

The living-SEB-app, a Shiny app for meta-synthesis and review of the SEB literature:**USER MANUAL**

Tommaso Feraco

Department of General Psychology, University of Padova

Author Note

Tommaso Feraco  <https://orcid.org/0000-0002-8920-5330>

Correspondence concerning this article should be addressed to Tommaso Feraco,
Department of General Psychology, University of Padova, Via Venezia 8, Padova, Veneto 35129,
Italy, Email: tommaso.feraco@unipd.it

The living-SEB-app, a Shiny app for meta-synthesis and review of the SEB literature:**USER MANUAL****Table of contents**

Introduction	3
Getting started	3
Running a meta-analysis of the associations between SEB domains and their correlates	3
Steps	3
Notes and tips	5
Running a meta-SEM using the metaSEM package	5
Steps	5
Notes and tips	7
Reviewing and exporting literature subsets	7
Steps	7
What the app computes (technical details)	8
Data export & reports	8
Reproducibility checklist	9
Troubleshooting & FAQ	9
Citation	9
Discussion	10
Limitations and Future Directions	10
Conclusion	10

The living-SEB-app, a Shiny app for meta-synthesis and review of the SEB literature

The livingSEBapp (<https://feracoshiny.shinyapps.io/livingSEBapp/>) is a free and open-source Shiny app for synthesizing peer-reviewed studies that used the Behavioral, Emotional, and Social Skills Inventory (BESSI). The app is designed as a rapid tool for data synthesis, effect-size visualization, and for building priors/hypotheses. It is **not** a replacement for a full systematic meta-analysis; results should be interpreted as a high-quality *rapid synthesis*. For a full meta-analysis, use the open data here as a starting point, add study-level moderators as needed, enrich the dataset with additional studies if needed, and tailor the analyses on your hypotheses.

Below, each function of the app is described to guide correct and efficient use.

Getting started

- **Home tab.** Read the overview and use the quick-start buttons to jump to **Meta-analysis**, **MetaSEM**, or **Review** pages.
- **Navigation.** You can switch tabs anytime; your selections persist within each section until you change them or click “Clean selection” (Review tab).
- **Data sources.** All analyses run on the *combined* dataset curated from BESSI studies and related correlates. The data should hopefully cover the full list of cross-sectional associations between SEB skills (as measured with the BESSI) and all other measures included in each study). Code and data are linked on the app’s GitHub.

Running a meta-analysis of the associations between SEB domains and their correlates

A core utility is the rapid synthesis of associations between the five SEB domains and any correlate measured in ≥2 independent samples. The app estimates **five separate meta-analytical associations** (one per SEB domain) with a **single chosen outcome** (Multivariate models are not currently supported in this section).

Steps

1. **Select the data of interest (Meta-analysis → Summary).**
 - In the sidebar, choose a **broad outcome grouping** and then a **specific outcome** within it.

By default, **SCHOOL** → *academic achievement* is selected. The full list of outcomes is in the matrix codebook (repository).

- Optionally filter by **age class** and **year of publication**.
- As you select, the **Samples included** table updates with study IDs, authors, year, age class, N, and DOI.

2. Run the meta-analysis (Meta-analysis → Meta results).

- Move to the *Meta results* tab and click **Run the meta-analysis**.
- The app computes Fisher-z correlations via `metafor::escalc(measure = "ZCOR")`, then fits **multilevel random-effects** models with `metafor::rma.mv()` using random intercepts for paper_id/matrix_id. Estimates are back-transformed to *r*.
- You'll see, for each SEB domain:
 - **r** and its **95% CI, SE, p**
 - **k** (number of effect sizes) and total **N** (sample size)
 - **τ^2** (between-study variance; summed across random components)
 - **Q, df**, and **Q p-value** (heterogeneity)
 - **Prediction interval (PI)** for *r* (in the results object and plot annotation)

3. Inspect a forest plot (Meta-analysis → Forest plots).

- Move to the *Forest plots* tab and choose a **skill** (SEB domain) and the app displays the forest plot for that domain–outcome association.

