STOCHASTIC INDEX NUMBERS:
A REVIEW

by

Kenneth W Clements
School of Economics and Commerce
The University of Western Australia

H Y Izan
School of Economics and Commerce, and
Faculty of Economics and Commerce
The University of Western Australia

and

E Antony Selvanathan
School of International Business
Griffith University

Abstract

The stochastic approach is a new way of viewing index numbers in which uncertainty and statistical ideas play a central role. Rather than just providing a single number for the rate of inflation, the stochastic approach provides the whole probability distribution of inflation. This paper reviews the key elements of the approach and then discusses some previously overlooked links with Fisher’s early work contained in his book The Making of Index Numbers. We then consider some more recent developments, including Diewert’s well-known critique of the stochastic approach, and provide responses to his criticisms. We also provide a review of Theil’s work on the stochastic approach, and present and extend Diewert’s work on this topic within the context of the Country Product Dummy method which measures price levels internationally.

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1. Introduction

There are two major streams in index-number theory. The first is the test approach whereby indexes are judged on their ability to satisfy certain criteria; this stream is associated with Fisher (1927) in particular. The economic theory of index numbers is the second stream and this deals with their foundations in utility theory; for a review, see Diewert (1981). A less well known methodology, but one which is now attracting considerable attention, is the stochastic approach (SA). When applied to the prices, the SA to index numbers treats the underlying (or “true”) rate of inflation as an unknown parameter to be estimated from the individual prices. That is, the individual prices are observed with error and the problem is a signal-extraction one of how to combine noisy prices so as to minimise the effects of measurement errors. Under certain circumstances, this approach leads to familiar index-number formulae such as Divisia, Laspeyres etc., but, as uncertainty plays a key role in the SA, their foundations differ markedly from the conventional deterministic approach. The SA provides not only a point estimate of the rate of inflation, but also its variance, the source of which is the divergence of the individual prices from a common trend, that is, the extent to which the structure of relative prices changes. Accordingly, the SA provides the intuitively plausible result that it is more difficult to obtain precise estimates of inflation when there are large changes in relative prices.

Krugman (1999) has likened the US inflationary process in the 1970s to the noise level in a restaurant:

“Once upon a time, … the US economy was like a trendy restaurant – one of those places where the tables are set close together and the ceiling seems custom-designed to maximise the din. What happens in that kind of environment is that everyone tries to talk above the background noise so as to be heard by his or her companions. But by talking louder, you yourself raise the noise level, forcing everyone else to talk louder, raising the noise level still further, and eventually everyone is shouting themselves hoarse. Substitute wage and price increases for speaking volume and inflation for the overall noise level, and you have a capsule analysis of the kind of inflationary spiral that the US faced in the 1970s.”

Although this instructive metaphor relates to the dynamics of inflation characterised by a wage-price spiral, it could equally well apply to the signal-extraction approach to measuring inflation. The conversation volume at an individual table is made up of some audible words which convey an intelligible message (the signal), plus some yelling (the noise) which does not. The measurement of the information content of all the messages in the restaurant then involves some form of filtering to minimise the distortive effects of the noise. This is exactly the basis of the SA in its decomposition.

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1 Balk (1995) provides a comprehensive survey of the test approach.
of the individual price changes into inflationary and relative-price components with some form of averaging procedure filtering out the distortionary impact of the latter.

The SA is also relevant to the conduct of monetary policy and inflation targeting in particular. A popular approach is for policy makers to exclude from the index the prices of volatile items such as food and energy, and use “core” or “underlying” inflation for the specification of inflation targets. In another approach, the Reserve Bank of Australia currently has a “soft” inflation target of 2-3 percent p.a. on average over the cycle. Both the exclusion of noisy prices from the index and the idea of a soft target could be given more satisfactory statistical foundations by employing the SA. The SA gives specific guidance regarding the weighting scheme employed in the index; such a scheme is comprehensive in that it deals with all items in the regime, rather than just setting to zero the weights of the volatile items. Regarding the specification of a soft target, the SA could be used to express the target as $X \pm 1.96$ standard errors, for example.

The SA originated in the work of Jevons and Edgeworth (see Frisch, 1936, for references), but then fell into obscurity, perhaps in part due to the criticism by Keynes (1930, pp. 85-88) that it was too rigid as the approach made no allowance for sustained changes in relative prices. For a history-of-thought review of stochastic index numbers, see Aldrich (1992) who attributes the introduction of the term “stochastic” in this context to Frisch (1936), and adopted by Allen (1975), to describe Edgeworth’s analysis. More recently, the SA has been rehabilitated by Balk (1980), Clements and Izan (1981, 1987), Crompton (2000), Giles and McCann (1994), Miller (1984), Ogwang (1995), Ong et al. (1999), Prasada Rao and Selvanathan (1992a, b), Prasada Rao et al. (2003), Selvanathan (1989, 1991, 1993) and Selvanathan and Prasada Rao (1992). This literature is still expanding and has been the subject of a book by Selvanathan and Prasada Rao (1994), who emphasise the versatility and usefulness of the approach, a review paper by Diewert (1995), which has a critical tone, and a response by Selvanathan and Prasada Rao (1999). Even more recently, papers have appeared by Diewert (2002, 2004) and Prasada Rao (2004) which extend the SA in new directions.

The above-mentioned critique by Diewert (1995), while unpublished, has been influential and widely cited as providing what some may see as a telling case against stochastic index numbers. In this paper, we provide an in-depth assessment of the criticisms and show how the majority can be answered satisfactorily. Section 2 of the paper provides an overview of stochastic index numbers, while Section 3 discusses some early ideas of Fisher (1927) that seem not to have been previously appreciated as having relevance to the stochastic approach. Our responses to Diewert’s criticisms are contained in Section 4. Theil’s (1967) stochastic approach is presented in Section 5. Section 6

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2 Related approaches involve using the median, other “trimmed” means, a dynamic factor index and averaging over longer horizons. See Bryan and Cecchetti (1993, 1994), Bryan et al. (1997) and Cecchetti (1997).
reviews and extends some recent results of Dievert (2002) which apply stochastic index number
theory to the problem of the measurement of price levels across countries. Concluding comments
are contained in Section 7.

2. What are Stochastic Index Numbers?

In this section, we provide a brief review of some of the basic results on stochastic index
numbers; this material is mainly based on Clements and Izan (1987). To make the exposition as
sharp and as clear as possible, we concentrate on the simplest possible cases. Although we only
consider prices, it should be clear that the SA applies also to quantities.

Let

\[ Dp_{it} = \log p_{it} - \log p_{i(t-1)} \]

be the log-change in the price of commodity \( i \) (\( i = 1, \ldots, n \))
from year \( t-1 \) to \( t \). Suppose that each price change is made up of a systematic part that is common
to all prices, \( \alpha_t \), and a zero-mean random component \( \varepsilon_{it} \),

\[ \alpha_t + \varepsilon_{it} \]

(2.1)

As the term \( \alpha_t \) equals \( E(Dp_{it}) \), it is interpreted as the common trend in all prices, or the underlying
rate of inflation. With this interpretation, the change in the relative price of good \( i \) is then
\( Dp_{it} - \alpha_t \). As equation (2.1) implies that \( Dp_{it} - \alpha_t = \varepsilon_{it} \) and as \( E(\varepsilon_{it}) = 0 \), it follows that the
expected value of the change in the \( i^{th} \) relative price is zero, which means that, on average, all
relative prices are constant. While this is obviously restrictive, the approach can be extended by
adding a commodity-specific parameter to (2.1), as will be discussed below.

Let the disturbances in (2.1) for \( i, j=1,\ldots,n, \varepsilon_{it} \), have variances and covariances of the form
\( \sigma_{ijt} \) and let \( \Sigma_i = \begin{bmatrix} \sigma_{ij} \end{bmatrix} \) be the corresponding \( n \times n \) covariance matrix. We write (2.1) for \( i = 1, \ldots, n \)
in vector form as

\[ \begin{bmatrix} Dp_{i1} \\ Dp_{i2} \\ \vdots \\ Dp_{in} \end{bmatrix} = \begin{bmatrix} \alpha_{i1} \\ \alpha_{i2} \\ \vdots \\ \alpha_{in} \end{bmatrix} + \begin{bmatrix} \varepsilon_{i1} \\ \varepsilon_{i2} \\ \vdots \\ \varepsilon_{in} \end{bmatrix} \]  

(2.2)

where \( Dp_i = [Dp_{it}] \), \( i=[1,\ldots,1]^t \), and \( \varepsilon_i = [\varepsilon_{it}] \) are all \( n \times 1 \) vectors. Application of GLS to
(2.2) yields the BLUE of \( \alpha_{i} \),

\[ \hat{\alpha}_i = \left( \alpha_i \right)^{-1} \left( \alpha_i \right)^{-1} Dp_i \]

(2.3)

with
As indicated above, $\varepsilon_{it}$ is interpreted as the change in the $i^{th}$ relative price. Suppose that $\varepsilon_{it}$ and $\varepsilon_{jt}$ for $i \neq j$ are independent and that the variance of $\varepsilon_{it}$ is inversely proportional to the corresponding budget share $w_{it}$,

\begin{equation}
\text{var} \varepsilon_{it} = \frac{\lambda_{i}^2}{w_{it}},
\end{equation}

where $\lambda_{i}$ is a parameter independent of $i$; and $w_{it} = p_{it}q_{it}/M_{t}$ is the $i^{th}$ budget share, with $q_{it}$ the quantity consumed of good $i$ in year $t$ and $M_{t} = \sum_{i=1}^{n} p_{it}q_{it}$ total expenditure. There are two justifications for specification (2.5), at least as an approximation: (i) As a commodity absorbs a large part of the overall economy (i.e., as its budget share rises), there is less scope for its relative price to vary as there is simply less amount of “all other goods” against which its price can change. This restriction on the scope for the relative price changes of a large good means that its variance is smaller. (ii) If we think in terms of optimal sampling of prices, it would make sense for the relevant statistical collection agency to devote more resources to sampling prices of the more “important” goods by obtaining more price quotations. One way of identifying the importance of a good is by the size of its budget share. Such a sampling procedure would again lead to smaller variances of the relative prices of the more important goods. Finally, note that as $\sum_{i=1}^{n} w_{it} \text{var} \varepsilon_{it} = n \lambda_{i}^2$, the parameter $\lambda_{i}^2$ is interpreted as proportional to a budget-share-weighted variance; this $\lambda_{i}^2$ can also be expressed as $\sum_{i=1}^{n} w_{it}^2 \text{var} \varepsilon_{it}$.

The above assumptions imply that $\text{cov}(\varepsilon_{it}, \varepsilon_{jt}) = \delta_{ij} \lambda_{i}^2 / w_{it}$, where $\delta_{ij}$ is the Kronecker delta ($\delta_{ij} = 1$ if $i = j$ , 0 otherwise), so that the $n \times n$ covariance matrix takes the form

\begin{equation}
\Sigma_{t} = \lambda_{t}^2 W_{t}^{-1}
\end{equation}

where $W_{t} = \text{diag}[w_{1t}, \ldots, w_{nt}]$. Equation (2.3) then becomes $\hat{\alpha}_{t} = (\mathbf{1}' W_{t} \mathbf{1})^{-1} \mathbf{1}' W_{t} \mathbf{D} p_{t}$.

