



BMJ Open is committed to open peer review. As part of this commitment we make the peer review history of every article we publish publicly available.

When an article is published we post the peer reviewers' comments and the authors' responses online. We also post the versions of the paper that were used during peer review. These are the versions that the peer review comments apply to.

The versions of the paper that follow are the versions that were submitted during the peer review process. They are not the versions of record or the final published versions. They should not be cited or distributed as the published version of this manuscript.

BMJ Open is an open access journal and the full, final, typeset and author-corrected version of record of the manuscript is available on our site with no access controls, subscription charges or pay-per-view fees (<http://bmjopen.bmj.com>).

If you have any questions on BMJ Open's open peer review process please email info.bmjopen@bmj.com

BMJ Open

Efficacy of psychotherapy, pharmacotherapy, or their combination in chronic depression: Study protocol for a systematic review and network meta-analysis using aggregated and individual patient data

Journal:	<i>BMJ Open</i>
Manuscript ID	bmjopen-2024-089356
Article Type:	Protocol
Date Submitted by the Author:	28-May-2024
Complete List of Authors:	Schramm, Elisabeth; University of Freiburg, Department of Psychiatry and Psychotherapy Elsaesser, Moritz; University of Freiburg, Department of Psychiatry and Psychotherapy, University Medical Center – University of Freiburg, Faculty of Medicine Müller, Julia; University of Freiburg, Department of Psychiatry and Psychotherapy Kwarteng, Nana-Adjoa; University of Freiburg, Institute of Medical Biometry and Statistics Evrenoglou, Theodoros; University of Freiburg, Institute of Medical Biometry and Statistics Cuijpers, Pim; VU University Amsterdam, Department of Clinical Psychology; Babeş-Bolyai University, International Institute for Psychotherapy Orestis, Efthimiou; University of Bern, Institute of Primary Health Care (BIHAM) Klein, Daniel; Stony Brook University, Department of Psychology Keller, Martin; The Warren Alpert Medical School of Brown University, Department of Psychiatry and Human Behavior Furukawa, Toshi; Kyoto University, Graduate School of Medicine and School of Public Health Nikolakopoulou, Adriani ; University of Freiburg, Institute of Medical Biometry and Statistics
Keywords:	Systematic Review, Meta-Analysis, Network Meta-Analysis, Depression & mood disorders < PSYCHIATRY, Adult psychiatry < PSYCHIATRY, PSYCHIATRY

SCHOLARONE™
Manuscripts

1
2
3
4
5
6
7
8
9
10
11
12
13
14
15
16
17
18
19
20
21
22
23
24
25
26
27
28
29
30
31
32
33
34
35
36
37
38
39
40
41
42
43
44
45
46
47
48
49
50
51
52
53
54
55
56
57
58
59
60

**Efficacy of psychotherapy, pharmacotherapy, or their combination
in chronic depression: Study protocol for a systematic review and
network meta-analysis using aggregated and individual patient data**

Schramm, E.^{1*}, Elsaesser, M.^{1*}, Mueller, J.¹, Kwarteng, N.², Evrenoglou, T.², Cuijpers, P.^{3,4}, Efthimiou, O.⁵, Klein, D. N.⁶, Keller, M. B.⁷, Furukawa, T. A.^{8*} & Nikolakopoulou, A.^{2*}

¹ Department of Psychiatry and Psychotherapy, Medical Center – University of Freiburg, Faculty of Medicine, University of Freiburg, Freiburg, Germany

² Institute of Medical Biometry and Statistics, Faculty of Medicine and Medical Center, University of Freiburg, Freiburg im Breisgau, Germany

³ Department of Clinical, Neuro and Developmental Psychology, WHO Collaborating Centre for Research and Dissemination of Psychological Interventions, Vrije Universiteit Amsterdam, Amsterdam, The Netherlands

⁴ Babeş-Bolyai University, International Institute for Psychotherapy, Cluj-Napoca, Romania

⁵ Institute of Primary Health Care (BIHAM), University of Bern, Bern, Switzerland

⁶ Department of Psychology, Stony Brook University, Stony Brook, New York, United States

⁷ Department of Psychiatry and Human Behavior, The Warren Alpert Medical School of Brown University, Providence, USA

⁸ Department of Health Promotion and Human Behavior, Kyoto University Graduate School of Medicine/School of Public Health, Kyoto, Japan

*shared first/senior authorship

Word count: 4657

***Correspondence:**
Prof Dr Elisabeth Schramm
Department of Psychiatry and Psychotherapy
University Medical Center – University of Freiburg,
Hauptstraße 5, DE–79104 Freiburg (Germany)
✉ elisabeth.schramm@uniklinik-freiburg.de

Abstract

Introduction: Chronic depression represents a common and highly disabling disorder. Several randomised controlled trials investigated the effectiveness of psychological, pharmacological, and combined treatments for chronic depression. This is the first overarching systematic review and network meta-analysis based on aggregated and individual patient data (IPD-NMA) comparing the efficacy and acceptability of various treatment options for all subtypes of chronic depression. Furthermore, individual demographic and clinical characteristics that predict or moderate therapy outcomes will be investigated.

Methods and analysis: A systematic literature search of the Cochrane Library, MEDLINE via Ovid, PsycINFO, Web of Science, and metapsy databases will be conducted to include all available samples from randomised controlled trials (RCTs) that investigated treatment effects in adult inpatients or outpatients with a primary diagnosis of chronic depression. The main outcome is depression severity measured on a continuous observer-rated scale for depression at six months post treatment (range 3-12 months). Two reviewers will independently screen and select eligible studies based on the pre-defined inclusion and exclusion criteria. Risk of bias will be assessed using Version 2 of the Cochrane risk-of-bias tool for randomized trials (Rob 2.0). Individual patient data (IPD) will be requested and incorporated in the network when provided. For studies which do not provide IPD, aggregate data (AD) will be extracted and incorporated in lieu of IPD for the network meta-analysis (NMA). An NMA comparing psychotherapies and a network meta-regression (NMR) estimating individualised treatment effects of psychotherapy will be implemented assuming a Bayesian framework. All models will be fitted in R with calls to JAGS. Empirical informative prior distributions will be used for model parameters where available, and non-informative priors will be used in cases where empirical priors are not available.

Ethics and dissemination: This IPD-NMA requires no ethical approval. All results will be disseminated as peer-reviewed publication in a leading journal in this field and presented at (inter)national scientific conferences.

88 Introduction

89 Chronic depression is a common and long-term disabling disorder[1]. Up to one third of all
90 depressive disorders take a chronic course[2] with the definition of chronic varying in the
91 literature regarding duration (between a minimum of one to three years), severity (dysthymia,
92 chronic major depression) and the type of course since the first onset (dysthymia with or
93 without superimposed major depressive episode(s), chronic major depression, or recurrent
94 major depression without inter-episode recovery[3]). In 2013, the DSM-5 merged these various
95 presentations into the rubric of persistent depressive disorders (PDD) with defined diagnostic
96 criteria and four coding specifiers. Unlike major depressive disorder, chronic depression tends
97 to have a more subtle but persistent presentation, leading to comparably more significant
98 impairments in functioning and more reduced quality of life[4]. Despite new developments of
99 treatment approaches in the past decades, chronic depression remains challenging to treat,
100 with varying degrees of success reported for different modalities. Common treatment
101 complicating factors such as comorbidity with mental and physical disorders, poor social
102 integration, early onset, or a history of early trauma lead to a high rate of treatment-
103 resistance[5].

104 Psychotherapy and pharmacotherapy are the two primary treatment modalities used to
105 manage chronic depression. Psychotherapy encompasses various approaches such as the
106 Cognitive Behavioral Analysis System of Psychotherapy (CBASP[6–8]) as the only model
107 specifically designed for chronic depression, and more general methods such as Cognitive-
108 Behavioral Therapy (CBT[9]), Interpersonal Therapy (IPT[10]), or Psychodynamic and
109 Psychoanalytic Therapy[11,12]. Pharmacotherapy, as the most commonly used treatment,
110 involves antidepressant medications including selective serotonin reuptake inhibitors (SSRIs),
111 serotonin-norepinephrine reuptake inhibitors (SNRIs), tricyclic antidepressants, and others, to
112 modulate neurotransmitter activity in the brain.

113 Furthermore, psychotherapy and pharmacotherapy are integrated into combination therapy to
114 potentially yield faster and better treatment outcomes. Despite the availability of these
115 treatment options, uncertainty remains regarding their relative efficacy and differential

response in managing chronic and treatment-resistant depression. While some studies suggest that combination therapy may be more effective than either psychotherapy or pharmacotherapy alone[13,14], others indicate that certain monotherapies are particularly suitable for specific patient profiles or severity levels[15]. The last comprehensive meta-analysis specifically about chronic depression dates from 2014 and recommends different approaches depending on the subtype of chronic depression[12]. However, within the last ten years, several new studies have appeared that have not yet been synthesized with the previous evidence (e.g. [16–22]). An individual participant data network meta-regression (NMR) investigated the efficacy of CBASP, pharmacotherapy or their combination and several effect moderators, but included only three studies[15,23]. In a fairly recent network meta-analysis on adult depression, a subgroup analysis showed superior effects for combined treatment versus psychotherapy or pharmacotherapy[14]. However, this subgroup analysis did not distinguish between chronic courses and treatment-resistant depression and did not consider long-term effects.

Therefore, an up-to-date comprehensive systematic review on the efficacy of psychotherapy, pharmacotherapy, and their combination in treating chronic depression is urgently needed. Our network meta-analysis (NMA) will incorporate all the available randomised controlled trials (RCTs) to capture the full breadth of evidence[24]. By analysing individual patient data (IPD) as well as aggregate data (AD; if IPD is not available) from a range of studies, we aim to obtain insights into the most effective treatment approaches, identify potential predicting and moderating factors, and offer guidance for clinicians and patients seeking optimal management strategies. The results of this meta-analysis may have significant implications for the individualised treatment of chronic depression, informing clinical practice and shaping future research in this field.

