

Multi-dimensional Error Measures for Ex Post Facto Forecast and Estimation Evaluations¹

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Abstract

This paper deals with an issue that appears to be unexplored by demographers. The issue is based on the question: "if one was not only doing ex post facto evaluations of estimates (or projections) of the total population but by characteristics such as race and geography, how would one summarize the summary measures of error?" The paper examines this issue in two dimensions using a standard summary measure of ex post facto error "Mean Algebraic Percent Error" (MALPE). It offers hypothetical, but empirical, examples of the issue using MALPE. It is aimed at answering the underlying question, which is "what is the nature of the relationship, if any, of summary errors of measure taken across different dimensions in conducting an ex post facto evaluation of the accuracy of a set of estimates or projections. The paper finds that relationships do exist and offers proof of these relationships both mathematically and empirically. It concludes by noting work to be done includes generalizing these findings to more than two dimensions and extending the findings to other commonly used summary measures of ex post facto error, such as Mean Absolute Percent Error (MAPE) and Root Mean Square Error (RMSE).

I. The Problem

Suppose we have done estimates back in 2009 for the year 2010 by race (using three race groups) for a state that has four counties. We now want to do an ex post facto evaluation of the accuracy of our estimates. Key to this evaluation is the summary measures of error for (1) Race, (2) Geography, and (3) for the race-geography jointly (i.e., the "overall MALPE"). As you will see below using the example of Mean Algebraic Percent Error (MALPE), the MALPE for race is not the same as the MALPE for geography and neither are the same as the MALPE for race-geography jointly. In table 1 below are the estimates (E_{ij}) and in Table 2 are the actual populations (P_{ij}). Table 3 contains the errors and Table 4 contains the proportionate errors ($(E_{ij}-P_{ij})/P_{ij}$).

Note that in each case, we have a 3 (race groups) by 4 (geography) table, with marginal totals for race in the fifth row and marginal totals for geography in the fourth column.

Table 1. 2010 Estimates (Eij) for the state by race and county

| | estimate | | | |
|-------|----------|-------|-------|-------|
| | race1 | race2 | race3 | total |
| geog1 | 305 | 100 | 55 | 460 |
| geog2 | 500 | 125 | 40 | 665 |
| geog3 | 86 | 10 | 5 | 101 |
| geog4 | 125 | 20 | 10 | 155 |
| total | 1016 | 255 | 110 | 1381 |

Table 2. 2010 Actual Population (Pij) Counts for the state by race and county

| | race1 | race2 | race3 | total |
|-------|-------|-------|-------|-------|
| geog1 | 291 | 98 | 45 | 434 |
| geog2 | 490 | 130 | 51 | 671 |
| geog3 | 85 | 8 | 4 | 97 |
| geog4 | 130 | 25 | 12 | 167 |
| total | 996 | 261 | 112 | 1369 |

Table 3. Numerical Error (Eij-Pij) in the 2010 Estimates by race and county

| | race1 | race2 | race3 | total |
|-------|-------|-------|-------|-------|
| geog1 | 14 | 2 | 10 | 26 |
| geog2 | 10 | -5 | -11 | -6 |
| geog3 | 1 | 2 | 1 | 4 |
| geog4 | -5 | -5 | -2 | -12 |
| total | 20 | -6 | -2 | 12 |

Table 4. Proportionate Error ((Eij-Pij)/Pij) in 2010 Estimates by race and county

| | race1 | race2 | race3 | total |
|-------|----------|----------|----------|----------|
| geog1 | 0.04811 | 0.020408 | 0.222222 | 0.059908 |
| geog2 | 0.020408 | -0.03846 | -0.21569 | -0.00894 |
| geog3 | 0.011765 | 0.25 | 0.25 | 0.041237 |
| geog4 | -0.03846 | -0.2 | -0.16667 | -0.07186 |
| total | 0.02008 | -0.02299 | -0.01786 | 0.008766 |

Using the Preceding data, here are the MALPEs for Race, Geography, and jointly by race-geography, respectively.