Since $\mathbf{1}' W_{t} \mathbf{1} = \sum_{i=1}^{n} w_{it} = 1$ and $\mathbf{1}' W_{t} \mathbf{D} p_{t} = \sum_{i=1}^{n} w_{it} \mathbf{D} p_{it}$, the above simplifies to

\begin{equation}
\hat{\alpha}_{t} = \sum_{i=1}^{n} w_{it} \mathbf{D} p_{it}.
\end{equation}
In words, the estimator of the underlying rate of inflation is a budget-share weighted average of the \( n \) price log-changes, an attractively simple result. In this index, more weight is given to those goods occupying a larger fraction of the consumer’s budget, which makes intuitive sense. Furthermore, if we reinterpret \( w_{it} \) as the arithmetic average of the observed budget shares in years \( t-1 \) and \( t \), the right-hand-side of (2.7) is the Divisia price index number, which has a number of desirable properties.\(^3\)

The Divisia index is a weighted first-order moment of the \( n \) price-changes \( Dp_{it}, \ldots, Dp_{nt} \). The corresponding second-order moment is the Divisia variance,

\[
\Pi_t = \sum_{i=1}^{n} w_{it} (Dp_{it} - DP_t)^2,
\]

where \( DP_t = \sum_{i=1}^{n} w_{it} Dp_{it} \) is the Divisia index. This variance measures the cross-commodity variance of relative prices; when all relative prices are unchanged, \( \Pi_t = 0 \). Under covariance specification (2.6), the variance of \( \hat{\alpha}_t \), defined in equation (2.4), becomes \( \lambda_t^2 (t' W_t t)^{-1} = \lambda_t^2 \), which can be estimated unbiasedly by

\[
\left[1/(n-1)\right] (Dp_t - \hat{\alpha}_t t)' W_t (Dp_t - \hat{\alpha}_t t) = \left[1/(n-1)\right] \sum_{i=1}^{n} w_{it} (Dp_{it} - DP_t)^2,
\]

which is proportional to the Divisia variance \( \Pi_t \). Accordingly, under (2.6) we have

(2.8) \[
\text{var} \hat{\alpha}_t = \frac{1}{n-1} \Pi_t.
\]

In words, the sampling variance of the inflation estimator is proportional to the variance of relative prices. When there is more dispersion of relative prices, i.e., when prices are changing more disproportionately, the sampling variance increases. This is also an attractive result which agrees with the intuitive idea that the underlying rate of inflation is in some sense less well defined when there are large changes in relative prices.

The above results can be extended to allow for more general specifications for the distribution of the disturbance terms. Crompton (2000) analyses White (1980)-type heteroscedasticity and derives analytical scalar expressions for the standard error of inflation under this formulation. Selvanathan and Prasada Rao (1999) consider a more general error covariance structure. Even with these extensions, the basic insight remains unchanged, viz., the standard error

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\(^3\) This is also known as the Törnqvist (1936)-Theil (1967) index.
of the estimate of the rate of inflation increases with the degree of variability of relative prices. Model (2.1) can also be extended to allow for relative price changes by adding a commodity-specific parameter $\beta_i$:

\begin{equation}
Dp_i = \alpha_i + \beta_i + \epsilon_i, \quad E(\epsilon_i) = 0, \quad \text{var} \epsilon_i = \lambda_i^2 / w_i,
\end{equation}

where $w_i$ is the sample mean of $w_i$. The new parameter $\beta_i$ is interpreted as the systematic part of the change in the $i$th relative price, $Dp_i - \alpha_i$. As the $\beta_i$ are not identified, Clements and Izan (1987) use the normalization that a budget-share-weighted average of the relative price changes is zero, $\sum_{i=1}^n w_i \beta_i = 0$.

The above presentation of the key elements of the SA has been in the context of a price index formulated in terms of changes over time. However, this is not essential as the SA is also equally applicable to indexes in levels, such as Laspeyres and Paasche. For details, see Selvanathan (1991, 1993) and Selvanathan and Prasada Rao (1994); also the recent results of Diewert (2002), considered in detail in Section 6, which show how the approach can be applied to yield a number of familiar index formulae in the context of the measurement of price levels across countries.\footnote{For other cross-country applications of the SA, see Selvanathan and Prasada Rao (1992).}

3. Irving Fisher and Stochastic Index Numbers

In his monumental book The Making of Index Numbers, Fisher (1927) introduced the “atomistic” or “test” approach to index-number theory. According to this approach, the quality of a particular index number is assessed by its ability to satisfy three primary tests. (i) The commodity-reversal test, which means that the value of the index should be invariant to an interchange (or reversal) in the order of any two commodities. (ii) The time-reversal test, whereby the price index at time $t$ with base-period 0, $P_{t0}$, should be the reciprocal of $P_{0t}$, the index at time 0 with base-period $t$. In other words, the product of the forward and backward indexes should be unity. (iii) The factor-reversal test, according to which the product of the price and quantity indexes should equal the observable ratio of values in the corresponding years. Tests (ii) and (iii) constitute “the two legs on which index numbers can be made to walk” (Fisher, 1927, p. xiii). One of the few indexes that satisfied the three criteria was the geometric mean of the Laspeyres and Paasche indexes, which came to be known as “Fisher’s ideal index”. Although Fisher’s name is not usually associated with the stochastic approach to index numbers, a rereading of his book reveals some
early contributions that are at least related to the approach, and can be thought of as providing some early clues and directions. This is perhaps not surprising given the breadth and depth of this influential book, which still commands a leading position in an area that has expanded enormously since Fisher’s time. In what follows, these early contributions are set out.

As discussed in the previous section, the stochastic approach emphasises the idea of index numbers as averages of the underlying component prices. Fisher also made this emphasis, as is clear from the following (Fisher, 1927, p. 2):

> There would be no difficulty … if all prices moved up in perfect unison or down in perfect unison. But since, in actual fact, the prices of different articles move very differently, we must employ some sort of compromise or average of their divergent movements.

If we look at prices as starting at any time from the same point, they seem to scatter or disperse like the fragments of a bursting shell. But, just as there is a definite centre of gravity of the shell fragments, as they move, so is there a definite average movement of the scattering prices. This average is the “index number.” Moreover, just as the center of gravity is often convenient to use in physics instead of a list of the individual shell fragments, so the average of the price movements, called their index number, is often convenient to use in economics.

An index number of prices, then, shows the average percentage change of prices from one point of time to another. The percentage change in the price of a single commodity from one time to another is, of course, found by dividing its price at the second time by its price at the first time. The ratio between these two prices is called the price relative of that one particular commodity in relation to those two particular times. An index number of the prices of a number of commodities is an average of their price relatives. (Fisher’s emphasis.)

The idea of using an index number to summarise the disparate movements in individual prices is also implicit in many of Fisher’s diagrams. Figure 1 provides an example in the form of a time-series plot of the 36 prices used by Fisher to illustrate the workings of 100+ types of indexes later in his book. In commenting on Figure 1, and its quantity counterpart, Fisher (1927, p. 14) in fact uses the language of the stochastic approach whereby the estimated rate of inflation (the change in the price index) is referred to as the “common trend” in prices:

> How it is possible to find a common trend for such widely scattered price relatives or quantity relatives? Will not there be as many answers to such a question as there are methods of calculation? Will not these answers vary among themselves 50 percent or 100 percent? The present investigation will show how mistaken is such a first impression.
The term "bias" is used in statistics to refer to the difference between the expected value of a sample statistic and its population counterpart. Fisher, however, uses the term to refer to something different: The deviation of a given index from the time- and factor– reversal criteria. Fisher (1927, pp. 108-9, 387-95) shows that his bias increases with the degree of dispersion of the underlying prices and quantities. Interestingly, this is reminiscent of one of the basic results of the stochastic approach whereby the uncertainty of the estimated rate of inflation is proportional to the standard deviation of relative prices; see equation (2.8) above, for example. Thus under the stochastic approach the overall rate of inflation is estimated less precisely in those periods when there is high variability of relative prices.
A further link with the stochastic approach is what Fisher calls the “probable error” involved in an index number. This seems to be Fisher’s response to the then-current criticism of index numbers as being unreliable. A flavour of this criticism is given in the opening paragraph of his book (Fisher, 1927, p. 1):

In 1896, in the Economic Journal, the Dutch economist, N. G. Pierson, after pointing out some apparently absurd results of index numbers, said: “The only possible conclusion seems to be that all attempts to calculate and represent average movements of prices, either by index numbers or otherwise, ought to be abandoned.” No economist would today express such an extreme view. And yet there lingers a doubt as to the accuracy and reliability of index numbers as a means of measuring price movement.

The term “probable error” is related to the sampling distribution of the mean, which can be explained as follows. Suppose a random variable $x$ is normally distributed with mean $\mu$ and standard deviation $\sigma$. The probability of drawing a value of $x$ that falls within the range $\mu \pm .6745 \times \sigma$ is then 50 percent. The quantity $.6745 \times \sigma$ is known as the probable error in measuring $x$; in other words, under normality, there is a chance of 1 in 2 of $x$ being $.6745$ standard deviations away from the mean. Next, we interpret $x$ as the sample mean of $n$ underlying observations $x_1, \ldots, x_n$ with standard deviation $s = \sqrt{\frac{1}{n-1} \sum_{i=1}^{n} (x_i - \bar{x})^2}$. Then the standard error of the mean is $s/\sqrt{n}$, which is an estimator of $\sigma$ above. Accordingly, the probable error of the mean is $.6745 \times s/\sqrt{n}$.

Fisher (1927, pp. 225-29) used his $n=13$ most preferred indexes for the period 1914-18 to compute their probable errors. Take as an example 1917, the year in which the probable error of the price indexes is largest at 0.128 percent. This error is to be compared to the value of the ideal price index (identified by Fisher as “Formula Number 353”) of 161.56, so that on average prices in 1917 were 61.6 percent higher than in 1913. The 50-percent confidence interval for this year is then 161.56±0.128 percent, or 161.35-161.77. Fisher (p. 228) was obviously excited by this incredibly low error in declaring “[w]e may, therefore, be assured that Formula 353… is able to correctly to measure the general trend in 36 dispersing relative prices… within less than one eighth of one percent!…” Fisher (p. 229) concludes triumphantly by writing: “As physicists or astronomers would say, the ‘instrumental error’ negligible. The old idea that among the difficulties in measuring price movements is the difficulty of finding a trustworthy mathematical method may now be dismissed once and for all.”

A skeptical assessment of the above approach is that the small error simply reflects the closeness of the 13 indexes considered. These indexes are all computed from exactly the same underlying data and, it could be argued, the differences in the algebraic forms of the indexes are
relatively minor as they all involve a geometric mean of one form or another. Indeed, in at least two cases, the index is defined as the geometric mean of two other indexes included in the list of 13. Nevertheless, it is clear that Fisher did have considerable early insights into the nature of the uncertainties associated with index numbers, a topic that is central to the stochastic approach. As mentioned above, in referring to the index value as the “common trend” in prices, Fisher himself thought of the index as the mean of the underlying price relatives. Accordingly, had Fisher applied probable error theory to the index itself, rather than the mean of the 13 different indexes, he would then have identified the estimation uncertainty of the index as reflecting the degree of relative price variability, and thus possibly been a major contributor to the development of the stochastic approach.