Objectives

The aim of this systematic review is to evaluate the efficacy of psychological, pharmacological, or combination treatments in chronic depression. To this end, the proposed systematic review will answer the following questions:

Protected by copyright, including for uses related to text and data mining, AI training, and similar technologies.
Erasmus Hogeschool

1. Which psychological, pharmacological, or combination treatments are most effective for chronic depression?
2. How do individual demographic and clinical characteristics predict or moderate individualised treatment recommendations?

Methods and analysis

Eligibility criteria

We will include data from RCTs that investigated treatment effects in adult (>18 years old) inpatients or outpatients with a primary diagnosis of chronic depression. Depressive courses are defined as chronic when lasting for at least two years including persistent depressive disorders (DSM-5), chronic major depression, double depression (dysthymia with superimposed major depressive episode), recurrent major depression with incomplete inter-episode recovery, dysthymia, or any corresponding conditions according to standard operationalized diagnostic criteria as the primary diagnosis. Comorbid disorders are allowed and all respective information will be collected.

The treatments of interest in our IPD-NMA include:

- Different types of psychotherapies (including Cognitive Behavioral Analysis System of Psychotherapy, Cognitive-Behavioral Therapy, Interpersonal Therapy, Modular-Based Psychotherapy, Metacognitive Training, Dialectical Behavior Therapy, Mindfulness-based Cognitive therapy, (Long-term) Psychoanalytic Psychotherapy, (Brief) Supportive Psychotherapy, Group Person-based Cognitive Therapy, Cognitive Therapy, Problem-solving Treatment, Schema Therapy).
- Antidepressant pharmacotherapy including any of the antidepressive agents licensed for the treatment of major depression in the country where the trial was conducted.
- Different types of psychotherapies mentioned above as an adjunct treatment to other treatments, e.g. treatment as usual (TAU), care as usual (CAU), exercise, counselling.
- Different types of psychotherapies mentioned above as an adjunct treatment to any type of antidepressant pharmacotherapy.

- If waitlist control is included in the screened studies, it will be used as the primary reference category when reporting relative treatment effects. CAU and TAU conditions will be compared with each other and examined with regard to their similarity. If substantial differences between TAU and CAU are detected, we will consider whether they need to be split into meaningful categories. Otherwise they will be merged and treated as one single treatment (i.e. TAU/CAU) in the statistical analyses.
- Only studies implementing face-to-face psychotherapy will be included; internet-based treatment studies will be excluded.

Information sources and search strategy

We will first conduct an electronic literature search of the Cochrane Library, MEDLINE via Ovid, PsycINFO, Web of Science, and metapsy databases with the keywords “psychotherapy”, “chronic depression” and all related terms. First searching will be conducted in March 2024. Searches will be re-run just before the final analyses and any further studies identified will be retrieved for inclusion. There will be no restriction for the publication period. As an example, this is the final search string for OVID:

((chroni*5 adj3 depress*).ti,ab. OR (chroni*5 adj3 MDD).ti,ab. OR exp Depressive Disorder, Treatment-Resistant/ OR exp Dysthymic Disorder/ OR (treatment-resistan* adj3 depress*).ti,ab. OR (treatment-resistan* adj3 MDD).ti,ab. OR (therapy-resistan* adj3 depress*).ti,ab. OR (therapy-resistan* adj3 MDD).ti,ab. OR (dysthymia or (dysthym* adj2 disorder*1)).ti,ab. OR (persist* adj2 depress*).ti,ab. OR (persist* adj2 MDD).ti,ab. OR (double depression).ti,ab.) AND (exp Psychotherapy/ or psychotherap*.ti,ab. OR (psychologi* adj3 treatmen*2).ti,ab. OR (psychologi* adj3 interventio*2).ti,ab. OR CBASP.ti,ab. OR ((cognitive adj2 therapy).ti,ab. OR (behavior* adj1 therapy).ti,ab. OR mindfulness.ti,ab. OR psychoanalytic.ti,ab. OR psychodynamic.ti,ab. OR (schema therapy).ti,ab.)

In accordance with the Cochrane handbook, preprints will be considered a potentially relevant source of study evidence and will therefore be assessed for eligibility. Backward and forward citation searches of included studies and relevant reviews will be performed. To find references

in the grey literature, we will contact authors of included studies and relevant conference abstracts of unpublished studies. Furthermore, we will use relevant mailing lists, for example from the Society for Psychotherapy Research or (inter)national professional associations, to draw attention to this project.

Study selection

At least two reviewers will independently screen the title and abstract of all records of the systematic literature search and select eligible studies based on the pre-defined inclusion and exclusion criteria. If both reviewers independently determine that a study may be eligible based on title and abstract screening, then a full-text article review will be completed. Disagreements between individual judgements will be resolved via discussion with a third reviewer. In case of ongoing disagreement, a meeting will be held with the complete study team involving all reviewers allowing them to present the reasoning for their judgement. Consensus should be reached after debate and decisions will be documented in written form.

Data extraction

The IPD of the originally established datasets will be requested from the authors of all eligible studies. One person will check the received data for completion. All obtained IPD will be cleaned, coded, and saved in appropriate files to make the data as uniform as possible. Afterwards, we will compare the published data of each dataset (i.e. numbers and percentages, or means and standard deviations of baseline demographics as well as clinical variables) with the summaries that we will obtain directly from the IPD. Any major inconsistencies will be discussed and followed-up. Corrections will be made as necessary. In cases where IPD are not provided within a reasonable time after the request, AD will be extracted from the publication and used instead. Two individuals will be responsible for performing the data extraction. One extractor will perform the data extraction while the second will perform the quality control and ensure that all data are properly extracted.

When extracting AD, the following data points will be collected:

- Study level data: Year of publication; validated depression scale(s) used in study.

- Participants: Mean age at baseline; mean age at first depressive episode; mean duration of chronic depression; proportion of participants with prior antidepressant treatment; proportion of participants with current antidepressant treatment; proportion of participants with prior psychotherapeutic treatment; other covariates (e.g. childhood maltreatment, loneliness, attachment style, personality, rejection sensitivity, mentalisation, emotion regulation, comorbidities, severity, interaction style, avoidance, social functioning, alexithymia, empathy).
- Outcomes per arm: mean follow-up duration; mean baseline, post-treatment, and follow-up depression score, and standard deviation; mean change from baseline depression score and standard deviation; number of participants at baseline, post-treatment, and follow-up; percentage of dropouts (treatment discontinuation); side effects; (serious) adverse events.

In case of repeated measures, the time point of each repeated measure will also be extracted. Disagreements between individual judgements will be resolved via discussion. In case of discrepancies, a meeting will be held involving both extractors allowing them to present the reasoning for their judgement. A consensus should be reached after debate and decisions should be documented in meeting minutes. Data will be recorded in excel spreadsheets.

Outcomes

The main outcome is depression severity measured on a continuous observer-rated scale for depression at six months post treatment (range 3-12 months). If the respective study reports results at two or more time points within this frame, we will prioritize the time point closest to six months. If time points are equidistant, we will use the later one. Where different scales such as the Montgomery-Asberg Depression Rating Scale (MADRS[25]), the Quick Inventory of Depressive Symptomatology - Clinician Rating (QIDS-C[26]) or different versions of the Hamilton Rating Scale for Depression (HAM-D/HRSD[27]) are reported, we will attempt to transform their respective scores to the 17-item HRSD score[28,29] as the most frequently utilized measure[30] using conversion procedures (e.g. <http://ids-qids.org/interpretation.html> or [31]).

254 Secondary outcomes are:

- 255 1. Treatment response, defined as 50% or greater reduction from baseline to study
256 endpoint in the study's primary observer-rated or self-rated depression scale.
- 257 2. Remission, defined as scoring below the validated thresholds of the study's primary
258 observer-rated or self-rated depression scale at endpoint.
- 259 3. Study drop-out for any reason, as a proxy measure of overall treatment acceptability.
- 260 4. Study drop-out due to side effects.
- 261 5. Depression severity as measured on a continuous self-rating scale for depression,
262 such as Beck Depression Inventory-II (BDI-II[32]) or Inventory of Depressive
263 Symptomatology, Self-Report (IDS-SR[33]). Different scales will be converted into BDI-
264 II using conversion tables and equipercentile linking[34].
- 265 6. Global Assessment of Functioning (GAF).
- 266 7. Social functioning, as measured by any validated measure for impaired social
267 functioning such as the Social Adjustment Scale-Self Report (SAS-SR[35]).
- 268 8. Quality of Life, as measured by any validated measure for life quality such as the World
269 Health Organization Quality of Life (WHOQOL[36]) or the Short Form 36 Health Survey
270 (SF-36[37,38]).
- 271 9. Side effects.
- 272 10. Adverse and serious adverse events.

273 ***Risk of Bias Assessment***

274 We will assess risk of bias (RoB) in the included studies using the Cochrane RoB 2 tool for
275 randomized trials[39]. The assessment will be done by two independent raters. If raters
276 disagree, the final rating will be made by consensus with involvement of another member of
277 the review group (if necessary). We will evaluate RoB in the following domains: Bias arising
278 from the randomization process, bias due to deviations from intended interventions, bias due
279 to missing outcome data, bias in measurement of the outcome, and bias in selection of the
280 reported result.

Please note that, when IPD are available, the RoB assessments can be different from those based on AD. For example, even when the original authors used alternative imputation methods for handling missing outcomes (e.g. last observation carried forward (LOCF)), we can apply different imputation methods more suitable for the current use case (e.g. multiple imputation) on the IPD. We will assess the RoB for the outcomes used for the primary outcome of our NMA.