Table 5. MALPE by Dimension
Dimension MALPE

| | |
|-------|---------|
| RACE | -0.0069 |
| GEOG | 0.0051 |
| JOINT | 0.0136 |

In percent terms we have Race MALPE = -.69%, Geog MALPE = .51% and joint MALPE = 1.36%

As you can see, not only is it the case that none of the MALPEs is equal to another, but there is no clear relationship among them.

Now imagine yourself in front of a group of “stakeholder” discussing these estimates and you are describing these MALPEs (and how accurate your estimates are) when you are asked “why are they different?” The answer is not apparent, likely because it appears that demographers have not yet addressed the issue of simultaneous measures of error on multiple dimensions of a forecast or set of estimates. What follows is an initial attempt to answer the question, “Why are they different?” As you will see, the answer I believe is found in a subtle way that has to do with weighting.

II. MALPE and Weighted MALPE (WMALPE)

First recall, from Table 5

| Dimension | MALPE |
|-----------|---------|
| RACE | -0.0069 |
| GEOG | 0.0051 |
| JOINT | 0.0136 |

Now compare each of the (unweighted) MALPEs from Table 5 with its respective counterpart, WMALPE, as shown in Table 6.

Table 6. WMALPE by Dimension

| | |
|-------|--------|
| RACE | 0.0088 |
| GEOG | 0.0088 |
| JOINT | 0.0088 |

The obvious difference is that for each dimension, WMALPE is the same, which I believe is the key to understanding the differences observed in Table 5.

First, let’s look at some definitions.

Let E_i = estimate (or projection) for population in category i (e.g., race, geography, etc.)

P_i = actual population in category i (e.g., race, geography, etc.)

k = number of categories

In general terms, MALPE and WMALPE can be expressed as follows:

$$\text{MALPE} = (1/k) * [\sum ((E_i - P_i) / P_i)] \quad [1]$$

$$\text{And since } (1/k) = [1/(\sum P_i/k)/\sum P_i] \quad [1.a]$$

$$\text{MALPE} = [1/(\sum P_i/k)/\sum P_i] * [\sum ((E_i - P_i)/P_i)] \quad [1.b]$$

$$= /(\sum (P_i/k)/\sum P_i) * [\sum ((E_i - P_i)/P_i)] \quad [1.c]$$

$$= [\sum ((E_i - P_i)/(k * P_i))] \quad [1.d]$$

$$= [\sum ((E_i)/(k * P_i))] - 1 \quad [1.e]$$

$$\text{WMALPE} = \sum [(P_i/\sum P_i) * ((E_i - P_i)/P_i)] \quad [2]$$

$$\text{And since WMALPE} = \sum [(P_i/\sum P_i) * ((E_i - P_i)/P_i)] \quad [2.a]$$

$$= \sum [E_i - P_i] / \sum P_i \quad [2.b]$$

$$= \sum E_i / \sum P_i - \sum P_i / \sum P_i \quad [2.c]$$

$$= [\sum E_i / \sum P_i] - 1 \quad [2.d]$$

Thus, we have

$$\text{MALPE} = [\sum ((E_i)/(k * P_i))] - 1 \quad [3]$$

$$\text{and } 1 + \text{MALPE} = [\sum ((E_i)/(k * P_i))] \quad [3.a]$$

$$\text{WMALPE} = [(\sum E_i) / (\sum P_i)] - 1 \quad [4]$$

$$\text{and } 1 + \text{WMALPE} = [(\sum E_i) / (\sum P_i)] \quad [4.a]$$

In other words:

MALPE can be found by subtracting 1.00 from the sum of the ratios of: the Estimated Population in category i to the Actual Population in category i times the number of categories (k).

1 + MALPE is the sum of the ratios of: the Estimated Population in category i to the Actual Population in category i times the number of categories (k).

WMALPE can be found by subtracting 1.00 from the sum of the estimated population divided by the sum of the actual population.

1+ WMALPE is the sum of the estimated population divided by the sum of the actual population.