Diewert (1995) has also commented on Fisher’s attempt to assess the precision of a price index. He points out that (Diewert, 1995, p. 22-23)

…the proponents of the test and economic approaches to index number theory use their favorite index number formula and thus provide a precise answer whether the price relatives are widely dispersed or not. Thus the test and economic approaches give a false sense of precision.

The early pioneers of the test approach addressed the above criticism. Their method works as follows: (i) decide on a list of desirable tests that an index number formula should satisfy; (ii) find some specific formulae that satisfy these tests (if possible); (iii) evaluate the chosen formulae with the data on hand and (iv) table some measure of the dispersion of the resulting index number computations (usually the range or standard deviation was chosen). The resulting measure of dispersion can be regarded as a measure of functional form error.

Diewert then goes on to describe Fisher’s application of this approach, as discussed above. Diewert also cites the work of Persons (1928) and Walsh (1921) who apply similar methods to assessing the reliability of price indices. Diewert (1995, p. 24) is not satisfied with the above approach and writes:

It is clear that there are some problems in implementing the above test approach to the determination of functional form error; i.e., what tests should we use and how many index number formulae should be evaluated in order to calculate the measure of dispersion? However, it is interesting to note that virtually all the above index number formulae suggested by Fisher, Persons and Walsh approximate each other to the second order around an equal price and quantity point.

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5 In Fisher’s Table 26 (p. 226) Formula Number 5307 is the geometric of Formulae 307 and 309; and $5323 = \sqrt{323 \times 325}$.

6 According to Stigler (1982), Jevons (1884, p. 157) attempted such an application but he “seems to have put only a rough faith in that result and did not repeat the attempt” (Stigler, 1982).
Note that the criticism implied in the last sentence of this quotation is another way of stating that the various formulae being compared have similar structure, which is almost the same as our point above that Fisher compared 13 geometric means of one form or another. In the subsequent section we shall return to Diewert (1995) and discuss his criticism of the stochastic approach in detail.

Model (2.9) draws a sharp distinction between a change in the price level \((\alpha_i)\) and a change in a relative price \((\beta_i)\). As they “wash out” in the aggregate \((\Sigma w_i \beta_i = 0)\), relative price changes are not inflationary. Fisher also advocated such a distinction as a way to avoid circular reasoning and to enhance clear thinking about the causes of inflation, as is illustrated by the following quotation (Fisher, 1920, p. 73):

> It is true that individual prices do react on one another in thousands of ways. But the several pushes and pulls among individual prices are not what raise them as a group. Such forces within the group could not move the group itself any more than a man can raise himself from the ground by tugging at his boot-straps. We cannot explain the rise or fall of a raft on the ocean by observing how one log in the raft is linked to the others and is pulled up or down by them. It is true that some prices rise more promptly than others and give the proximate reason for raising the others. The whole raft of prices is bound together and its parts creak and groan to make the needed adjustments. But such readjustments between prices do not explain why the whole raft of prices has risen. (Fisher’s emphasis.)

4. **The Diewert Critique**

The stochastic approach has attracted the attention of Erwin Diewert, a leading expert in index numbers. In a major review paper, Diewert (1995) places the SA in historical context in a masterful fashion and cites the early work on the topic by Bowley, Edgeworth and Jevons in particular. He then goes on to make four specific criticisms of the SA in its modern version; below we discuss each in turn.\(^7\)

**Criticism 1: The Variance Assumptions are not Consistent with the Facts**

We return to the basic model given by equation (2.1) and note that the error term \(\varepsilon_{it}\) is interpreted as the change in the \(i^{th}\) relative price. Equation (2.5) postulates that the variances of relative prices are inversely proportional to the corresponding budget shares. Accordingly, the prices of those commodities that are more (less) important in the consumer’s budget are less (more) variable. Diewert argues that this assumption does not agree with the observed behaviour of prices. In support of his position, Diewert cites the evidence presented in Clements and Izan (1987, p. 345) who conceded the point. Diewert (1995, pp.15-16) also argues:

\(^7\) For an earlier response to Diewert’s criticisms, see Selvanathan and Prasada Rao (1999).
...[F]ormal statistical tests are not required to support the common observation that the food and energy components of the consumer price index are more volatile than many of the remaining components. Food has a big share while energy has a small share... volatility of price components is simply not highly correlated with the corresponding expenditure shares.

While the arguments presented below equation (2.5) supporting the idea that the variance specification (2.5) is not totally implausible, as indicated above the results of Clements and Izan (1987) reject this specification. Such a rejection does not mean that the entire SA has to be abandoned however, as this particular variance specification is only one of a multitude of possibilities. That is to say, variance specification (2.5) is just one way of parametrising the \( \Sigma_t \) in the general expression for the variability of the index, equation (2.4). Put slightly differently, result (2.8) is a case of the general result (2.4); the special case is based on assumption (2.5)

To illustrate how the SA is able to deal with difference specifications of \( \Sigma_t \), consider three other special cases. First, suppose the \( n \) prices are independent so that \( \Sigma_t \) is a diagonal matrix with \( \sigma_{11}, \ldots, \sigma_{nn} \) on the main diagonal. To set out the implications of this case, let \( x_{it} = \sigma_{it}^{-1} / S_t \), where \( S_t = \sum_{i=1}^{n} \sigma_{it}^{-1} \). Application of equations (2.3) and (2.4) then yields

\[
\hat{\alpha}_t = \sum_{i=1}^{n} x_{it} Dp_t, \quad \text{var} \hat{\alpha}_t = S_t^{-1}.
\]

Here we see that the estimated rate of inflation is still a weighted average of the \( n \) price changes, but now the weights are \( x_{it} \), which are proportional to the reciprocals of the variances of the respective relative prices \( \sigma_{it}^{-1} \). By construction, the weights \( x_{it} \) are all positive fractions and have a unit sum. Accordingly, more weight in the price index is accorded to lower variance prices, which is a sensible property. The second member of equation (4.1) reveals that the variance of the price index equals the inverse of \( S_t \), the sum of the reciprocals of the \( n \) variances \( \sigma_{11}, \ldots, \sigma_{nn} \). As the term \( S_t \) is an inverse measure of noise in the system, it follows that when there is more
variability in relative prices, $S_t^{-1}$ is larger, and the price index is estimated with less precision. In other words, at times when prices move in a highly disproportionate manner, the overall rate of inflation is less well defined conceptually and this is reflected in the higher estimation uncertainty in its measurement. Again, this is a sensible property. The above example is Diewert’s (1995) neo-Edgeworthian model, with some minor modifications.

As a second example, suppose that at time $t$ relative prices have a common variance $\sigma_t^2$ and a common correlation coefficient $\rho_t$, so that the covariance matrix now takes the form $\Sigma_t = \sigma_t^2 [(1-\rho_t)I + \rho_t u']$, where $I$ is an identity matrix and $u$ is an $n$-vector of unit elements. Application of equations (2.3) and (2.4), as before, yields

$$\hat{\alpha}_t = \frac{1}{n} \sum_{i=1}^{n} D p_i, \quad \text{var} \hat{\alpha}_t = \sigma_t^2 \left[ \rho_t + \frac{1-\rho_t}{n} \right].$$

Result (4.2) shows that in the equicorrelated case, the estimated rate of inflation is an unweighted average of the price changes, while its variance is increasing in the correlation $\rho_t$. If prices are independent $\rho_t = 0$ and $\text{var} \hat{\alpha}_t = \sigma_t^2/n$, while if they are perfectly correlated $\text{var} \hat{\alpha}_t = \sigma_t^2$.

The final example is a mixture of the two specifications of $\Sigma_t$ considered above. For ease of notation, we drop the $t$ subscript from $\Sigma = [\sigma_{ij}]$, and write it as $\Sigma = D (I + \lambda) D$, where $D$ is a diagonal matrix with the standard deviation of the $n$ prices on the matrix diagonal, $\sigma_{11}^{1/2}, \ldots, \sigma_{nn}^{1/2}$; and $\lambda = [\lambda_{ij}]$ is an $n \times n$ symmetric matrix with diagonal elements zero and with $(i,j)$th off-diagonal element the relevant correlation, $\lambda_{ij} = \sigma_{ij}/\sqrt{\sigma_{ii} \sigma_{jj}}$. Recall the result that $(I - \lambda)^{-1} = I + \lambda + \lambda^2 + \ldots \approx I + \lambda$, if the elements of $\lambda$ are not “too large”. The approximation $(I - \lambda)^{-1} \approx I + \lambda$ implies that $(I + \lambda)^{-1} \approx (I - \lambda)$, so that

$$\Sigma^{-1} \approx D^{-1} (I - \lambda) D^{-1}.$$
This inverse has the useful property that it can be expressed as the sum of two parts, (i) a variance component, \( \mathbf{D}^{-1} \mathbf{D}^{-1} = [\sigma_{ii}^{-1}] \); and (ii) a component related to the covariances,
\[-\mathbf{D}^{-1} \mathbf{\lambda} \mathbf{D}^{-1} = -\left[ \frac{1}{2} \sigma_{ii}^{-1/2} \sigma_{ji}^{-1/2} \right] = -\left[ \sigma_{ij} / \sigma_{ii} \sigma_{jj} \right],
\] which is a measure of the lack of independence among the \( n \) prices.

Define \( \mathbf{\lambda}^* \) as the \( n \times n \) matrix \( \mathbf{D}^{-1} \mathbf{\lambda} \mathbf{D}^{-1} \), with (i, j)th element \( \lambda_{ij}^* \), and \( \lambda_{ij}^* = \sum_{i=1}^{n} \lambda_{ij}^* \) as the sum of the elements in the \( j \)th column of \( \mathbf{\lambda}^* \). Now consider the fraction
\[(4.4) \quad y_i = \frac{\frac{\sigma_{ii}^{-1} - \lambda_{ii}^*}{\sum_{j=1}^{n} (\sigma_{jj}^{-1} - \lambda_{jj}^*)}},
\]
which satisfies \( \sum_{i=1}^{n} y_i = 1 \). The fraction \( y_i \) is larger when (i) the \( i \)th variance is lower and (ii) the \( i \)th column sum is lower, which will be the case when the \( i \)th relative price is less correlated with the others. It is well known from portfolio theory that an asset whose return is highly correlated with the other assets will, cet. par., not receive a large weight in an efficiently diversified portfolio as it tends to just “duplicate” the others. In other words, there is little point in holding an asset that is a linear combination of others. The fraction (4.4) possesses a similar property: If we consider \( y_i \) as a function of \( \lambda_{ii}^* \), \( y_i(\lambda_{ii}^*) \), when the \( i \)th relative price is independent of the other prices, \( \lambda_{ii}^* = 0 \) and \( y_i(0) \propto \sigma_{ii}^{-1} \); this is the case in the first example above. Then as the \( i \)th price becomes more and more correlated with the others, the fraction \( y_i \) falls.

If we replace \( \Sigma_{i}^{-1} \) in equations (2.3) and (2.4) and use (4.3) we obtain (see Appendix for details)
\[(4.5) \quad \hat{a}_i = \sum_{i=1}^{n} y_i \mathbf{D} \mathbf{p}_i, \quad \text{var} \hat{a}_i = \frac{1}{\sum_{i=1}^{n} (\sigma_{ii}^{-1} - \lambda_{ii}^*)}.
\]
In words, the estimated rate of inflation is again a weighted average of the \( n \) price changes. But now the weights are \( y_i \) which are related to the variances and covariances in a manner discussed above. The variance of the inflation rate is now inversely related to the amount of independent noise in the system.

