

Analysis of the Grip Strength Data Using Anti-Diagonal Symmetry Models

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Abstract

For the analysis of square contingency tables with the same row and column ordinal classifications, this article proposes new models which indicate the structures of symmetry with respect to the anti-diagonal of the table. Also, this article gives a simple decomposition in 3×3 contingency table using the proposed models. The proposed models are applied to grip strength data.

Keywords

Anti-Diagonal, Decomposition, Grip Strength Data, Square Contingency Table, Symmetry

1. Introduction

Consider the data in **Table 1**. **Table 1** is the data of grip strength of 805 male examinees aged 15 - 18 at high schools in Japan, which visited Tokyo University of Science, Open Campus, August, examined in 2011-2015. In **Table 1** the row variable is the right hand muscle strength level and the column variable is the left hand muscle strength level. The category in **Table 1** means muscle strength level compared with other people of one's age and sex. Generally, for such data with similar classifications, many observations tend to fall (or near) the main diagonal cells. For the data in **Table 1**, 73% of observations concentrate in the main diagonal. Thus, the independence between classifications is unlikely to hold. Therefore, we are interested in whether or not there is a structure of symmetry with respect to the main diagonal in the table.

For the analysis of an $r \times r$ square contingency table with the same ordinal row and column classifications, let p_{ij} denote the probability that an observation will fall in the i th row and j th column of the table ($i = 1, \dots, r; j = 1, \dots, r$). Bowker [1] proposed the symmetry model, defined by

$$p_{ij} = \psi_{ij} \quad (i = 1, \dots, r; j = 1, \dots, r),$$

where $\psi_{ij} = \psi_{ji}$ (see also Martin and Pardo [2]; Kolassa and Bhagavatula [3]; Tahata and Tomizawa [4]). This

Table 1. Grip strength test of 805 male examinees aged 15 - 18 at high schools in Japan, examined in 2011-2015. (The parenthesized values are MLEs of expected frequencies under the AMH model).

Right hand grip strength level	Left hand grip strength level			Total
	Excellent (1)	Good (2)	Poor (3)	
Excellent (1)	74 (77.00)	89 (85.90)	3 (3.00)	166 (165.90)
Good (2)	10 (10.81)	444 (444.00)	93 (96.48)	547 (551.29)
Poor (3)	0 (0.00)	23 (21.39)	69 (66.41)	92 (87.81)
Total	84 (87.81)	556 (551.29)	165 (165.90)	805 (805.00)

model states that the probability that an observation will fall in the (i,j) th cell of the table is equal to the probability that it falls in the (j,i) th cell. Namely, this model describes a structure of symmetry with respect to the main diagonal of the table. Stuart [5] proposed the marginal homogeneity model, defined by

$$p_i = p_{\cdot i} \quad (i = 1, \dots, r),$$

where $p_i = \sum_{t=1}^r p_{it}$ and $p_{\cdot i} = \sum_{s=1}^r p_{si}$. This model states that the row marginal distribution is identical to the column marginal distribution. Read [6] considered the global symmetry model, defined by

$$\sum_{i < j} p_{ij} = \sum_{i > j} p_{ij}.$$

This model states that the probability that an observation will fall in one of the upper-right triangle cells above the main diagonal of the table is equal to the probability that it falls in one of the lower-left triangle cells below the main diagonal.

For the data in **Table 1**, we see that many observations fall in the upper-right triangle cells above the main diagonal. Thus, the models for symmetry between classifications are unlikely to hold. Then, the symmetry with respect to the anti-diagonal may hold for the data in **Table 1**. Note that the probabilities for the anti-diagonal cells are $p_{1r}, p_{2,r-1}, \dots, p_{r1}$ for the $r \times r$ table. When the number of the categories is 3, *i.e.*, $r = 3$, (such as the data in **Table 1**), the anti-diagonal cells are p_{13}, p_{22} and p_{31} . Thus, we are interested in proposing new models for symmetry with respect to the anti-diagonal, which would hold for the data in **Table 1**.

The present paper proposes three models and gives a simple decomposition using the proposed models in 3×3 contingency table. Also it illustrates new models with the grip strength data in **Table 1**.