NOTE: The re-expressions found in [3] and [4] represent short-cut ways to calculate MALPE and WMALPE, respectively.

Using the re-expressions in [3] and [4] the following relationship can be seen between WMALPE and MALPE. The proof is found in Appendix A.

$$\text{WMALPE} = \text{MALPE} + [(\sum E_i)/(\sum P_i)] - [\sum (E_i/kP_i)] \quad [5]$$

$$\text{MALPE} = \text{WMALPE} - [(\sum E_i)/(\sum P_i)] + [\sum (E_i/kP_i)] \quad [6]$$

The proof for [5] is found in Appendix A. Since [6] follows naturally from [5], no proof for it is needed, although the relationship shown in [6] is repeated in the Appendix following the proof for [5].

WMALPE can be viewed as MALPE + the sum of the estimated Population divided by the sum of actual population from which is subtracted the sum of the (1 = 1 to k) ratios of estimated populations to the actual populations with the latter weighted by a constant, k

Although I have not yet done a proof, it appears clear from [4] that WMALPEs for two dimensions (e.g., race and geography) will not only equal one another but also the joint WMALPE. This extends to n dimensions, such that all n WMALPEs = one another and their joint WMALPE. It appears that we also have general statements ([5] and [6]) for the relationship between a MALPE on a given dimension and its equivalent WMALPE. With this knowledge in hand, I now turn to specifying the relationships between a set of dimensional MALPEs, their joint MALPE and the corresponding WMALPE.

To be more specific, again using the two example dimensions of race and geography (but noting that this results can be generalized to more than 3 columns and 4 rows), we have

$$\text{RMALPE} = \text{WMALPE} - [(\sum E_i)/(\sum P_i)] + [\sum (E_i/k*P_i)] \quad [7]$$

$$\text{GMALPE} = \text{WMALPE} - [(\sum E_j)/(\sum P_j)] + [\sum (E_j/(L*P_j))] \quad [8]$$

and

$$\text{JMALPE} = \text{WMALPE} - [(\sum E_{ij})/(\sum P_{ij})] + [\sum (E_{ij}/(m*P_{ij}))] \quad [9]$$

The Proof for Equation [9] is found in Appendix B.

where k = number of columns

L = number of rows

M = number of cells (k*L)

Using the preceding equations, we have from our example data in tables 1 and 2, the respective values for RMALPE, GMALPE, and JMALPE:

$$\begin{aligned}
 \text{RMALPE} &= \text{WMALPE} - \frac{[(\sum E_i)/(\sum P_i)]}{0.0088} + \frac{[\sum (E_i/(k \cdot P_i))]}{1.008765522} - 0.993078224 - 0.0069 \\
 \text{GMALPE} &= \text{WMALPE} - \frac{[(\sum E_j)/(\sum P_j)]}{0.0088} + \frac{[\sum (E_j/(L \cdot P_j))]}{1.005086696} - 1.005086696 - 0.0051 \\
 \text{JMALPE} &= \text{WMALPE} - \frac{[(\sum E_{ij})/(\sum P_{ij})]}{0.0088} + \frac{[\sum (E_{ij}/(m \cdot P_{ij}))]}{1.013636434} - 1.013636434 - 0.0136
 \end{aligned}$$

As can be seen in comparison with the values shown in Table 5, the values for RMALPE, GMALPE, and JMALPE derived using equations [7], [8], and [9], respectively, are the same.

III. The Relationship between Dimensional MALPEs and the Joint MALPE

In this section, I first will cover the relationship between the dimensional MALPEs and then the relationship between them, on the one hand, and the joint MALPE, on the other.

III.A The Relationship between Dimensional MALPEs.

Although this can be generalized to more than two dimensions and to more than the number of categories in the preceding example data set using Race (3 categories) and Geography (4 categories), I will stick with this example for now. As you can see in Table 1, the estimates for RACE are summarized by the “marginal total” for each of the 3 columns (i) and the estimates for GEOG are summarized by the “marginal total” for each of the 4 the rows (j).

