## Probability of agreement

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# Background

- Correlation coefficients have a long history: Pearson, Spearman, Kendall, ....
- A new coefficient of correlation Chatterjee, S. (2020), Journal of the American Statistical Association, https:

//doi.org/10.1080/10.1080/01621459.2020.1758115.

- The Hellinger correlation Geenens, G. & de Micheaux, P. L. (2020), Journal of the American Statistical Association, https://doi.org/10.1080/01621459.2020.1791132.
- Concordance correlation coefficient (CCC) Lin. L. I. (1989). A concordance correlation coefficient to evaluate reproducibility. *Biometrics, 45, 255–268.* Lin, L., Hedayat, A. S., Sinha, B., & Yang, M. (2002). Statistical methods in assessing agreement: models, issues, and tools. *Journal of the American Statistical Association* 97, 257–270.

# Background

Pandit, V. & Schuller, B. (2020). The many-to-many mapping between the concordance correlation coefficient, and the mean square error. https://arxiv.org/abs/1902.05180:

 $MSE_1 < MSE_2$  does not imply  $CCC_1 > CCC_2$  (counterintuitive).

# Probability of agreement

Stevens, N. T., Steiner, S. H., & MacKay, R. J. (2017). Assessing agreement between two measurement systems: An alternative to the limits of agreement approach. *Statistical Methods in Medical Research* 26, 2487–2504.

The probability of agreement is related to the coverage probability (Lin *et al.*, 2002).

Model in Chan, L. K. & Mak, T. K. (1979). Likelihood estimation of a linear structural relationship with replication. *Journal of the Royal Statistical Society B* 41, 263–268 (not cited).

Stevens, N. T., Steiner, S. H., & MacKay, R. J. (2018). Comparing heteroscedastic measurement systems with the probability of agreement. *Statistical Methods in Medical Research* 27, 3420–3435.

## Extensions

1 Stevens *et al.* (2018) assume that  $M_{ijk} \sim N(0, \sigma_{ij}^2)$ , with  $\sigma_{ij}^2 = \sigma_j^2(s_i) = (\omega_j + \tau_j s_i)^2$ .

Based on the replicates  $Y_{ij1}, \ldots, Y_{ijr}$ , r > 1, the variances  $\sigma_{ij}^2$  can be estimated and a maximum pseudo-likelihood solution might be proposed.

The replicates can be balanced (r) or unbalanced  $(r_{ij})$ .

#### Extensions

2 Model for several systems

$$Y_{i1k} = S_i + M_{i1k},$$
(1)  
$$Y_{ijk} = \alpha_j + \beta_j S_i + M_{ijk},$$
(2)

for k = 1, ..., r (replicates), j = 2, ..., m (systems), and i = 1, ..., n (subjects), noticing that  $\alpha_1 = 0$  and  $\beta_1 = 1$ .  $S_i \sim N(\mu, \sigma_s^2)$  and errors M as in Stevens *et al.* (2017). 3 Marginal probability of agreement

$$heta_j^{(M)} = P(|Y_{ij} - Y_{i1}| \le c_j), ext{ for } j = 2, \dots, m.$$
 (3)

#### 4 Conditional probability of agreement

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$$heta_j^{(C)} = P(|Y_{ij} - Y_{i1}| \le c_j \mid Y_{il} = y_l : l \notin \{1, j\}), \quad (4)$$
  
or  $j = 2, \dots, m.$ 

#### Extensions

#### 5 Maximal probability of agreement For m = 2 systems ( $\alpha_2 = \alpha$ , $\beta_2 = \beta$ , and $c_2 = c$ ),

$$Y_{i1k} = S_i + M_{i1k}, \qquad (5)$$

$$\psi(Y_{i2k}) = \alpha + \beta S_i + M_{i2k}, \tag{6}$$

for k = 1, ..., r (replicates) and i = 1, ..., n (subjects).

$$\theta^{(\max)}(s) = \sup_{\psi} P(|\psi(Y_{i2}) - Y_{i1}| \le c \mid S_i = s).$$
(7)

$$\theta^{(\max)} = \sup_{\psi} P(|\psi(Y_{i2}) - Y_{i1}| \le c).$$
(8)

Thank you!