4. Export a brief report (Meta-analysis → Report).

- Download an **HTML report** that includes your selections, descriptive tables, meta-analytic results, and plots.
- You can also download the **meta-analysis dataset (CSV)** from the Summary tab for offline work.

Notes and tips

- If filters yield **no data**, the results tab will indicate this. Broaden your filters or pick a different outcome.
- Interpret **PIs** for expected dispersion of future samples; **CIs** for precision of the mean effect.
- The **total N** is computed across unique matrices and may differ across skills when studies report different subsets.

Running a meta-SEM using the metaSEM package

This section builds a **meta-analytic correlation matrix** across selected variables and fits **one-stage MASEM** models using `metaSEM`.

Steps

1. Select variables and get coverage (MetaSEM → Summary).

- In the sidebar, select any combination of **SEB domains**, **SEB facets**, **Big Five traits**, and **outcomes**. Optional filters: **age class** and **year**.
- Click **Get info**. The app assembles the data and shows:
 - **K correlations** (how many independent samples contribute to each pair).
A red banner appears if any **k = 0** pairs exist; please remove variables so every pair has coverage. As a general rule of thumb, use `metaSEM` if each pair of correlations has
 - * 2 studies = minimum (barely usable, only fixed-effect style).
 - * 3–5 studies = acceptable for exploratory models, but limited power.
 - * 10+ studies per correlation = good practice for reliable meta-SEM.
 - **Sample size per correlation (N)**.
 - **Studies included** table.
- Warnings appear if some correlations have **k < 3** or **N < 1000** (interpret with caution).

2. Compute the pooled correlation matrix (MetaSEM → SEM matrix).

- Move to the *SEM matrix* tab and click **Get the meta-matrix**. The app runs `metaSEM:::tssem1()` to pool the correlation structure.
- View the **estimated correlation matrix** (rounded) and a **corrplot**.
You can **download** the correlation plot (PNG).

3. Specify and fit a model (MetaSEM → metaSEM).

- Move to the *metaSEM* tab and insert your **lavaan syntax** in the text area (e.g., `y ~ x1 + x2`). If you are not familiar with lavaan or R, please see <https://lavaan.ugent.be/tutorial/syntax1.html>.
 - You can fit both regression models and measurement models.
 - The app **auto-augments** your model to:
 - * Fix exogenous variances correctly,
 - * And saturate the residual covariances.
 - If you do not want the model to be saturated, unclick the **checkbox**.
 - * You can now specify the covariances you want to estimate using `~~`.
 - * If the model is not saturated, you can obtain fit indices.
- Click **Fit the specified SEM model** to estimate a **one-stage MASEM** (`osmasem`) with a **diagonal random-effects** structure (`(..., RE.type = "Diag")`).
- **Model fit** panel:
 - If the model is **saturated (df = 0)**, fit indices aren't informative; focus on parameter estimates.
 - Otherwise, the app shows **CFI, TLI, RMSEA, SRMR**.
- **Model estimates** table includes:
 - Paths and variances with **Estimate, SE, z, p, 95% CI**, plus a **τ (sd)** column for random-effect variance and its transformation to SD where applicable.

- **Residual variance / R²** table lists each variable's residual variance and **R² = 1 – residual variance** (endogenous variables first).

4. Export a brief report (MetaSEM → Report).

- Download an **HTML report** with your selections, K/N tables, pooled matrix, model specification (original + augmented), fit indices, parameter table with τ , and R² table.

Notes and tips

- Do not proceed with models when the **K table** shows **k = 0** pairs or **k < 5**; adjust the variable set.
- Large models with many variables can be slow and sometimes unstable; start small and build up.
- Use the **auto-saturate** option if you are not familiar with SEM models. Saturated residuals can mask model misfit but will give interpretable estimates.

Reviewing and exporting literature subsets

The **Review** tab lets you filter the curated literature by **broad topics** and **specific topics**.

Steps

1. Choose topics.

- Select one or more **BROAD topics** and/or **Specific topics**.
- A dynamic **Filtered based on broad topic** selector appears to help you pick specific topics that belong to the chosen broad topics.
- The specific topic selection is nested in the broad topic selection: If you select a broad topic, specific topics will not show additional papers but will only filter the selection more (if that specific topic exist in the broader topic/s).

