

Online appendix

More sensitivity analysis

This is a continuation of the sensitivity analysis of section 10 when $\rho = 0.5$ and $\rho = 0.9$. If you compare these Tables with Tables 4 (p. on page 12) and 5 (p. on page 12) you will find few differences.

Table 9: Sensitivity analysis. True distribution centered on the uniform distribution ($\rho = 0.9$).

Variability		Coefficient				
Guessing*	True*	Cohen-Fleiss	Fleiss	Cohen	BP	Cohen-BP
None	None	0.0e+00	0.0e+00	0.0e+00	1.1e-02	2.2e-02
	Low	2.2e-02	2.2e-02	2.2e-02	1.2e-02	8.3e-02
	High	3.3e-01	3.3e-01	3.3e-01	1.8e-02	4.6e-01
Low	None	2.8e-05	3.4e-04	2.4e-04	1.4e-02	3.1e-02
	Low	2.3e-02	2.3e-02	2.3e-02	1.4e-02	9.0e-02
	High	3.4e-01	3.4e-01	3.4e-01	2.2e-02	4.7e-01
High	None	2.3e-04	2.8e-03	1.9e-03	3.1e-02	7.1e-02
	Low	2.4e-02	2.7e-02	2.6e-02	3.1e-02	1.3e-01
	High	3.4e-01	3.5e-01	3.4e-01	3.7e-02	4.7e-01

*Variability of the true distributions: Baseline: True distribution is uniform.

*Variability of the guessing distributions. Baseline: All guessing distributions are equal to the true distribution.

Table 10: Sensitivity analysis. True distribution centered on the marginal guessing distribution ($\rho = 0.9$).

Variability		Coefficient				
Guessing*	True*	Cohen-Fleiss	Fleiss	Cohen	BP	Cohen-BP
None	None	0.0e+00	0.0e+00	0.0e+00	0.0e+00	0.0e+00
	Low	3.9e-03	4.0e-03	3.9e-03	0.0e+00	1.2e-02
	High	7.7e-02	7.8e-02	7.7e-02	0.0e+00	1.7e-01
Low	None	5.5e-05	4.4e-05	2.8e-05	4.4e-05	0.0e+00
	Low	3.7e-03	3.8e-03	3.8e-03	7.3e-04	1.2e-02
	High	7.5e-02	7.6e-02	7.5e-02	2.5e-03	1.7e-01
High	None	8.3e-04	5.9e-04	4.0e-04	6.3e-04	0.0e+00
	Low	3.6e-03	4.5e-03	4.1e-03	2.8e-03	1.2e-02
	High	7.4e-02	7.6e-02	7.5e-02	1.1e-02	1.7e-01

*Variability of the true distributions: Baseline: True distribution equals the marginal guessing distribution.

*Variability of the guessing distributions. Baseline: All guessing distributions are equal.

Exact variance of \hat{p}_a and $\hat{\nu}_{BP}$

In this section we derive the exact, non-asymptotic variance of \hat{p}_a and the Brennan-Prediger coefficient $\hat{\nu}_{BP}$.

Proposition 9. *The variance of*

$$\hat{p}_a = n^{-1} \sum_{i=1}^n \binom{R}{2}^{-1} \sum_{r>r'} 1[X_{ir} = X_{ir'}]$$

is $\text{Var } \hat{p}_a = n^{-1}(p_{2a} - p_a^2)$. Here

$$p_{2a} = \binom{R}{2}^{-2} \sum_{r_1>r'_1} \sum_{r_2>r'_2} \mu_{r_1 r'_1 r_2 r'_2},$$

and

$$\mu_{r_1 r'_1 r_2 r'_2} = P(X_{ir_1} = X_{ir'_1}, X_{ir_2} = X_{ir'_2}).$$

Table 11: Sensitivity analysis. True distribution centered on the uniform distribution ($\rho = 0.5$).

Variability		Coefficient				
Guessing*	True*	Cohen–Fleiss	Fleiss	Cohen	BP	Cohen–BP
None	None	0.0e+00	0.0e+00	0.0e+00	1.1e-02	2.3e-02
	Low	2.2e-02	2.2e-02	2.2e-02	1.2e-02	8.2e-02
	High	3.4e-01	3.4e-01	3.4e-01	1.9e-02	4.6e-01
Low	None	0.0e+00	3.9e-04	2.5e-04	1.5e-02	3.2e-02
	Low	2.3e-02	2.3e-02	2.3e-02	1.6e-02	9.2e-02
	High	3.3e-01	3.4e-01	3.4e-01	2.3e-02	4.6e-01
High	None	0.0e+00	3.0e-03	1.9e-03	3.8e-02	8.8e-02
	Low	2.4e-02	2.7e-02	2.6e-02	4.0e-02	1.4e-01
	High	3.4e-01	3.5e-01	3.4e-01	4.7e-02	4.7e-01

*Variability of the true distributions: Baseline: True distribution is uniform.

*Variability of the guessing distributions. Baseline: All guessing distributions are equal to the true distribution.

Table 12: Sensitivity analysis. True distribution centered on the marginal guessing distribution ($\rho = 0.5$).

Variability		Coefficient				
Guessing*	True*	Cohen–Fleiss	Fleiss	Cohen	BP	Cohen–BP
None	None	0.0e+00	0.0e+00	0.0e+00	6.60e-02	1.44e-01
	Low	2.19e-01	2.19e-01	2.19e-01	4.89e-02	3.24e-01
	High	5.95e-01	5.95e-01	5.95e-01	4.88e-02	6.33e-01
Low	None	7.30e-05	3.81e-04	2.84e-04	4.90e-02	1.08e-01
	Low	2.99e-02	3.02e-02	3.01e-02	5.05e-02	1.65e-01
	High	3.47e-01	3.47e-01	3.47e-01	5.38e-02	4.83e-01
High	None	6.00e-04	3.13e-03	2.33e-03	6.48e-02	1.47e-01
	Low	3.14e-02	3.36e-02	3.28e-02	6.47e-02	1.91e-01
	High	3.47e-01	3.52e-01	3.48e-01	6.71e-02	4.93e-01

*Variability of the true distributions: Baseline: True distribution equals the marginal guessing distribution.

