RF and Microwave Power Sensor Calibration by Direct Comparison Transfer

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1. Introduction

Many instruments can be used to measure radio frequency (RF) and microwave power. The most accurate one is a power sensor with a meter. The accuracy of RF and microwave power measurement depends on the accuracy of power sensor calibration. This chapter provides calibration methods of RF and microwave power sensor with system setup, modeling, equations, and analyses in different representations, traceability and measurement uncertainty evaluations. From the simple direct comparison transfer method, to the different improvements, and then the general analytical models, the methods provided in this chapter are useful for the guided-wave power sensor calibration in frequency range of a few MHz to several hundred GHz.

2. Microwave power standard and traceability of power sensor calibration

Although the chapter is focusing on the power sensor calibration, it has to talk about first the primary microwave power standard so that the traceability of power sensor calibration is clearly defined.

According to the definition, primary standard is derived directly in terms of base units of the International System of Units (SI). Now the prevalently accepted primary microwave power standard is the calorimeter or microcalorimeter, which is a substitution type of primary standard based on heat measurement (Brunetti & Vremera, 2003; Clague, 1995; Cui, X. & Crowley, T. P. (2011); Famton, 1990; JCGM 200:2008; Oldfield, 1989).

The primary microwave power standard determines the effective efficiency and calibration factor through DC power substitution to realize the traceability to SI units. Power is measured in terms of heat capacity and rate of temperature rise. When microwave power is applied to a terminating device or load through transmission line, the microwave energy is absorbed and converted to heat energy, causing the load temperature to change. Similarly, when DC power is applied to the same load, the DC energy is converted to heat, causing the load temperature to change. When the temperature change caused by the DC power is equivalent to that caused by the microwave power, the DC power can be used to precisely determine the corresponding microwave power. This is the principle of DC power substitution. The substitution technique obviates the need for detailed knowledge of heat losses and thermal capacities.
The terminating device may not react in the same way for microwave and DC power absorption, so the effective efficiency \( \eta_e \) is used to perform the correction. In equation (1), microwave power \( P_{HF} \) absorbed by the terminating device is calculated by dividing the substituted DC power \( P_{DC} \) by the effective efficiency \( \eta_e \):

\[
\eta_e = \frac{P_{DC}}{P_{HF}}
\]  (1)

Since the effective efficiency is independent of the mismatch correction, the calibration factor \( K \) is used to describe both the effective efficiency \( \eta_e \) and mismatch \( \Gamma \) as follows:

\[
K = \eta_e \times (1 - |\Gamma|^2)
\]  (2)

The calibration factor \( K \) is generally used at the time of calibration transfer from a reference standard to an unknown microwave power sensor. It is the focus in the following sections.

The measurement uncertainty is a non-negative parameter characterizing the dispersion of the effective efficiency \( \eta_e \) or the calibration factor \( K \) being attributed to the standards. The uncertainty is evaluated using “law of propagation of uncertainty” following “Guide to the expression of Uncertainty in Measurement” (GUM) (JCGM 100:2008). Evaluation of measurement and calibration uncertainty by Monte Carlo Method (MCM) is to use Monte Carlo simulation in the uncertainty evaluation of output quantities based on “uncertainty probability distribution propagation” (JCGM 101:2008). The following sections will cover both methods for the measurement uncertainty evaluations.

The value of a primary standard is disseminated to a secondary standard through calibration or comparison. Then the reference standard and working power sensors will be calibrated for use. The measurement results through such relations as unbroken chain of calibrations establish the metrological traceability, each contributing to the measurement uncertainty. The traceability is illustrated in Fig. 1. Here for reference purpose we deliberately provide not only the hierarchy, but also the uncertainties typically related. The real uncertainties depend on the frequency band and each laboratory conditions.

![Fig. 1. Dissemination of primary standard to end user- Traceability Chart](www.intechopen.com)
3. Power sensor calibration by direct comparison transfer

3.1 Modeling

The calibration of RF and microwave power sensor is to transfer the effective efficiency or calibration factor from a primary standard to a secondary standard or from a secondary standard to a reference standard or from a reference standard to a power sensor. The parameter transfer is through comparison, or calibration one against the other. The simplest and the most obvious method to calibrate a power sensor against a reference standard is to connect each in turn to a stable power source, as illustrated in Fig. 2.

Fig. 2. Calibration of power sensor by the method of simple direct comparison transfer

Generally power from a source, \( P_i \), with reflection coefficient \( \Gamma_i \), incident to a load with reflection coefficient \( \Gamma_L \), can be expressed as follows (Agilent, 2003; Engen, 1993; Mial, 2007):

\[
P_l = P_{Z_0} \frac{1}{|1 - \Gamma G \Gamma_L|^2}
\]

and a reflected power \( P_r \)

\[
P_r = P_{Z_0} \frac{\Gamma_L^2}{|1 - \Gamma G \Gamma_L|^2}
\]

For Fig. 2, the power dissipated to the reference power standard \( P_{Std} \), can be derived as

\[
P_{Std} = P_l - P_{r,Std} = P_{Z_0} \frac{1 - |\Gamma_{Std}|^2}{|1 - \Gamma G \Gamma_{Std}|^2}
\]

And the power dissipated to the power sensor to be calibrated (DUT), \( P_{DUT} \), is as

\[
P_{DUT} = P_l - P_{r,DUT} = P_{Z_0} \frac{1 - |\Gamma_{DUT}|^2}{|1 - \Gamma G \Gamma_{DUT}|^2}
\]

where \( P_{Z_0} \) is the power available to a load with characteristic impedance \( Z_0 \). When idealized source with available power \( P_a \) has no internal impedance \( Z_0 \), \( P_{Z0} = P_a (1 - |\Gamma_c|^2) \).

As shown in equation (1), the effective efficiency of a power sensor is a ratio. \( P_{DC} \) is DC or low frequency equivalent power, generating the same heat effect as the high frequency power being measured. For calibration of a power sensor, that is, to transfer the effective efficiency of a reference standard, \( \eta_{Std} \), to a power sensor to be calibrated (DUT), \( \eta_{DUT} \), it can be derived:
\[ \eta_{\text{DUT}} = \frac{P_{\text{DC,DUT}}}{P_{\text{RF,DUT}}} = \frac{P_{\text{DC,Std}}}{P_{\text{RF,Std}}} \frac{1 - |\Gamma_{\text{DUT}}|^2}{1 - |\Gamma_{\text{Std}}|^2} \times \frac{1 - |\Gamma_{G_{\text{DUT}}}|^2}{1 - |\Gamma_{G_{\text{Std}}}|^2} \]

So the calibration equation of effective efficiency for the method of Fig. 2 is expressed as:

\[ \eta_{\text{DUT}} = \eta_{\text{Std}} \times \frac{P_{\text{DC,DUT}}}{P_{\text{DC,Std}}} = \frac{P_{\text{DC,DUT}}}{P_{\text{DC,Std}}} \frac{1 - |\Gamma_{\text{DUT}}|^2}{1 - |\Gamma_{\text{Std}}|^2} \times \frac{1 - |\Gamma_{G_{\text{DUT}}}|^2}{1 - |\Gamma_{G_{\text{Std}}}|^2} \]

(3)

where \( \eta_{\text{DUT}} \) is the effective efficiency of DUT sensor.

This calibration equation involves three factors:

\[ \eta_{\text{DUT}} = \eta_{\text{Std}} \times P_{\text{RATIO}} \times M_1 M_2 \]

\( \eta_{\text{Std}} \) is the effective efficiency of a standard sensor. It comes from a national metrology institute, a calibration laboratory, or a manufacturer with traceability. \( P_{\text{RATIO}} \) is an equivalent DC or low frequency power ratio, depending on the system setup. \( M_1 M_2 \) is the mismatch factor.

