

SUPTECH WORKSHOP III

Background Session II - Measures of statistical association

Measures of association – motivation

We are often interested in modeling relationships:

- is A related to B?
- does A cause B?
- do A and B usually coincide?
- what implications does the occurrence of A have on B? (conditional probability)

Probabilistic independence: Two events A and B are independent $(A \perp B)$ if and only if their joint probability equals the product of their probabilities.

$$P(A \cap B) = P(A)P(B) \iff P(A) = \frac{P(A \cap B)}{P(B)} = P(A \mid B)$$

Independence and correlation

Pearson product-moment correlation coefficient:

$$\rho_{X,Y} = \operatorname{corr}(X,Y) = \frac{\operatorname{cov}(X,Y)}{\sigma_X \sigma_Y} = \frac{\operatorname{E}[(X - \mu_X)(Y - \mu_Y)]}{\sigma_X \sigma_Y}$$

where ${\rm E}$ is the expected value operator, ${\rm cov}$ means covariance, and ${\rm corr}$ is a the correlation coefficient.

Pearson correlation coefficient



Pearson correlation coefficient



Correlation vs causality

For any two correlated events, A and B, the different possible relationships include:

- A causes B (direct causation);
- B causes A (reverse causation);
- A and B are consequences of a common cause, but do not cause each other;
- A and B both cause C, which is (explicitly or implicitly) conditioned on;
- A causes B and B causes A (bidirectional or cyclic causation);
- A causes C which causes B (indirect causation);
- There is no connection between A and B; the correlation is a coincidence.

Correlation vs causality

Thus there can be **no conclusion** made regarding the existence or the direction of a cause-and-effect relationship only from the fact that A and B are correlated.

Determining whether there is an actual cause-and-effect relationship requires further investigation, even when the relationship between A and B is statistically significant, a large effect size is observed, or a large part of the variance is explained.

Spearman's rank correlation coefficient is a **nonparametric** measure of rank correlation .

It assesses how well the relationship between two variables can be described using a **monotonic** function.

For a sample of size n, the scores X_i and Y_i are converted to ranks $\operatorname{rg} X_i, \operatorname{rg} Y_i$, and r_s is computed from:

$$r_s = \rho_{\mathrm{rg}_X, \mathrm{rg}_Y} = \frac{\mathrm{cov}(\mathrm{rg}_X, \mathrm{rg}_Y)}{\sigma_{\mathrm{rg}_X} \sigma_{\mathrm{rg}_Y}}$$



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Partial correlation

Partial correlation measures the degree of association between two random variables, with the effect of a set of controlling random variables removed.

$$\rho_{XY \cdot Z} = \frac{\rho_{XY} - \rho_{XZ} \rho_{ZY}}{\sqrt{1 - \rho_{XZ}^2} \sqrt{1 - \rho_{ZY}^2}}$$

If we define the precision matrix $P = (p_{ij}) = \Omega^{-1}$, we have:

$$\rho_{X_i X_j} \cdot \mathbf{v}_{\{X_i, X_j\}} = -\frac{p_{ij}}{\sqrt{p_{ii} p_{jj}}}$$

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Cross-quantilogram (Han et al., 2016)

Denote $(p_t, s_t), t \in \mathbb{Z}$ a bivariate time series, with $F_{p,t}(\cdot)$ and $F_{s,t}(\cdot)$ the conditional distribution functions and $q_{p,t}(\alpha_p) = \inf\{x, F_{p,t}(x) \ge \alpha_p\}, q_{s,t}(\alpha_s) = \inf\{x, F_{s,t}(x) \ge \alpha_s\}$ the corresponding quantiles.

Given a lag $k \in \mathbb{Z}$, we want to analyze the dependence between the events $p_t \leq q_{p,t}(\alpha_p)$ and $s_{t-k} \leq q_{s,t}(\alpha_s)$. Denote $\psi_a(u) = I(u < 0) - a$, where I is an indicator function. The cross-quantilogram for (α_s, α_p) at lag k is then defined as

$$\rho_{(\alpha_s,\alpha_p)}(k) = \frac{E\left[\psi_{\alpha_p}\left(p_t - q_{p,t}(\alpha_p)\right)\psi_{\alpha_s}\left(s_{t-k} - q_{s,t-k}(\alpha_s)\right)\right]}{\sqrt{E\left[\psi_{\alpha_p}^2\left(p_t - q_{p,t}(\alpha_p)\right]}\sqrt{E\left[\psi_{\alpha_s}^2\left(s_{t-k} - q_{s,t-k}(\alpha_s)\right)\right]}}$$

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Sources and further reading

https://en.wikipedia.org/wiki/Pearson_correlation_coefficient

https://en.wikipedia.org/wiki/Correlation_and_dependence

https://en.wikipedia.org/wiki/Correlation_does_not_imply_causation

https://en.wikipedia.org/wiki/Spearman's_rank_correlation_coefficient

https://en.wikipedia.org/wiki/Kendall_rank_correlation_coefficient

Han, H., Linton, O., Oka, T., & Whang, Y. J. (2016). The cross-quantilogram: Measuring quantile dependence and testing directional predictability between time series. Journal of Econometrics, 193(1), 251-270.