Data synthesis

Measurements of depression severity are often reported on validated continuous observer-rated scales for depression. If all selected studies report depression severity using scales with defined conversion factors, we will attempt to transform them into the 17-item HAM-D and use the mean difference (MD) between treatment groups in the specific study as the effect measure for the primary outcomes. However, if depression severity is measured via an alternative scale, we will standardize all the study-specific means and synthesize the data by using the standardized mean difference (SMD) as our effect measure. For binary outcomes, odds ratios (OR) will be used as effect measure. In such cases, studies reporting zero events in all treatment arms will be excluded from the analysis. Studies that provide IPD but do not report outcomes for more than 50% of the participants, will be excluded from the analysis of the relevant outcome. In studies providing IPD with a missing outcome for less than or equal to 50% of the participants, we will use multilevel models to borrow information across studies and perform multiple imputations at the network level, which involves borrowing information across studies while allowing for heterogeneity and clustering in a multilevel imputation model[40]. AD reporting studies that do not report a given outcome will be excluded from the analysis of the relevant outcome.

Transitivity checks

The validity of our NMA relies on the validity of its core assumption, namely the assumption of transitivity. This requires that all the characteristics of the included studies that act as effect modifiers have a balanced distribution across all treatment comparisons. To perform such

assessments, we plan to examine the distribution of the extracted effect modifiers using boxplots (for continuous characteristics) and barplots (for categorical characteristics)

Network Meta-Analysis

To capture the clinical and methodological differences that inevitably occur among the included studies we will fit random-effects models. For simplicity we will assume a common heterogeneity parameter across all the available treatment comparisons.

If IPD is provided for all studies, IPD-NMAs will be conducted for the overall treatment effects and the overall treatment adherence. All analyses will be conducted in R with calls to JAGS via the `rjags` package for the implementation of Bayesian hierarchical models using vague priors for all location parameters (effect sizes and regression coefficients).

The IPD-NMA will be implemented based on the model described by Saramago et al. (model 2), and adapted as needed depending on whether the outcome is continuous or binary[41].

We will assume the following model for the case of a two-arm study:

$$y_{ijk} = \begin{cases} \mu_{jb}, & \text{if } k = b \\ \mu_{jb} + \delta_{jbk}, & \text{if } k \neq b \end{cases}$$

$$\delta_{jbk} \sim N(d_{bk}, \tau^2)$$

$$d_{kq} = d_{kb} - d_{qb}$$

Equation 1

Where y_{ijk} denotes the observed outcome measure, i , j , and k are the patient, study, and treatment indices, respectively. b denotes an arbitrarily chosen reference treatment. q represents a treatment in the network which is neither k or b . μ_{jb} is the mean outcome in the reference group in study j , and δ_{jbk} represents the average relative treatment effect between treatment b and treatment k . δ_{jbk} is assumed to be normally distributed across studies around a mean of d_{bk} , with a variance of random effects τ^2 , assumed to be common for all comparisons. For binary outcomes, y_{ijk} will be assumed to come from a Bernoulli distribution

and δ_{jbk} will correspond to the log odds ratio of treatment k and treatment b. For multi-arm trials the equation above will need to be extended to multivariate normal distributions. If IPD is not available for one or more of the included studies, we will use the respective published AD. Available IPD will be synthesized together with AD from studies for which IPD is not available. The IPD/AD-NMA will be implemented in a three-step Bayesian hierarchical model in R, with calls to JAGS for the implementation of Bayesian hierarchical models to estimate the overall treatment effects and the overall treatment adherence. Models will use informative priors for heterogeneity when available. If informative priors are not available for heterogeneity and for all other location parameters, vague priors will be used. The model, based on Saramago et al. (model 3) is similar to Equation 1[41]. However, in the IPD/AD-NMA for binary outcomes, y_{ijk} is assumed to come from a binomial distribution for AD or a Bernoulli distribution for IPD. d_{bk} is determined using both the AD and IPD and the associated heterogeneity considers the heterogeneity across both sources of data. Finally, across treatments effect estimates and their standard errors will be considered in order to calculate the final relative ranking of the different competing treatments. To do so, we will rely on the ranking metric defined according to the Surface Under the Cumulative Ranking curve (SUCRA) of each treatment[42].

Consistency checks and heterogeneity estimation

To assess the statistical manifestation of transitivity, namely the consistency assumption, we plan to employ both local and global checks for consistency. For local checks, we will implement the Separating Indirect from Direct Evidence (SIDE) approach to evaluate the difference between direct and indirect estimates for each treatment comparison in the network that provides both of this sources of information, Global checks will be implemented using the design-by-treatment interaction model.

Network Meta-Regression (NMR)

The literature suggests many candidates for prognostic factors (variables associated with response regardless of the treatment) and for effect modifiers (variables associated with

differential response depending on the treatment) in the treatment of depression. However, we will only include variables in our analyses if they are particularly pertinent in the differential treatment of chronic depression in the context of psychological and pharmacological treatments. The variables will first be limited by their availability in the included original studies. When several variables that measure similar aspects are available, the research team will discuss and reach consensus on the most important predictors and decide on which should be included in the model. For systematically missing covariates, missing data will be described and reasons for missing data will be explored. If 30% of a variable's data is missing, we will consider imputation. If such a scenario arises, we will assess the appropriateness of multiple imputation methods and explore the impact of missing data on conclusions about the comparative effects on the primary outcome in sensitivity analyses. We will fit a penalized regression model, e.g. Bayesian LASSO or ridge regression. As in the network meta-analysis, models will use empirical priors for heterogeneity when available and non-informative priors (e.g. $N(0,100^2)$) for all location parameters otherwise. The NMR model will assume independent, unrelated treatment-covariate interactions for each treatment comparison between treatment effects and covariates.

If IPD is provided for all included studies, the IPD-NMR will be implemented based on the model described by Jansen et al. (model 1)[43]. We will use the following model, in which only one covariate is shown for simplicity:

$$y_{ijk} = \begin{cases} \mu_{jb} + \beta_{0j}x_{ij}, & \text{if } k = b \\ \mu_{jb} + \delta_{jbk} + \beta_{0j}x_{ij} + \beta_{1bk}x_{ij}, & \text{if } k \neq b \end{cases}$$

$$\delta_{jbk} \sim N(d_{bk}, \tau^2)$$

$$d_{kq} = d_{kb} - d_{qb}$$

$$\beta_{1,kq} = \beta_{1,kb} - \beta_{1,qb}$$

Equation 2

Where i, j, k, q and b are as defined for Equation 1. x_{ij} is the value of the covariate for individual i in study j , β_{0j} is the study-specific estimated prognostic effect of x , and β_{1bk} is the interaction between x_{ij} and the relative treatment effect b vs k . In this model, μ_{jb} denotes the mean

outcome for the control group of the study when the covariate value is 0, and δ_{jbk} denotes the effect size between treatment b and treatment k in a study when the covariate value is 0. Binary outcomes and multi-arm studies will be handled as in the case of IPD-NMA. If IPD is not available for a subset of the included studies, we will use the respective published AD and synthesize them together with IPD providing studies. The IPD/AD-NMA will be implemented based on the model described by Jansen et al (model 2)[43] as follows:

IPD:

$$y_{ijk} = \begin{cases} \mu_{jb} + \beta_{0j}x_{ij}, & \text{if } k = b \\ \mu_{jb} + \delta_{jbk} + \beta_{0j}x_{ij} + \beta_{1bk}x_{ij}, & \text{if } k \neq b \end{cases}$$

AD:

$$y_{ijk} = \begin{cases} \mu_{jb}, & \text{if } k = b \\ \mu_{jb} + \delta_{jbk} + \beta_{1bk}x_j, & \text{if } k \neq b \end{cases}$$

$$\delta_{jbk} \sim N(d_{bk}, \tau^2)$$

$$d_{kq} = d_{kb} - d_{qb}$$

$$\beta_{1,kq} = \beta_{1,kb} - \beta_{1,qb}$$

Equation 3

Where i, j, k, q , and b are as defined for Equation 1. For the IPD portion, x_{ij} , β_{0j} , β_{1bk} , μ_{jb} , δ_{jbk} , and y_{ijk} are as defined in Equation 2. For the AD subset, x_j is the average of the covariate in study j and β_{1bk} is the interaction of x_j for treatment b relative to treatment k for the aggregated covariate. The definition for μ_{jb} is the effect size for the control group of the study, and δ_{jbk} represents the effect size between treatment b and treatment k in a study when the covariate value is 0. For binary outcomes, y_{ijk} and δ_{jbk} will be handled as in the case of IPD/AD-NMA. To enable estimation using both IPD and AD, this model assumes that the treatment-by-covariate interaction regression coefficients, β_{1bk} , are the same for individual level covariates from IPD and study level covariates for AD. We will use the estimated parameters of the IPD-NMA or IPD/AD-NMA model to create a prediction model in order to provide personalised predictions according to the covariates considered. The model will take patient-specific values for covariates as inputs and provide a prediction of the outcome under each treatment of interest. Initial nodes will be defined as

stated in the eligibility criteria of the protocol. If not enough data exist for making predictions for psychotherapy types, we will merge nodes to

- Any type of psychotherapy
- Antidepressant pharmacotherapy
- Any type of psychotherapies as an adjunct (TAU, CAU, exercise, counselling, electroconvulsive therapy)
- Any type of psychotherapy as an adjunct treatment to any type of antidepressant pharmacotherapy
- Waiting list, CAU, TAU either alone or as an adjunct to any type of antidepressant pharmacotherapy

To assess the predictive performance of this model, we will use an internal-external cross-validation method. This involves taking one study out at a time, developing the model using the remaining studies, and testing the model on the left-out study. Then, we will assess the model performance in terms of calibration and discrimination for benefit, and decision accuracy, using recently developed methods[44–46].