2. New Models and a Simple Decomposition

Firstly, we propose a model defined by

$$p_{ij} = \psi_{ij} \quad (i = 1, \dots, r; j = 1, \dots, r),$$

where $\psi_{ij} = \psi_{j^*i^*}$. The symbol “*” denotes $i^* = r + 1 - i$. This model states that the probability that an observation will fall in the (i,j) th cell of the table is equal to the probability that it falls in the (j^*,i^*) th cell. Namely, this model indicates the structure of symmetry with respect to the anti-diagonal of the table. We shall refer to this model as the anti-diagonal symmetry (AS) model. Note that the AS model is a special case of the reverse conditional symmetry model, proposed by Tomizawa [7].

Secondly, we propose a model defined by

$$\sum_{i+j < r+1} p_{ij} = \sum_{i+j > r+1} p_{ij}.$$

Let X and Y denote the row and column variables, respectively. Then, this model is also expressed as

$$\Pr(X + Y < r + 1) = \Pr(X + Y > r + 1).$$

We shall refer to this model as the anti-diagonal global symmetry (AGS) model. Finally, we propose a model defined by

$$p_i = p_{i^*} \quad (i = 1, \dots, r).$$

This model states that the row marginal distribution is identical to the column marginal distribution in reverse order. We shall refer to this model as the anti-diagonal marginal homogeneity (AMH) model.

We obtain the following theorem.

Theorem 1. *When $r = 3$, the AS model holds if and only if both the AGS and AMH models hold.*

Proof. If the AS model holds, then the AGS and AMH models hold. Assuming that both the AGS and AMH models hold, then we shall show that the AS model holds. If the AMH and AGS models hold, then we have $p_1 = p_3$ (i.e., $p_{11} + p_{12} = p_{23} + p_{33}$), $p_2 = p_2$ (i.e., $p_{21} + p_{23} = p_{12} + p_{32}$) and $p_{11} + p_{12} + p_{21} = p_{23} + p_{32} + p_{33}$. Thus, we see $p_{11} = p_{33}$, $p_{12} = p_{23}$ and $p_{21} = p_{32}$. Namely, the AS model holds. The proof is completed.

Note that this theorem does not hold when $r \geq 4$. Let n_{ij} denote the observed frequency in the (i, j) th cell of the table $(i = 1, \dots, r; j = 1, \dots, r)$ with $n = \sum \sum n_{ij}$, and let m_{ij} denote the corresponding expected frequency. Assume that $\{n_{ij}\}$ have a multinomial distribution. The maximum likelihood estimates (MLEs) of $\{m_{ij}\}$ under the AS and AGS model, are expressed as the closed-forms as follows:

1) The MLE of m_{ij} under the AS model is

$$\hat{m}_{ij} = \begin{cases} \frac{n_{ij} + n_{j^*i^*}}{2} & (i \neq j^*) \\ n_{ij} & (i = j^*). \end{cases}$$

2) The MLE of m_{ij} under the AGS model is

$$\hat{m}_{ij} = \begin{cases} \frac{L+R}{2L} n_{ij} & (i \neq j^*, i+j < r+1) \\ \frac{L+R}{2R} n_{ij} & (i \neq j^*, i+j > r+1) \\ n_{ij} & (i = j^*), \end{cases}$$

where

$$L = \sum_{i+j < r+1} \sum n_{ij}, \quad R = \sum_{i+j > r+1} \sum n_{ij}.$$

The MLEs of $\{m_{ij}\}$ under the AMH model could be obtained using the Newton-Raphson method in the log-likelihood equation. Let $G^2(M)$ denote the likelihood ratio chi-squared statistic for testing goodness-of-fit of model M . For the AS model, $\{p_{ij}\}$ are determined by $r(r-1)/2$ of $\{\psi_{ij}, i+j < r+1\}$, $r-1$ of $\{\psi_{ii^*}\}$ for anti-diagonal cells (since $\sum \sum p_{ij} = 1$), thus a total of $(r+2)(r-1)/2$. Therefore, the number of degrees of freedom (df) for testing goodness-of-fit of the AS model is $(r^2 - 1) - (r+2)(r-1)/2 = r(r-1)/2$. Similarly, the numbers of df for testing goodness-of-fit of the AGS and AMH model are 1 and $r-1$, respectively. Note that when $r \geq 4$, the number of df for the AS model is greater than the sum of numbers of df for the AGS and AMH models, and when $r = 3$, it is equal to the sum of them.

