WENSLO (Weights by ENvelope and SLOpe) ALWAS (Aczel-Alsina Weighted Assessment) Method

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Summary

- WENSLO: objective method to calculate weights of criteria based on the envelope/slope ratio, using linear normalization, accumulation of values (concept inspired by Grey theory), calculation of the envelope (accumulated Euclidean distance) and slope. Final weight: additive envelope/slope ratio normalization.
- WENSLO: Normalizes → accumulates → calculates slope/envelope → obtains QJ → normalizes on w_i.
- ALWAS: The aggregation/ranking method based on nonlinear norms Aczel–Alsina — defines two strategies (weighted average and weighted geometric) and integrates both through a function with stabilization parameters (φ, θ) and parameter ξ of the Aczel–Alsina family. The output is a score of 0–1 per alternative, used to sort.
- ALWAS: standardizes → calculates R⁽¹⁾ and R⁽²⁾ via Aczel–Alsina functions → integrates with Initial matrix (ℜ): Build the matrix "m x n" where "m" are the alternatives and "n" are the criteria. ζ_{ij} values represent the performance of alternative "i" in criterion "j"(19) → obtains scores and ranking; parameters ξ, φ, θ control behavior/sensitivity.

WENSLO

1. **Initial matrix (\Re):** Build the matrix "m x n" where "m" are the alternatives and "n" are the criteria. ζ_{ij} values represent the performance of alternative "i" in criterion "j".

$$\Re(A, C) = \begin{bmatrix} A/C & C_1 & C_2 & \dots & C_j \\ \text{target maxmin maxmin} & \dots & \text{maxmin} \\ A_1 & \zeta_{11} & \zeta_{12} & \dots & \zeta_{1j} \\ A_2 & \zeta_{21} & \zeta_{22} & \dots & \zeta_{2j} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ A_i & \zeta_{i1} & \zeta_{i2} & \dots & \zeta_{mn} \end{bmatrix}$$

2. **Linear normalization**: Transform the initial matrix into a normalized matrix "Z" using linear normalization by sum (Eq. 2). The result is "z_{ij}" values between 0 and 1, making the criteria dimensionless and comparable.

$$z_{ij} = rac{\zeta_{ij}}{\sum_{i=1}^{m} \zeta_{ij}}$$

Produces matrix $Z=[z_{ij}]$ with $0 < z_{ij} < 1$

3. Class Range Δz_j (Sturges rule): Use Sturges' Rule (Eq. 5) to calculate the amplitude of the interval for each criterion "j". This step is fundamental to "graduate" the axis of alternatives, treating the sequence of normalized values of a criterion as a time series where the "distance" between two consecutive alternatives is fixed and equal to Δz_j .

$$\Delta z_j = \frac{\max_i z_{ij} - \min_i z_{ij}}{1 + 3.322 \log(m)}$$

- used to graduate the axis and treat alternatives as "points in time".
 - 4. **Slope (Criterion Inclination) tan \phij**: Calculate the tangent of the angle of inclination of the hypotenuse of the right triangle formed by the artificial accumulation of the data (Eq. 7). This slope represents the "ideal" or average growth rate of the criterion.

$$\tan \varphi_j = \frac{\sum_{i=1}^m z_{ij}}{(m-1) \cdot \Delta z_i} \quad \forall j \in [1, 2, \dots, n].$$

5. **Criterion envelope** E_j: Calculate the total length of the poly line (zig-zag) that connects the points of the ordered normalized values of each criterion (Eq. 8). This envelope captures the volatility and actual behavior of the criterion through alternatives.

$$E_j = \sum_{i=1}^{m-1} \sqrt{(z_{i+1,j} - z_{i,j})^2 + \Delta z_j^2}$$

$$\forall j \in [1, 2, \dots, n]$$

6. **Envelope/slope ratio** (q_i): Calculate the ratio of the envelope (real volatility) to the slope (average trend) for each criterion (Eq. 9). A criterion

with a high envelope (high volatility) and low slope (low upward trend) will have a high ratio, indicating greater importance in the decision process.

$$q_j = rac{E_j}{ an arphi_j}$$

$$\forall j \in [1, 2, \dots, n]$$

7. Calculation of the Final Weights of the Criteria (w_i): Normalize the ratios q_i to obtain the final weights (Eq. 10).

$$w_j = \frac{q_j}{\sum_{j=1}^n q_j}$$

$$\forall j \in [1, 2, \dots, n]$$

Interpretations: High envelope + low slope \rightarrow heavier weight.