As defined before, let RMALPE = MALPE for RACE

and GMALPE = MALPE for GEOG

JMALPE = MALPE for RACE & GEOG JOINTLY. i.e., the overall ERROR, which is

$$\sum (E_{ij} - P_{ij}) / (P_{ij})$$

WMALPE = weighted MALPE

where WRMALPE = Weighted RACE MALPE

WGMALPE = Weighted GEOG MALPE

WJMALPE = Weighted JOINT (overall both i and j) MALPE

[An important note is that WRMALPE = WGMALPE = WJMALPE. I use this in actually deriving the relationships shown below between RMALPE and GMALPE. The algebra for the derivations is tedious and not shown here. The details are in the attached excel file, along with numerical examples and definitions of RMALPE, GMALPE, and the WMALPEs].

Then

$$\text{RMALPE} = [\sum(\text{Ei}/\text{k}*\text{Pi})] - [\sum(\text{Ej}/\text{L}*\text{Pj})] + \text{GMALPE} \quad [10]$$

$$\text{GMALPE} = [\sum(\text{Ej}/\text{L}*\text{Pj})] - [\sum(\text{Ei}/\text{k}*\text{Pi})] + \text{RMALPE} \quad [11]$$

where Ei = Estimate for Race Group i

k = number of i categories (columns)

Pi = Actual number for Race Group i (Census)

Ej = Estimate for geography group j

L = number of j categories (rows)

We can interpret these relationships as follows:

(1) RMALPE is equal to GMALPE +

(the sum of: (the estimates by racial group, which are in the i column marginals of the estimates table divided by (the product of corresponding population numbers for each racial group, which are in the i column marginals of the population table and the number (k) of column marginals (racial groups)))

-

(the sum of: (the estimates by geography, which are in the j row marginals of the estimates table divided by (the product of the corresponding population numbers for each

geographic group in the j row marginals of the population table and the number (L) of row marginals (geographic groups))).

and

(2) GMALPE is equal to RMALPE +

(the sum of: (the estimates by geography, which are in the j row marginals of the estimates table divided by (the product of the corresponding population numbers for each geographic group in the j row marginals of the population table and the number (L) of row marginals (geographic groups)))

-

(the sum of: (the estimates by racial group, which are in the i column marginals of the estimates table divided by (the product of corresponding population numbers for each racial group, which are in the i column marginals of the population table and the number (k) of column marginals (racial groups))).

Using the data in Tables 1 and 2, we find that the empirical data results confirm the statements in equations [10] and [11]. Recall, again, from Table 5 that

| Dimension | MALPE |
|-----------|---------|
| RACE | -0.0069 |
| GEOG | 0.0051 |
| JOINT | 0.0136 |

Using equations [10] and [11] on the data in tables 1 and 2, we get the results shown in Table 7.

Table 7. Relationship Between RMALPE and GMALPE

RACE = Dimension i
 GEOG = Dimension j
 JOINT= Dimension ij

| | | | | | | | |
|------|---------|--------------------|---|--------------------|---|----------|---------|
| RACE | MALPE = | $[\sum(E_i/kP_i)]$ | - | $[\sum(E_j/kP_j)]$ | + | GMALPE = | RMALPE |
| | | 0.993078 | | 1.005087 | | 0.0051 | -0.0069 |
| GEOG | MALPE = | $[\sum(E_j/kP_j)]$ | - | $[\sum(E_i/kP_i)]$ | + | RMALPE = | GMALPE |
| | | 1.005087 | | 0.993078 | | -0.0069 | 0.0051 |

As you can see in Table 7, the empirical results correspond to what is expected from equations [10] and [11] as given in Table 5.

III.b The Relationship between Dimensional MALPEs and the Joint MALPE.