Subgroup analyses will be performed for study level characteristics defined above. Study level characteristics will be explored using meta-regression in a Bayesian framework with informative priors where available or vague priors in all other cases.

Meta-biases

We will examine the existence of possible small study effects and publication bias both in terms of the pairwise and network level. Pairwise examinations will be held by using the contour-enhanced funnel plots for each pairwise comparison with at least 10 studies available while for examinations at the network level we will use the comparison adjusted funnel plot and plot all the study-specific treatment effects in a unique graph. In order to evaluate potential availability bias with regards to the available participant level covariates included in the model, covariate effects will be estimated in IPD studies and compared with AD studies. If considerable differences are found, this will be discussed in the final publication.

Confidence in meta-analytical estimates

To rate the quality of the best available evidence and provide a comprehensive summary, the Grading of Recommendations, Assessment, Development and Evaluation (GRADE) approach will be used. The quality of evidence will be assessed across the domains of risk of bias, consistency, directness, precision, and publication bias. Additional domains may be considered where appropriate.

When evaluating the confidence in the NMA for the primary outcomes, we will use the six domain CINeMA framework[47], which considers within-study bias, across-studies bias, indirectness, imprecision, heterogeneity and incoherence when evaluating the confidence in the NMA. Since there is no current consensus regarding the definition of clinically important effect size for change in depression severity in chronic depression, the clinically important effect size for clinical equivalence will be based on statistical difference.

Ethics and dissemination

As a systematic review of published studies, this IPD-NMA requires no ethical approval. In the event that changes to the protocol occur, changes will be reported in protocol amendments in PROSPERO. All results will be disseminated as peer-reviewed publication in a leading journal in this field and presented at (inter)national scientific conferences.

Discussion

This protocol describes a systematic review and network meta-analysis (NMA) on the efficacy of psychotherapy, pharmacotherapy or their combination in the treatment of chronic depression. As the most recent comparable systematic reviews on chronic depression are several years old, do not combine IPD and AD, and do not include various important studies[12,15], an up-to-date overview is urgently needed. To include all available RCTs and maximize statistical power, our NMA approach will allow direct and indirect comparisons based on AD as well as IPD.

In some cases, IPD meta-analysis has demonstrated the potential to produce better quality, more precise, and more reliable results than meta-analysis of only AD[48–51]. However, it is

Protected by copyright, including for uses related to text and data mining, AI training, and similar technologies.
Erasmus Hogeschool

often difficult to access IPD, and subgroup analyses of interest within IPD often will be small[41,43,52]. In order to maximize the use of the available information and attempt to reduce the overall uncertainty, complex methods that allow the inclusion of both IPD and AD can be used to evaluate overall treatment effects, irrespective of subgroups of interest. Then, regression techniques within the combined IPD and AD can enable the examination of the association between treatment effects and potential covariates.

Despite the planned combination of AD and IPD, the expected number of applicable studies fulfilling the defined inclusion and exclusion criteria is small. Thus, any subgroup analysis performed from the identified studies is also expected to be limited by small numbers and therefore to be low powered. The updated NMA will improve our understanding of the therapeutic landscape. However, uncertainty in the estimation of outcomes will likely persist. It is important to note that despite attempts to access the individual patient data of all identified studies, the possibility that some studies will not be able to provide the IPD remains. This could lead to an availability bias with regards to the available participant level covariates included in the model. Furthermore, since the NMA is based on existing studies, the analysis can only examine potential predictors or effect modifiers that are measured in the original studies. For aggregate data, this data availability is further limited to the variables that are reported in the original publication. Therefore, it is currently unclear whether all variables of interest defined in the protocol will be incorporated in the NMR model. The use of mean differences would enhance the clinical interpretability of the meta-analysis results if all scales used in the original studies are convertible to the HAM-D. However, if alternative scales are used, it will be necessary to use standardized mean differences to obtain estimates of treatment efficacy, which will limit the clinical interpretability of the results.

This systematic review and meta-analysis will focus on the inclusion of randomized controlled trials (RCTs), since they are considered less susceptible to known and unknown confounders due to the randomization and have well defined criteria, treatments, and endpoints. We plan to complete the systematic literature search, obtain aggregated or individual participant data from the relevant studies, and conduct the statistical analyses by the end of 2024. The

publication will be submitted to an international peer-reviewed journal by mid 2025. The code behind our results will be made accessible through a public GitHub repository. These codes could then potentially be used to create an interactive tool (e.g. an R-Shiny application) that will allow using our prediction model and visualize its results without requiring any programming expertise for users. Ultimately, this systematic review and meta-analysis aims to help elucidate and estimate the impact of treatment alternatives, and associated predictive and effect modifying factors, on chronic depression to enable customized treatment strategies for patients experiencing chronic depression.

Declarations

Availability of data and materials

All data sets are exclusively made available to the review team in the context of this IPD-NMA. On request, they can be obtained from the corresponding authors of the original studies.

Patient and public involvement

Not applicable.

Competing interests

E. Schramm received modest book royalties and honoraria for workshops and presentations related to the CBASP. T. A. Furukawa reports personal fees from Boehringer-Ingelheim, Daiichi Sankyo, DT Axis, Kyoto University Original, Shionogi, SONY and UpToDate, and a grant from DT Axis and Shionogi, outside the submitted work; In addition, T. A. Furukawa has a patent 7448125 and a pending patent 2022-082495, and intellectual properties for Kokoro-app licensed to Mitsubishi-Tanabe. No further disclosures were reported.

Funding

This meta-analysis is funded by the Deutsche Forschungsgemeinschaft (DFG, German Research Foundation) – Project-ID 499552394 – SFB 1597. The DFG is not involved in any other aspect of the project, such as the design of the project’s protocol and analysis plan, the collection and analyses. The funder will have no input on the interpretation or publication of the study results.

Author Contributions

ES, ME, JM, NK, TE and AN drafted the manuscript. ES and AN obtained funding. All authors contributed to the development of the selection criteria, the risk of bias assessment strategy and data extraction criteria. ES, ME and JM developed the search strategy. NK, TE, PC, OE, TAF and AN provided statistical expertise. All authors provided feedback and approved the final manuscript.

Acknowledgements

We acknowledge financial support by the Open Access Publication Fund of the University of Freiburg. Furthermore, we thank Sina Wilhelm, Magdalena Hammen, Marion Braun, and Hannah Piosczyk for their valuable contributions to the systematic literature search.

References

- Schramm E, Klein DN, Elsaesser M, *et al.* Review of dysthymia and persistent depressive disorder: history, correlates, and clinical implications. *Lancet Psychiatry*. 2020;7:801–12.
- Murphy JA, Byrne GJ. Prevalence and correlates of the proposed DSM-5 diagnosis of Chronic Depressive Disorder. *J Affect Disord*. 2012;139:172–80.
- Angst J, Gamma A, Rössler W, *et al.* Long-term depression versus episodic major depression: Results from the prospective Zurich study of a community sample. *Journal of Affective Disorders*. 2009;115:112–21.
- Köhler S, Wiethoff K, Ricken R, *et al.* Characteristics and differences in treatment outcome of inpatients with chronic vs. episodic major depressive disorders. *J Affect Disord*. 2015;173:126–33.
- McIntyre RS, Alsuwaidan M, Baune BT, *et al.* Treatment-resistant depression: definition, prevalence, detection, management, and investigational interventions. *World Psychiatry*. 2023;22:394–412.
- McCullough JP. *Treatment for chronic depression. Cognitive behavioral analysis system of psychotherapy*. NY: Guilford Press. 2000.
- McCullough JP. Treatment for chronic depression using Cognitive Behavioral Analysis System of Psychotherapy (CBASP). *J Clin Psychol*. 2003;59:833–46.
- McCullough J. Characteristics of the Optimal Cognitive Behavioral Analysis System of Psychotherapy (CBASP) Therapist Role. *Frontiers in Psychiatry*. 2021;11. doi: 10.3389/fpsy.2020.609954
- Beck AT. *Cognitive Therapy of Depression*. Guilford Press 1979.