Similarly, calibration equation of calibration factor for the method of Fig. 2 is expressed as follows, considering equation (2):

\[ K_{\text{DUT}} = K_{\text{Std}} \times \frac{P_{\text{DC,DUT}}}{P_{\text{DC,Std}}} = \frac{P_{\text{DC,DUT}}}{P_{\text{DC,Std}}} \frac{1 - |\Gamma_{\text{DUT}}|^2}{1 - |\Gamma_{\text{Std}}|^2} \times \frac{1 - |\Gamma_{G_{\text{DUT}}}|^2}{1 - |\Gamma_{G_{\text{Std}}}|^2} \]

(4)

And the calibration factor transfer equation can also be expressed as three factors:

\[ K_{\text{DUT}} = K_{\text{Std}} \times P_{\text{RATIO}} \times M \]

Notice that \( M \) is the mismatch factor for the calibration factor calibration transfer.

### 3.2 Uncertainty evaluation

From calibration transfer equation of effective efficiency (3) and that of calibration factor (4), if DUT power sensor has identical reflection coefficient as that of reference power standard, which means \( \Gamma_{\text{DUT}} = \Gamma_{\text{Std}} \), each of them absorbs exactly the same amount of power from the source. Then by power ratio measurement it transfers the effective efficiency, \( \eta_\epsilon \), or calibration factor, \( K \), from standard to DUT, to complete the calibration.

But the actual reflection coefficients of power sensors being compared usually differ significantly from one another (refer to Table 1). The signal generator reflection coefficient cannot dismiss easily. The complex items inside (3) and (4), i.e. the mismatch factors, cannot be neglected in accurate power sensor calibration with small uncertainty.
In recognition of the standards adopted internationally, the GUM (JCGM 100:2008) is considered as the basic technique to evaluate the uncertainty of measurement. The method proposed in the GUM is based on the law of propagation of uncertainty which is essentially the first-order Taylor series approximation of calibration equation, such as equation (4) of calibration factor. The partial differentiation of the output estimate with respect to the input estimates, which is termed sensitivity coefficient, is interpreted as a description of how the output estimate varies with changes in the values of the input estimates. The following equation is the definition of combined standard uncertainty $u_c(y)$ which includes both type A ($u_a(x)$) and type B ($u_b(x)$) uncertainties when the mathematical model is of the form $y = f(x_1, x_2, \ldots)$:

$$u_c(y)=\sqrt{u_a^2(x)+\sum_{i=1}^{N}\left(\frac{\partial f}{\partial x_i}\right)^2u_b^2(x_i)}$$  \hspace{1cm} (5)

Considering that the setup in Fig. 2 is the simple direct comparison transfer method, we present the uncertainty evaluation equations in different implementations, aiming at providing laboratories more realistic choices. The realistic magnitudes for reflection coefficients of instruments are used for the calculation and comparison. The specifications of reflection coefficients of the instruments quoted from different products at some frequency points are listed in Table 1.

<table>
<thead>
<tr>
<th>product</th>
<th>10 GHz</th>
<th>18 GHz</th>
<th>50 GHz</th>
<th>75GHz(wg)</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>X</td>
<td>Y</td>
<td>Z</td>
<td>X</td>
</tr>
<tr>
<td>sensor (Standard)</td>
<td>0.07</td>
<td>0.01</td>
<td>0.10</td>
<td>0.07</td>
</tr>
<tr>
<td>sensor (DUT)</td>
<td>0.11</td>
<td>0.13</td>
<td>0.16</td>
<td>0.11</td>
</tr>
<tr>
<td>signal generator</td>
<td>0.30</td>
<td>0.33</td>
<td>0.30</td>
<td>0.23</td>
</tr>
</tbody>
</table>

Table 1. Typical reflection coefficients of power sensors and signal generators

### 3.2.1 The simplest evaluation of measurement uncertainty

The simplest way in the calibration transfer of calibration factor from standard to DUT sensor is to simplify equation (4) to the following equation:

$$K_{DUT} = K_{Std} \times \frac{P_I}{P_S} = \text{Ratio Factor}$$  \hspace{1cm} (6)

In which only the ratio factor is considered and set the mismatch factor $M$ equal to 1. The real values of reflection coefficients are considered for the uncertainty budget only. In this case, the mismatch factor $M$ for calibration factor $K_{DUT}$ is expressed as:

$$M = \frac{M_U}{M_S}$$

where $M_U$ and $M_S$ are given by:

$$M_U = |1 - \Gamma_G \Gamma_{DUT}|^2, \quad M_S = |1 - \Gamma_G \Gamma_{Std}|^2$$
The associated uncertainties are calculated with the following equations (Agilent, 2003; Shan et al., 2010a):

\[ u(M_X) = \sqrt{2} |\Gamma_c \parallel \Gamma_X|; \quad X = Std, DUT \]  

(7)

According to GUM for the law of propagation of uncertainties, the sensitivity coefficients are partial differentiations with respect to the individual variables in equation (6):

\[
\begin{align*}
\frac{\partial K_{DUT}}{\partial K_{Std}} &= \frac{P_{Ul}}{P_S} = \frac{K_{DUT}}{K_{Std}} \\
\frac{\partial K_{DUT}}{\partial P_{Ul}} &= \frac{K_{Std}}{P_S} = \frac{K_{DUT}}{P_{Ul}} \\
\frac{\partial K_{DUT}}{\partial P_S} &= -\frac{K_{Std} \times P_{Ul}}{P_S}^2 = (-1) \frac{K_{DUT}}{P_S} 
\end{align*}
\]  

(8)

In which the expression is prepared for relative uncertainty representation since the combined standard uncertainty \( u_c(y) \) can be expressed as an estimated relative combined variance \( \left( \frac{u_c(y)}{y} \right)^2 = \sum_{i=1}^{N} p_i^2 \left( \frac{u(x_i)}{x_i} \right)^2 \) when the mathematical model is of the form \( Y = cX_1^{P_1} X_2^{P_2} ... X_N^{P_N} \).

The uncertainty budget is listed in Table 2 at frequency 18 GHz with type N connector. The calculations of mismatch uncertainties are based on Table 1 best and worst specifications.

It can be seen from Table 2 that the simplest evaluation of measurement uncertainty method is not an accurate method to obtain a small uncertainty. But it is useful for calibration laboratories with simple measurement set up as illustrated in Fig. 2. When this uncertainty value meets the demand, it is acceptable for industrial applications. In some evaluation experiments, it is also a practical method. Note that the value and uncertainty of a reference standard comes from the calibration of national metrology institutes or other calibration laboratories if the laboratory provides the calibration service using the method.

| 18 GHz based on best specifications |  
|---|---|---|---|---|---|
| Quant. | Estim. | Standard uncertainty | Probability distribution | Sensitivity coefficient | Uncertainty contribution |
| \( X_i \) | \( x_i \) | \( u(x_i) \) | | \( c_i \) | \( u_c(y) \) |
| \( K_S \) | 0.9894 | 0.0012 | normal | 1.0137 | 0.0012 |
| \( M_S \) | 1.0000 | 0.0098 | U-shaped | 1.0000 | 0.0098 |
| \( M_U \) | 1.0000 | 0.0195 | U-shaped | 1.0000 | 0.0195 |
| \( P_U \) | 1.0158 | 0.0018 | normal | 0.9873 | 0.0018 |
| \( P_S \) | 1.0021 | 0.0004 | normal | -1.0008 | -0.0004 |
| \( y=K_U \) | 1.0029 | | | | 0.0219 |

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### Table 2. Uncertainty budget at 18 GHz for simplest evaluation of uncertainty for Fig. 2 measurement setup. Uncertainties $u_i$ are at one standard deviation. Powers are measured in mW.