Other possible specifications of the covariance matrix are clearly possible. For example, we could merge these above two examples into one by having different prices having different variances, while at the same time being correlated. Or, following Crompton (2000), we could
simply let $\Sigma_i$ evolve in a fairly arbitrary way and apply White’s (1980) heteroskedastic-consistent approach to the prices, possibly after weighting. The key point is that the precise specification of $\Sigma_i$ is not the fundamental aspect of the SA. While the form that $\Sigma_i$ takes obviously affects the results, still the key idea is to think of the rate of inflation as the underlying common trend in prices and to estimate the trend by some type of mean of the $n$ price changes.

**Criticism 2: The Budget Shares Serve Two Distinct Purposes**

We return to the model (2.9) which allows for sustained changes in relative prices. In this model, the commodity-specific parameters $\beta_i$ are identified by the following normalisation:

$$\sum_{i=1}^{n} w_i \beta_i = 0 .$$

As $\beta_i$ is interpreted as the expected change in the relative price of good $i$, rule (4.6) states that a weighted average of such relative price changes is zero. By their very nature, changes in relative prices must “wash out” when we consider all $n$ commodities simultaneously in the sense that not all relative prices can increase, nor can they all decrease. A relative price involves the comparison of the nominal price of the good in question with some form of an index of all $n$ prices, a comparison which takes the form of the difference between the price and the index when we use logarithms. As the index is a logarithmic mean of the $n$ prices, the relative price is just like the deviation of the nominal price from its mean. The sum of such deviations from the unweighted mean is zero, while the weighted sum of the deviations from the weighted mean is zero (when the two sets of weights coincide). As the price index (2.7) is a budget-share-weighted mean, it can be seen that the normalisation rule (4.6) is entirely natural. Despite its attractive interpretation, it should nevertheless be acknowledge that other normalisations are possible.

It can be seen that the budget shares $w_i$ play a role in two places, (i) the normalisation (4.6) and (ii) the variance specification (2.5) which also applies to the extended model (2.9). Diewert objects to $w_i$ playing these two roles simultaneously. To understand clearly the basis for this objection, we need to introduce the mean of the $i$th budget share in the $T$ periods of the sample, $w_{iT}, \ldots, w_{iT}$:

$$\bar{w}_i = \frac{1}{T} \sum_{t=1}^{T} w_{it} .$$
It is the means of the budget shares that are used in the empirical implementation of the model. Making the appropriate changes in equation numbers and notation, Diewert’s (1995, p. 12, 15, 19 and 20) criticism is contained in the following three quotations:

The restriction (4.6) says that a share weighted average of the specific commodity price trends $\beta_i$ sums to zero, a very reasonable assumption since the parameter $\alpha_i$ contains the general period $t$ trend. What is not so reasonable, however, is the assumption that the $w_i$ which appear in (4.6) are the same as the $w_i$ which appear in (2.5).

...[T]he $w_i$ defined by (4.7) depend on the prices $p_{it}$ and hence the “fixed” weights $\overline{w_i}$ which appear in (2.5) and (4.6) are not really independent of the price relatives $p_{it}/p_{it-1}$. Hence the applicability of model (2.9) when the $w_i$ are defined by (4.7) is in doubt.

...[T]hese models forces the same weights $w_i$ to serve two distinct purposes and it is unlikely that their weights could be correct for both purposes. In particular, their expenditure-based weights are unlikely to be correct for the first purpose [the variance specification (2.5)] (which is criticism 1 again).

Diewert describes as “very reasonable” normalisation (4.6), which is the one of the two places that $w_i$ appears. Accordingly, his Criticism 2 is really an objection to the use of $w_i$ in the variance specification (2.5), which is just his Criticism 1. This partial duplication of these two criticisms is made explicit by Diewert in the last sentence of the last quotation given above. To the extent that Criticism 2 coincides with Criticism 1, we have nothing more to say in response in addition to our response to Criticism 1 given above. However, the second of the three quotations above deals with something different, the dependence of the mean budget shares $\overline{w_i}$ on the price relative $p_{it}/p_{it-1}$, or the log-change $Dp_{it}$. In most countries, budget shares tend to change quite modestly over time, so we feel that treating them as constants for this purpose would not be a major problem in a time-series context. But going across countries, budget shares changes dramatically; food, for example, accounts for substantially less than 10 percent of the budget in the richest countries, while it absorbs more than 50 percent in the poorest. An alternative specification that avoids the problem is

$$nw_{it}Dp_{it} = \alpha_i + \beta_i + \varepsilon_{it},$$
with the new normalisation $\sum_{i=1}^{n} \beta_i = 0$. Now the budget shares only appear on the right-hand side of equation (4.8). If, for the purpose of illustration, we assume that the disturbances in model (4.8), $\varepsilon_{it}$, have a common variance, the least-squares estimator of the inflation parameter $\alpha_i$ is $\sum_{i=1}^{n} w_{it} Dp_n$, the same as that given in equation (2.7).\textsuperscript{11}

**Criticism 3: Stochastic Index Numbers Age**

In most countries the CPI takes the Laspeyres form. After its release such an index does not change with the passage of time as new information becomes available on subsequent values of prices and expenditure patterns. Accordingly, as the Laspeyres index is fixed for all time, we could say that it possesses an “ageless” property. Many other popular index numbers also share this property. It should be noted that aging refers to the effect on the index value of the receipt of new data that become available with the passage of time, and not to the impact of pre-existing data being subsequently revised.

Consider the regression equation $y_t = \alpha + \beta x_t + \varepsilon_t$, $t = 1, ..., T$. As the least-squares estimates of the coefficients $\alpha$ and $\beta$ depend on all of the $T$ observations, they will take different values when we obtain an additional data point and use the $T+1$ observations. As stochastic index numbers can be cast as regression coefficients, when additional data become available in the future and we re-estimate with the expanded data set, in general past index values will change. Thus in general stochastic index numbers are subject to aging.\textsuperscript{12} The aging process associated with the stochastic approach is Diewert’s third criticism. In his words (Diewert, 1995, p. 20):

…[Stochastic] price indexes are not invariant to the number of periods $T$ in the sample.

Referring to Balk (1980), Diewert notes that this could be a problem of practical importance for statistical agencies.

Diewert is completely correct in noting the aging problem. But just how significant is this problem? Although as mentioned above, aging and data revisions are conceptually distinct, there is a sense in which they are similar. Quarterly national account data are notorious for their revisions: Although it would be very unusual for a recession (two consecutive quarters of negative growth in

\textsuperscript{11} Model (4.8) is related to the work of Voltaire and Stack (1980).

\textsuperscript{12} The reason for including the “in general” qualifier is that it is possible to devise simple cases in which stochastic indices do not age. For example, model (2.2) yields as the estimated rate of inflation $\sum_{i=1}^{n} w_{it} Dp_n$, as indicated by equation (2.7), and this expression is ageless in the above sense. Of course, the agelessness of equation (2.7) disappears if we replace the observed budget shares $w_{it}$ with their sample means $\bar{w}_i$ as these change as more data accumulate.
real GDP) to be subsequently revised away, such revisions can still be nontrivial. Users of national accounts data seem to have learnt to live with this problem, and there is little evidence of any lack of demand for the data – if anything, demand has intensified from financial markets, government users and business economists. The high demand for these imperfect data is clear from the first part of the Abstract of a paper from researchers at the Reserve Bank of Australia (Stone and Wardrop, 2002):

Quarterly national accounts data are amongst the most important and eagerly awaited economics information available, with estimates of recent growth regarded as a key summary indicator of the current health of the Australian economy. Official estimates of quarterly output are, however, subject to uncertainty and subsequent revision. Hence, the official estimates of quarterly national account aggregates, with which policy-makers must work, may in practice be an inaccurate guide to their ‘true’ values, not just initially but even for some time after the event.

The problem of aging associated with the stochastic approach is also analogous to the use in economic policy of any concept that is not directly observable, but can be estimated with data under certain assumptions. Examples included the natural rate of unemployment (or the output gap), the underlying rate of inflation and the equilibrium exchange rate. These three concepts have proven to be valuable policy tools, but as they are all derived from econometric estimates of one form or another, they are subject to aging in exactly the same manner as stochastic index numbers. If we have been able to live with the measurement problems surrounding the natural rate etc., then perhaps the same principle can apply to the stochastic approach.

As a final response to the criticism of aging, we would emphasise that the seriousness of this problem is to be compared with the benefit that of stochastic index numbers bring. In using information on the dispersion of relative prices, stochastic index numbers come with measures of estimation uncertainty. No other index numbers provide these measures.

Criticism 4: Relative Price Changes are Not Accounted for and All Prices are Given the Same Weight.

This is Keynes’ criticism to the very early work on stochastic price indexes of Jevons and Edgeworth which involved an unweighted geometric mean of the n price relatives, 

$$\left[ \prod_{i=1}^{n} \left( \frac{p_i}{p_{i0}} \right) \right]^{1/n} .$$

In making this criticism, Diewert (1995, p. 21) quotes Keynes (1930, p. 30):

The hypothetical change in the price level, which would have occurred if there had been no changes in relative prices, is no longer relevant if relative prices have in fact changed – for the change in relative prices has in itself affected the price level.
I conclude, therefore, that the unweighted (or rather the randomly weighted) index number of prices – Edgeworth’s ‘indefinite’ index number -- ...has no place whatever in a rightly conceived discussion of the problems of price levels.

Diewert (1995, p. 21) then adds:

Criticism four can be restated as follows. The early statistical approaches of Jevons and Edgeworth ... treated each price relative as an equally valid signal of the general inflation rate: the price relative for pepper is given by the same weight as the price relative for bread. This does not seem to be reasonable to “Keynesians” if the quantity of pepper consumed is negligible.

Such a criticism is clearly valid in the context of the above unweighted geometric mean. Diewert acknowledges, however, that the problem is addressed, at least in part, by the newer versions of stochastic index numbers that (i) introduce commodity weighting and (ii) allow for systematic changes in relative prices. In Diewert’s (1995, p. 21) view, model (2.9), with two modifications, deals adequately with Criticism 4. First, the variance assumption (2.5) needs to be made more reasonable, which is Criticism 1 above. Second, the constant commodity parameters $\beta_i$ should be replaced by a set of period-specific parameters $\beta_{it}$. While Diewert notes that this would result in too many parameters to be identified, he offers no suggestions how to proceed.

With the exception of the two modification noted in the above paragraph, Criticism 4 does not apply to our work.

Summary

We acknowledge the above criticisms, especially the first and second, as being constructive and provocative. We accept Criticism 1 about the variance assumption and we indicated above how to proceed with more palatable alternatives. Criticism 2 deals with the budget shares serving two purposes. In part, this criticism overlaps with the first. To answer the non-overlapping part of Criticism 2 we introduced a new way of formulating the basic model of the stochastic approach. The third criticism, that stochastic index numbers are subject to aging, is true and we are unable to offer any modifications to the approach that avoid this problem. We argued that like aging of human beings, it is something that has to be lived with (as in the old adage, “if it can’t be cured, it has to be endured”), and that the problem is not confined to the stochastic approach. Finally, as Criticism 4 does not apply to our work, we have nothing to respond to on this count.