- 10 Klerman GL, Weissman MM, Rounsaville BJ, *et al.* *Interpersonal psychotherapy of depression*. NY: Basic Books 1984.
- 11 Jobst A, Brakemeier E-L, Buchheim A, *et al.* European Psychiatric Association Guidance on psychotherapy in chronic depression across Europe. *Eur Psychiatry*. 2016;33:18–36.
- 12 Kriston L, von Wolff A, Westphal A, *et al.* Efficacy and acceptability of acute treatments for persistent depressive disorder: a network meta-analysis. *Depress Anxiety*. 2014;31:621–30.
- 13 Negt P, Brakemeier E-L, Michalak J, *et al.* The treatment of chronic depression with cognitive behavioral analysis system of psychotherapy: a systematic review and meta-analysis of randomized-controlled clinical trials. *Brain Behav*. 2016;6:e00486.
- 14 Cuijpers P, Noma H, Karyotaki E, *et al.* A network meta-analysis of the effects of psychotherapies, pharmacotherapies and their combination in the treatment of adult depression. *World Psychiatry*. 2020;19:92–107.
- 15 Furukawa TA, Efthimiou O, Weitz ES, *et al.* Cognitive-Behavioral Analysis System of Psychotherapy, Drug, or Their Combination for Persistent Depressive Disorder: Personalizing the Treatment Choice Using Individual Participant Data Network Metaregression. *PPS*. 2018;87:140–53.
- 16 Schramm E, Kriston L, Zobel I, *et al.* Effect of Disorder-Specific vs Nonspecific Psychotherapy for Chronic Depression: A Randomized Clinical Trial. *JAMA Psychiatry*. 2017;74:233–42.
- 17 Rief W, Bleichhardt G, Dannehl K, *et al.* Comparing the Efficacy of CBASP with Two Versions of CBT for Depression in a Routine Care Center: A Randomized Clinical Trial. *Psychother Psychosom*. 2018;87:164–78.
- 18 Schramm E, Kriston L, Elsaesser M, *et al.* Two-Year Follow-Up after Treatment with the Cognitive Behavioral Analysis System of Psychotherapy versus Supportive Psychotherapy for Early-Onset Chronic Depression. *PPS*. 2019;88:154–64.
- 19 Elsaesser M, Feige B, Kriston L, *et al.* Longitudinal Clusters of Long-Term Trajectories in Patients with Early-Onset Chronic Depression: 2 Years of Naturalistic Follow-Up after Extensive Psychological Treatment. *Psychotherapy and Psychosomatics*. 2023;1–9.
- 20 Elsaesser M, Herpertz S, Piosczyk H, *et al.* Modular-based psychotherapy (MoBa) versus cognitive-behavioural therapy (CBT) for patients with depression, comorbidities and a history of childhood maltreatment: study protocol for a randomised controlled feasibility trial. *BMJ Open*. 2022;12:e057672.
- 21 Schramm E, Elsaesser M, Jenkner C, *et al.* Algorithm-based modular psychotherapy vs. cognitive-behavioral therapy for patients with depression, psychiatric comorbidities and early trauma: a proof-of-concept randomized controlled trial. *World Psychiatry*. 2024;23:257–66.
- 22 Elsaesser M, Furukawa TA, Schramm E. Why one answer is not always enough: Reply to “CBASP may not be superior to other treatments for chronic depression.” *J Affect Disord*. 2022;312:144–5.
- 23 Furukawa TA, Schramm E, Weitz ES, *et al.* Cognitive-Behavioural Analysis System of Psychotherapy (CBASP), a drug, or their combination: differential therapeutics for persistent

- depressive disorder: a study protocol of an individual participant data network meta-analysis. *BMJ Open*. 2016;6:e011769.
- 24 Efthimiou O, Debray TPA, van Valkenhoef G, *et al*. GetReal in network meta-analysis: a review of the methodology. *Research Synthesis Methods*. 2016;7:236–63.
- 25 Montgomery SA, Asberg M. A new depression scale designed to be sensitive to change. *Br J Psychiatry*. 1979;134:382–9.
- 26 Rush AJ, Trivedi MH, Ibrahim HM, *et al*. The 16-Item quick inventory of depressive symptomatology (QIDS), clinician rating (QIDS-C), and self-report (QIDS-SR): a psychometric evaluation in patients with chronic major depression. *Biological Psychiatry*. 2003;54:573–83.
- 27 Carrozzino D, Patierno C, Fava GA, *et al*. The Hamilton Rating Scales for Depression: A Critical Review of Clinimetric Properties of Different Versions. *PPS*. 2020;89:133–50.
- 28 Hamilton M. A rating scale for depression. *J Neurol Neurosurg Psychiatry*. 1960;23:56–62.
- 29 Hamilton M. Development of a Rating Scale for Primary Depressive Illness. *British Journal of Social and Clinical Psychology*. 1967;6:278–96.
- 30 Veal C, Tomlinson A, Cipriani A, *et al*. Heterogeneity of outcome measures in depression trials and the relevance of the content of outcome measures to patients: a systematic review. *The Lancet Psychiatry*. 2024;11:285–94.
- 31 Leucht S, Fennema H, Engel RR, *et al*. Translating the HAM-D into the MADRS and vice versa with equipercentile linking. *Journal of Affective Disorders*. 2018;226:326–31.
- 32 Beck AT, Steer RA, Brown GK. *Manual for the Beck Depression Inventory-II*. San Antonio, TX: Psychological Corporation 1996.
- 33 Rush AJ, Gullion CM, Basco MR, *et al*. The Inventory of Depressive Symptomatology (IDS): psychometric properties. *Psychological Medicine*. 1996;26:477.
- 34 Furukawa TA, Reijnders M, Kishimoto S, *et al*. Translating the BDI and BDI-II into the HAMD and vice versa with equipercentile linking. *Epidemiology and Psychiatric Sciences*. 2020;29:e24.
- 35 Gameroff MJ, Wickramaratne P, Weissman MM. Testing the Short and Screener versions of the Social Adjustment Scale – Self-report (SAS-SR). *Int J Methods Psychiatr Res*. 2011;21:52–65.
- 36 ANGERMEYER C. WHOQOL-100 und WHOQOL-BREF. *Handbuch für die deutschsprachige Version der WHO Instrumente zur Erfassung der Lebensqualität*. Published Online First: 2000.
- 37 Ware JE, Sherbourne CD. The MOS 36-item short-form health survey (SF-36). I. Conceptual framework and item selection. *Med Care*. 1992;30:473–83.
- 38 McHorney CA, Ware JE, Raczek AE. The MOS 36-Item Short-Form Health Survey (SF-36): II. Psychometric and clinical tests of validity in measuring physical and mental health constructs. *Med Care*. 1993;31:247–63.
- 39 Sterne JAC, Savović J, Page MJ, *et al*. RoB 2: a revised tool for assessing risk of bias in randomised trials. *BMJ*. 2019;366:l4898.

Quartagno M, Grund S, Carpenter J. jomo: A Flexible Package for Two-level Joint Modelling Multiple Imputation. *The R Journal*. 2019;11:205.

Saramago P, Sutton AJ, Cooper NJ, *et al*. Mixed treatment comparisons using aggregate and individual participant level data. *Statistics in Medicine*. 2012;31:3516–36.

Salanti G, Ades AE, Ioannidis JPA. Graphical methods and numerical summaries for presenting results from multiple-treatment meta-analysis: an overview and tutorial. *Journal of Clinical Epidemiology*. 2011;64:163–71.

Jansen JP. Network meta-analysis of individual and aggregate level data. *Research Synthesis Methods*. 2012;3:177–90.

van Klaveren D, Steyerberg EW, Serruys PW, *et al*. The proposed ‘concordance-statistic for benefit’ provided a useful metric when modeling heterogeneous treatment effects. *Journal of Clinical Epidemiology*. 2018;94:59–68.

Efthimiou O, Hoogland J, Debray TPA, *et al*. Measuring the performance of prediction models to personalize treatment choice. *Stat Med*. 2023;42:1188–206.

Hoogland J, Efthimiou O, Nguyen TL, *et al*. Evaluating individualized treatment effect predictions: a model-based perspective on discrimination and calibration assessment. 2023. <http://arxiv.org/abs/2209.06101> (accessed 27 May 2024).

Nikolakopoulou A, Higgins JPT, Papakonstantinou T, *et al*. CINeMA: An approach for assessing confidence in the results of a network meta-analysis. *PLOS Medicine*. 2020;17:e1003082.

Balduzzi S, Rücker G, Schwarzer G. How to perform a meta-analysis with R: a practical tutorial. *BMJ Ment Health*. 2019;22:153–60.

Kanters S, Karim ME, Thorlund K, *et al*. When does the use of individual patient data in network meta-analysis make a difference? A simulation study. *BMC Medical Research Methodology*. 2021;21:21.

Lambert PC, Sutton AJ, Abrams KR, *et al*. A comparison of summary patient-level covariates in meta-regression with individual patient data meta-analysis. *Journal of Clinical Epidemiology*. 2002;55:86–94.

Riley RD, Dias S, Donegan S, *et al*. Using individual participant data to improve network meta-analysis projects. *BMJ Evidence-Based Medicine*. 2023;28:197–203.

Sutton AJ, Kendrick D, Coupland C a. C. Meta-analysis of individual- and aggregate-level data. *Statistics in Medicine*. 2008;27:651–69.

BMJ Open

Efficacy of psychotherapy versus pharmacotherapy, or their combination, in chronic depression: study protocol for a systematic review and network meta-analysis using aggregated and individual patient data

Journal:	<i>BMJ Open</i>
Manuscript ID	bmjopen-2024-089356.R1
Article Type:	Protocol
Date Submitted by the Author:	28-Jan-2025
Complete List of Authors:	Schramm, Elisabeth; University of Freiburg, Department of Psychiatry and Psychotherapy Elsaesser, Moritz; University of Freiburg, Department of Psychiatry and Psychotherapy, University Medical Center – University of Freiburg, Faculty of Medicine Müller, Julia; University of Freiburg, Department of Psychiatry and Psychotherapy Kwarteng, Nana-Adjoa; University of Freiburg, Institute of Medical Biometry and Statistics Evrenoglou, Theodoros; University of Freiburg, Institute of Medical Biometry and Statistics Cuijpers, Pim; VU University Amsterdam, Department of Clinical Psychology; Babeş-Bolyai University, International Institute for Psychotherapy Orestis, Efthimiou; University of Bern, Institute of Primary Health Care (BIHAM) Klein, Daniel; Stony Brook University, Department of Psychology Keller, Martin; The Warren Alpert Medical School of Brown University, Department of Psychiatry and Human Behavior Furukawa, Toshi; Kyoto University, Graduate School of Medicine and School of Public Health Nikolakopoulou, Adriani ; University of Freiburg, Institute of Medical Biometry and Statistics
Primary Subject Heading:	Mental health
Secondary Subject Heading:	Mental health
Keywords:	Systematic Review, Meta-Analysis, Network Meta-Analysis, Depression & mood disorders < PSYCHIATRY, Adult psychiatry < PSYCHIATRY, PSYCHIATRY



1
2
3
4
5
6
7
8
9
10
11
12
13
14
15
16
17
18
19
20
21
22
23
24
25
26
27
28
29
30
31
32
33
34
35
36
37
38
39
40
41
42
43
44
45
46
47
48
49
50
51
52
53
54
55
56
57
58
59
60