<table>
<thead>
<tr>
<th>18 GHz</th>
<th>based on worst specifications</th>
</tr>
</thead>
<tbody>
<tr>
<td>Quant.</td>
<td>Estim.</td>
</tr>
<tr>
<td>$X_i$</td>
<td>$x_i$</td>
</tr>
<tr>
<td>$K_S$</td>
<td>0.9894</td>
</tr>
<tr>
<td>$M_S$</td>
<td>1.0000</td>
</tr>
<tr>
<td>$M_U$</td>
<td>1.0000</td>
</tr>
<tr>
<td>$P_U$</td>
<td>1.0158</td>
</tr>
<tr>
<td>$P_S$</td>
<td>1.0021</td>
</tr>
<tr>
<td>$y = K_U$</td>
<td>1.0029</td>
</tr>
</tbody>
</table>

#### 3.2.2 Measurement uncertainty improvement with mismatch correction

To improve the calibration accuracy and uncertainty evaluation, we have to perform a mismatch correction, i.e., the complex reflection coefficients have to be considered in the calibration model. The complex value is either representing in term of magnitude and phase or its real and imaginary components. In the majority of engineering applications, the magnitude and phase representation (Polar coordinates) is generally preferred because this representation bears a direct relationship to physical phenomena affecting the measurement process (Ridler & Salter, 2002). For example, phase relates directly to the electrical path length of a signal, and magnitude relates directly to the signal attenuation. The same cannot be said for the representation of real and imaginary components (Cartesian coordinates). If the scales are used to depict the different representations, the real and imaginary axes in the complex plane extend infinitely ($\pm \infty$). It is the same as the scale is used to depict all real numbers which is routinely for the arithmetic operations. While the scales are used to represent magnitude and phase each possess a significant difference. The magnitude has a lower bound of zero below which values cannot exist, and phase is represented conventionally on a cyclical scale, either from $-180^0$ to $+180^0$ or from $0^0$ to $360^0$.

With mismatch correction added in computing the calibration factor using equation (4), the calculation of sensitivity coefficients involves the partial differentiations with respect to complex reflection coefficients. In the following sections, we separately provide the derived sensitivity coefficients in both Cartesian and Polar coordinate representations for practical measurement uncertainty solution and application; and also examples for their associated uncertainties are included. The derived sensitivity coefficients in both representations have been numerically appreciated by making use of MATLAB version R2010a, symbolic differentiation in Math Toolbox.

#### 3.2.2.1 Cartesian representation of sensitivity coefficients for equation (4)

Representing calibration factor $K_{DUT}$ of equation (4) with magnitude and phase components, it becomes

$$K_{DUT} = K_{Std} \times P_{RATIO} \times M = K_{Std} \times \frac{P_{DUT}}{P_{Std}} \times \frac{1 + |\Gamma_{DUT}|^2 |\Gamma_G|^2 - 2 |\Gamma_{DUT}| |\Gamma_G| \cos(\theta_{DUT} + \theta_G)}{1 + |\Gamma_{Std}|^2 |\Gamma_G|^2 - 2 |\Gamma_{Std}| |\Gamma_G| \cos(\theta_{Std} + \theta_G)}$$

(9)
Let $M_N$ represent the numerator and $M_D$ denominator of mismatch $M$:

$$M_N = |1 - \Gamma_{DUT} \Gamma_G|^2 = 1 + |\Gamma_{DUT} \Gamma_G|^2 - 2 |\Gamma_{DUT} \Gamma_G| \cos(\theta_{DUT} + \theta_G)$$

$$M_D = |1 - \Gamma_{Std} \Gamma_G|^2 = 1 + |\Gamma_{Std} \Gamma_G|^2 - 2 |\Gamma_{Std} \Gamma_G| \cos(\theta_{Std} + \theta_G)$$

Let $A = |\Gamma_G|$, $B = |\Gamma_{DUT}|$, $C = |\Gamma_{Std}|$, $D = \cos(\theta_{DUT} + \theta_G)$, $E = \cos(\theta_{Std} + \theta_G)$, and $Ratio\ Factor = K_{Std} \times \frac{P_{DUT}}{P_{Std}}$.

Then,

$$K_{DUT} = K_{Std} \times \frac{P_{DUT}}{P_{Std}} = Ratio\ Factor \times M = Ratio\ Factor \times \frac{1 + A^2 B^2 - 2 ABD}{1 + A^2 C^2 - 2 ACE}$$

(10)

According to GUM for the law of propagation of uncertainties, the sensitivity coefficients are partial differentiations with respect to the individual variables in equation (10), total 9 items. The derived sensitivity coefficients are as follows:

1. The sensitivity coefficient for $K_{Std}$

$$\frac{\partial K_{DUT}}{\partial K_{Std}} = \frac{K_{DUT}}{K_{Std}}$$

(11)

2. The sensitivity coefficient for $P_{DUT}$

$$\frac{\partial K_{DUT}}{\partial P_{DUT}} = \frac{K_{DUT}}{P_{DUT}}$$

(12)

3. The sensitivity coefficient for $P_{Std}$

$$\frac{\partial K_{DUT}}{\partial P_{Std}} = -\frac{K_{DUT}}{P_{Std}}$$

(13)

4. The sensitivity coefficient for $|\Gamma_{Std}|$ ($=C$)

$$\frac{\partial K_{DUT}}{\partial |\Gamma_{Std}|} = -\frac{K_{DUT}}{M_D} \times 2 A(AC - E)$$

(14)

5. The sensitivity coefficient for $\theta_{Std}$

$$\frac{\partial K_{DUT}}{\partial \theta_{Std}} = -\frac{K_{DUT}}{M_D} \times 2 AC \sin(\theta_{Std} + \theta_G)$$

(15)

6. The sensitivity coefficient for $|\Gamma_{DUT}|$ ($=B$)

$$\frac{\partial K_{DUT}}{\partial |\Gamma_{DUT}|} = Ratio\ Factor \times \frac{1}{M_D} \times 2 A(AB - D)$$

(16)
7. The sensitivity coefficient for $\theta_{DUT}$

$$\frac{\partial K_{DUT}}{\partial \theta_{DUT}} = \text{Ratio Factor} \times \frac{2AB}{M_D} \sin(\theta_{DUT} + \theta_C) \quad (17)$$

8. The sensitivity coefficient for $|\Gamma_G| (= A)$

$$\frac{\partial K_{DUT}}{\partial |\Gamma_G|} = \text{Ratio Factor} \times \frac{2B}{M_D} \times (AB - D) - \frac{MN \times 2C}{(M_D)^2} \times (AC - E) \quad (18)$$

9. The sensitivity coefficient for $\theta_G$

$$\frac{\partial K_{DUT}}{\partial \theta_G} = \text{Ratio Factor} \times \{ \frac{2AB}{M_D} \sin(\theta_{DUT} + \theta_C) - \frac{M}{M_D} \times 2AC \sin(\theta_{Std} + \theta_C) \} \quad (19)$$

3.2.2.2 Polar representation of sensitivity coefficients for equation (4)

Representing calibration factor $K_{DUT}$ of equation (4) with real and imaginary components, the following denotation is used to denote the real and imaginary components of reflection coefficient:

- $a = \text{Re}\{\Gamma_{Std}\} = \Gamma_{Std-\text{Re}}$
- $b = \text{Im}\{\Gamma_{Std}\} = \Gamma_{Std-\text{Im}}$
- $c = \text{Re}\{\Gamma_{DUT}\} = \Gamma_{DUT-\text{Re}}$
- $d = \text{Im}\{\Gamma_{DUT}\} = \Gamma_{DUT-\text{Im}}$
- $e = \text{Re}\{\Gamma_G\} = \Gamma_{G-\text{Re}}$
- $f = \text{Im}\{\Gamma_G\} = \Gamma_{G-\text{Im}}$

Then the calibration model becomes:

$$K_{DUT} = K_{Std} \times P_{RATIO} \times M1$$

$$= \text{Ratio Factor} \times \frac{1+2df-2ce+c^2e^2+d^2e^2+c^2f^2+d^2f}{1+2bf-2ae+a^2e^2+b^2e^2+a^2f^2+b^2f^2} \quad (20)$$

Let $M1_N$ represent the numerator and $M1_D$ denominator of mismatch $M1$ in real and imaginary format:

$$M1_N = |1 - \Gamma_{DUT}\Gamma_G|^2 = 1+2df-2ce+c^2e^2+d^2e^2+c^2f^2+d^2f$$

$$M1_D = |1 - \Gamma_{Std}\Gamma_G|^2 = 1+2bf-2ae+a^2e^2+b^2e^2+a^2f^2+b^2f^2$$

Similarly, according to GUM for the law of propagation of uncertainties, the sensitivity coefficients are partial differentiation with respect to each input quantities in equation (20), total 9 items. The derived sensitivity coefficients are as follows:

1. The sensitivity coefficients for $K_{Std}$, $P_{DUT}$, and $P_{Std}$ are the same as those in equations (11), (12) and (13). But here $K_{DUT}$ should use equation (20) instead.
2. The sensitivity coefficient for $\Gamma_{Std-\text{Re}} (= a)$ can be evaluated to obtain

$$\frac{\partial K_{DUT}}{\partial \Gamma_{Std-\text{Re}}} = \frac{K_{DUT}}{M1_D} \times (2e - 2ae^2 - 2af^2) \quad (21)$$
3. The sensitivity coefficient for $\Gamma_{\text{Std-Img}} (= b)$

$$\frac{\partial K_{\text{DUT}}}{\partial \Gamma_{\text{Std-Img}}} = \frac{K_{\text{DUT}}}{M1_D} \times \{(2f + 2be^2 + 2bf^2)\}$$  \hspace{1cm} (22)

4. The sensitivity coefficient for $\Gamma_{\text{DUT-Re}} (= c)$

$$\frac{\partial K_{\text{DUT}}}{\partial \Gamma_{\text{DUT-Re}}} = \text{Ratio Factor} \times \frac{2e + 2ce^2 + 2cf^2}{M1_D}$$  \hspace{1cm} (23)

5. The sensitivity coefficient for $\Gamma_{\text{DUT-Img}} (= d)$

$$\frac{\partial K_{\text{DUT}}}{\partial \Gamma_{\text{DUT-Img}}} = \text{Ratio Factor} \times \frac{2f + 2de^2 + 2df^2}{M1_D}$$  \hspace{1cm} (24)

6. The sensitivity coefficient for $\Gamma_{G-\text{Re}} (= e)$

$$\frac{\partial K_{\text{DUT}}}{\partial \Gamma_{G-\text{Re}}} = \text{Ratio Factor} \times \frac{2c + 2e^2 + 2d^2 - (M1)\{2a - 2a^2 e - 2b^2 e\}}{M1_D}$$  \hspace{1cm} (25)

7. The sensitivity coefficient for $\Gamma_{G-\text{Img}} (= f)$

$$\frac{\partial K_{\text{DUT}}}{\partial \Gamma_{G-\text{Img}}} = \text{Ratio Factor} \times \frac{2d + 2e^2 f + 2d^2 f - (M1)\{2b + 2a^2 f + 2b^2 f\}}{M1_D}$$  \hspace{1cm} (26)

3.2.2.3 Example

With the same data as used in Table 2, the uncertainty budget is listed in Table 3 with mismatch corrections considered in the calibration equation.

<table>
<thead>
<tr>
<th>Quantity</th>
<th>Estim. $x_i$</th>
<th>Standard uncertainty</th>
<th>Probability distribution</th>
<th>Sensitivity coefficient $c_i$</th>
<th>Uncertainty contribution $u_i(y)$</th>
</tr>
</thead>
<tbody>
<tr>
<td>$K_{\text{Std}}$</td>
<td>0.9894</td>
<td>0.0012</td>
<td>normal</td>
<td>0.9996</td>
<td>0.0012</td>
</tr>
<tr>
<td>$P_{\text{DUT}}$</td>
<td>1.0158</td>
<td>0.0018</td>
<td>normal</td>
<td>0.9737</td>
<td>0.0018</td>
</tr>
<tr>
<td>$P_{\text{Std}}$</td>
<td>1.0021</td>
<td>0.0004</td>
<td>normal</td>
<td>-0.9870</td>
<td>-0.0004</td>
</tr>
<tr>
<td>$\Gamma_{\text{DUT-Img}}$</td>
<td>0.0600</td>
<td>0.0120</td>
<td>normal</td>
<td>-0.4613</td>
<td>-0.0055</td>
</tr>
<tr>
<td>$\Gamma_{\text{DUT-Phase}}$</td>
<td>3.1416</td>
<td>1.5709</td>
<td>normal</td>
<td>0.4581</td>
<td>0.0027</td>
</tr>
<tr>
<td>$\Gamma_{\text{Std-Img}}$</td>
<td>0.0300</td>
<td>0.0060</td>
<td>normal</td>
<td>0.0000</td>
<td>0.0000</td>
</tr>
<tr>
<td>$\Gamma_{\text{Std-Phase}}$</td>
<td>3.1416</td>
<td>1.5709</td>
<td>normal</td>
<td>0.0000</td>
<td>0.0000</td>
</tr>
<tr>
<td>$\Gamma_{\text{Gmag}}$</td>
<td>0.2300</td>
<td>0.0460</td>
<td>normal</td>
<td>-0.0606</td>
<td>-0.0028</td>
</tr>
<tr>
<td>$\Gamma_{\text{Gphase}}$</td>
<td>3.1416</td>
<td>1.5709</td>
<td>normal</td>
<td>0.0000</td>
<td>0.0000</td>
</tr>
<tr>
<td>$y = K_U$</td>
<td>0.9890</td>
<td></td>
<td></td>
<td></td>
<td>0.0071</td>
</tr>
</tbody>
</table>

18 GHz based on best specifications
Table 3. Uncertainty budget at 18 GHz for measurement uncertainty improvement with mismatch correction for Fig. 2 measurement setup. Uncertainties $u_i$ are at one standard deviation. Powers are measured in mW.

Compare the result in Table 3 with that in Table 2, the uncertainty is improved by mismatch correction. In the calculation, the magnitude uncertainty of reflection coefficient is assumed to be 40% of its value and the phase uncertainty is assumed to be $180^\circ$ and value is $\pi$ for all. In terms of computation cost, it is the same for both Cartesian and Polar coordinate representations. Note that additional uncertainties should also be included such as the connector repeatability, noise, cable flexibility, drift, linearity and frequency error when they are not negligible in practical application.

### 3.3 Discussions on simple direct comparison transfer

#### 3.3.1 The pros and cons for simple direct comparison transfer

The above analyses based on the method illustrated in Fig. 2 are the simple direct comparison transfer calibration method. It is the most basic microwave power transfer technique. The system setup is fast, easy and simple.

But it is only for relatively rough measurement, since the reflection coefficient of a generator $\Gamma_G$ is more difficult to measure than that of a passive load. $\Gamma_G$ varies with time and frequency. Many signal generators and amplifiers have non-linear output impedance. There are several developments on the measurement methods for $\Gamma_G$ (Shimaoka et al., 2006; Torok et al., 2001) which is beyond the discussion of this chapter.

#### 3.3.2 Improvement of direct comparison transfer by inserting passive components

To obviate the $\Gamma_G$, an intermediate component is proposed to use to increase the accuracy. If put an attenuator or an isolator on the source output, the reflection coefficient of a generator is improved. But low power is a problem for inserting an attenuator and being limited in frequency range is a problem for isolators (Rumfelt & Elwell, 1967).
A leveling circuit or ratio measurement can avoid the troubles. Resistive power splitter or directional coupler is normally used as the intermediate component and constructing the leveling circuit.

### 3.3.3 Splitter vs divider vs Tee: Scattering parameter matrix and suitable applications

Before going further analyses, it is time to say a few words about the power splitter (Johson, 1975), power divider and Tee since even some experienced engineers misuse them and not clear of the differences among them fundamentally. The physical and mathematical expressions are given here.

As illustrated in Fig. 3 for a) power divider; b) power splitter; and c) Tee, there are three resistors for power divider; two resistors for power splitter; and no resistors for Tee.

The three resistors in power divider are \( \frac{Z_0}{3} \) each; for 50 ohm system, they are 16-2/3 Ω. Two resistors in power splitter are \( Z_0 \) each.

The power splitter has fixed input at port one; power divider and Tee inputs are exchangeable (bi-directional).

Substantially the scattering parameter matrix or S-parameter matrix are different for these three 3-port components as shown under the physical structures in Fig. 3; in which the S-parameter matrixes are for ideal cases.