While Diewert has been a leading supporter of other approaches to index-number theory, the tone of his comments indicate that he is certainly not hostile to the stochastic approach. For example, in the closing part of his paper Diewert (1995, p. 30) writes:
...[P]erhaps this diversity is a good thing. The new stochastic approach to index numbers has at least caused this proponent of the test and economic approaches to think more deeply about the foundations of the subject.

As Diewert (2002, 2004) has subsequently written papers that significantly extend the stochastic approach, it seems that he has more recently adopted an even more positive attitude to the approach, as is confirmed by the following exchange of correspondence. In an email to Diewert, Clements (2003) wrote:

I have just read your very interesting paper "Weighted Country Product Dummy Variable Regressions and Index Number Formulae" [Diewert, 2002]. As you point out, it is closely related to stochastic index numbers, and as an advocate of that approach, I was pleased to see your support for the approach.

The next day, Diewert (2003) responded as follows:

Yes, it is a bit surprising that I have moved into the ranks of the stochastic index number fans! Of course, I really liked Theil's explanation for the Törnqvist Theil index using weighting. I have mostly been critical of unweighted stochastic approaches. I guess my current line of research is to pursue alternative weighting schemes to see if I can generate traditional index number formulae so in a sense, it is not all that different from what you, Prasada and Se[...]vanathan have been doing, but I have been getting my heteroskedastic variances using weighting and representativity in the marketplace arguments rather than assumptions about the variance of error terms... but in the end, there is not a lot of difference in the approaches.

In the next section we present Theil’s explanation mentioned by Diewert.

5. Theil’s Approach

In the above discussion, it is assumed that the log-change in the $i^{th}$ price is stochastic; in the simplest case of model (2.1) under normality and assumption (2.5), each price change has the same mean, $Dp_i \sim N(\alpha_i, \sigma_i^2)$, with $\sigma_i = \lambda_i^2 / w_i$. While this type of parametric formulation leads to the standard error and confidence intervals of the index, it is clear from the above discussion of the first element of the Diewert Critique that there is considerable scope for views to differ about the appropriateness of this specification. A different stochastic approach, due to Theil (1967), avoids the problem by proceeding along the lines of descriptive statistics. This section sets out this approach.

Theil (1967, p. 135) describes the problem to be considered as follows:

Suppose we have price and quantity data for the individual commodities in two different regions or in two different periods; can we then argue in any

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13 We thank Erwin Diewert for granting permission to quote from this correspondence.
meaningful way that the price level in the second region (or period) is, say, 10 percent higher than that in the first? Can we do the same thing for the quantities? Note that both questions refer to averages: The prices in the second region (or period) exceed those of the first, on the average, by a certain percentage (similarly for the quantities). We may also be interested in the variation of individual price and quantity ratios around such averages. For some commodities the price in the first region may exceed the price in the second by very much more than the average indicates; for other commodities the converse may be true. A Ford is cheaper in the United States compared with its price in India, but household assistance is comparatively much cheaper in India. Such dispersion problems will also be considered in this chapter. (Theil’s emphasis.)

The use of explicit statistical language in this quotation is to be noted, as is the very title of the famous chapter from which it comes, viz., “A Statistical Approach to the Problem of Price and Quantity Comparisons”. A reinforcement of this same point is provided later in the chapter when Theil (1967, p. 158) states without apology that “the approach of this chapter has its roots in statistics rather than economics”.

Rather than considering the evolution of prices over time, we follow Theil and now move to prices in different countries, so that the objective is to measure the price level in one country relative to that in another. We write $p_{ic}, q_{ic}$ for the price and quantity consumed of good $i$ ($i = 1,...,n$) in country $c$ ($c = 1,2$), and $w_{ic} = p_{ic} q_{ic} / M_c$ for corresponding budget share, with $M_c = \sum_{i=1}^n p_{ic} q_{ic}$ total expenditure in country $c$. An attractive way of measuring the level of prices in country 2 relative to 1 is via the Törnqvist (1936)-Theil (1967) index

$$\log P_{21} = \sum_{i=1}^n w_{i21} \log \left( \frac{p_{i2}}{p_{i1}} \right),$$

where $w_{i21} = (1/2)(w_{i2} + w_{i1})$ is the arithmetic average of the budget share of good $i$ countries 1 and 2. Index (5.1) is a weighted average of the logarithmic relative prices, where the weights are $w_{i21}$. The weight $w_{i21}$ can be viewed as the budget share of good $i$ that pertains in a third “neutral” country located mid way between 1 and 2. This third country is neutral with respect to 1 and 2 as its consumption basket, as measured by the budget shares $w_{i21}$, is an unweighted average of that in the other two countries; in other words, 1 and 2 are both equally represented in $w_{i21}$, a property that has democratic attributes. But there is an even more compelling reason to use neutral country weights, rather than those for either country 1 or 2, $w_{i1}, w_{i2}$. This choice ensures that the index is symmetric in 1 and 2, so that if for example, the price level in country 2 is a multiple 1.2 of that in
country 1, then the price level in 1 relative to 2 will be $1/1.2$, as can be seen from equation (5.1) if we interchange the 2 and 1 country subscripts. Thus index (5.1) satisfies Fisher’s time-reversal test in a cross-country context.

Theil (1967, pp. 136-37) provides an ingenious justification for index (5.1) along the following sampling lines. For convenience, write the relative price $\log(p_{12}/p_{11})$ as $r_{121}$, and consider a discrete random variable $R_{21}$ which can take the values $r_{121},\ldots,r_{n21}$. To derive the probabilities attached to these $n$ possible realisations, suppose that prices are drawn at random from this distribution such that the each dollar of expenditure in the neutral country has an equal chance of being selected. This means that the probability of drawing $r_{121}$ is $w_{121}$, which is nonnegative and possesses a unit sum. Accordingly, the expected value of $R_{21}$ is $E(R_{21}) = \sum_{i=1}^{n} w_{i21} \times r_{i21}$, which coincides with index (5.1). This sampling framework thus shows that the Törnqvist-Theil index has the interpretation as the expected value of the distribution of logarithmic relative prices.

As emphasised by Diewert (2004), Theil’s approach is appealing as it does not require any assumptions about the stochastic nature of the individual prices, nor any distributional assumptions. It can thus be considered as a nonparametric stochastic approach. In Diewert’s (2004, pgs. 23, 26) words:

> Theil’s stochastic approach is a nice one: the logarithm of the price index is simply the mean of the discrete probability distribution of the log price ratios and it is not necessary to make any assumptions about the exact distribution of the error terms. (Diewert’s emphasis.)

> The main advantage of [Theil’s] approach is that it is completely nonparametric; i.e., we do not have to make problematical assumptions as to what the “true” distribution of log price relatives is: the distribution is simply the empirical population distribution and we take the mean of this (weighted) log price distribution as our preferred summary measure of this distribution…

For an extension of Theil’s bilateral approach to the multilateral case, see Diewert (2004).

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14 Theil (1967) also measures the dispersion of the distribution of prices by the corresponding weighted variance $\Sigma_{i=1}^{n} w_{i21} \left( \log(p_{i2}/p_{i1}) - \log P_{21} \right)^2$, which is the cross-country version of what was referred to as the “Divisia variance” in Section 2 above. It is clear that the larger is this variance, the less reliable will be the index (5.1), but this idea was not formally developed by Theil. Theil (1967, p. 155) refers to $\log P_{21}$ defined by (5.1) as the “price level”, while he describes the weighted variance as a measure of the differences in “price structure”.
The Weighted Country-Product-Dummy Model

In a recent paper, Diewert (2002) contributes to stochastic index number theory by extending Summers’ (1973) country-product-dummy (CPD) methodology, which was originally formulated for making international comparisons of prices within the hedonic regression framework whereby the only characteristic of the commodity is the commodity itself. Diewert modifies this approach by introducing weights, thereby showing that some well-known index numbers emerge when the weights are chosen appropriately. Diewert (2002, p. 9) describes this unification of the CPD approach and more familiar methods from index-number theory as follows:

At first glance, it seems that the Country Product Dummy method… for comparing prices between countries (or time periods) is totally unrelated to traditional index number methods for making comparisons between countries or time periods. However, … if the unweighted Country Product Dummy … regressions are replaced with suitable weighted counterparts, then the resulting measures of prices changes are very closely related to traditional bilateral index number formulae. (Diewert’s emphasis.)

In what follows, we first summarise Diewert’s work and provide some additional interpretative material, and then discuss some additional aspects of his results.

Diewert’s Results

Let \( p_{ic} \) be the price of commodity \( i \) (\( i = 1, \ldots, n \)) in country \( c \) (\( c = 1, \ldots, C \)). Consider a logarithmic decomposition of this price:

\[
(6.1) \quad \log p_{ic} = \lambda_c + \mu_i + \varepsilon_{ic},
\]

where \( \lambda_c \) and \( \mu_i \) are country- and commodity-specific components of the price. The term \( \varepsilon_{ic} \) is an independently-distributed stochastic error with zero mean and variance \( \sigma^2 / a_{ic}^2 \), where the term \( a_{ic} \) is a function (to be specified) relating to good \( i \) in country \( c \). Several features of model (6.1) should be mentioned. First, if we adopt the normalisation \( \lambda_1 = 0 \), \( \lambda_c \) can then be interpreted as the expected value of the price level in country \( c \) relative to that in 1; that is, \( \lambda_c = E[\log(p_{ic}/p_{i1})] \), \( i = 1, \ldots, n \). To interpret this parameter further, suppose that for some country \( c \), initially all prices are the same as those in country 1, so that \( p_{ic} = p_{i1}, i = 1, \ldots, n \). In this

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15 Diewert (2002) states that part of his paper “can be viewed as a specialization of [Prasada] Rao (2002) to the two-country case”.

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situation, clearly the price levels in the two countries coincide, so that \( \lambda_c = 0 \). Next, suppose that the currency of \( c \) is redenominated such that each new currency unit is now worth two old currency units, and that all prices, expressed in terms of the new currency, fall by one half, so that the structure of relative prices remains unchanged. This means that the price level has fallen by 50 percent and \( \lambda_c \) now takes the value \(-\log 2 \approx -0.69\), which again illustrates how this parameter measures the price level in \( c \). Second, consider the role of the parameter \( \mu_i \). Note that 
\[
\log p_{ic} - \lambda_c = \log \left[ \frac{p_{ic}}{\exp(\lambda_c)} \right]
\]
is the logarithm of the price of \( i \) in country \( c \) deflated by the price level in that country. This means that we can write 
\[
\mu_i = E \{ \log \left[ \frac{p_{ic}}{\exp(\lambda_c)} \right] \},
\]
which shows that \( \mu_i \) is the expected value of this deflated price, with this expectation being the same for all \( C \) countries. Accordingly, we could call \( \mu_i \) the “relative price parameter for good \( i \)”. A third aspect of model (6.1) is that if the \( \varepsilon_{ic} \) are normal, then the prices are distributed lognormally, which is reasonable and ensures that they are always positive. Finally, a useful interpretation of the error in model (6.1), due to Prasada Rao (2004), follows from expressing it as
\[
\varepsilon_{ic} = \log \left[ \frac{p_{ic}/\exp(\lambda_c)}{\exp(\mu_i)} \right].
\]

As mentioned above, the term \( p_{ic}/\exp(\lambda_c) \) is the price of commodity \( i \) in country \( c \) deflated by that country’s price level. Alternatively, the term can be interpreted as this price expressed in terms of a common currency, namely that of the base country, country 1, for which the price level is normalised at unity (as \( \lambda_1 = 0 \), \( \exp(\lambda_1) = 1 \)). In the above equation this common-currency price is then compared to the common price of this good in all countries \( \exp(\mu_i) \). Accordingly, the error \( \varepsilon_{ic} \) is the logarithmic deviation of the common-currency price in country \( c \) from the world price of the good in question.