Efficacy of psychotherapy versus pharmacotherapy, or their combination, in chronic depression: study protocol for a systematic review and network meta-analysis using aggregated and individual patient data

Schramm, E.^{1*}, Elsaesser, M.^{1*}, Mueller, J.¹, Kwarteng, N.², Evrenoglou, T.², Cuijpers, P.^{3,4}, Efthimiou, O.⁵, Klein, D. N.⁶, Keller, M. B.⁷, Furukawa, T. A.^{8*} & Nikolakopoulou, A.^{2*}

¹ Department of Psychiatry and Psychotherapy, Medical Center – University of Freiburg, Faculty of Medicine, University of Freiburg, Freiburg, Germany

² Institute of Medical Biometry and Statistics, Faculty of Medicine and Medical Center, University of Freiburg, Freiburg im Breisgau, Germany

³ Department of Clinical, Neuro and Developmental Psychology, WHO Collaborating Centre for Research and Dissemination of Psychological Interventions, Vrije Universiteit Amsterdam, Amsterdam, The Netherlands

⁴ Babeş-Bolyai University, International Institute for Psychotherapy, Cluj-Napoca, Romania

⁵ Institute of Primary Health Care (BIHAM), University of Bern, Bern, Switzerland

⁶ Department of Psychology, Stony Brook University, Stony Brook, New York, United States

⁷ Department of Psychiatry and Human Behavior, The Warren Alpert Medical School of Brown University, Providence, USA

⁸ Department of Health Promotion and Human Behavior, Kyoto University Graduate School of Medicine/School of Public Health, Kyoto, Japan

*Shared first/senior authorship

Word count: 4851

***Correspondence to:**

Prof Dr Elisabeth Schramm
Department of Psychiatry and Psychotherapy
University Medical Center – University of Freiburg,
Hauptstraße 5, DE–79104 Freiburg (Germany)
✉ elisabeth.schramm@uniklinik-freiburg.de

Abstract

Introduction: Chronic depression represents a common and highly disabling disorder. Several randomised controlled trials investigated the effectiveness of psychological, pharmacological, and combined treatments for chronic depression. This is the first overarching systematic review and network meta-analysis based on aggregated and individual patient data comparing the efficacy and acceptability of various treatment options for all subtypes of chronic depression. Furthermore, individual demographic and clinical characteristics that predict or moderate therapy outcomes will be investigated.

Methods and analysis: A systematic literature search of the Cochrane Library, MEDLINE via Ovid, PsycINFO, Web of Science, and metapsy databases will be conducted from database inception without language restrictions to include all available samples from randomised controlled trials (RCTs) that investigated the efficacy of psychotherapy vs. pharmacotherapy, or their combination in adult inpatients or outpatients with a primary diagnosis of chronic depression. Exclusively internet-based treatment studies will be excluded. The main outcome is depression severity measured on a continuous observer-rated scale for depression at six months post treatment (range 3-12 months). Two reviewers will independently screen and select eligible studies based on the pre-defined inclusion and exclusion criteria. Risk of bias will be assessed using Version 2 of the Cochrane risk-of-bias tool for randomized trials (Rob 2.0). Individual patient data (IPD) will be requested and incorporated in the network when provided, as it is the gold standard of evidence. For studies which do not provide individual patient data (IPD), aggregate data (AD) will be extracted and incorporated in lieu of IPD for the network meta-analysis (NMA), strengthening the evidence base and leveraging all existing evidence regardless of data availability restrictions. An NMA comparing psychotherapies and a network meta-regression (NMR) estimating individualised treatment effects of psychotherapy will be implemented assuming a Bayesian framework. All models will be fitted in R with calls to JAGS. Empirical informative prior distributions will be used for model parameters where available, and non-informative priors will be used in cases where empirical priors are not available.

Protected by copyright, including for uses related to text and data mining, AI training, and similar technologies.
Erasmus Hogeschool

Ethics and dissemination: This IPD-NMA requires no ethical approval. All results will be disseminated as peer-reviewed publication in a leading journal in this field and presented at (inter)national scientific conferences.

Study registration: PROSPERO, CRD42024526755 (17 April 2024).

Keywords: Chronic Depression; Persistent Depressive Disorders; Psychological Treatment; Psychotherapy; Pharmacological Therapy; Combination Treatment; Systematic Review; Meta-Analysis; Network Meta-analysis; Individual Patient Data; Individual Participant Data.

Strengths and limitations of this study

- To include all available randomised controlled trials and maximise statistical power, we will perform a network meta-analysis (NMA) and synthesise evidence based on both aggregated data (AD) and individual participant data (IPD).
- The results will offer guidance about the most effective treatment approaches for clinicians and patients seeking optimal individualised management strategies.
- A network meta-regression (NMR) model adjusted in terms of individual demographic and clinical characteristics, that impact therapy outcomes, will be fitted to yield individualised treatment recommendations.
- The IPD-NMA will not be able to examine variables that have not been measured in the original studies.
- The evidence base may yield different forms of sparse data such as sparse networks, small number of studies, and small subgroups, which will likely lead to persisting uncertainty around synthesized estimates of primary and secondary outcomes.

INTRODUCTION

Chronic depression is a common and long-term disabling disorder[1]. Up to one third of all depressive disorders take a chronic course[2] with the definition of chronic varying in the literature regarding duration (between a minimum of one to three years), severity (dysthymia, chronic major depression) and the type of course since the first onset (dysthymia with or without superimposed major depressive episode(s), chronic major depression, or recurrent major depression without inter-episode recovery[3]). In 2013, the DSM-5 merged these various presentations into the rubric of persistent depressive disorders (PDD) with defined diagnostic criteria present for at least 2 years and four coding specifiers. Unlike major depressive disorder, chronic depression tends to have a more subtle but persistent presentation, leading to comparably more significant impairments in functioning and more reduced quality of life[4]. Despite new developments of treatment approaches in the past decades, chronic depression remains challenging to treat, with varying degrees of success reported for different modalities. Common treatment complicating factors such as comorbidity with mental and physical disorders, poor social integration, early onset, or a history of early trauma lead to a high rate of treatment-resistance[5].

Psychotherapy and pharmacotherapy are the two primary treatment modalities used to manage chronic depression. Psychotherapy encompasses various approaches such as the Cognitive Behavioral Analysis System of Psychotherapy (CBASP[6–8]) as the only model specifically designed for chronic depression, and more general methods such as Cognitive-Behavioral Therapy (CBT[9]), Interpersonal Therapy (IPT[10]), or Psychodynamic and Psychoanalytic Therapy[11,12]. Pharmacotherapy, as the most commonly used treatment, involves antidepressant medications including selective serotonin reuptake inhibitors (SSRIs), serotonin-norepinephrine reuptake inhibitors (SNRIs), tricyclic antidepressants, and others, to modulate neurotransmitter activity in the brain.

Furthermore, psychotherapy and pharmacotherapy are integrated into combination therapy to potentially yield faster and better treatment outcomes. Despite the availability of these treatment options, uncertainty remains regarding their relative efficacy and differential

Protected by copyright, including for uses related to text and data mining, AI training, and similar technologies.
Erasmus Hogeschool

response in managing chronic and treatment-resistant depression. While some studies suggest that combination therapy may be more effective than either psychotherapy or pharmacotherapy alone[13,14], others indicate that certain monotherapies are particularly suitable for specific patient profiles or severity levels[15]. The last comprehensive meta-analysis specifically about chronic depression dates from 2014 and recommends different approaches depending on the subtype of chronic depression[12]. However, within the last ten years, several new studies have appeared that have not yet been synthesized with the previous evidence (e.g. [16–22]). An individual participant data network meta-regression (NMR) investigated the efficacy of CBASP, pharmacotherapy or their combination and several effect moderators, but included only three studies[15,23]. In a fairly recent network meta-analysis on adult depression, a subgroup analysis showed superior effects for combined treatment versus psychotherapy or pharmacotherapy[14]. However, this subgroup analysis did not distinguish between chronic courses and treatment-resistant depression and did not consider long-term effects.

Therefore, an up-to-date comprehensive systematic review on the efficacy of psychotherapy vs. pharmacotherapy, and their combination in treating chronic depression is urgently needed. Our network meta-analysis (NMA) will incorporate all available randomised controlled trials (RCTs) to capture the full breadth of evidence[24]. By analysing individual patient data (IPD) as well as aggregate data (AD; if IPD is not available) from a range of studies, we aim to obtain insights into the most effective treatment approaches, identify potential predicting and moderating factors, and offer guidance for clinicians and patients seeking optimal management strategies. The results of this meta-analysis may have significant implications for the individualised treatment of chronic depression, informing clinical practice and shaping future research in this field.

Objectives

The aim of this systematic review is to evaluate the efficacy of psychological vs. pharmacological, or combination treatments in chronic depression. To this end, the proposed systematic review will answer the following questions:

1. Which psychological vs. pharmacological, or combination treatments are most effective for chronic depression?
2. How do individual demographic and clinical characteristics predict or moderate individualised treatment recommendations?

METHODS AND ANALYSIS

Eligibility criteria

We will include data from RCTs that investigated treatment effects in adult (>18 years old) inpatients or outpatients with a primary diagnosis of chronic depression. Depressive courses are defined as chronic when lasting for at least two years (according to the definition of the classification systems of DSM-5 and ICD-11) including persistent depressive disorders (DSM-5), chronic major depression, double depression (dysthymia with superimposed major depressive episode), recurrent major depression with incomplete inter-episode recovery, dysthymia, or any corresponding conditions according to standard operationalized diagnostic criteria as the primary diagnosis. Comorbid disorders are allowed and all respective information will be collected.

