwave fields. There have been extraordinary recent advances in computer-aided design (CAD) of microwave circuits. They have been implemented in software tools and are being applied to microwave filters simulation, modeling and design using CAD, the costs for design and tuning can be reduced greatly. CAD can provide more accurate design, reduces the labor intensiveness and decreases the time from design to production. Furthermore, if the materials used are expensive, the first-pass design or less iteration afforded by CAD will reduce the extra cost of materials and other factors necessary for developing a satisfactory prototype [1,3].

Using artificial intelligence techniques in the field of microwaves has produced powerful results [4-8]. Artificial neural network is one of these techniques, it can be used for modeling nonlinear multidimensional relationships. The evaluation time of a neural network model is also fast. For these reasons, neural networks have been used for various modeling and design applications including passive microwave circuits, in reference [5] the authors designed a low-pass filter by neural network in cut-off frequency of 0.5 GHz and 0.7 GHz, in reference [6] the authors designing microstrip high-pass filter by neural network and focused their work in cut-off frequency of 1.5 GHz and in reference [7] the designing of stepped impedance microstrip low-pass filters using artificial neural network at 1.8 GHz is achieved by the authors. In [9] the design and analysis of stub microstrip band-pass filter at mid-band frequency 1.8 GHz is done using ANN and in [10] a presented design approach for a λ/2 resonator bandpass filter by using the artificial neural network modeling technique when the input variables are the dimensions of filter.

In this paper we present our contribution to the simulation of low-pass and band-pass microstrip filters using artificial neural networks for all operating frequencies, in
order to develop a precise neural network model for simulation of microstrip filters. We have generated a database using a program that simulates the microstrip filter where the inputs are the filter dimensions, all range of operating frequency, the height of substrate and the relative dielectric constant, the outputs are the transmission (S21) and the reflection (S11) coefficients. This database is to train our neural network model such as the input variables are filter dimensions and the output is frequency response. This paper is organized into three sections: section 1 presents the design process of the low-pass and band-pass microstrip filters, section 2 focuses on modeling the filter and formulation of the problem and section 3 is dedicated to the presentation of simulation results.

2. DESIGN PROCESS OF FILTERS

The design of microstrip filters involves two main steps. The first is to select an appropriate low-pass prototype (in case of other filters: band-pass, high-pass,...etc, the transformation to a low-pass is done firstly), the choice of the type of response (Butterworth, Chebyshev), including the pass-band ripple and the number of reactive elements, will depend on the required specifications. The element values of the low-pass prototype filter, which are usually normalized to make a source impedance \( g_0 = 1 \) and a cut-off frequency \( f_c = 1 \), are then transformed to the \( L-C \) elements for the desired cut-off frequency and source impedance, which is normally 50 \( \Omega \) for microstrip filters. Having obtained a suitable lumped-element filter design, the next main step in the design of microstrip filters is to find an appropriate microstrip realization that approximates the lumped-element filter. A relatively easy way to implement low-pass filters in microstrip line is using alternating sections of very high and very low characteristic impedance lines. Such filters are usually referred to as stepped-impedance, or hi-Z, low-Z filters, and are popular because they are easier to design and take up less space than a similar low-pass filter using other technology [2]. For the band-pass filters, the parallel coupled line is one of the popular techniques used in the design of band-pass filters, it use half or quarter wavelength line resonators, the adjacent resonators are parallel to each other along half of their length [1]. In our study two examples of filters are chosen:

A. Example 1: Low-pass filter
Cut-off frequency \( f_c = 1 \text{GHz} \).
Maximal attenuation in bandwidth \( A_{\text{max}} = 0.1 \text{ dB} \).
Load impedance equal source impedance \( (Z_0=Z_L=50 \text{ Ohm}) \).

B. Example 2: Band-pass filter
Fractional bandwidth \( \text{FBW} = 10\% \)
Center frequency \( f_0 = 2.45 \text{ GHz} \).
Maximal attenuation in bandwidth \( A_{\text{max}} = 0.5 \text{ dB} \).
Load impedance equal source impedance \( (Z_0=Z_L=50 \text{ Ohm}) \).

For example 1, we used the Butterworth approximation for the synthesis of the parameters of the desired filter. We find the order of the filter equal to 3 [2]. Figure 1 shows the low-pass prototype filter with normalized element values.

![Fig. 1. Low-pass prototype filter.](image1)

The design of this filter on a relative dielectric constant substrate \( \varepsilon_r = 10.8 \) and a height \( h = 1.27 \) gives us the following dimensions: \( W_0 = 0.2 \text{ mm}, W_c = 4 \text{ mm}, L_1 = 9.81 \text{ mm} \) and \( L_c = 7.11 \text{ mm} \).

Figure 2 shows the layout of stepped-impedance microstrip low-pass filter.