Power splitter is used in leveling or ratio measurement; power divider is used for simple power division. Tee is used in low frequency case and in those non-critical measurements.

![Fig. 3. Three 3-port components and their corresponding scattering parameter matrix](image)

### 4. Coaxial splitter based power sensor calibration by direct comparison transfer

#### 4.1 Benefit

The coaxial splitter based calibration setup is illustrated in Fig. 4. By using coaxial splitter in the direct comparison transfer method of RF and microwave power sensor calibration, the measurement accuracy improves from source mismatch effect and the load / device under test (DUT) mismatch effects. The explanations are as follows:
a. Improving source match is achieved by holding effective source output power constant. When a leveling loop or ratio sensor is employed, port 2 of the power splitter becomes the effective source output. $\Gamma_{EG}$ is the equivalent source reflection coefficient rather than $\Gamma_G$ as illustrated in Fig. 5(a). It solves the measurement trouble of $\Gamma_G$ for signal generator by obtaining passive component splitter S-parameters. The well known equivalent source reflection coefficient equation for power splitter

$$\Gamma_{EG} = S_{22} - S_{21} \frac{S_{32}}{S_{31}}$$

will be further discussed in section 6.

![Fig. 4. Coaxial splitter based power measurement setup](image)

b. As illustrated in Fig. 5(b), since $S_{21} = S_{31}$, both arms 2 and 3 experience the same variation in input power, the leveling loop can compensate for any changes in effective source’s output power.

c. As illustrated in Fig. 5(c) the leveling loop compensates for load variations as well. Since $S_{22} = S_{32}$ it means that any changes in output power caused by the load reflection is also seen in the monitoring arm, permitting the leveling loop to compensate for these variations.

The source mismatch effect can be read through the monitoring arm, and so do the load / DUT mismatch effects. The system setup and mathematical model is established based on the complex reflection coefficients. Analyses have been provided through their S-parameter matrix to perform full mismatch corrections (Weidman, 1996; Juroshek, 2000; Ginley, 2006; Crowley, 2006; Shan et al., 2008; Shan et al., 2010b). Here we provide models for different cases in the application.

Since the splitter is not perfect, we seek to characterize them so that the mismatch error can be minimized by mathematical correction. The computation involved is not trivial, but the benefits of the technique are considerable, in that accurate measurements are made across a broad band without the need for mechanical adjustment.
4.2 Modeling

Models of the three cases are considered for the system setup illustrated in Fig. 4.

Case 1 – obtain $\eta_{DUT}$ from $\eta_{Std}$:

$$\eta_{DUT} = \eta_{Std} \times P_{RATIO} \times MM = \eta_{Std} \times \frac{P_{DUT}}{P_{Std}} \times \frac{P_{Std}}{P_{3DUT}} \times \frac{1 - |\Gamma_{Std}|^2}{1 - |\Gamma_{DUT}|^2} \times \frac{|1 - \Gamma_{DUT}\Gamma_{EG}|^2}{|1 - \Gamma_{Std}\Gamma_{EG}|^2} \quad (28)$$

The $\Gamma_{Std}$, $\Gamma_{DUT}$ and $\Gamma_{EG}$ are the complex value reflection coefficients for the standard, DUT and equivalent signal generator respectively as indicated in Fig. 4. The power ratio is different from equation (3), which is caused by using splitter with monitoring arm. And here $\Gamma_{EG}$ is used instead of $\Gamma_{G}$, different from simple direct comparison transfer. The value of $\Gamma_{EG}$ is obtained by equation (27).

Case 2 – obtain $K_{DUT}$ from $K_{Std}$:

$$K_{DUT} = K_{Std} \times \frac{P_{DUT}}{P_{Std}} \times \frac{P_{Std}}{P_{3DUT}} \times \frac{|1 - \Gamma_{DUT}\Gamma_{EG}|^2}{|1 - \Gamma_{Std}\Gamma_{EG}|^2} \quad (29)$$

Case 3 – obtain $K_{DUT}$ from $\eta_{Std}$:

$$K_{DUT} = \eta_{Std} \times \frac{P_{DUT}}{P_{Std}} \times \frac{P_{Std}}{P_{3DUT}} \times \frac{1 - |\Gamma_{Std}|^2}{|1 - \Gamma_{Std}\Gamma_{EG}|^2} \times \frac{|1 - \Gamma_{DUT}\Gamma_{EG}|^2}{|1 - \Gamma_{Std}\Gamma_{EG}|^2} \quad (30)$$

Case 3 is the practical application calibration equation in national metrology institutes. The obtained value from a primary standard is effective efficiency $\eta_{Std}$, the customer DUT calibration factor requests $K_{DUT}$.

In the following sections, we separately provide both Cartesian and Polar representations for practical solution and application.

4.3 Polar model

The models for the above three cases in terms of magnitude and phase is derived and expressed as follows:
RF and Microwave Power Sensor Calibration by Direct Comparison Transfer

\[ \eta_{DUT} = \eta_{Std} \times \frac{P_{DUT}}{P_{Std}} \times \frac{P_{SSD}}{P_{DUTSSD}} \times \frac{1}{1 - \left| \Gamma_{Std} \right|^2} \times \frac{1 + \left| \Gamma_{STD} \right|^2 \left| \Gamma_{EC} \right|^2 - 2 \left| \Gamma_{DUT} \right| \left| \Gamma_{EC} \right| \cos(\theta_{DUT} + \theta_{EC})}{1 + \left| \Gamma_{STD} \right|^2 \left| \Gamma_{EC} \right|^2 - 2 \left| \Gamma_{Std} \right| \left| \Gamma_{EC} \right| \cos(\theta_{Std} + \theta_{EC})} \] (31)

\[ K_{DUT} = K_{Std} \times \frac{P_{DUT}}{P_{Std}} \times \frac{P_{SSD}}{P_{DUTSSD}} \times \frac{1 + \left| \Gamma_{DUT} \right|^2 \left| \Gamma_{EC} \right|^2 - 2 \left| \Gamma_{DUT} \right| \left| \Gamma_{EC} \right| \cos(\theta_{DUT} + \theta_{EC})}{1 + \left| \Gamma_{Std} \right|^2 \left| \Gamma_{EC} \right|^2 - 2 \left| \Gamma_{Std} \right| \left| \Gamma_{EC} \right| \cos(\theta_{Std} + \theta_{EC})} \] (32)

\[ K_{DUT} = \eta_{Std} \times \frac{P_{DUT}}{P_{Std}} \times \frac{P_{SSD}}{P_{DUTSSD}} \times (1 - \left| \Gamma_{Std} \right|^2) \times \frac{1 + \left| \Gamma_{STD} \right|^2 \left| \Gamma_{EC} \right|^2 - 2 \left| \Gamma_{DUT} \right| \left| \Gamma_{EC} \right| \cos(\theta_{DUT} + \theta_{EC})}{1 + \left| \Gamma_{STD} \right|^2 \left| \Gamma_{EC} \right|^2 - 2 \left| \Gamma_{Std} \right| \left| \Gamma_{EC} \right| \cos(\theta_{Std} + \theta_{EC})} \] (33)

4.4 Uncertainty based on polar representation by Monte Carlo method

If using GUM to evaluate the uncertainty of measurement, it is based on propagation of uncertainties which is similar to analyses in previous section with more items. The mismatch uncertainty part is similar, just replace \( F_G \) with \( F_{EG} \). It is seen from previous section that the sensitivity coefficients are quite tedious for complex value involved models although only first order summation of uncertainties are used. The derived sensitivity coefficients with partial differentiations with respect to each variable for the above models (31), (32) and (33) are obtained and have been numerically appreciated by making use of MATLAB. Here we discuss the Monte Carlo simulation Method (MCM). Then we compare the uncertainties by two methods and discuss the findings.