As you would suspect, a joint MALPE can be expressed in terms of each of the dimensional MALPEs. Continuing with our example of an ex post facto evaluation of error on two dimensions, Race and Geography, we will describe the relationship of RMALPE to JMALPE and the relationship of GMALPE to JMALPE. First, note that we can express JMALPE in terms of the Weighted MALPE (which remember are all equal: the dimensional WMALPEs are equal to one another and to the JMALPE).

$$\text{JMALPE} = \text{WMALPE} - \left[\frac{\sum E_{ij}}{\sum P_{ij}} \right] + \left[\sum \left(\frac{E_{ij}}{m \cdot P_{ij}} \right) \right] \quad [12]$$

With the help of equation [12] and some manipulation of terms, I first express JMALPE in terms of RMALPE [13] and then in terms of GMALPE [14]:

$$\text{JOINT MALPE} = \left[\sum \left(\frac{E_{ij}}{m \cdot P_{ij}} \right) \right] - \left[\sum \left(\frac{E_i}{k \cdot P_i} \right) \right] + \text{RMALPE} \quad [13]$$

$$\text{JOINT MALPE} = \left[\sum \left(\frac{E_{ij}}{m \cdot P_{ij}} \right) \right] - \left[\sum \left(\frac{E_j}{L \cdot P_j} \right) \right] + \text{GMALPE} \quad [14]$$

where, as before

k = number of columns

L = number of rows

M = number of cells (k*L)

Using equations [13] and [14] on the data in tables 1 and 2, we get the results shown in Table 8.

Table 8. The relationship between Dimensional MALPES and their Joint MALPE

| | | | | | | | |
|---------------|--|---|--|---|---------|---|--------|
| JOINT MALPE = | $\left[\sum \left(\frac{E_{ij}}{m \cdot P_{ij}} \right) \right]$ | - | $\left[\sum \left(\frac{E_j}{L \cdot P_j} \right) \right]$ | + | GMALPE | = | JMALPE |
| | 1.013636434 | | 1.005086696 | | 0.0051 | | 0.0136 |
| JOINT MALPE = | $\left[\sum \left(\frac{E_{ij}}{m \cdot P_{ij}} \right) \right]$ | - | $\left[\sum \left(\frac{E_i}{k \cdot P_i} \right) \right]$ | + | RMALPE | = | JMALPE |
| | 1.013636434 | | 0.993078224 | | -0.0069 | | 0.0136 |

As you can see in Table 8, the empirical results correspond to what is expected from equations [12], as given in Table 5.

To summarize,

(1) JMALPE is equal to RMALPE +

the sum of: (the estimates by the 12 race-geography groups in the estimates table, divided by (the product of the corresponding population numbers for each of the 12 race-geographic groups of the population table and the number (m=12) of cells comprising the 12 race-geographic groups))). -

(the sum of: (the estimates by racial group, which are in the i column marginals of the estimates table divided by (the product of corresponding population numbers for each racial group, which are in the i column marginals of the population table and the number (k=3) of column marginals (racial groups)))

and

(2) JMALPE is equal to GMALPE +

the sum of: (the estimates by the 12 race-geography groups in the estimates table, divided by (the product of the corresponding population numbers for each of the 12 race-geographic groups of the population table and the number (m=12) of cells comprising the 12 race-geographic groups))). -

(the sum of: (the estimates by geography, which are in the j row marginals of the estimates table divided by (the product of the corresponding population numbers for each geographic group in the j row marginals of the population table and the number (L=4) of row marginals (geographic groups)))

As a way of summarizing these results to-date, here is my answer to someone who asks why the MALPEs in Table 5 are different and, given these differences, how they are related to one another:

“Let me answer this question by looking first at JMALPE, which is .0136. One might think that it is an average of RMALPE (-.0069) and GMALPE (.0051), but it is not. The relationship between, on the one hand, RMALPE and GMALPE, and, on the other, JMALPE is a bit more complicated in that it involves the implicit “weights” stemming from the estimates and populations for the 3 race groups and the 4 geographic areas, which together translate into 12 (4 *3) race-geography groups. We have to take into account these groupings as well as the estimates and actual populations within them.