The treatments of interest in our IPD-NMA include:

- Different types of psychotherapies (including Cognitive Behavioral Analysis System of Psychotherapy, Cognitive-Behavioral Therapy, Interpersonal Therapy, Modular-Based Psychotherapy, Metacognitive Training, Dialectical Behavior Therapy, Mindfulness-based Cognitive therapy, (Long-term) Psychoanalytic Psychotherapy, (Brief) Supportive Psychotherapy, Group Person-based Cognitive Therapy, Cognitive Therapy, Problem-solving Treatment, Schema Therapy).
- Antidepressant pharmacotherapy as a comparator to psychotherapy including any of the antidepressive agents licensed for the treatment of major depression in the country where the trial was conducted.
- Different types of psychotherapies mentioned above as an adjunct treatment to other treatments, e.g. treatment as usual (TAU), care as usual (CAU), exercise, counselling.

- Different types of psychotherapies mentioned above as an adjunct treatment to any type of antidepressant pharmacotherapy (combination treatment).
- If waitlist control is included in the screened studies, it will be used as the primary reference category when reporting relative treatment effects. CAU and TAU conditions will be compared with each other and examined with regard to their similarity. If substantial differences between TAU and CAU are detected, we will consider whether they need to be split into meaningful categories. Otherwise they will be merged and treated as one single treatment (i.e. TAU/CAU) in the statistical analyses.
- Only studies implementing face-to-face psychotherapy will be included; exclusively internet-based treatment studies will be excluded.

Information sources and search strategy

We will first conduct an electronic literature search of the Cochrane Library, MEDLINE via Ovid, PsycINFO, Web of Science, and metapsy databases with the keywords “psychotherapy”, “chronic depression” and all related terms. Initial searching will be conducted in March 2024. Searches will be re-run just before the final analyses and any further studies identified will be retrieved for inclusion. There will be no restriction for the publication period. As an example, this is the final search string for OVID:

((chroni*5 adj3 depress*).ti,ab. OR (chroni*5 adj3 MDD).ti,ab. OR exp Depressive Disorder, Treatment-Resistant/ OR exp Dysthymic Disorder/ OR (treatment-resistan* adj3 depress*).ti,ab. OR (treatment-resistan* adj3 MDD).ti,ab. OR (therapy-resistan* adj3 depress*).ti,ab. OR (therapy-resistan* adj3 MDD).ti,ab. OR (dysthymia or (dysthym* adj2 disorder*1)).ti,ab. OR (persist* adj2 depress*).ti,ab. OR (persist* adj2 MDD).ti,ab. OR (double depression).ti,ab.) AND (exp Psychotherapy/ or psychotherap*.ti,ab. OR (psychologi* adj3 treatmen*2).ti,ab. OR (psychologi* adj3 interventio*2).ti,ab. OR CBASP.ti,ab. OR ((cognitive adj2 therapy).ti,ab. OR (behavior* adj1 therapy).ti,ab. OR mindfulness.ti,ab. OR psychoanalytic.ti,ab. OR psychodynamic.ti,ab. OR (schema therapy).ti,ab.)

In accordance with the Cochrane handbook, preprints will be considered a potentially relevant source of study evidence and will therefore be assessed for eligibility. Backward and forward citation searches of included studies and relevant reviews will be performed. To find references in the grey literature, we will contact authors of included studies and relevant conference abstracts of unpublished studies. Furthermore, we will use relevant mailing lists, for example from the Society for Psychotherapy Research or (inter)national professional associations, to draw attention to this project.

Study selection

At least two reviewers will independently screen the title and abstract of all records of the systematic literature search and select eligible studies based on the pre-defined inclusion and exclusion criteria. If both reviewers independently determine that a study may be eligible based on title and abstract screening, then a full-text article review will be completed. Disagreements between individual judgements will be resolved via discussion with a third reviewer. In case of ongoing disagreement, a meeting will be held with the complete study team involving all reviewers allowing them to present the reasoning for their judgement. Consensus should be reached after debate and decisions will be documented in written form.

Data extraction

The IPD of the originally established datasets will be requested from the authors of all eligible studies. One person will check the received data for completion. All obtained IPD will be cleaned, coded, and saved in appropriate files to make the data as uniform as possible. Afterwards, we will compare the published data of each dataset (i.e. numbers and percentages, or means and standard deviations of baseline demographics as well as clinical variables) with the summaries that we will obtain directly from the IPD. Any major inconsistencies will be discussed and followed-up. Corrections will be made as necessary. In cases where IPD are not provided within a reasonable time after the request, AD will be extracted from the publication and used instead. Two individuals will be responsible for performing the data

Protected by copyright, including for uses related to text and data mining, AI training, and similar technologies.
Erasmus Hogeschool

extraction. One extractor will perform the data extraction while the second will perform the quality control and ensure that all data are properly extracted.

When extracting AD, the following data points will be collected:

- Study level data: Year of publication; validated depression scale(s) used in study.
- Participants: Mean age at baseline; mean age at first depressive episode; mean duration of chronic depression; proportion of participants with prior antidepressant treatment; proportion of participants with current antidepressant treatment; proportion of participants with prior psychotherapeutic treatment; other covariates (e.g. childhood maltreatment, loneliness, attachment style, personality, rejection sensitivity, mentalisation, emotion regulation, comorbidities, severity, interaction style, avoidance, social functioning, alexithymia, empathy).
- Outcomes per arm: mean follow-up duration; mean baseline, post-treatment, and follow-up depression score, and standard deviation; mean change from baseline depression score and standard deviation; number of participants at baseline, post-treatment, and follow-up; percentage of dropouts (treatment discontinuation); side effects; (serious) adverse events.

In case of repeated measures, the time point of each repeated measure will also be extracted.

Disagreements between individual judgements will be resolved via discussion. In case of discrepancies, a meeting will be held involving both extractors allowing them to present the reasoning for their judgement. A consensus should be reached after debate and decisions should be documented in meeting minutes. Data will be recorded in excel spreadsheets.

Outcomes

The main outcome is depression severity measured on a continuous observer-rated scale for depression at six months post treatment (range 3-12 months). If the respective study reports results at two or more time points within this frame, we will prioritize the time point closest to six months. If time points are equidistant, we will use the later one. Where different scales such as the Montgomery-Asberg Depression Rating Scale (MADRS[25]), the Quick Inventory of Depressive Symptomatology - Clinician Rating (QIDS-C[26]) or different versions of the

Hamilton Rating Scale for Depression (HAM-D/HRSD[27]) are reported, we will attempt to transform their respective scores to the 17-item HRSD score[28,29] as the most frequently utilized measure[30] using conversion procedures (e.g. <http://ids-qids.org/interpretation.html> or [31]).

Secondary outcomes are:

1. Treatment response, defined as 50% or greater reduction from baseline to study endpoint in the study's primary observer-rated or self-rated depression scale.
2. Remission, defined as scoring below the validated thresholds of the study's primary observer-rated or self-rated depression scale at endpoint.
3. Study drop-out for any reason, as a proxy measure of overall treatment acceptability.
4. Study drop-out due to side effects.
5. Depression severity as measured on a continuous self-rating scale for depression, such as Beck Depression Inventory-II (BDI-II[32]) or Inventory of Depressive Symptomatology, Self-Report (IDS-SR[33]). Different scales will be converted into BDI-II using conversion tables and equipercentile linking[34].
6. Global Assessment of Functioning (GAF).
7. Social functioning, as measured by any validated measure for impaired social functioning such as the Social Adjustment Scale-Self Report (SAS-SR[35]).
8. Quality of Life, as measured by any validated measure for life quality such as the World Health Organization Quality of Life (WHOQOL[36]) or the Short Form 36 Health Survey (SF-36[37,38]).
9. Side effects.
10. Adverse and serious adverse events.

Risk of bias assessment

We will assess risk of bias (RoB) in the included studies using the Cochrane RoB 2 tool for randomized trials[39,40]. The assessment will be done by two independent raters who have successfully completed an official RoB 2 workshop by Cochrane Germany. If raters disagree, the final rating will be made by consensus with involvement of another member of the review

Protected by copyright, including for uses related to text and data mining, AI training, and similar technologies. Erasmushogeschool

group (if necessary). We will evaluate RoB in the following domains: Bias arising from the randomization process, bias due to deviations from intended interventions, bias due to missing outcome data, bias in measurement of the outcome, and bias in selection of the reported result. Please note that, when IPD are available, the RoB assessments can be different from those based on AD. For example, even when the original authors used alternative imputation methods for handling missing outcomes (e.g. last observation carried forward (LOCF)), we can apply different imputation methods more suitable for the current use case (e.g. multiple imputation) on the IPD. We will assess the RoB for the outcomes used for the primary outcome of our NMA.

Data synthesis

Measurements of depression severity are often reported on validated continuous observer-rated scales for depression. If all selected studies report depression severity using scales with defined conversion factors, we will attempt to transform them into the 17-item HAM-D and use the mean difference (MD) between treatment groups in the specific study as the effect measure for the primary outcomes. In studies where multiple scales are reported, all scales reported in the articles will be extracted and those that can be converted to HAM-D will be used for the analysis. In cases where multiple scales are reported and more than one can be converted to HAM-D, or if neither scale can be converted, the scale that is more commonly used across studies will be selected for consistency and comparability. However, if depression severity is measured via an alternative scale, we will standardize all the study-specific means and synthesize the data by using the standardized mean difference (SMD) as our effect measure. For binary outcomes, odds ratios (OR) will be used as effect measure. In such cases, studies reporting zero events in all treatment arms will be excluded from the analysis. Studies that provide IPD but do not report outcomes for more than 50% of the participants, will be excluded from the analysis of the relevant outcome. In studies providing IPD with a missing outcome for less than or equal to 50% of the participants, we will use multilevel models to borrow information across studies and perform multiple imputations at the network level, which involves borrowing information across studies while allowing for heterogeneity and clustering

in a multilevel imputation model[41]. AD reporting studies that do not report a given outcome will be excluded from the analysis of the relevant outcome.