MCM is based on the propagation of distribution proposed by GUM supplement 1 (JCGM, 101:2008) instead of the GUM propagation of uncertainty method. The MCM allows one to get rid of much of the calculation of partial derivatives where analytical expressions are complex. The MCM evaluates measurement uncertainty by setting a probability function to each input quantity in the measurement equation. From a series of numerical calculations, probability density function (pdf) of the output function is obtained and the standard uncertainty is evaluated from this pdf.

The steps for applying MCM are summarized as follows:

Step 1. Select the number \( N \) of Monte Carlo Trials (iterations) to be made, with a given model \( Y \), such as using equation (32). A value of \( 10^6 \) is often expected to deliver a 95% coverage interval for the output quantity.

Step 2. Generate \( N \) samples, by means of sampling from each input pdf. This \( N \) samples could be stored as a row vector in the program.

Step 3. For each of the input vector, form the output model \( Y \), such as using equation (32).

Step 4. Sort the \( N \) evaluated values into a strict increasing order. Here, strict increasing requires all output values of \( Y \) to be unique. A good random generator in the program should satisfy this condition.

Step 5. With the sorted values, create a histogram for an approximation of the output pdf.

Step 6. Evaluate the estimate of the output model, given as \( y \), and the standard uncertainty of \( y \), given as \( u(y) \).

Step 7. Form a coverage interval for \( Y \), given a required probability \( p \). If the output pdf is symmetric, then the probabilistically symmetric interval is equal to the shortest coverage interval. Otherwise, shortest coverage interval has to be used.
The Monte Carlo numerical simulation is performed through a program developed using MATLAB software. It allows estimating the measurement uncertainties based on the mathematical models.

4.5 Measurement system setup and type N connection result

The realization and implementation of coaxial splitter based power sensor calibration by direct comparison transfer shown in Fig. 4 is illustrated in Fig. 6 for the type N connector system setup (up to 18 GHz).

The calibration results of calibration factor of a power sensor with the measurement uncertainties are illustrated in Fig. 7. The DUT sensor is calibrated by a power standard (CN mount) in the frequency range 50 MHz to 18 GHz using the direct comparison transfer method with Weinschel 1870A splitter. The effective efficiency of power standard was measured using a micro-calorimeter. The calibration factor is calculated using equation (33). The expanded uncertainties are less than 0.9% for the frequency range at coverage factor of \( k = 2 \).

Compared with the results obtained in section 3, it is shown that with insertion of the passive splitter with monitoring arm, measurement uncertainties are much improved. And such measurement transfer system adds little measurement uncertainty when transfer the value from standard to DUT.

Uncertainty budget table is listed in Table 4 for the calibration system at frequency of 8 GHz.

Fig. 6. Type N connection realization of splitter based power sensor calibration by direct comparison transfer
RF and Microwave Power Sensor Calibration by Direct Comparison Transfer

Fig. 7. Measurement results by direct comparison method for power sensor 8481A vs CN.

Table 4. Uncertainty budget at 8 GHz for Fig. 4 measurement setup. Uncertainties $u_i$ are at one standard deviation. Powers are measured in mW.

<table>
<thead>
<tr>
<th>Quant.</th>
<th>Estim.</th>
<th>Standard uncertainty</th>
<th>probability distribution</th>
<th>Sensitivity coefficient</th>
<th>Uncertainty contribution</th>
</tr>
</thead>
<tbody>
<tr>
<td>$X_i$</td>
<td>$x_i$</td>
<td>$u(x_i)$</td>
<td></td>
<td>$c_i$</td>
<td>$u_i(y)$</td>
</tr>
<tr>
<td>$\eta_{\text{Std}}$</td>
<td>0.9650</td>
<td>0.00165</td>
<td>normal</td>
<td>1.00617</td>
<td>0.00166</td>
</tr>
<tr>
<td>$P_{\text{Std}}$</td>
<td>0.9774</td>
<td>0.00036</td>
<td>normal</td>
<td>-0.99343</td>
<td>-0.00036</td>
</tr>
<tr>
<td>$P_{\text{DUT}}$</td>
<td>0.9886</td>
<td>0.00171</td>
<td>normal</td>
<td>0.98219</td>
<td>0.00168</td>
</tr>
<tr>
<td>$P_{3\text{Std}}$</td>
<td>1.0000</td>
<td>0.00010</td>
<td>normal</td>
<td>0.97099</td>
<td>0.00010</td>
</tr>
<tr>
<td>$P_{3\text{DUT}}$</td>
<td>1.0000</td>
<td>0.00010</td>
<td>normal</td>
<td>-0.97099</td>
<td>-0.00010</td>
</tr>
<tr>
<td>$\Gamma_{\text{Stdmag}}$</td>
<td>0.0466</td>
<td>0.00750</td>
<td>normal</td>
<td>-0.14611</td>
<td>-0.00110</td>
</tr>
<tr>
<td>$\Gamma_{\text{StdPhase}}$</td>
<td>-1.4228</td>
<td>0.18328</td>
<td>normal</td>
<td>-0.00269</td>
<td>-0.00049</td>
</tr>
<tr>
<td>$\Gamma_{\text{DUTmag}}$</td>
<td>0.0047</td>
<td>0.00750</td>
<td>normal</td>
<td>-0.07607</td>
<td>-0.00057</td>
</tr>
<tr>
<td>$\Gamma_{\text{DUTPhase}}$</td>
<td>2.8563</td>
<td>1.57088</td>
<td>normal</td>
<td>0.00012</td>
<td>0.00020</td>
</tr>
<tr>
<td>$\Gamma_{\text{EGmag}}$</td>
<td>0.0414</td>
<td>0.00751</td>
<td>normal</td>
<td>-0.07154</td>
<td>-0.00054</td>
</tr>
<tr>
<td>$\Gamma_{\text{EGphase}}$</td>
<td>-2.5226</td>
<td>0.18381</td>
<td>normal</td>
<td>-0.00257</td>
<td>-0.00047</td>
</tr>
</tbody>
</table>

$y = K_U \quad 0.9678 \quad 0.0029$

$\eta_{\text{DUT}} = \eta_{\text{Std}} \times \frac{P_{\text{DUT}}}{P_{\text{Std}}} \times \frac{P_{3\text{Std}}}{P_{3\text{DUT}}} \times \frac{1-A^2-B^2}{1-C^2-D^2} \times \frac{1+2DF-2CE+C^2E^2+D^2E^2+C^2F^2+D^2F}{1+2BF-2AE+A^2E^2+B^2E^2+A^2F^2+B^2F^2}$ \hspace{1cm} (34)

$K_{\text{DUT}} = K_{\text{Std}} \times \frac{P_{\text{DUT}}}{P_{\text{Std}}} \times \frac{P_{3\text{Std}}}{P_{3\text{DUT}}} \times \frac{1+2DF-2CE+C^2E^2+D^2E^2+C^2F^2+D^2F}{1+2BF-2AE+A^2E^2+B^2E^2+A^2F^2+B^2F^2}$ \hspace{1cm} (35)

$K_{\text{DUT}} = \eta_{\text{Std}} \times \frac{P_{\text{DUT}}}{P_{\text{Std}}} \times \frac{P_{3\text{Std}}}{P_{3\text{DUT}}} \times (1-A^2-B^2) \times \frac{1+2DF-2CE+C^2E^2+D^2E^2+C^2F^2+D^2F}{1+2BF-2AE+A^2E^2+B^2E^2+A^2F^2+B^2F^2}$ \hspace{1cm} (36)
where for all three cases, \( \Gamma_{std} = A + jB, \quad \Gamma_{DUT} = C + jD, \quad \Gamma_{EG} = E + jF \).

### 4.7 Measurement system for 2.4mm connection and comparison

The measurement setup for 2.4mm connector (up to 50 GHz) is illustrated in photo in Fig. 8. The calibration result is illustrated in Fig. 9. For uncertainty evaluation, the parameters used in the MCM are listed in Table 5. And the two columns in Table 6 compare the results obtained from GUM and MCM for selected frequency points.

Using Monte Carlo method, we generate a graphical approximation in the form of a histogram of the probability density function of the output quantity, represented in Fig. 10(a) with the frequency at 50 GHz. The uncertainty differences from the two methods for different frequency range are illustrated in Fig. 10(b). In this case, the uncertainties obtained by different methods are quite close. They can be used as the verification of the measurement uncertainty evaluations.