Once we do this, we can express JMALPE as a function both of RMALPE and of GMALPE, as follows.

$$JMALPE = RMALPE + [\sum(E_{ij}/(12*P_{ij}))] - [\sum(E_i/(3*P_i))]$$

$$0.0136 = -0.0069 + 1.013636434 - 0.993078224$$

$$JMALPE = GMALPE + [\sum(E_{ij}/(12*P_{ij}))] - [\sum(E_j/(4*P_j))]$$

$$0.0136 = 0.0051 + 1.013636434 - 1.005086696$$

Thus, JMALPE is found by adding to RMALPE the sum of: (the estimates by the 12 race-geography groups in the estimates table, divided by (the product of the corresponding population numbers for each of the 12 race- geographic groups of the population table and the number ($m=12$) of cells comprising the 12 race-geographic groups))) and then subtracting from RMALPE the sum of: (the estimates by racial group, which are in the i column marginals of the estimates table divided by (the product of corresponding population numbers for each racial group, which are in the i column marginals of the population table and the number ($k=3$) of column marginals (racial groups))).

In terms of GMALPE, JMALPE is found by adding to GMALPE the sum of: (the estimates by the 12 race-geography groups in the estimates table, divided by (the product of the corresponding population numbers for each of the 12 race- geographic groups of the population table and the number ($m=12$) of cells comprising the 12 race-geographic groups))) and then subtracting from GMALPE the sum of: (the estimates by geography, which are in the j row marginals of the estimates table divided by (the product of the corresponding population numbers for each geographic group in the j row marginals of the population table and the number ($L=4$) of row marginals (geographic groups))).

RMALPE is related to GMALPE by adding to GMALPE, the sum of: (the estimates by geography, which are in the j row marginals of the estimates table divided by (the product of the corresponding population numbers for each geographic group in the j row marginals of the population table and the number (L) of row marginals (geographic groups))) and subtracting from RMALPE the sum of: (the estimates by racial group, which are in the i column marginals of the estimates table divided by (the product of corresponding population numbers for each racial group, which are in the i column marginals of the population table and the number (k) of column marginals (racial groups))).

Finally, GMALPE is related to RMALPE as follows. To get GMALPE we would add to RMALPE, the sum of: (the estimates by geography, which are in the j row marginals of the estimates table divided by (the product of the corresponding population numbers for each geographic group in the j row marginals of the population table and the number (L) of row marginals (geographic groups))) and then subtract from RMALP, the sum of: (the estimates by racial group, which are in the i column marginals of the estimates table divided by (the product of corresponding population numbers for each racial group, which are in the i column marginals of the population table and the number (k) of column marginals (racial groups))).

IV. Some Concluding Thoughts

It appears that in two dimensions (e.g., race and geography), we can answer the question: "if one was not only doing ex post facto evaluations of estimates (or projections) of the total population but by characteristics such as race and geography, how would one summarize the summary measures of error?" The paper has examined this issue in two dimensions using a standard summary measure of ex post facto error "Mean Algebraic Percent Error" (MALPE). The paper finds that relationships do exist not only between the MALPEs on two dimensions, but between them and their "joint" (overall) MALPE. offers proofs of these relationships both mathematically and empirically. Like most findings, these suggest additional research.

1. The "weighting" of MALPE shares some features with a common demographic technique, "direct standardization." However, it is not the same. The common features may, however, yield a readily interpretable "decomposition" of the difference between MALPE on one dimension (e.g., Race) and on another (e.g., geography) for the same set of estimates/forecasts (Kitagawa, 1955; Das Gupta, 1978).
2. It is intriguing that WMALPE, which is, in fact, a weighted mean, ends up being the same as the "overall percent error." In the hypothetical population, the overall percent error is $((E - P)/P) = (1320 - 1369)/1369 = -49/1369 = -0.0358$. This is the same result we get with WMALPE (the weighted mean percent error).
3. What about MAPE and, especially, Root Mean Square Error (RMSE), both other common summary measure of error? Exploring RMSE may prove fruitful because it uses Sum of Squared Errors, a concept ubiquitous in statistics and an important tool in the analysis of "error."
4. Can the relationships for MALPE be extended to more than two dimensions? The results suggest that this will be the case, but both mathematical proofs and corresponding empirical examples are needed.

