

# The Average Odds Ratios as a Technique for Categorical Simplification

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# Introduction

- Mobility tables express the association of fathers and children in terms of social class.
- This association is considered an index of equality of opportunities (post facto).
- Open societies vs Closed Societies: More Fluid vs More Rigid.
- Log-linear models were adopted with marginal assumptions and log-multiplicative interaction structure.

## Uniform Association Model: Unidiff Equation

The unidiff model, formulated by Xie and Goldthorpe & Erikson (1992):

$$\log F_{ijk} = \mu + \lambda_i^R + \lambda_j^C + \lambda_k^C + \lambda_{ik}^{RL} + \lambda_{jk}^{RL} + \beta_{jk}^{RL}$$

Where:

- $F_{ijk}$ : Expected frequency in cell  $(i, j, k)$ .
- $\mu$ : Overall mean.
- $\lambda_i^R$ : Row marginal effect.
- $\lambda_j^C$ : Column marginal effect.

## Uniform Association Model: Unidiff Equation (cont.)

- $\lambda_k^C$ : Layer marginal effect.
- $\lambda_{ik}^{RL}$ : Row-layer interaction.
- $\lambda_{jk}^{RL}$ : Column-layer interaction.
- $\beta_k$ : Strength by layer.
- $\phi_{ij}$ : Association parameter.

**Stata users:** Two ados: Unidiff Pisati & Udiff Ben Jann. 20 m.

## RC-II Model

The RC-II model, formulated by Goodman:

$$\log F_{ij} = \lambda + \lambda_i + \lambda_j + \sum_{m=1}^M \phi_m \mu_m \nu_{im}$$

Where:

- $F_{ij}$ : Expected frequency for cell  $(i, j)$  in the contingency table.
- $\lambda$ : Overall constant.
- $\lambda_i$ : Main effect for row  $i$ .
- $\lambda_j$ : Main effect for column  $j$ .

## RC-II Model (cont.)

- $\phi_m \mu_m \nu_{im}$ : Interaction effects, with  $M$  being the number of latent dimensions,  $\mu_m$  the latent scores for row  $i$ , and  $\nu_{jm}$  the latent scores for column  $j$ .

**STATA:** No ados. Extremely difficult estimation. For one table: John Hendrickx.

## Mutual Information Index

Based on information theory principles:

$$I(X_i; Y) = \sum_i \sum_j p_{ij} \log \left( \frac{p_{ij}}{p_i p_j} \right)$$

Where:

- $I(X_i; Y)$ : Mutual Information Index between variables  $X$  and  $Y$ .
- $p_{ij}$ : Joint probability of being in cell  $(i, j)$ .
- $p_i$ : Marginal probability of row  $i$ .
- $p_j$ : Marginal probability of column  $j$ .

**Stata Users:** Mindex Ado by Ben Jann. 6/10.

## Average Odds Ratios: AGOR

Uses log-linear models (e.g., mlogit) to compute Average Global Odds Ratios (AGOR):

$$\text{AGOR} = \frac{1}{(I-1)(J-1)} \sum_{i=1}^{I-1} \sum_{j=1}^{J-1} \frac{\left( \frac{F_{ij} F_{i+1,j+1}}{F_{i,j+1} F_{i+1,j}} \right)}{\left( \frac{F_{ij} F_{i+1,j+1}}{F_{i,j+1} F_{i+1,j}} \right)}$$

Where:

- AGOR = Average Global Odds Ratio.
- $I$ : Number of rows in the mobility table.
- $J$ : Number of columns in the mobility table.

## Average Odds Ratios: AGOR (cont.)

- $F_{ij}$ : Observed frequency in cell  $(i, j)$ .
- $F_{i+1,j+1}$ : Odds ratio for the  $2 \times 2$  subtable formed by cells  $(i, j), (i, j + 1), (i + 1, j)$ , and  $(i + 1, j + 1)$ .

Average Global ORs calculated for different mobility scenarios. Mean AGOR for Clase Trabajadora vs. Clase de Servicio: 1.416 (SE: [incomplete]).

## Alternative Average Odds Ratios

An alternative formulation using a simple mean of odds ratios:

$$\theta_{ij} = \left( \frac{F_{ij} F_{i+1,j+1}}{F_{i,j+1} F_{i+1,j}} \right), \quad \bar{\theta} = \frac{1}{N} \sum_{i=1}^{I-1} \sum_{j=1}^{J-1} \theta_{ij}$$

When we have two means of odds ratios, we can subject them to a t-test if we have the errors.