![Figure 8](image1.png)

**Fig. 8.** 2.4mm connection realization of splitter based power sensor calibration by direct comparison transfer

![Figure 9](image2.png)

**Fig. 9.** 50 MHz to 50 GHz using the direct comparison transfer. The expanded uncertainties are less than 4% for the frequency range at \( k=2 \).
Table 5. Probability distribution parameters assigned to the individual input quantities at 50 GHz

<table>
<thead>
<tr>
<th>$X_i$</th>
<th>Distribution</th>
<th>Parameters</th>
</tr>
</thead>
<tbody>
<tr>
<td>$\eta_{Std}$</td>
<td>$N(\mu, \sigma^2)$</td>
<td>$0.9047$</td>
</tr>
<tr>
<td>$P_{DUT}$</td>
<td>$N(\mu, \sigma^2)$</td>
<td>$8.4400e-04$</td>
</tr>
<tr>
<td>$P_{Std}$</td>
<td>$N(\mu, \sigma^2)$</td>
<td>$8.6200e-04$</td>
</tr>
<tr>
<td>$P_{3DUT}$</td>
<td>$N(\mu, \sigma^2)$</td>
<td>$9.9800e-04$</td>
</tr>
<tr>
<td>$P_{3Std}$</td>
<td>$N(\mu, \sigma^2)$</td>
<td>$1.0000e-03$</td>
</tr>
<tr>
<td>$</td>
<td>\Gamma_{DUT}</td>
<td>$</td>
</tr>
<tr>
<td>$\theta_{DUT}$</td>
<td>$N(\mu, \sigma^2)$</td>
<td>$2.7501$</td>
</tr>
<tr>
<td>$</td>
<td>\Gamma_{EG}</td>
<td>$</td>
</tr>
<tr>
<td>$\theta_{EG}$</td>
<td>$N(\mu, \sigma^2)$</td>
<td>$2.6222$</td>
</tr>
<tr>
<td>$</td>
<td>\Gamma_{Std}</td>
<td>$</td>
</tr>
<tr>
<td>$\theta_{Std}$</td>
<td>$N(\mu, \sigma^2)$</td>
<td>$2.9504$</td>
</tr>
</tbody>
</table>

Table 6. Compares the results of different uncertainty evaluation methods for selected frequency points

<table>
<thead>
<tr>
<th>Frequency = 50MHz</th>
<th>GUM</th>
<th>MCM</th>
</tr>
</thead>
<tbody>
<tr>
<td>Mean</td>
<td>1.00854</td>
<td>1.008492</td>
</tr>
<tr>
<td>Stdev</td>
<td>0.008106</td>
<td>0.0076298</td>
</tr>
<tr>
<td>Lower Lim</td>
<td>0.992328</td>
<td>0.9935218</td>
</tr>
<tr>
<td>Upper Lim</td>
<td>1.024753</td>
<td>1.023428</td>
</tr>
<tr>
<td>Coverage Int</td>
<td>0.016212</td>
<td>0.0149531</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Frequency = 50GHz</th>
<th>GUM</th>
<th>MCM</th>
</tr>
</thead>
<tbody>
<tr>
<td>Mean</td>
<td>0.874599</td>
<td>0.8728157</td>
</tr>
<tr>
<td>Stdev</td>
<td>0.01635</td>
<td>0.0154997</td>
</tr>
<tr>
<td>Lower Lim</td>
<td>0.841899</td>
<td>0.8423992</td>
</tr>
<tr>
<td>Upper Lim</td>
<td>0.907299</td>
<td>0.9031398</td>
</tr>
<tr>
<td>Coverage Int</td>
<td>0.032699</td>
<td>0.0303703</td>
</tr>
</tbody>
</table>

Table 6. Compares the results of different uncertainty evaluation methods for selected frequency points
5. Generic models for power sensor calibration

5.1 Modeling with signal flow graph

For the sake of general purpose of application and analyses, in this section we discuss the three port structure expressed in signal flow graph. A flow graph method is particularly helpful in understanding a complex network relying on S-parameter matrix. Its correspondence with the physical behavior of the circuits allows simplification through well-founded approximations with no loss of physical insight (Bryant, 1993).

Referring to Fig. 11(a), the central part is signal flow graph for a general three port components, most commonly used are splitter and coupler. Three two-port networks, G, A and B, are added for the generic analysis. The signal source connected to port 1 is denoted by \( b_s \).
5.2 Non-touching theory for the modeling analyses

The full power of flow graphs is evident only when they are combined with the non-touching loop rule. It is then possible to find the transfer function between any two nodes of a network almost by inspection. The complex wave functions $a_i, b_i$ are the nodal points of signal flow along the paths designated by the arrows.

There are several special cases which are practically useful. We are going to discuss the following three cases.

5.3 Case study 1: Special case of the generic model comparing with section 4

When removing all the three two-port networks, G, A and B, the generic model is the special case as discussed in section 4. The generic model signal flow graph for a 3-port structure is then simplified as Fig. 11(b) (Wong, 2002).

The steps to evaluate the flow graph are summarized as follows:

There are two paths from $b_x$ to $a_x$

$$P_1 = S_{21} \cdot k_2$$
$$P_2 = S_{31} \cdot \Gamma_r \cdot S_{32} \cdot k_2$$

The transfer function is,

$$\frac{a_x}{b_x} = \frac{S_{21} \cdot k_2 [1 - (\Gamma_r \cdot S_{33})]}{1 - \sum L(1) + \sum L(2) - \sum L(3) + ...} + \frac{S_{31} \cdot \Gamma_r \cdot S_{32} \cdot k_2 [1 - 0]}{1 - \sum L(1) + \sum L(2) - \sum L(3) + ...}$$

There are two paths from $b_x$ to $a_y$

$$P_1 = S_{31} \cdot k_3$$
$$P_2 = S_{21} \cdot \Gamma_d \cdot S_{32} \cdot k_3$$

The transfer function is,

$$\frac{a_y}{b_x} = \frac{S_{31} \cdot k_3 [1 - (\Gamma_d \cdot S_{22})]}{1 - \sum L(1) + \sum L(2) - \sum L(3) + ...} + \frac{S_{21} \cdot \Gamma_d \cdot S_{32} \cdot k_3 [1 - 0]}{1 - \sum L(1) + \sum L(2) - \sum L(3) + ...}$$

$$\frac{a_x}{a_y} = \frac{a_x}{b_x} \times \frac{b_x}{a_y}$$

Taking the ratio of $a_x$ with respect to $a_y$:

$$\frac{a_x}{a_y} = \frac{a_x}{b_x} \times \frac{b_x}{a_y} = \frac{k_2}{k_3} \left( \frac{S_{21}}{S_{31}} \right) \left( 1 + \frac{\Gamma_r \cdot S_{33} - \frac{S_{31} \cdot S_{32}}{S_{21}}}{1 - \Gamma_d \cdot S_{22}} \right) \left( 1 - \frac{S_{33} - \frac{S_{31} \cdot S_{32}}{S_{21}}}{1 - \Gamma_d \cdot S_{22}} \right)$$

$$= \left( \frac{k_2}{k_3} \right) \left( \frac{S_{21}}{S_{31}} \right) \left( 1 - \frac{\Gamma_r \cdot S_{33} - \frac{S_{31} \cdot S_{32}}{S_{21}}}{1 - \Gamma_d \cdot S_{22}} \right)$$

Here $k_2$ and $k_3$ as illustrated in Fig. 11 are some unknown terms related to the cable and connector leakage, connection repeatability, drift, linearity and frequency error. And $\Gamma_d$ and $\Gamma_r$ are the so called equivalent source match terms of port 3 and port 2 respectively.

When the signal flow graph is used in power sensor calibration, recall that the power measured by a sensor $P_m$ is actually $|b_i|^2$ times calibration factor $K$, i.e.

$$P_m = |b_i|^2 \times K$$
where $|b_i|^2$ is the power delivered to port $i$ and $K$ is the calibration factor.