Endnotes

1. This paper stems from a question George Hough asked me, which was "if one was not only doing ex post facto evaluations of estimates (or projections) of the total population but by characteristics such as race and geography, how would one summarize the summary measures of error (e.g., MALPE by Race, MALPE by geography and MALPE for the total population)?" This question immediately got my attention and I contacted demographers I knew who were experienced in doing ex post evaluations of estimates and projections. Their responses were similar to mine, not only had not one of them ever thought of it, but also none of them knew of anybody who had dealt with this issue. In addition to the question posed by George Hough, others that have provided helpful

feedback on earlier drafts of this paper include, Jack Baker, Matt Kaneshiro, Stan Smith, Richard Verdugo, Paul Voss, and Paul Wilson,

References

Das Gupta, P. 1978. "A general method of decomposing a difference between two rates into several components." *Demography* 15(1): 99-112.

Kitagawa, E. 1955. "Components of a difference between two rates." *Journal of the American Statistical Association* 50 (272): 1168-1194.

APPENDIX A.

$$\text{PROOF: MALPE} = \text{WMALPE} + [\sum((E_i/(k*P_i)))] - [(\sum E_i)/(\sum P_i)]$$

We know from Equation [1.e] that

$$\text{MALPE} = [\sum((E_i)/(k*P_i))] - 1 \quad [\text{A.1}]$$

And from Equation [2.d] we know that

$$\text{WMALPE} = [\sum E_i/\sum P_i] - 1 \quad [\text{A.2}]$$

Since we want to know how WMALPE is related to MALPE, let

$$\text{WMALPE} = \text{MALPE} + X \quad [\text{A.3}]$$

Which is equal to

$$[\sum E_i/\sum P_i] - 1 = [\sum((E_i)/(k*P_i))] - 1 + X \quad [\text{A.3.1}]$$

Rearranging terms to isolate X, we get

$$[\sum E_i/\sum P_i] - 1 - [\sum((E_i)/(k*P_i))] + 1 = X \quad [\text{A.3.2}]$$

Which reduces to

$$[\sum E_i/\sum P_i] - [\sum((E_i)/(k*P_i))] = X \quad [\text{A.3.3}]$$

And substituting X into A.3 we get

$$\text{WMALPE} = \text{MALPE} + [\sum E_i/\sum P_i] - [\sum((E_i)/(k*P_i))] \quad [\text{A.4}]$$

Which means that

$$\text{MALPE} = \text{WMALPE} + [\sum((E_i)/(k*P_i))] - [\sum E_i/\sum P_i] \quad [\text{A.5}]$$

And we have our proof that

$$\text{MALPE} = \text{WMALPE} + [\sum((E_i)/(k*P_i))] - [(\sum E_i)/(\sum P_i)]$$

APPENDIX B.

$$\text{PROOF: } \text{JMALPE} = \text{WMALPE} - [(\sum E_{ij})/(\sum P_{ij})] + [\sum (E_{ij}/(m * P_{ij}))]$$

This is for only two dimensions, i and j, with i=1 through k and j=1 through L

Since we know from equation [1.b] that on any given single dimension (e.g., i=1 through k) that

$$\text{MALPE} = [1/(\sum P_i/k)/\sum P_i] * [\sum (E_i - P_i)/P_i] \quad [\text{B.1}]$$

And from Appendix A, we know that

$$\text{WMALPE} = \text{MALPE} + [(\sum E_i)/(\sum P_i)] - [\sum (E_i/k P_i)] \quad [\text{B.2}]$$

Since

$$\text{MALPE} = [\sum ((E_i)/(k * P_i))] - 1 \quad [\text{B.3}]$$

and

$$\text{WMALPE} = [\sum E_i/\sum P_i] - 1 \quad [\text{B.4}]$$

We can generalize [B.1] to two dimensions

$$\text{MALPE} = [1/(\sum P_{ij}/k * L)/\sum P_{ij}] * [\sum (E_{ij} - P_{ij})/P_{ij}] \quad [\text{B.5}]$$