We have \( n \times C \) prices and \( n + C - 1 \) parameters to be estimated in model (6.1). In the two-country case, we consider the estimation of \( \lambda_2 \) and \( \mu_1 \) as a weighted least-squares problem by minimising \( \Sigma_{i=1}^{n} \Sigma_{c=1}^{2} a_{ic} \varepsilon_{ic}^2 \), or
\[
\Sigma_{i=1}^{n} a_{i1} \left( \log p_{i1} - \mu_1 \right)^2 + \Sigma_{i=1}^{n} a_{i2} \left( \log p_{i2} - \lambda_2 - \mu_1 \right)^2.
\]
While this weighting scheme is a direct consequence of the error structure set out in the previous paragraph, an alternative perspective that the weight accorded to each price in an index of all prices should reflect its economic importance, as measured by $a_{ik}$. Thus for example if expenditure on food is twice as that on clothing, then the price of the former good should be weighted twice as heavily as that of the latter; we will illustrated this idea below in the case in which the terms $a_{ik}$ are related to the budget shares.

It can be shown that the WLS estimator of $\lambda_2$ is

$$\hat{\lambda}_2 = \sum_{i=1}^{n} s_i \log \left( \frac{p_{i2}}{p_{i1}} \right),$$

where the weight $s_i$ is defined as

$$s_i = \frac{a_{i1} a_{i2} / (a_{i1} + a_{i2})}{\sum_{j=1}^{n} a_{j1} a_{j2} / (a_{j1} + a_{j2})} = \frac{h(a_{i1}, a_{i2})}{\sum_{j=1}^{n} h(a_{j1}, a_{j2})},$$

with $h(a_{i1}, a_{i2}) = \left[ (1/2)(a_{i1}^{-1} + a_{i2}^{-1}) \right]^{-1} = 2(a_{i1} a_{i2})/(a_{i1} + a_{i2})$ the harmonic mean of $a_{i1}$ and $a_{i2}$. As $0 \leq s_i \leq 1$ and $\sum_i s_i = 1$, the index $\hat{\lambda}_2$ is a share-weighted average of the logarithms of the price ratios $p_{i2}/p_{i1}$. As countries 1 and 2 appear in a symmetric fashion in equations (6.2) and (6.3), it can be seen that if the two countries are interchanged, then the corresponding WLS estimator of the price level of country 1 relative to country 2 is the negative of $\hat{\lambda}_2$. This means that the index $\hat{\lambda}_2$ satisfies Fisher’s time-reversal test.

It can also be shown that the WLS estimator of the relative price parameter $\mu_i$ is

$$\hat{\mu}_i = b_i \log p_{i1} + (1 - b_i) \left( \log p_{i2} - \hat{\lambda}_2 \right),$$

where $b_i = a_{i1} / (a_{i1} + a_{i2})$ is a positive fraction and $\hat{\lambda}_2$ is the estimated price level, as defined in equation (6.2). This equation reveals that the estimator of the relative price of good $i$ is a weighted average of the prices of the good in the two countries, where the weights are inversely proportional to the error variances. The apparent asymmetry in the way the prices in the two countries are expressed is due entirely to the normalisation that the price level in country 1 is taken to be unity, or in logarithmic terms $\lambda_1 = 0$. Accordingly, each of the two price terms on the right of equation
(6.4), \( \log p_{i2} \) and \( \left( \log p_{i2} - \hat{\lambda}_2 \right) \), is interpreted as a relative price, the nominal price of the good in terms of the price of all goods in the country in question.

Now we consider some special cases of the specification of the weights \( a_{i1} \) and \( a_{i2} \) which lead to interesting results. First, if we set \( a_{i1} = a_{i2} = a \), a constant (for all \( i \)) in both countries, then the weight given by equation (6.3) becomes \( 1/n \), so that equation (6.2) is simplified to an unweighted average of the price ratios \( p_{i2}/p_{i1} \),

\[
(6.5) \quad \frac{1}{n} \sum_{i=1}^{n} \log \left( \frac{p_{i2}}{p_{i1}} \right),
\]

which is the Jevons (1865) price index. In this case, the estimator of the relative price parameter of equation (6.4) also becomes an unweighted average of the two relative prices as \( b_i = 1/2 \).

Let \( q_{ic} \) be the quantity consumed of \( i \) in \( c \), and \( M_c = \sum_{i=1}^{n} p_{ic} q_{ic} \) be total expenditure in country \( c \), so that \( w_{ic} = p_{ic} q_{ic}/M_c \) is the \( i \)th budget share in \( c \). Next, if we set \( a_{i} = w_{ic} \), then the index (6.2) becomes a weighted average of the logarithms of the price ratios \( p_{i2}/p_{i1} \), with weights that are functions of the budget shares. It can be shown that this expression is an approximation to the Törnqvist (1936)-Theil (1967) index. Moreover, if we write \( w_i = (w_{i1} + w_{i2})/2 \) for the arithmetic average of the budget shares of good \( i \) in the two countries and set \( a_{i1} = a_{i2} = w_i \), then the weight (6.3) takes the form \( s_i = w_i \), and equation (6.2) becomes the exact Törnqvist-Theil index, \( \hat{\lambda}_2 = \sum_{i=1}^{n} w_i \log \left( p_{i2}/p_{i1} \right) \). Given that the Törnqvist-Theil index is a superlative index (Diewert, 1976), it appears that specifying the weights to be the arithmetic average of the budget shares is a desirable choice.

Model (6.1) has the logarithm of the price on the left-hand index. Some further interesting results emerge if we consider the associated multiplicative model (with the error term suppressed):

\[
(6.6) \quad p_{ic} = \alpha_c \beta_i,
\]

where \( \alpha_c = \exp(\lambda_c) \), with \( \alpha_1 = 1 \), and \( \beta_i = \exp(\mu_i) \). If we now weight by quantities consumed \( q_{ic} \), the WLS problem is to minimise
However the solution for $\alpha_2$ suffers from the fatal flaw that it is not invariant to changes in the units of measurement of commodities. Diewert’s solution is to use the following power transformation of equation (6.6) with the nonzero parameter $\rho$, $p_{ic}^\rho = \alpha_i^\rho \beta_i^\rho$. Defining $\gamma_c^\rho = \alpha_c^\rho$, $\delta_i^\rho = \beta_i^\rho$, we write the transformed equation as $p_{ic}^\rho = \gamma_c^\rho \delta_i^\rho$, with $\gamma_1 = 1$. Then if we divide both sides of the previous equation by $\gamma_c$ and define $\phi_c = 1/\gamma_c = 1/\alpha_c^\rho$, we can write $\phi_c p_{ic}^\rho = \delta_i$, with $\phi_1 = 1$. The counterpart to the weighted sum of squares (6.7) is then

$$
\sum_{i=1}^n q_{il} (p_{il} - \beta_i)^2 + \sum_{i=1}^n q_{i2} (p_{i2} - \alpha_2 \beta_i)^2.
$$

This weighting scheme can be justified if we take $p_{ic}^\rho$ to be independently distributed with mean and variance of $\gamma_c \delta_i$ and $\gamma_c^2 \sigma^2/q_{ic}$, respectively.

Equation (6.8) then leads to the following WLS estimator:

$$
\hat{\phi}_2 = \frac{\sum_{i=1}^n h(q_{il}, q_{i2})(p_{il} \times p_{i2})^\rho}{\sum_{j=1}^n h(q_{jl}, q_{j2})(p_{j2})^{2\rho}},
$$

where $h(q_{il}, q_{i2})$ is the harmonic mean of the consumption of commodity $i$ in countries 1 and 2. It can be shown that $\hat{\phi}_2$ is not invariant to changes in the units of measurement unless $\rho = 1/2$.

With this value of $\rho$, the implied estimator of $\alpha_2 = \gamma_2^{1/\rho} = \phi_2^{1/\rho}$ is

$$
\hat{\alpha}_2 = \left[ \frac{\sum_{i=1}^n h(q_{il}, q_{i2}) p_{i2}}{\sum_{j=1}^n h(q_{jl}, q_{j2})(p_{jl})^{1/2}} \right]^2.
$$

According to this expression, the square root of the estimated price level in country 2 is the ratio of two sets of total consumption expenditure. An identical consumption basket is priced in both the numerator and denominator of this ratio; that basket involves the harmonic mean of the
consumption of each good in the two countries. The numerator of the ratio is the cost in country 2 of this basket, as this country’s prices are used to evaluate the cost of the basket. The denominator is the cost of the same basket evaluated at the geometric mean of the prices in the two countries. It can therefore be seen that country 2’s price level involves a comparison of a weighted average of its prices with a similar weighted average of world prices, the latter defined in a geometric mean sense. That is, if we make the (heroic!) assumption that quantities are all expressed in similar units, so that

$$H = \sum_{i=1}^{n} h(q_{i1}, q_{i2})$$

is a well-defined total, and then write $k_i = h(q_{i1}, q_{i2})/H$ for the share of good $i$ in this total and $p_j^* = \left(\frac{p_{ji} \times p_{j2}}{\sqrt{\alpha_2}}\right)^{1/2}$ for the world price of good $j$, we have

$$\sqrt{\alpha_2} = \frac{\sum_{i=1}^{n} k_i \ p_{i2}}{\sum_{j=1}^{n} k_j \ p_j^*}.$$  

(6.10)

A second interpretation of index (6.9) is as follows. Write the numerator of (6.9) as

$$\sum_{i=1}^{n} h(q_{i1}, q_{i2})p_{i2} = \sum_{i=1}^{n} h(q_{i1}, q_{i2})\left(\frac{p_{i2}}{p_{i1} \times p_{i2}}\right)^{1/2} \frac{p_{i2}}{\left(\frac{p_{i1} \times p_{i2}}{\sqrt{\alpha_2}}\right)^{1/2}}$$

(6.11)

$$= \sum_{i=1}^{n} h(q_{i1}, q_{i2})p_i^* \left(\frac{p_{i2}}{p_i}\right).$$

The expression on the second line of this equation is a weighted sum of the prices in country 2 relative to world prices $p_i^*$, each weight being expenditure on harmonic mean consumption of the relevant good evaluated at world prices. Accordingly, defining $H' = \sum_{j=1}^{n} h(q_{j1}, q_{j2})p_j^*$ and $k'_i = h(q_{i1}, q_{i2})p_i^*/H'$ as the corresponding total expenditure and the associated budget share of good $i$, it follows from equations (6.9) and (6.11) that

$$\sqrt{\alpha_2} = \sum_{i=1}^{n} k'_i \left(\frac{p_{i2}}{p_i}\right).$$  

(6.12)

Equation (6.12) gives rise to a more attractive interpretation of the index as it does not involve the assumption that quantities are expressed in comparable units. Note that using the definition of
world prices \( p_i = (p_{i1} \times p_{i2})^{1/2} \), equation (6.12) can also be expressed as \( \sqrt{\alpha_2} = \sum_{i=1}^{n} k'_i \left( \frac{p_{i2}}{p_{i1}} \right)^{1/2} \), so that

\[
\hat{\alpha}_2 = \left[ \sum_{i=1}^{n} k'_i \left( \frac{p_{i2}}{p_{i1}} \right)^{1/2} \right]^2.
\]

This clearly shows that the index possesses the appropriate homogeneity properties.