2. View results.

- The **Filtered studies** table shows items matching your filters, with a banner indicating the number of matching records. Helpful messages appear when there are zero matches.

3. Reset / export.

- Click **Clean selection** to clear all topic filters.
- Click **Download the filtered data** to export the current subset as **UTF-8 CSV**.

What the app computes (technical details)

- **Effect sizes.** Correlations are transformed to Fisher-z via `metafor::escalc(measure = "ZCOR")`; models are fit on z and back-transformed to r .
- **Meta-analytic model (Meta-analysis).** `rma.mv(yi, V = vi, random = ~ 1 | paper_id / matrix_id, test = "t")`.
 - Reported: point estimate, 95% CI, SE, p; τ^2 (sum of variance components), **Q**, **df**, **Q p-value**, and **prediction interval**.
- **Pooled matrix (MetaSEM).** `tssem1()` aggregates correlations across studies.
- **One-stage MASEM.** `osmasem()` with **diagonal random effects**; τ reported per parameter where random effects are estimated.
- **R².** Computed from residual variances in the fitted model ($R^2 = 1 - \text{residual variance}$), bounded at 0.

Data export & reports

- **Meta-analysis data (CSV).** The exact dataset used to compute the meta-analysis.
- **MetaSEM data (ZIP CSV/RDS).** The exact datasets used to compute the metaSEM or pooled correlations.
- **Correlation plot (PNG).** High-resolution (2000×2000 px, 300 dpi) of the pooled correlation matrix.
- **Review subset (CSV).** Filtered literature table.

- **Reports (HTML).** `report_meta.html` and `report_masem.html` include selections, methods, tables, and figures.
- **Bibliography.** Reports include a **filtered .bib** of the cited studies based on your current data.

Reproducibility checklist

- Save your **filters** and **selected outcome/variables**.
- Download the **CSV/RDS/ZIP** outputs alongside the **HTML reports**.
- For full reproducibility, clone the GitHub repository and note your **R** and **package versions** when running locally.

Troubleshooting & FAQ

- **No results / disabled download.** You'll see a friendly message; broaden filters or change the outcome/topic.
- **K table shows zeros.** Remove variables so every pair has coverage; otherwise, MASEM is not identified.
- **Saturated SEM (df = 0).** Fit indices are not meaningful; focus on estimates and CIs.
- **Buttons do nothing.** Ensure you clicked **Get info** (MetaSEM) or **Run the meta-analysis** (Meta-analysis) first.
- **Large models are slow.** Reduce the number of variables, fit the matrix first, then gradually add paths. Alternatively, download the data and code and run locally (faster).

Citation

If you use this app or dataset, please cite:

Feraco, T. (2025). *The Living SEB Project: A Living Database for Review and Meta-Analysis of Social, Emotional, and Behavioral Skills*.

https://github.com/feracotomaso/living_SEB_review

A BibTeX entry is:

```
@Misc{livingSEB,
  author      = {Tommaso Feraco},
  title       = {The Living SEB Project: A Living Database for Review and Meta-Analysis},
  year        = {2025},
  howpublished = {GitHub repository},
  url         = {https://github.com/feracotommaso/living_SEB_review}
}
```

Discussion

The app enables rapid, transparent synthesis of BESSI-related findings and structural models. It is ideal for:

- **Scoping evidence** and visualizing effect sizes,
- **Generating priors/hypotheses** for confirmatory work,
- **Literature research and review** on SEB skills.
- **Teaching** meta-analytic and MASEM concepts with real data.

Limitations and Future Directions

- Extend meta-analysis to **facet-level** skills.
- Add **moderator/meta-regression** capabilities (study-level covariates).
- Expand the **Review** schema (e.g., study design, region) for finer filtering.
- Consider **multivariate meta-analysis** in the Meta-analysis section.
- Extend to non-crossectional analyses

Conclusion

The livingSEBapp lowers the barrier to evidence synthesis in SEB research by combining curated data, robust meta-analytic methods, and an accessible interface. It complements, but does not replace, full systematic meta-analysis.