![Fig. 2. The layout of stepped-impedance microstrip filter.](image2)

For example 2, we used the Chebyshev approximation for the synthesis the parameters of the desired filter. We find the order of the filter equal to 3 [2]. Figure 5 shows the low-pass prototype filter with normalized element values.
The design of this filter on a relative dielectric constant substrate $\varepsilon_r = 4.2$ and a height $h = 1.58\text{ mm}$ gives us the following dimensions: $W_1 = W_4 = 2.54\text{ mm}$, $W_2 = W_3 = 3.53\text{ mm}$, $S_1 = S_4 = 0.046\text{ mm}$, $S_2 = S_3 = 0.431\text{ mm}$, $l_1 = l_4 = 17.25\text{ mm}$ and $l_2 = l_3 = 17.02\text{ mm}$. Figure 4 shows the layout of parallel coupled line band-pass filter.

After the design of the filters, the simulation is done using an electromagnetic simulator such as IE3D, CST, etc [11,12].

3. PROBLEM FORMULATION

Most filters are comprised of linear elements or components, linear simulations based on the network or circuit analyses are simple and fast for computer-aided analysis. Linear simulations analyze frequency responses of microwave filters based on their analytical circuit models. The filter shown in figure 2 consists of three section transmission line, and the scattering matrix (ABCD) of a section transmission line is given by [1]:

$$\begin{bmatrix} A & B \\ C & D \end{bmatrix} = \prod_{i=1}^{n} \begin{bmatrix} A & B \\ C & D \end{bmatrix}$$

(5)

And the transmission ($S_{21}$) and reflection ($S_{11}$) coefficients are given by:

$$S_{21} = \frac{2}{A + B / Z_0 + CZ_0 + D}$$

(6)

$$S_{11} = \frac{A + B / Z_0 - CZ_0 - D}{A + B / Z_0 + CZ_0 + D}$$

(7)

The synthesis of these filters can be considered as a function of $Z_r$, $Z_c$, $\theta_l$ and $\theta_c$ for example 1 and $Z_{0e}$, $Z_{0o}$, $\theta_{0e}$ and $\theta_{0o}$ for example 2. Using the equations they calculate the width and length of the microstrip line and the width, length and spacing of parallel-coupled line from the characteristic impedance and electrical lengths [2], we can calculate the transmission and reflection coefficients using the filter dimensions, substrate characteristics and operating frequency.

The first step is to generate a database to train the neural network model, where the inputs are the filter dimensions, the relative dielectric constant and the operating frequency, the outputs are the transmission and reflection coefficients. The next step is to validate the neural network model by a simulation of filters.
Tables 1 and 2 show the search space of the filter of low-pass and band-pass filters to generate the database.

**TABLE 1. SEARCH SPACE OF LOW-PASS FILTER**

| Dimensions | Min(mm) | Max(mm) |
|------------|---------|---------|
| W₁         | 0.1     | 0.3     |
| L₁         | 6       | 10      |
| W₂         | 2       | 6       |
| L₂         | 5       | 8       |

**TABLE 2. SEARCH SPACE OF BAND-PASS FILTER**

| Dimensions | Min(mm) | Max(mm) |
|------------|---------|---------|
| W₁         | 2       | 3       |
| L₁         | 0.04    | 0.05    |
| S₁         | 16      | 18      |
| W₂         | 3       | 4       |
| L₂         | 0.4     | 0.5     |
| S₂         | 16      | 18      |

The architects of our neural network model for the simulation of low-pass and band-pass filters as shown in figure 5 and figure 6.

**4. RESULTS AND DISCUSSION**

Our contribution in this paper is the design of two microstrip filters by a linear simulator and the proposed artificial neural network model, a low-pass filter with a cutoff frequency of 1Ghz and a band-pass filter with a centre frequency of 2.45Ghz and fractional bandwidth of 10%, these filters shown in figure 7 and figure 8.

After training the neural network model by the database, these filters are simulated using the formulas (6) and (7), and then simulated by the proposed model.

Our database contains 400 filters for each application to train the neural network model.
TABLE 3. THE COEFFICIENTS S21 AND S11 VALUES.