Notwithstanding its appealing interpretation, index (6.9) does not satisfy the important time reversal test. In order to overcome this problem, Diewert (2002) uses Fisher’s (1927) “rectification procedure”, which can be explained as follows. If we now take country 2 as the base country (that is, replace the normalization \( \alpha_i = 1 \) with \( \alpha_2 = 1 \)) and repeat the minimisation problem, we obtain the following estimator for \( \alpha_1 \):

\[
\hat{\alpha}_1 = \left[ \frac{\sum_{i=1}^{n} h(q_{i1}, q_{i2}) p_{i1}}{\sum_{j=1}^{n} h(q_{j1}, q_{j2}) \left( p_{j1} \times p_{j2} \right)^{1/2}} \right]^2,
\]

which is exactly the same as the right-hand side of equation (6.9) except that the prices of country 1 now replace those of country 2 in the numerator. The geometric mean of \( \hat{\alpha}_2 \) and \( 1/\hat{\alpha}_1 \) is then used as an estimator of the price level of country 2 relative to that of country 1, which we write as \( \hat{\alpha} = \sqrt{\hat{\alpha}_2 / \hat{\alpha}_1} \). Accordingly,

\[
(6.13) \quad \hat{\alpha} = \frac{\sum_{i=1}^{n} h(q_{i1}, q_{i2}) p_{i2}}{\sum_{j=1}^{n} h(q_{j1}, q_{j2}) p_{j1}} = \sum_{i=1}^{n} k''_i \left( \frac{p_{i2}}{p_{i1}} \right),
\]

where \( k''_i = h(q_{i1}, q_{i2}) p_{i2} / H'' \) is another budget share, with \( H'' = \sum_{j=1}^{n} h(q_{j1}, q_{j2}) p_{j1} \) total expenditure, now defined as the cost of the harmonic mean basket evaluated at country 1 prices. Index (6.13) is again a weighted average of the prices in country 2 relative to those in country 1, with weights that now involve harmonic mean consumption valued at country 1 prices.

Expression (6.13) is especially rich in its implications. Not only does it satisfy homogeneity and time reversal, but it coincides with the Geary (1958)-Khamis (1970) bilateral index number
formula, which was also considered by Fisher (1927). An interesting special case emerges if we replace the quantities consumed in each country, $q_{i1}$ and $q_{i2}$, with “world consumption”, now defined as the arithmetic average of the respective quantity consumed in the two countries, $q_{i1}^* = q_{i2}^* = \frac{1}{2}(q_{i1} + q_{i2})$. Then, harmonic mean consumption becomes world consumption, $h(q_{i1}^*, q_{i2}^*) = \frac{1}{2}(q_{i1} + q_{i2})$, and index (6.13) becomes the Marshall (1887)-Edgeworth (1925) bilateral index:

$$
\frac{\sum_{i=1}^{n}(1/2)(q_{i1} + q_{i2})p_{i2}}{\sum_{j=1}^{n}(1/2)(q_{j1} + q_{j2})p_{j1}}
$$

Alternatively, if rather than the arithmetic mean, we define world consumption as the geometric mean, so that $q_{i1}^* = (q_{i1} \times q_{i2})^{1/2}$, $q_{i2}^* = (q_{i1} \times q_{i2})^{1/2}$, expression (6.13) now takes the form of the Walsh (1901) index:

$$
\frac{\sum_{i=1}^{n}(q_{i1} \times q_{i2})^{1/2}p_{i2}}{\sum_{j=1}^{n}(q_{j1} \times q_{j2})^{1/2}p_{j1}}
$$

Finally, setting $q_{i1}^* = q_{i2}^* = 1$, $i = 1, ..., n$, yields the unweighted index due to Dutot (1738):

$$
(6.14)
\frac{\frac{1}{n}\sum_{i=1}^{n}p_{i2}}{\frac{1}{n}\sum_{j=1}^{n}p_{j1}}
$$

This is the arithmetic analogue of Jevons unweighted geometric index (5.5).

Diewert’s (2002) insightful results show how the weighted and unweighted indexes of the price level in one country are related to one another, and he concludes that (p. 9)

These unweighted indexes [equations (6.5) and (6.14)] can be very far from their weighted counterparts. Thus the main conclusion we draw from this note is that in running Country Product Dummy regressions or hedonic regressions in the time series context, it is very important to run appropriately weighted version of these regressions in order to obtain more accurate estimates of price levels. (Diewert’s emphasis.)
Further Analysis of the CPD Model

The “economic approach” generates index numbers from the consumer’s utility function, the cost function or some transform thereof, while according to the “test approach” indexes must satisfy some fundamental desirable properties. In contrast to these two approaches, it is clear that the various indexes discussed in the above subsection are derived from a weighted least-squares problem, which is turn is associated with an underlying regression model relating prices to country and commodity dummy variables.

The WLS foundations of the indexes mean that under the stated assumptions, they have desirable properties not considered by the traditional approaches. First, the (weighted) sum of squared deviations of the observed prices from the underlying economic model is a minimum, which implies that the corresponding index has a “best fit” property. Second, as the index can be expressed in the form of a regression coefficient, the index itself is a random variable. This is in marked contrast to other approaches in which the index number is purported to be deterministic, so that estimation uncertainty plays no role. Thus under the stochastic approach, there is a whole probability distribution of the index, which could be characterised by its moments and certain ranges, in the usual manner. Third, a stochastic price index is a best linear unbiased estimator of the underlying price level. The “best” part of BLUE is particularly attractive as it means that (among the class of linear unbiased estimators) the index is minimum variance. Diewert’s (1976) “superlative” index number approach evaluates the performance of a given index by investigating its abilities to approximate an unknown true index\(^{16}\); it is in this sense that Diewert uses the expression “more accurate estimates of the price level” in the above quotation. While this aspect of superlative index numbers seems to be not unrelated to the minimum variance property of stochastic indexes, it still seems sufficiently different to justify treating the two approaches as members of different species.

As the estimated price level is a random variable under the stochastic approach, it is natural to use the tools of statistical inference to test hypotheses. We could ask for example, Is the price level in country 2 greater than that in country 1 (which amounts to the parametric restriction \(\alpha_2 = 0\))? To conduct hypothesis testing we need the standard error of the estimated price level, as noted by Diewert (2002, p. 2):

> The main advantage of the CPD method for comparing prices across countries over traditional index number methods is that we can obtain standard errors for the country price levels …. This advantage of the stochastic approach to

\(^{16}\)More precisely, an index is described as “superlative” if it is exact for a flexible underlying aggregator function (Diewert, 1976).
index number theory was stressed by Summers (1973) and more recently by Selvanathan and [Prasada] Rao (1994). (Diewert’s emphasis.)

As such standard errors are readily available as part of the regression output, they can be readily computed numerically. But in certain simplified cases, it is possible to derive straightforward algebraic expressions that yield considerable insight into the nature of the problem, as we shall now show.

We write model (6.1) for \( c = 1, 2 \) as

\[
\log p_{i1} = \lambda_i + \mu_i + \varepsilon_{i1}, \quad \log p_{i2} = \lambda_2 + \mu_i + \varepsilon_{i2},
\]

so that \( \log \left( \frac{p_{i2}}{p_{i1}} \right) = \lambda_2 + (\varepsilon_{i2} - \varepsilon_{i1}) \), as \( \lambda_i = 0 \). Substituting the right-hand side of this equation in (6.2) yields the following expression for the estimator of country 2’s price level:

\[
\hat{\lambda}_2 = \frac{\sum_{i=1}^{n} s_i \left[ \lambda_2 + (\varepsilon_{i2} - \varepsilon_{i1}) \right]}{s},
\]

where \( s_i \) is the positive fraction defined by equation (6.3). Under the assumption that the errors \( \varepsilon_{ic} \) are independent over countries and commodities with \( \text{var} \varepsilon_{ic} = \sigma^2 / a_{ic}^2 \), we have

\[
\text{var} \hat{\lambda}_2 = \sigma^2 \left[ \sum_{i=1}^{n} s_i^2 \left( \frac{1}{a_{i1}^2} + \frac{1}{a_{i2}^2} \right) \right] = 2\sigma^2 \left[ \sum_{i=1}^{n} s_i^2 h\left( a_{i1}^2, a_{i2}^2 \right) \right],
\]

where \( h\left( a_{i1}^2, a_{i2}^2 \right) \) is the harmonic mean of \( a_{i1}^2 \) and \( a_{i2}^2 \).

The parameter \( \sigma^2 \) is a basic measure of noise in the system in the sense that the each error variance is proportional to its value: \( \text{var} \varepsilon_{ic} = \sigma^2 / a_{ic}^2 \). To further interpret \( \sigma^2 \), consider a weighted logarithmic variance of prices in country \( c \):

\[
(6.16) \quad \sum_{i=1}^{n} a_{ic}^2 (\log p_{ic} - \lambda_c - \mu_i)^2 = \sum_{i=1}^{n} a_{ic}^2 \varepsilon_{ic}^2.
\]

Under the interpretation of \( \varepsilon_{ic} \) as the logarithmic deviation of the common-currency price of \( i \) in \( c \) from the world counterpart, equation (6.16) is a weighted sum of squares of the \( n \) such deviations.
As \( \text{var} \varepsilon_{ic} = \sigma^2/a_{ic}^2 \), the expectation of the right-hand side of equation (6.16) is \( n \sigma^2 \). Accordingly, expression (6.15) reveals that the variance of the price level estimator increases with the amount of noise there is in the system, which in turn measures the average weighted squared deviation of all the common-currency prices from world prices, as \( \sigma^2 = E\left[\frac{1}{n} \sum_{i=1}^{n} a_{ic}^2 \varepsilon_{ic}^2 \right] \). This interpretation agrees with the long-held idea (previously mentioned in Section 2) that the overall level of prices is in some sense less well defined when there are substantial exogenous shocks to the structure of relative prices.

The term in square brackets on the far right side of equation (6.15) indicates that the dependence of the variance of the price level on the parameters \( a_{ic} \) is rather complex. For some additional insight, consider the Jevons index (6.5), which pertains to the unweighted case where \( a_{i1} = a_{i2} = a \), a constant. In this situation, \( s_i = 1/n \) and \( h(a_{i1}^2, a_{i2}^2) = a^2 \), so that equation (6.15) becomes

\[
\text{var} \hat{\lambda}_2 = 2\sigma^2 \left[ \frac{\sum_{i=1}^{n} (1/n)^2}{a^2} \right] = \frac{2(\sigma/a)^2}{n}.
\]

Here the variance is directly proportional to \( \sigma^2 \), as before, and inversely proportional to \( a^2 \) and \( n \), the number of commodities.