We shall consider the comparison between two nested models. Suppose that model M_2 is a special case of model M_1 ; that is, M_2 is simpler than M_1 , so when M_2 holds, necessarily M_1 also holds. For testing that model M_2 holds assuming that model M_1 holds, we can use the likelihood ratio statistic $G^2(M_2 | M_1)$ which is the difference between the $G^2(M_2)$ and $G^2(M_1)$. When model M_1 holds, this statistic has an asymptotic chi-squared distribution with df being equal to the difference between the df for M_2 and for M_1 .

3. An Example

Consider the data in **Table 1** again. All the AS, AGS and AMH models fit these data well, yielding the likelihood

ratio statistic $G^2(AS) = 5.53$ with 3 df, $G^2(AGS) = 0.40$ with 1 df, and $G^2(AMH) = 0.64$ with 2 df, respectively. Since the AS model is a special case of the AGS model, we shall test the hypothesis that the AS model holds assuming that the AGS model holds. Since $G^2(AS | AGS) = G^2(AS) - G^2(AGS) = 5.13$ with 2 df being the difference between the numbers of df for the AS and AGS models, this hypothesis is accepted at the 0.05 significance level. Thus, the AS model would be preferable to the AGS model. Similarly, since the AS model is a special case of the AMH model, we shall test the hypothesis that the AS model holds assuming that the AMH model holds. Since $G^2(AS | AMH) = G^2(AS) - G^2(AMH) = 4.89$ with 1 df being the difference between the numbers of df for the AS and AMH models, this hypothesis is rejected at the 0.05 significance level. Therefore, the AMH model would be preferable to the AS model.

Under the AMH model, the probability that an examinee's right hand grip strength level is "Excellent (1)", is estimated to be equal to the probability that an another examinee's left hand grip strength level is "Poor (3)". Also, the probability that an examinee's right hand grip strength level is "Poor (3)", is estimated to be equal to the probability that an examinee's left hand grip strength level is "Excellent (1)".

4. Concluding Remarks

The decomposition of the AS model into the AGS and AMH models, given by Theorem 1, would be useful for seeing the reason for its poor fit when the AS model fits the 3×3 data poorly, and it should be considered for ordinal categorical data because all the AS, AGS and AMH models are not invariant under arbitrary same permutations of row and column categories.

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References

- [1] Bowker, A.H. (1948) A Test for Symmetry in Contingency Tables. *Journal of the American Statistical Association*, **43**, 572-574. <http://dx.doi.org/10.1080/01621459.1948.10483284>
- [2] Martin, N. and Pardo, L. (2010) A New Measure of Leverage Cells in Multinomial Loglinear Models. *Communications in Statistics - Theory and Methods*, **39**, 517-530. <http://dx.doi.org/10.1080/03610920903139991>
- [3] Kolassa, J.E. and Bhagavatula, H.G. (2012) Accurate Approximations to the Distribution of a Statistic Testing Symmetry in Contingency Tables. *Institute of Mathematical Statistics*, **8**, 181-189. <http://dx.doi.org/10.1214/11-imscol1812>
- [4] Tahata, K. and Tomizawa, S. (2014) Symmetry and Asymmetry Models and Decompositions of Models for Contingency Tables. *SUT Journal of Mathematics*, **50**, 131-165.
- [5] Stuart, A. (1955) A Test for Homogeneity of the Marginal Distributions in a Two-Way Classification. *Biometrika*, **42**, 412-416. <http://dx.doi.org/10.1093/biomet/42.3-4.412>
- [6] Read, C.B. (1977) Partitioning Chi-Square in Contingency Table: A Teaching Approach. *Communications in Statistics-Theory and Methods*, **6**, 553-562. <http://dx.doi.org/10.1080/03610927708827513>
- [7] Tomizawa, S. (1986) Four Kinds of Symmetry Models and Their Decompositions in a Square Contingency Table with Ordered Categories. *Biometrical Journal*, **28**, 387-393. <http://dx.doi.org/10.1002/bimj.4710280402>



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