# Odds Ratios Comparison

**Table:** Odds Ratios Comparison Between Countries

<b>Statistic</b>	<b>Value</b>
Mean OR for country 1	2.4119868
Mean OR for country 2	2.0455212
Difference	0.36646562
Standard Error of difference	0.17482364
t-statistic	2.0962018
p-value	0.03606429

## Stata Program: Setup

```
1 * Stata Meeting, Sevilla (2025)
2 * Average Odds Ratios
3
4 * Load the dataset from the specified path
5 use "C:\Users\imarq\Downloads\Cyprus_2019.dta"
```

## Stata Program: Country 1 Analysis (1/3)

```
1 * COUNTRY 1 ANALYSIS
2
3 * 1. Run multinomial logit model
4 * This fits a multinomial logistic regression model with
5 * egp3 as the dependent variable
6 * and egp3f as a categorical predictor, restricted to
7 * country==1, using outcome 1 as base
8 mlogit egp3 i.egp3f if country==1, baseoutcome(1)
9 matrix b = e(b) * Stores the model coefficients in
10 matrix b
11
12 * 2. Define outcomes and categories
13 * Defines local macros for equation names and category
14 * numbers
15 local eqs `"" "IIIab" "Clases_trabajadoras__IVabc_V_VI_"
16 " "
```

## Stata Program: Country 1 Analysis (2/3)

```
1 * 3. Calculate Odds Ratios
2 * Loops through equations and categories to compute odds
   ratios using exponentiation
3 local i = 1
4 foreach eq of local eqs {
5     foreach cat of local cats {
6         scalar or`i' = exp(_b[`eq':`cat'].e gp3f])
7         local i = `i' + 1
8     }
9 }
10
11 * 4. Calculate mean OR
12 * Computes the average of the four odds ratios
13 scalar sum_or = or1 + or2 + or3 + or4
14 scalar mean_or = sum_or / 4
```

## Stata Program: Country 1 Analysis (3/3)

```
1 * 6. Calculate SE of mean OR
2 * Computes variance and standard error for the mean odds
   ratio
3 scalar se_or1 = or1 * se_1
4 scalar se_or2 = or2 * se_2
5 scalar se_or3 = or3 * se_3
6 scalar se_or4 = or4 * se_4
7
8 scalar var_or1 = se_or1^2
9 scalar var_or2 = se_or2^2
10 scalar var_or3 = se_or3^2
11 scalar var_or4 = se_or4^2
12
13 scalar var_mean_or = (var_or1 + var_or2 + var_or3 +
   var_or4) / (4^2)
14 scalar se_mean_or = sqrt(var_mean_or)
```

## Stata Program: Country 2 Analysis (1/3)

```
1 * COUNTRY 2 ANALYSIS
2
3 * 1. Run multinomial logit model
4 * Fits the model for country 2
5 mlogit egp3 i.egp3f if country==2, baseoutcome(1)
6 matrix b = e(b)
7
8 * 2. Define outcomes and categories
9 * Same categories as country 1
10 local eqs `""' "IIIab" "Clases_trabajadoras__IVabc_V_VI_"
11      `"'
12
13 local cats "2 3"
14
15 * 3. Calculate Odds Ratios
16 * Computes odds ratios for country 2
17 local i = 1
```

## Stata Program: Country 2 Analysis (2/3)

```
1 * 4. Calculate mean OR
2 * Computes the average odds ratio for country 2
3 scalar sum_or2 = or2_1 + or2_2 + or2_3 + or2_4
4 scalar mean_or2 = sum_or2 / 4
5
6 * 5. Calculate standard errors
7 * Extracts standard errors for country 2 coefficients
8 local i = 1
9 foreach eq of local eqs {
10     foreach cat of local cats {
11         scalar se2_`i' = _se[`eq':`cat'].egp3f]
12         local i = `i' + 1
13     }
14 }
15
16 * 6. Calculate SE of mean OR
```

## Stata Program: Comparison Between Countries

```
1 * COMPARISON BETWEEN COUNTRIES
2 * Calculates the difference in means and performs a
   t-test
3 scalar diff_means = mean_or - mean_or2
4 scalar se_diff = sqrt(se_mean_or^2 + se_mean_or2^2)
5 scalar t_stat = diff_means/se_diff
6 scalar p_value = 2*(1-normal(abs(t_stat)))
7
8 * DISPLAY KEY RESULTS
9 * Displays the comparison results
10 display "Mean OR for country 1: " mean_or
11 display "Mean OR for country 2: " mean_or2
12 display "Difference: " diff_means
13 display "Standard Error of difference: " se_diff
14 display "t-statistic: " t_stat
15 display "p-value: " p_value
```