A second interpretation of the parameter \( \sigma^2 \) is due to Diewert (2004). Model (6.1) implies that the price of good \( i \) in country \( c \) relative to that in some other country \( d \) is

\[
p_{ic}/p_{id} = \exp(\hat{\lambda}_c - \hat{\lambda}_d) \times \exp(\varepsilon_{ic} - \varepsilon_{id}),
\]

which shows that if the errors vanish then all \( n \) prices are proportional in the two countries being compared, with \( \exp(\hat{\lambda}_c - \hat{\lambda}_d) \) the factor of proportionality. Thus the difference in the logarithmic errors, \( \varepsilon_{ic} - \varepsilon_{id} \), is a natural measure of disproportionality, or dissimilarity, of the relative price of good \( i \). One measure of the extent to which the overall structure of relative prices differs across countries \( c \) and \( d \) is the weighted sum of squared errors (6.16), which refers to country \( c \), less that for country \( d \). This difference for the two countries has expectation \( 2n\sigma^2 \). Accordingly, the term \( \sigma^2 \) on the right-hand side of equation (6.15) is interpreted as reflecting the relative dissimilarity in the structure of prices in the two countries, so that the variance of the price level rises with the degree of dissimilarity.\(^{17}\)

\(^{17}\) More generally, Diewert (2004) recommends using the variance, \( s^2 \), of the residuals for all countries and commodities from the CPD model as a measure of dissimilarity of the structure of relative prices; then the transform \( s^2/(1 + s^2) \) lies between 0 (no dissimilarity) and 1 (maximum dissimilarity).
We conclude this subsection with a brief note on additional items of the previous literature. The idea of employing weights within the CPD model has been also considered by others. Theil and Suhm (1981) and Voltaire and Stack (1980) extend the original approach of Summers (1973); and within the context of stochastic index numbers, Balk (1980), Clements and Izan (1981, 1987) and Selvanathan and Prasada Rao (1994) also use a similar weighted approach.

Summary

The paper by Diewert discussed above clearly illustrates the richness of the stochastic approach and how it can be integrated with more traditional index-number theory. Further new results along the lines suggested by Diewert can be expected to emerge in the future. For example, Diewert himself (2004) uses the same basic framework to analyse much more complex issues associated with the measurement of prices levels across countries, and it would seem likely that these ideas will soon be incorporated in real-world applications. Maybe some international institution will commence using these measures and publish them on a routine basis. For further related recent research on CPD methods, see Coondoo et al. (2004), Prasada Rao (2004), and the references included in Diewert (2004).

7. Concluding Comments

In recent years there has been a noticeable increase in professional interest in index numbers. There are a variety of reasons for this development, including an acceleration of the availability of new goods on the market, the enhanced quality of existing goods, increased emphasis on greater variety of consumption baskets, substantial improvements in information processing capabilities, and breakthroughs in index-number theory that have practical implications for statistical agencies. The stochastic approach to index numbers has shared in this development, and has attracted considerable attention in its own right. In contrast to the traditional approaches -- the economic theory approach and the test approach -- stochastic index numbers are generated by minimising a weighted sum of squared deviations from a regression line, so that they have a “best fit” status and uncertainty and statistical ideas play a central role. Rather than just providing one number for the rate of inflation, the stochastic approach provides the whole probability distribution of inflation; in particular as standard errors are available for the indexes, it is possible to carry out conventional hypothesis testing, something not possible with traditional approaches. Interestingly,
a number of familiar traditional index formulae can be generated by the stochastic approach, so that they can be given new interpretations derived from statistical foundations.

In this paper we have reviewed the basic elements of the stochastic approach, provided a link with Fisher’s (1927) early work, assessed Diewert’s (1995) well-known critique of the approach, presented Theil’s (1967) methodology, and summarised and extended Diewert’s (2002) recent work that applies the approach to measuring price levels across countries. We believe the stochastic approach to be very rich in its implications and that it offers considerable further research opportunities. Examples include:

- The measurement of total factor productivity.
- International comparisons of real incomes involving purchasing power parities.
- The testing of parity conditions in international finance involving prices, exchange rates and interest rate.
- The theory and measurement of monetary conditions. For example, the Reserve Bank of New Zealand used to employ a “monetary conditions index”, a weighted average of interest rates and the exchange rate. It would be possible to formulate such an index with the stochastic approach to endow it with firmer analytical foundations.
- Modelling the term structure of interest rates.

While out of necessity our review has not been exhaustive, four other approaches related to stochastic index numbers should be mentioned. First, Blankmeyer (1990) considers prices and quantities simultaneously in the following model:

$$\log M_{st} = \alpha + \beta_p + \beta_q + \text{disturbance}$$

where $$M_{st} = \sum_{i=1}^{n} p_{si} q_{si}$$ is expenditure on the quantities consumed in period $$t$$ evaluated at the prices prevailing in period $$s$$, $$\exp(\beta_p)$$ is an index of prices in $$s$$, $$\exp(\beta_q)$$ is a quantity index in $$t$$, and $$\exp(\alpha)$$ is a proportionality constant. This model for $$s, t = 1, ..., T$$ contains $$1 + 2T$$ unknown parameters to be estimated with $$T^2$$ “observations”, but these observations cannot all be independent.

The second approach is the repeat sales regression model:

$$\log\left(\frac{p_{ht}}{p_{hs}}\right) = \alpha - \alpha + \text{disturbance},$$

where $$p_{hs}$$ is the price of house $$h$$ ($$h = 1, ..., H$$) that was initially sold in period $$s$$ and then again later in $$t$$ at $$p_{ht}$$ ($$s = 1, ..., T$$); and $$\exp(\alpha)$$ is the value of the index in period $$t$$. The attraction of this approach is that it deals with the intermittent nature of house sales, as well as the heterogeneity of the quality of houses by considering the price of a given house at two different times, so that the fixed effect (the quality of the house) is differenced out. This approach, originally due to Bailey et al. (1963), has been extended by Shiller (1991) in several directions including weighting by house values. For an extensive discussion of recent research in this area, see Thibodeau (1997).

The third approach is Stockman’s (1988) well-known decomposition of the growth in output of sector $$i$$ in country $$c$$, $$D_{i\text{tot}}$$, which is formulated as

$$D_{i\text{tot}} = \alpha_i + \beta_i + \gamma_i + \text{disturbance},$$

where $$\alpha_i$$ is the growth in sector $$i$$ specific to country $$c$$; $$\beta_i$$ is the growth in $$i$$ in period $$t$$ common to all countries; and $$\gamma_i$$ is the growth in all sectors in country $$c$$ in $$t$$. The terms $$\alpha_i$$ and $$\beta_i$$ can be interpreted as representing technical change and/or shocks to sector $$i$$ which occur in country $$c$$ in all periods (for $$\alpha_i$$) and in period $$t$$ for all countries (for $$\beta_i$$). The term $$\gamma_i$$ represents aggregate disturbances (policy or otherwise) that affect all sectors in country $$c$$ in period $$t$$. From the perspective of the stochastic approach, it would be useful to consider weighting schemes to reflect the relative importance of sectors, countries and time periods.

Finally, Feenstra and Reinsdorf (2003) consider an integration of the stochastic and economic approaches by treating as random some of the parameters of underlying utility function, as well as the prices, and illustrate the basic idea with the CES and translog cases.

See Miller (1984) and Ong et al. (1999) for some preliminary attempts.
• The impact on the price index of adding additional information by (i) including in the index the prices of new commodities; (ii) using other sources of new information such as money market equilibrium in the form of the quantity theory equation of exchange; and (iii) employing the factor reversal criterion which involves the simultaneous use of price and quantities data to derive index numbers.\footnote{For some results along these lines, see Clements and Selvanathan (2001).}
APPENDIX

In this Appendix we derive result (4.5). We need to evaluate equations (2.3) and (2.4) with the covariance matrix \( \Sigma \) specified as \( \Sigma = D(I + \lambda)D \). Using result (4.3), the first part of the right-hand side of equation (2.3) is

\[
\left[ t' \Sigma^{-1} t \right]^{-1} = \left[ t' D^{-1} (I - \lambda) D^{-1} t \right]^{-1} = \left[ t' \left( D^{-2} t - D^{-1} \lambda D^{-1} t \right) \right]^{-1}
\]

\[
= \left[ \sum_{i=1}^{n} \left( \sigma_{ii}^{-1} - \sum_{j=1}^{n} \sigma_{ij}^{-1/2} \sigma_{ii} \sigma_{jj}^{1/2} \right) \right]^{-1},
\]

where \( \sigma_{ii} \) is the variance of the \( i \)th relative price; and \( \lambda_{ij} \) as the \((i, j)\)th element of \( \lambda \). Define \( \lambda_{ij}^* = \sigma_{ii}^{-1/2} \sigma_{ij} \sigma_{jj}^{1/2} = \sigma_{ij} / \sigma_{ii} \sigma_{jj} \), and \( \lambda_{ii}^* = \sum_{j=1}^{n} \lambda_{ij}^* \) as the \(i\)th row sum of the \( n \times n \) matrix \( \lambda^* = [\lambda_{ii}^*] \). As \( \lambda^* \) is symmetric, each row sum is equal to the corresponding column sum, \( \lambda_{ii}^* = \lambda_{ii}^* \), \( i = 1, ..., n \), where \( \lambda_{ii}^* = \sum_{j=1}^{n} \lambda_{ii}^* \). It then follows that

(A1)

\[
\left[ t' \Sigma^{-1} t \right]^{-1} = \left[ \sum_{i=1}^{n} \left( \sigma_{ii}^{-1} - \lambda_{ii}^* \right) \right]^{-1}.
\]

The second part of the right-hand side of equation (2.3) is

\[
t' \Sigma_t^{-1} Dp_t = t' \left[ \left( D^{-1} - \lambda D^{-1} \right) Dp_t \right] = t' \left[ D^{-2} - D^{-1} \lambda D^{-1} \right] Dp_t
\]

\[
= t' \left( D^{-2} - \lambda^* \right) Dp_t.
\]

As \( t'D^{-2} = [\sigma_{11}^{-1}, ..., \sigma_{nn}^{-1}] \) and \( t'\lambda^* \) is the \( 1 \times n \) vector of the column sums, \( \lambda_{ii}^*, ..., \lambda_{nn}^* \), we have

(A2)

\[
t' \Sigma_t^{-1} Dp_t = \sum_{i=1}^{n} \left( \sigma_{ii}^{-1} - \lambda_{ii}^* \right) Dp_{it}.
\]

The combination of (A1) and (A2) then yields

(A3)

\[
\hat{\alpha}_t = \left[ t' \Sigma_t^{-1} t \right]^{-1} t' \Sigma_t^{-1} Dp_t = \sum_{i=1}^{n} y_i Dp_{it},
\]

where \( y_i = (\sigma_{ii}^{-1} - \lambda_{ii}^*) / \sum_{j=1}^{n} (\sigma_{jj}^{-1} - \lambda_{jj}^*) \). From equations (2.4) and (A1) we have

(A1.4)

\[
\text{var} \hat{\alpha}_t = \frac{1}{\sum_{i=1}^{n} (\sigma_{ii}^{-1} - \lambda_{ii}^*)}.
\]

Equations (A3) and (A4) are result (4.5) of the text.
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