*Variability of the guessing distributions. Baseline: All guessing distributions are equal.

Proof. Define $\mu'_{rr} = P(X_{ir} = X_{ir'})$. Then

$$p_a = \binom{R}{2}^{-1} \sum_{r>r'} P(X_{ir} = X_{ir'}) = \binom{R}{2}^{-1} \sum_{r>r'} \mu'_{rr}$$

by definition and thus,

$$p_a^2 = \left[\binom{R}{2}^{-1} \sum_{r>r'} \mu'_{rr} \right]^2 = \binom{R}{2}^{-2} \sum_{r_1>r'_1} \sum_{r_2>r'_2} \mu_{r_1r'_1} \mu_{r_2r'_2}.$$

To find the variance we must calculate

$$\begin{aligned} E\hat{p}_a^2 &= n^{-2} \binom{R}{2}^{-2} \sum_{i_1, i_2} \sum_{r_1 > r'_1} \sum_{r_2 > r'_2} E \{1[X_{i_1 r_1} = X_{i_1 r'_1}]1[X_{i_2 r_2} = X_{i_2 r'_2}]\}, \\ &= n^{-2} \binom{R}{2}^{-2} \sum_{i_1, i_2} \sum_{r_1 > r'_1} \sum_{r_2 > r'_2} P(X_{i_1 r_1} = X_{i_1 r'_1}, X_{i_2 r_2} = X_{i_2 r'_2}). \end{aligned}$$

Looking at $P(X_{i_1 r_1} = X_{i_1 r'_1}, X_{i_2 r_2} = X_{i_2 r'_2})$, we notice that, if $i_1 \neq i_2$ it equals $P(X_{i_1 r_1} = X_{i_1 r'_1}, X_{i_2 r_2} = X_{i_2 r'_2}) = \mu_{r_1 r'_1} \mu_{r_2 r'_2}$ due to independence. This happens $n(n-1)$ times when we sum over i_1, i_2 . On the other hand, if $i_1 = i_2$, it equals $\mu_{r_1 r'_1 r_2 r'_2}$. This happens n times when we sum over i_1, i_2 .

It follows that

$$E\hat{p}_a^2 = (1 - n^{-1})p_a^2 + n^{-1} \binom{R}{2}^{-2} \sum_{r_1 > r'_1} \sum_{r_2 > r'_2} \mu_{r_1 r'_1 r_2 r'_2},$$

and the variance equals

$$\text{Var } \hat{p}_a = n^{-1}(E\hat{p}_a^2 - p_a^2) = n^{-1}(p_{2a} - p_a^2),$$

as claimed. \square

The sample Brennan–Prediger coefficient $\hat{\nu}_{BP} = (\hat{p}_a - C^{-1})/(1 - C^{-1})$ is easy to deal with, as it's a linear function of \hat{p}_a . As shown in Proposition 23, the sample agreement \hat{p}_a has exact variance $\text{Var}(\hat{p}_a) = n^{-1}(p_{2a} - p_a^2)$, where p_a is the true agreement and $p_{2a} = \binom{R}{2}^{-2} \sum_{r_1 > r'_1} \sum_{r_2 > r'_2} P(X_{i r_1} = X_{i r'_1}, X_{i r_2} = X_{i r'_2})$. The value of p_{2a} is easy to estimate using the sample estimator \hat{p}_{2a} . Moreover, the expectation of \hat{p}_{2a}^2 is $\frac{n-1}{n}p_a^2 + \frac{1}{n}p_{2a}$, hence a debiased estimator of $n^{-1}(p_{2a} - p_a^2)$ is

$$\hat{\text{Var}}\hat{p}_a = (n-1)^{-1}(\hat{p}_{2a} - \hat{p}_a^2),$$

provided $\hat{p}_{2a} > \hat{p}_a^2$. This inequality is rarely broken, even in small samples.

Using the expression for $\text{Var } \hat{p}_a$, it follows that the exact variance of the estimated Brennan–Prediger coefficient $\hat{\nu}$ is

$$\text{Var } \hat{\nu} = n^{-1}(p_{2a} - p_a^2) \frac{C^2}{(C-1)^2}, \quad (6.8)$$

and $\sqrt{n}(\hat{\nu} - \nu)$ is asymptotically normal by the delta method. A natural plug-in estimator of $\text{Var } \hat{\nu}$ is then

$$\hat{\text{Var}}\hat{\nu} = (n-1)^{-1}(\hat{p}_{2a} - \hat{p}_a^2) \frac{C^2}{(C-1)^2}$$

The usual estimator of $\text{Var } \hat{p}_a$ is $n^{-1}(n-1)^{-1} \sum (\hat{p}_a(i) - \hat{p}_a)^2$, where $\hat{p}_a(i)$ is the probability of agreement restricted to the i th item. This estimator is consistent since $\hat{p}_a = \frac{1}{n} \sum_{i=1}^n \hat{p}_a(i)$. We would recommend this estimator of the variance over the plug-in estimator for three reasons. First, it performs better than the plugin estimator in several scenarios. Second, it is always non-negative, being zero if and only if all \hat{p}_{ai} s are equal. Finally, it is considerably less computationally expensive.