R-Shiny tool to pool Kaplan Meier curves from single-arm studies using a distribution-free approach

Presented at: R for HTA consortium 2024

Presented by: Shubhram Pandey





Situation assessment and solution

Existing methods

Existing methods often pool survival probabilities reported at the same time point, treating them as proportions and combining them using fixed-effect univariate or random-effect methods. However, these methods may have limitations, such as not accounting for correlations between survival probabilities at different times.

Challenges

Make strong assumptions about the data, such as linearity or specific distributions, which may not always be supported, and the survival curves produced may not be non-increasing

Solution

- It proposes a distribution-free method to derive a summary survival curve using a product-limit estimator combined with an extension of the DerSimonian and Laird methodology
- It does not require assumptions about the shape of the survival curve or the distribution of random effects



Advantages

Distribution-Free Approach

It does not require assumptions about the shape of the survival curve or the distribution of the random effects. This makes it more flexible and broadly applicable compared to parametric approaches.

Handling of Censored Data

Similar to the Kaplan-Meier estimator, the method appropriately handles right-censored data, ensuring that all available information is used without making strong assumptions about the underlying survival distributions.

Incorporation of Between-Study Heterogeneity

By extending the DerSimonian and Laird methodology to multiple outcomes, the method accounts for heterogeneity between studies, providing a more accurate summary survival curve.

Output

Output from this method is a summary survival curve, which anyone can leverage for further analysis including the extrapolations of pooled Kaplan-Meier curve according to NICE TSD 14 and 21



Disadvantages

Approximation Issues

The method uses transformations and approximations (e.g., arcsine transformation) that may not always be perfectly accurate, particularly in case of rare events reported in Kaplan-Meier curve

Assumption of Independence

While the method accounts for between-study heterogeneity, it assumes that the conditional survival probabilities within a study are independent. This may not always be the case, especially if there are underlying correlations not captured by the model

Dependence on Published Data

The accuracy of the summary survival curve relies on the quality and completeness of the published survival curves from the included studies.

Missing or inaccurately reported data can affect the results.

Complexity of Implementation

The method involves advanced statistical techniques and computations, which may be challenging for practitioners without specialized statistical training or software capabilities.



Workflow

Secured by Auth0, it requires users to authenticate themselves

Input

A Pseudo-IPD data for each Kaplan-Meier curve in a data frame format with three columns:

- Time
- Event
- Label, i.e., Study Name or treatment name

Pooled Kaplan-Meier curve using Random and fixed effects model

Median time and restricted mean for the pooled curve using random and fixed effect model

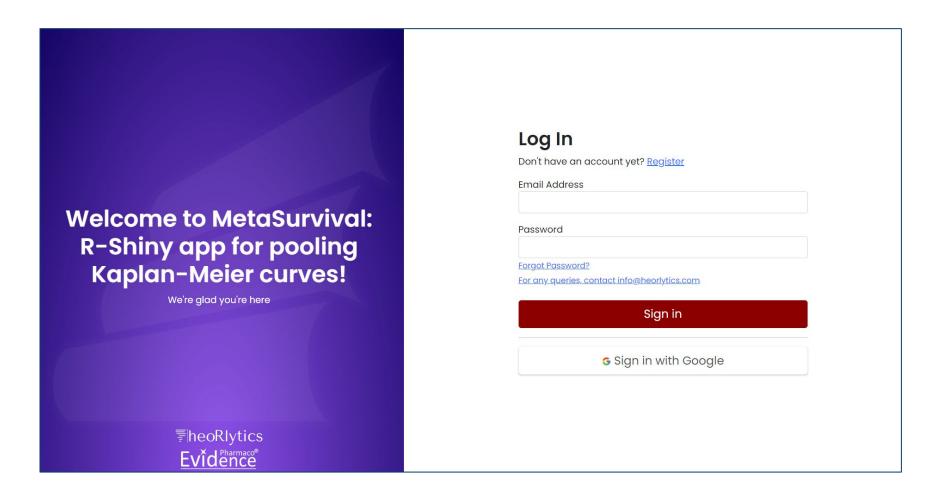
Pooled Survival proportions using random and fixed effect model

Outputs

The goodness of fit statistics, i.e., Heterogeneity Q, H-square and I-square



Interface



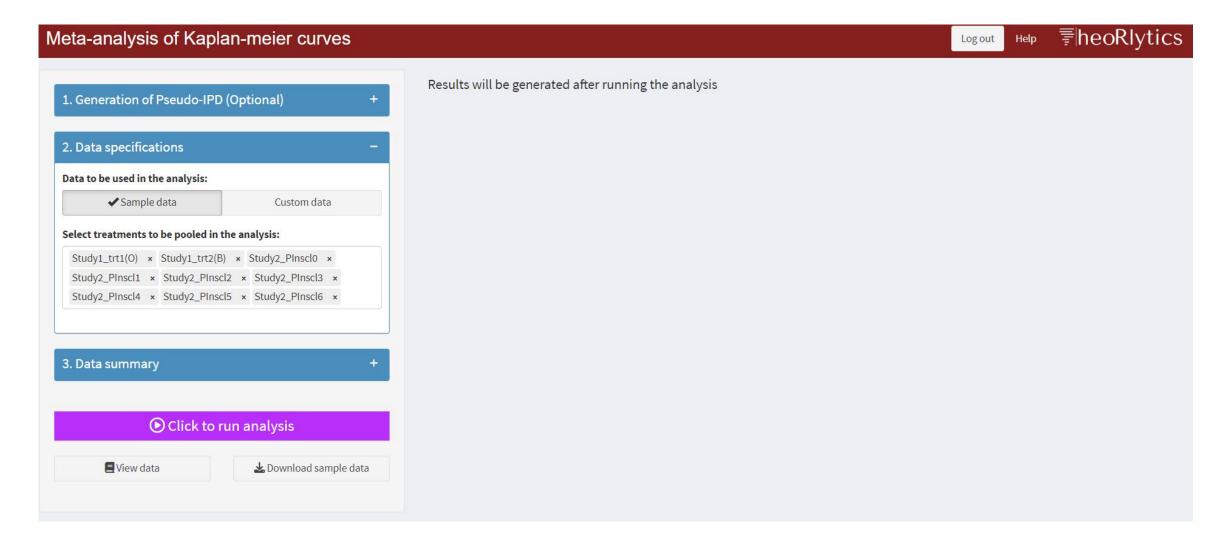
Scan the QR to open the application



Go to https://metasurvival.su rvlytics.com/



After Authentication





After Running Analysis