Recall that in the measurement setup shown in Fig. 4, port 3 of the splitter is connected to a power sensor and meter, serving as a monitoring arm. Port 2 is a test arm, alternatively connected with standard and DUT power sensor. Power is measured at both port 2 and port 3.

When DUT is connected to the port 2, from equation (38) we can obtain the following ratio:

$$\frac{P_{DUT}}{P_{3DUT}} = \frac{|b_{2DUT}|^2 \times K_{DUT}}{|b_{3DUT}|^2 \times K_{3DUT}}$$

$P_{DUT}$ and $P_{3DUT}$ are the powers measured at port 2 when DUT power sensor connected and that at port 3 respectively. $K_{DUT}$ and $K_{3DUT}$ are the calibration factors corresponding to the respective DUT sensor and sensor connected at port 3. Then

$$K_{DUT} = K_{3DUT} \times \frac{P_{DUT}}{P_{3DUT}} \times \frac{|b_{3DUT}|^2}{|b_{2DUT}|^2} \tag{39}$$

Alternatively connection with port 2 is replaced with the standard power sensor. Port 3 remains as the monitoring arm. With the changes made at port 2, the similar ratio can be obtained, $\frac{P_{Std}}{P_{3Std}} = \frac{|b_{2Std}|^2 \times K_{Std}}{|b_{3Std}|^2 \times K_{3Std}}$ and

$$K_{Std} = K_{3Std} \times \frac{P_{Std}}{P_{3Std}} \times \frac{|b_{3Std}|^2}{|b_{2Std}|^2} \tag{40}$$

Since $K_{3Std} = K_{3DUT}$ for port 3 remains as the monitoring arm no matter port 2 connected with DUT or standard sensor, ratio of equation (39) and equation (40) yields

$$K_{DUT} = K_{Std} \times \frac{P_{DUT}}{P_{3DUT}} \times \frac{|b_{3DUT}|^2}{|b_{2DUT}|^2} \times \frac{P_{3Std}}{P_{Std}} \times \frac{|b_{3Std}|^2}{|b_{2Std}|^2} \tag{41}$$

Link to the transfer function derived in equation (37), and let $a_x = b_2$ and $a_y = b_3$

$$\frac{b_2}{b_3} = a_x \over a_y = \left(\frac{k_2}{k_3}\right) \frac{S_{11}}{S_{31}} \left(\frac{1-\Gamma_{DUT}F_{D}}{1-\Gamma_{DUT}F_{2}}\right) \tag{42}$$

Further modifications are made to equation (41) to specify the transfer function for the alternate connections of the standard power sensor and the DUT power sensor to port 2 of the splitter while port 3 remains as the monitoring arm. Then:

$$K_{DUT} = K_{Std} \times \frac{P_{DUT}}{P_{3DUT}} \times \frac{P_{3Std}}{P_{Std}} \times \left(\frac{k_{2Std}}{k_{2DUT}}\right)^2 \times \left(\frac{1-\Gamma_{DUT}F_{D}}{1-\Gamma_{DUT}F_{2}}\right)^2 \tag{43}$$

Compare the derived equation (43) from signal flow graph Fig. 11(b) with equation (29) derived from Fig. 4, the only difference is $\left(\frac{k_{2Std}}{k_{2DUT}}\right)^2$ term. This term accounts for connector leakage, repeatability and so on. It is the Type A uncertainty shown in equation (5). The term is not explicitly expressed in the calibration equation (29). From different analytical method, we derive the same power sensor calibration equations.
5.4 Case study 2: Simultaneous comparison method analyses

Another way of using a resistive power splitter is the simultaneous comparison method. The reference standard and DUT power sensors to be calibrated are connected to the two output arms of the power splitter simultaneously. It seems an easy way of connection, no alternative connection of reference standard and DUT at port 2 of splitter as the above analyses. And there is application in some calibration laboratories.

With signal flow graph the transfer function can be easily obtained and reveal some findings. From equation (38), when DUT is connected to the port 2 and reference standard is connected to the port 3, we can obtain the following ratio:

$$\frac{P_{DUT}}{P_{3Std}} = \frac{|b_{2DUT}|^2 \times K_{DUT}}{|b_{3Std}|^2 \times K_{3Std}}$$

$P_{DUT}$ and $P_{3Std}$ are the powers measured at DUT power sensor connected to port 2 and the standard power sensor connected to port 3 respectively. $K_{DUT}$ and $K_{3Std}$ are the calibration factors corresponding to the respective sensors. Then

$$K_{DUT} = K_{3Std} \times \frac{P_{DUT}}{P_{3Std}} \times \frac{|b_{3Std}|^2}{|b_{2DUT}|^2}$$

(44)

Substitute equation (42) and simplify,

$$K_{DUT} = K_{Std} \times \frac{P_{DUT}}{P_{3Std}} \times \left( \frac{k_{3Std}}{k_{2DUT}} \right)^2 \times \frac{|s_{31}|^2}{|s_{21}|^2} \times \frac{1-r_{DURF}r_{es}}{1-r_{3Std}r_{es}}$$

(45)

Compare equation (45) with equation (43), correction for mismatch when using the simultaneous comparison method is more complicated than for the direct comparison transfer method in which one port is used to monitor the signal level and the power sensors to be compared are connected in turn to the other output port. It can be seen from equation (45) that the mismatches at both output ports of the power splitter are needed to be taken into account. Unless perfect symmetry can be assumed, it is necessary to determine for each output port the value of $r_3$ and $r_2$. For this reason the simultaneous comparison method is not often used for measurement where the highest precision is required. However the method is quite convenient if mismatch errors are not to be corrected for, provided that errors due to asymmetry are eliminated by interchanging the two power sensors to be compared.

5.5 Case study 3: Extend the model to waveguide application

As proved by equation (43), equation (29) actually can be extended to use for any three port structure, not only the coaxial splitter based calibration system. For waveguide power sensor calibration, the directional coupler replaces splitter as the media components. The same equation then is used for WR15 (50 GHz - 75 GHz) sensor calibration. The system setup is illustrated in Fig. 12.

6. Source mismatch determination and uncertainty

As mentioned in the beginning of section 4, the splitter provides small value of effective source reflection coefficient $\Gamma_{EC}$ at its output port 2 when used in a leveling or ratio measurement mode. And $\Gamma_{EC}$ can be calculated from its complex S-parameters by equation (27):
One possible way of determining $\Gamma_{EG}$ is to measure the splitter’s complex S-parameters and calculate its value from equation (27). The measurement uncertainty in the S-parameters is then evaluated (EURAMET, 2011). Reference (Ridler and Salter, 2001) presented the law of propagation of uncertainty using matrix notation, treating the complex quantities with real and imaginary evaluation.

Since the calculation method from S-parameters is sensitive to small measurement errors, several different measurement methods were proposed, such as the “passive open circuit” and “active open circuit” method (Moyer, 1987), and the direct calibration method (Juroshek, 1997). In the direct calibration method, the splitter is connected through ports 1 and 3 to a VNA. This effectively gives a new one port VNA at splitter port 2. This new VNA is calibrated using a one port calibration algorithm, e.g. short-open-load. The $\Gamma_{EG}$ is then obtained as one of the three one-port VNA error terms. References (Rodriguez, 2000; Yhland & Stenarson, 2007) assessed the measurement uncertainty and traceability in power splitter effective source reflection coefficient. Reference (Furrer, 2007) compared direct calibration method with the calculation method. It seems that the similar results were obtained.

7. Conclusion

From simple direct comparison transfer method, to coaxial splitter based direct comparison transfer, and then to the general models with signal flow graph analyses, the chapter has meticulously discussed RF and microwave power sensor calibration methods. The models and equations provided target for practical usage. The examples and case studies have shown the practical applications. The traceability and measurement uncertainty with GUM and MCM have provided in details. The calibration models and methods described are useful for the coaxial and waveguide power sensor calibrations. The general models can be further developed for different case analyses in future studies.
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9. References


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