Important notes (WENSLO)

- Process Validation (Accumulation): The method is based on Grey Systems Theory. Validation is done by comparing the actual accumulation of the normalized data with an artificial linear accumulation (the hypotenuse). The high correlation (r ≈ 1) and the low mean square error (MSE ≈ 0) between the two sequences (Table XI) confirm that the hypotenuse is a valid representation for calculating the slope (tan φj).
- The accumulation of z_{ij} (cumulative *sequence*) is used to reduce volatility and facilitate slope calculation see eq. (11) In the article and concept of artificial accumulation (hypotenuse) vs. real.
- Validation of the artificial approximation was performed with MSE and correlation coefficient, e.g., for criterion C₁₂ MSE=0.0013 and r=0.9981; mean r≈0.9747, mean MSE≈0.0126 indicates good adherence and validity of the procedure.

ALWAS

- 1. **Home matrix** ζ_{ij} : Use the original decision matrix $\Re = [\zeta_{ij}]$ (the same as in step 1 of WENSLO).
- 2. **Standardization**: transforms ζ_{ij} in standardized values using the article rule (eq. (16)), with different treatment for criteria such as benefit (B) and cost (C) and preserving the proportion of the original values (Eq. 16). This step is different from WENSLO's linear normalization.

$$\Re^s = [\hat{\zeta}_{ij}]_{m \times n}$$

$$\hat{\zeta}_{ij} \!=\! \begin{cases} \hat{\zeta}_{ij} = \frac{\zeta_{ij}}{\zeta_j'}, \text{ if } j \in \mathcal{B} \\ \hat{\zeta}_{ij} = -\frac{\zeta_{ij}}{\zeta_j'} + \max_{1 \leq i \leq m} \left(\frac{\zeta_{ij}}{\zeta_j'}\right) + \min_{1 \leq i \leq m} \left(\frac{\zeta_{ij}}{\zeta_j'}\right), \text{ if } j \in \mathcal{C} \end{cases}$$

- 3. **Define Weighted Aczel–Alsina Strategies**: Apply two nonlinear aggregation operators based on the triangular Aczel-Alsina norms, which introduce a flexibility parameter ξ ($\xi \ge 1$) to simulate different attitudes of the decision-maker (e.g., risk aversion).
 - Aczel-Alsina Weighted Average Strategy: Ri⁽¹⁾ξ (Aczel-Alsina weighted averaging) eq. (17).
 - Aczel-Alsina Weighted Geometric Strategy R⁽²⁾ξ (geometric) eq. (18).

where $f(\hat{\zeta}_{ij})=\hat{\zeta}_{ij}/\sum_{i}\hat{\zeta}_{ij}$ and the weights $\mathbf{w}_{\mathbf{j}}$ are those obtained by WENSLO.

$$\mathbb{R}_{i}^{(1)\xi} = \sum_{j=1}^{n} \hat{\zeta}_{ij} \left(1 - e^{-\left(\sum_{j=1}^{n} w_{j} \left(-\ln\left(1 - f\left(\hat{\zeta}_{ij}\right)\right)\right)^{\xi}\right)^{1/\xi}} \right)$$

$$\mathbb{R}_{i}^{(2)\xi} = \sum_{j=1}^{n} \hat{\zeta}_{ij} \cdot e^{-\left(\sum_{j=1}^{n} w_{j} \left(-\ln\left(f\left(\hat{\zeta}_{ij}\right)\right)\right)^{\xi}\right)^{1/\xi}}$$

4. **Integration of strategies** (Si): Combine the two previous strategies $R_{\xi}^{(1)} e R_{\xi}^{(2)}$ in a single integrated evaluation measure using a nonlinear function that incorporates two stabilization parameters, φ (which balances the importance between the two strategies) and θ (Eq. 19). The final ranking is obtained by ordering the alternatives in descending order by the values of Si.

$$\begin{split} \mathfrak{F}_{i} &= \frac{\mathbb{R}_{i}^{(1)\xi} + \mathbb{R}_{i}^{(2)\xi}}{1 + \left\{\phi\left(\frac{1 - f\left(\mathbb{R}_{i}^{(1)\xi}\right)}{f\left(\mathbb{R}_{i}^{(1)\xi}\right)}\right)^{\theta} + (1 - \phi)\left(\frac{1 - f\left(\mathbb{R}_{i}^{(2)\xi}\right)}{f\left(\mathbb{R}_{i}^{(2)\xi}\right)}\right)^{\theta}\right\}^{1/\theta}} \\ & \text{where } \theta > 0, \ \phi \geq 0, \ f(\mathbb{R}_{i}^{(1)\xi}) = \mathbb{R}_{i}^{(1)\xi}/\mathbb{R}_{i}^{(1)\xi} + \mathbb{R}_{i}^{(2)\xi} \ , \ \text{and} \\ f(\mathbb{R}_{i}^{(2)\xi}) &= \mathbb{R}_{i}^{(2)\xi}/\mathbb{R}_{i}^{(1)\xi} + \mathbb{R}_{i}^{(2)\xi}. \end{split}$$

Sensitivity Analysis - Parameters and sensitivity (ALWAS)

ξ (xi) controls "nonlinearity". The authors recommend low values (e.g., ξ =
 1) for good separation; In the experimentation of the article, large variations (ξ ≥ 36) leveled scores and reversed orders (e.g., change

between A1 and A6 to high ξ). It is recommended to explore $1 \le \xi \le 5$ for useful breakdown.