Let JMALPE stand for a MALPE on 2 dimensions, i and j and $m = k * L$, so we can re-express [B.5] as

$$\text{JMALPE} = [1/(\sum P_{ij}/m)/\sum P_{ij}] * [\sum (E_{ij} - P_{ij})/P_{ij}] \quad [\text{B.6}]$$

Using Equations [1.a] through [1.e] in conjunction with [B.6], we have

$$\text{Since } (1/m) = [1/(\sum P_{ij}/m)/\sum P_{ij}] \quad [\text{B.6.1}]$$

$$\text{JMALPE} = [1/(\sum P_{ij}/m)/\sum P_{ij}] * [\sum (E_{ij} - P_{ij})/P_{ij}] \quad [\text{B.6.2}]$$

$$= [1/(\sum (P_{ij}/m)/\sum P_{ij})] * [\sum (E_{ij} - P_{ij})/P_{ij}] \quad [\text{B.6.3}]$$

$$= [\sum((E_{ij}-P_{ij})/(m*P_{ij}))] \quad [B.6.4]$$

$$= [\sum((E_{ij})/(m*P_{ij}))] - 1 \quad [B.6.5]$$

$$WJMALPE = \sum [(P_{ij}/\sum P_{ij})*((E_{ij}-P_{ij})/P_{ij})] \quad [B.7]$$

NOTE: WJMALPE also equals WMALPE on any dimension (i or j)

And since

$$WJMALPE = \sum [(P_{ij}/\sum P_{ij})*((E_{ij}-P_{ij})/P_{ij})] \quad [B.7.1]$$

$$= \sum[E_{ij}-P_{ij}]/\sum P_{ij} \quad [B.7.2]$$

$$= \sum E_{ij}/\sum P_{ij} - \sum P_{ij}/\sum P_{ij} \quad [B.7.3]$$

$$= [\sum E_{ij}/\sum P_{ij}] - 1 \quad [B.7.4]$$

Thus, we have

$$JMALPE = [\sum((E_{ij})/(m*P_{ij}))] - 1 \quad [B.8]$$

$$\text{and } 1 + JMALPE = [\sum((E_{ij})/(m*P_{ij}))] \quad [B.8.1]$$

$$WJMALPE = [(\sum E_{ij})/(\sum P_{ij})] - 1 \quad [B.9]$$

$$\text{and } 1 + JWJMALPE = [(\sum E_{ij})/(\sum P_{ij})] \quad [B.9.1]$$

Since we want to know how WJMALPE is related to JMALPE, let

$$WJMALPE = JMALPE + X \quad [B.10]$$

Which is equal to

$$[\sum E_{ij}/\sum P_{ij}] - 1 = [\sum((E_{ij})/(m*P_{ij}))] - 1 + X \quad [B.10.1]$$

Rearranging terms to isolate X, we get

$$[\sum E_{ij}/\sum P_{ij}] - 1 - [\sum((E_{ij})/(m*P_{ij}))] + 1 = X \quad [B.10.2]$$

Which reduces to

$$[\sum E_{ij}/\sum P_{ij}] - [\sum((E_{ij})/(m*P_{ij}))] = X \quad [B.10.3]$$

And substituting X into [B.10.3] we get

$$WJMALPE = JMALPE + [\sum E_{ij}/\sum P_{ij}] - [\sum((E_{ij})/(m*P_{ij}))] \quad [B.11]$$

Which means that

$$JMALPE = WJMALPE + [\sum((E_{ij})/(m*P_{ij}))] - [\sum E_{ij}/\sum P_{ij}] \quad [B.11.1]$$

And we have our proof that

$$JMALPE = WJMALPE + [\sum(E_{ij}/(m*P_{ij}))] - [(\sum E_{ij})/(\sum P_{ij})]$$