| Frequency (GHz) | S21(dB) simulator | S21(dB) NNM | S11(dB) simulator | S11(dB) NNM |
|----------------|-------------------|-------------|-------------------|-------------|
| 0.1            | -0.0083           | -0.0812     | -0.0083           | -0.0812     |
| 0.2            | -0.0293           | -0.0281     | -0.0293           | -0.0281     |
| 0.3            | -0.0518           | -0.0516     | -0.0518           | -0.0516     |
| 0.4            | -0.0610           | -0.0601     | -0.0610           | -0.0601     |
| 0.5            | -0.0429           | -0.0429     | -0.0429           | -0.0429     |
| 0.6            | 0.0109            | 0.0116      | 0.0109            | 0.0116      |
| 0.7            | 0.0978            | 0.0980      | 0.0978            | 0.0980      |
| 0.8            | 0.1986            | 0.1975      | 0.1986            | 0.1975      |
| 0.9            | 0.2715            | 0.2706      | 0.2715            | 0.2706      |
| 1.0            | 0.2514            | 0.2519      | 0.2514            | 0.2519      |
| 1.1            | 0.0584            | 0.0598      | 0.0584            | 0.0598      |
| 1.2            | -0.3762           | -0.3754     | -0.3762           | -0.3754     |
| 1.3            | -1.0769           | -1.0754     | -1.0769           | -1.0754     |
| 1.4            | -2.0994           | -2.0878     | -2.0994           | -2.0878     |
| 1.5            | -3.0981           | -3.0993     | -3.0981           | -3.0993     |
| 1.6            | -4.2586           | -4.2587     | -4.2586           | -4.2587     |
| 1.7            | -5.4207           | -5.4219     | -5.4207           | -5.4219     |
| 1.8            | -6.5361           | -6.5371     | -6.5361           | -6.5371     |
| 1.9            | -7.5750           | -7.5782     | -7.5750           | -7.5782     |
| 2.0            | -8.5210           | -8.5222     | -8.5210           | -8.5222     |
| 2.1            | -9.3663           | -9.3657     | -9.3663           | -9.3657     |
| 2.2            | -10.1080          | -10.1086    | -10.1080          | -10.1086    |
| 2.3            | -10.7459          | -10.7505    | -10.7459          | -10.7505    |
| 2.4            | -11.2812          | -11.2743    | -11.2812          | -11.2743    |
| 2.5            | -11.7158          | -11.7199    | -11.7158          | -11.7199    |
| 2.6            | -12.0518          | -12.0531    | -12.0518          | -12.0531    |
| 2.7            | -12.2912          | -12.2901    | -12.2912          | -12.2901    |
| 2.8            | -12.4355          | -12.4361    | -12.4355          | -12.4361    |
| 2.9            | -12.4865          | -12.4824    | -12.4865          | -12.4824    |
| 3.0            | -12.4453          | -12.4429    | -12.4453          | -12.4429    |

Table 4. The coefficients s21 and s11 values.

| Frequency (GHz) | S21(dB) simulator | S21(dB) NNM | S11(dB) simulator | S11(dB) NNM |
|----------------|-------------------|-------------|-------------------|-------------|
| 2.0            | -14.5967          | -14.5938    | -0.0026           | -0.0026     |
| 2.1            | -13.6051          | -13.6093    | -0.0041           | -0.0041     |
| 2.2            | -12.5649          | -12.5677    | -0.0067           | -0.0067     |
| 2.3            | -11.4630          | -11.4621    | -0.0111           | -0.0110     |
| 2.4            | -10.2826          | -10.2820    | -0.0192           | -0.0192     |
| 2.5            | -9.0170           | -9.0171     | -0.0344           | -0.0345     |
| 2.6            | -7.6548           | -7.6544     | -0.0649           | -0.0650     |
| 2.7            | -6.1852           | -6.1856     | -0.1296           | -0.1296     |
| 2.8            | -4.6130           | -4.6127     | -0.2765           | -0.2765     |
| 2.9            | -3.0852           | -3.0855     | -0.6331           | -0.6331     |
| 3.0            | -1.4795           | -1.4797     | -1.5308           | -1.5308     |
| 3.1            | -0.4345           | -0.4346     | -3.0708           | -3.0708     |
| 3.2            | 0.0314            | 0.0315      | -9.2126           | -9.2126     |
| 3.3            | -0.0218           | -0.0216     | -10.0119          | -10.0119    |

Based on the values shown in Table 3 and Table 4, the difference between the results of linear simulator and obtained by the proposed model is very small in order of 0.0005dB for all range of operating frequency.

Figure 9 and Figure 10 shows the magnitude of transmission and reflection coefficients obtained by the simulator and neural network for low-pass and band-pass filters.
In figure 9 and figure 10, the results of proposed model is similar to the linear simulator in pass-band and stop-band frequency.

The design of these filters by the linear simulator takes a 5 minutes in PC of CPU 2.70GHz, RAM 8 Go and using Matlab R2015 , and the same design take 30 seconds when used the proposed model.

The developed model provide a better result in term of transmission and reflection coefficients for all points in operating frequency when multiple publication focused only in cutoff frequency.

5. CONCLUSION

In this work a developed neural network model for designing microstrip low-pass and band-pass filters is presented. The strong point of this proposed model that works in pass-band/stop-band frequency and provide an accurate results and easy to use for designing microstrip filters. The future work is applied this model for different technologies and many device like antenna and coupler.

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