• ϕ e θ : ϕ regulates the relative weight between the two strategies; ϕ =0.5 and θ =1 were used as the initial solution. Changes in ϕ (below ~0.46) caused small trades between some positions (e.g. A4 vs A5). θ affects stability, but not major rearrangements in the study.

Validation and comparison (in the article)

- WENSLO: Comparison with other methods of objective weighting (Entropy, CRITIC, SIDev). WENSLO showed a high Spearman correlation (>96%) with all of them, confirming the credibility of the weights.
- ALWAS: was compared with MABAC, TOPSIS, WASPAS, TODIM high degree of correlation; small differences appeared in TOPSIS for intermediate positions.

Limitations pointed out in the article

 The method does not incorporate uncertainty (noise/intervals/missing values) authors suggest extension with uncertainty theory (fuzzy, intervals, etc.).

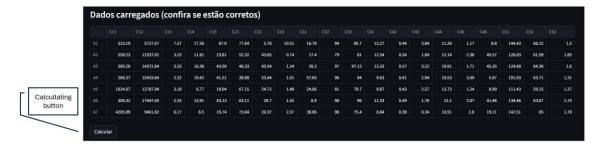
Computational Tool

1 – Main Screen

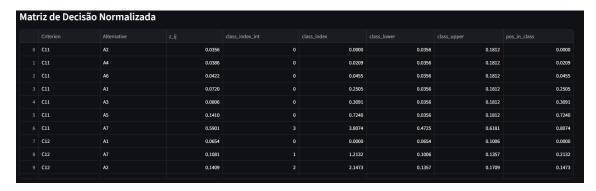
Web site: <u>WENSLO + ALWAS Tool · Streamlit</u> Web site: WENSLO + ALWAS Tool · Streamlit https://wenslo-alwastool.streamlit.app



2 - Uploaded Data



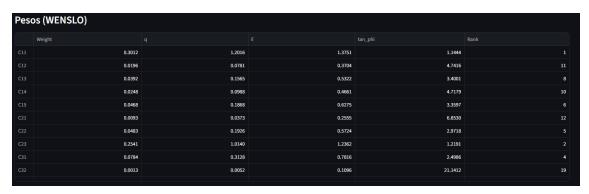
3 - Normalized Matrix



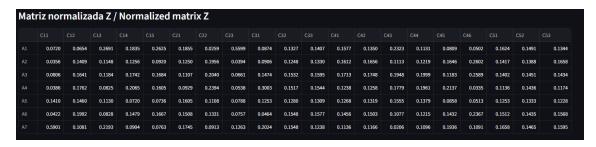
4 - Summary of graduation by criteria



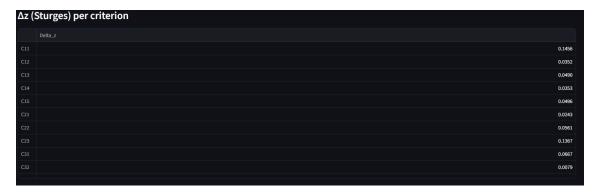
5 – Weights (WENSLO)



6 - Normalized Matrix Z



$7 - \Delta z$ (Sturges)



8 - Envelope E and tan_phi



9 - ALWAS Matrix



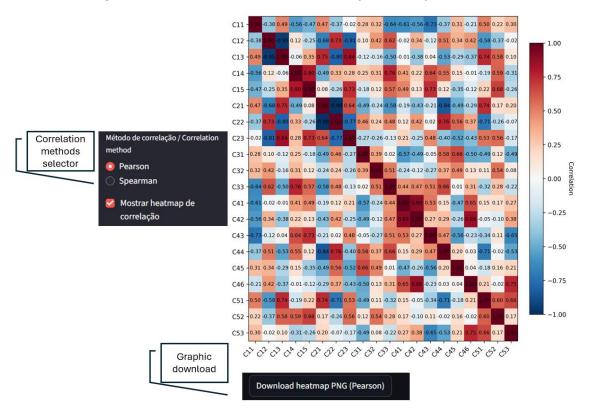
10 - Ranking (ALWAS)



11 - Accumulation validation



12 – Heatmap — Correlation between criteria (Pearson)



13 - Real vs Artificial Accumulation per criterion