Transitivity checks

The validity of our NMA relies on the validity of its core assumption, namely the assumption of transitivity. This requires that all the characteristics of the included studies that act as effect modifiers have a balanced distribution across all treatment comparisons. To perform such assessments, we plan to examine the distribution of the extracted effect modifiers using boxplots (for continuous characteristics) and barplots (for categorical characteristics)

Network meta-analysis

To capture the clinical and methodological differences that inevitably occur among the included studies we will fit random-effects models. For simplicity we will assume a common heterogeneity parameter across all the available treatment comparisons.

If IPD is provided for all studies, IPD-NMAs will be conducted for the overall treatment effects and the overall treatment adherence. All analyses will be conducted in R with calls to JAGS via the `rjags` package for the implementation of Bayesian hierarchical models using vague priors for all location parameters (effect sizes and regression coefficients).

The IPD-NMA will be implemented based on the model described by Saramago et al. (model 2), and adapted as needed depending on whether the outcome is continuous or binary[42].

We will assume the following model for the case of a two-arm study:

$$y_{ijk} = \begin{cases} \mu_{jb}, & \text{if } k = b \\ \mu_{jb} + \delta_{jbk}, & \text{if } k \neq b \end{cases}$$

$$\delta_{jbk} \sim N(d_{bk}, \tau^2)$$

$$d_{kq} = d_{kb} - d_{qb}$$

Equation 1

Where y_{ijk} denotes the observed outcome measure, i , j , and k are the patient, study, and treatment indices, respectively. b denotes an arbitrarily chosen reference treatment. q represents a treatment in the network which is neither k or b . μ_{jb} is the mean outcome in the

reference group in study j , and δ_{jbk} represents the average relative treatment effect between treatment b and treatment k . δ_{jbk} is assumed to be normally distributed across studies around a mean of d_{bk} , with a variance of random effects τ^2 , assumed to be common for all comparisons. For binary outcomes, y_{ijk} will be assumed to come from a Bernoulli distribution and δ_{jbk} will correspond to the log odds ratio of treatment k and treatment b . For multi-arm trials the equation above will need to be extended to multivariate normal distributions.

If IPD is not available for one or more of the included studies, we will use the respective published AD. Available IPD will be synthesized together with AD from studies for which IPD is not available. The IPD/AD-NMA will be implemented in a three-step Bayesian hierarchical model in R, with calls to JAGS for the implementation of Bayesian hierarchical models to estimate the overall treatment effects and the overall treatment adherence. Models will use informative priors for heterogeneity when available. If informative priors are not available for heterogeneity and for all other location parameters, vague priors will be used. The model, based on Saramago et al. (model 3) is similar to Equation 1[42]. However, in the IPD/AD-NMA for binary outcomes, y_{ijk} is assumed to come from a binomial distribution for AD or a Bernoulli distribution for IPD. d_{bk} is determined using both the AD and IPD and the associated heterogeneity considers the heterogeneity across both sources of data.

Finally, across treatments effect estimates and their standard errors will be considered in order to calculate the final relative ranking of the different competing treatments. To do so, we will rely on the ranking metric defined according to the Surface Under the Cumulative Ranking curve (SUCRA) of each treatment[43].

Consistency checks and heterogeneity estimation

To assess the statistical manifestation of transitivity, namely the consistency assumption, we plan to employ both local and global checks for consistency. For local checks, we will implement the Separating Indirect from Direct Evidence (SIDE) approach to evaluate the difference between direct and indirect estimates for each treatment comparison in the network

that provides both of this sources of information, Global checks will be implemented using the design-by-treatment interaction model.

Network meta-regression (NMR)

The literature suggests many candidates for prognostic factors (variables associated with response regardless of the treatment) and for effect modifiers (variables associated with differential response depending on the treatment) in the treatment of depression. However, we will only include variables in our analyses if they are particularly pertinent in the differential treatment of chronic depression in the context of psychological and pharmacological treatments. The variables will first be limited by their availability in the included original studies. When several variables that measure similar aspects are available, the research team will discuss and reach consensus on the most important predictors and decide on which should be included in the model. For systematically missing covariates, missing data will be described and reasons for missing data will be explored. If 30% of a variable's data is missing, we will consider imputation. If such a scenario arises, we will assess the appropriateness of multiple imputation methods and explore the impact of missing data on conclusions about the comparative effects on the primary outcome in sensitivity analyses. We will fit a penalized regression model, e.g. Bayesian LASSO or ridge regression. As in the network meta-analysis, models will use empirical priors for heterogeneity when available and non-informative priors (e.g. $N(0,100^2)$) for all location parameters otherwise. The NMR model will assume independent, unrelated treatment-covariate interactions for each treatment comparison between treatment effects and covariates.

If IPD is provided for all included studies, the IPD-NMR will be implemented based on the model described by Jansen et al. (model 1)[44]. We will use the following model, in which only one covariate is shown for simplicity:

$$y_{ijk} = \begin{cases} \mu_{jb} + \beta_{0j}x_{ij}, & \text{if } k = b \\ \mu_{jb} + \delta_{jbk} + \beta_{0j}x_{ij} + \beta_{1bk}x_{ij}, & \text{if } k \neq b \end{cases}$$

$$\delta_{jbk} \sim N(d_{bk}, \tau^2)$$

$$d_{kq} = d_{kb} - d_{qb}$$

$$\beta_{1,kq} = \beta_{1,kb} - \beta_{1,qb}$$

Equation 2

Where i, j, k, q and b are as defined for Equation 1. x_{ij} is the value of the covariate for individual i in study j , β_{0j} is the study-specific estimated prognostic effect of x , and β_{1bk} is the interaction between x_{ij} and the relative treatment effect b vs k . In this model, μ_{jb} denotes the mean outcome for the control group of the study when the covariate value is 0, and δ_{jbk} denotes the effect size between treatment b and treatment k in a study when the covariate value is 0.

Binary outcomes and multi-arm studies will be handled as in the case of IPD-NMA.

If IPD is not available for a subset of the included studies, we will use the respective published AD and synthesize them together with IPD providing studies. The IPD/AD-NMA will be implemented based on the model described by Jansen et al (model 2)[44] as follows:

IPD:

$$y_{ijk} = \begin{cases} \mu_{jb} + \beta_{0j}x_{ij}, & \text{if } k = b \\ \mu_{jb} + \delta_{jbk} + \beta_{0j}x_{ij} + \beta_{1bk}x_{ij}, & \text{if } k \neq b \end{cases}$$

AD:

$$y_{ijk} = \begin{cases} \mu_{jb}, & \text{if } k = b \\ \mu_{jb} + \delta_{jbk} + \beta_{1bk}x_j, & \text{if } k \neq b \end{cases}$$

$$\delta_{jbk} \sim N(d_{bk}, \tau^2)$$

$$d_{kq} = d_{kb} - d_{qb}$$

$$\beta_{1,kq} = \beta_{1,kb} - \beta_{1,qb}$$

Equation 3

Where i, j, k, q , and b are as defined for Equation 1. For the IPD portion, x_{ij} , β_{0j} , β_{1bk} , μ_{jb} , δ_{jbk} , and y_{ijk} are as defined in Equation 2. For the AD subset, x_j is the average of the covariate in study j and β_{1bk} is the interaction of x_j for treatment b relative to treatment k for the aggregated covariate. The definition for μ_{jb} is the effect size for the control group of the study, and δ_{jbk} represents the effect size between treatment b and treatment k in a study when the

covariate value is 0. For binary outcomes, y_{ijk} and δ_{jbk} will be handled as in the case of IPD/AD-NMA. To enable estimation using both IPD and AD, this model assumes that the treatment-by-covariate interaction regression coefficients, β_{1bk} , are the same for individual level covariates from IPD and study level covariates for AD.

We will use the estimated parameters of the IPD-NMA or IPD/AD-NMA model to create a prediction model in order to provide personalised predictions according to the covariates considered. The model will take patient-specific values for covariates as inputs and provide a prediction of the outcome under each treatment of interest. Initial nodes will be defined as stated in the eligibility criteria of the protocol. If not enough data exist for making predictions for psychotherapy types, we will merge nodes to

- Any type of psychotherapy
- Antidepressant pharmacotherapy
- Any type of psychotherapies as an adjunct (TAU, CAU, exercise, counselling, electroconvulsive therapy)
- Any type of psychotherapy as an adjunct treatment to any type of antidepressant pharmacotherapy
- Waiting list, CAU, TAU either alone or as an adjunct to any type of antidepressant pharmacotherapy

To assess the predictive performance of this model, we will use an internal-external cross-validation method. This involves taking one study out at a time, developing the model using the remaining studies, and testing the model on the left-out study. Then, we will assess the model performance in terms of calibration and discrimination for benefit, and decision accuracy, using recently developed methods[45–47].

Subgroup analyses will be performed for study level characteristics defined above. Study level characteristics will be explored using meta-regression in a Bayesian framework with informative priors where available or vague priors in all other cases. If subgroup analyses cannot be undertaken due to small number of studies, we will conduct separate pairwise meta-analyses for the small subgroups and provide descriptive statistics to offer insight regarding

the subgroup. In the particular case where subgroup analyses result in disconnected sub-networks, i.e. there are pairs of treatments that neither direct nor indirect estimates can be derived, we will employ a component NMA (cNMA) in an attempt to connect the disconnected networks consisting the small subgroup and facilitate analysis.

Meta-biases

We will examine the existence of possible small study effects and publication bias both in terms of the pairwise and network level. Pairwise examinations will be held by using the contour-enhanced funnel plots for each pairwise comparison with at least 10 studies available while for examinations at the network level we will use the comparison adjusted funnel plot and plot all the study-specific treatment effects in a unique graph. In order to evaluate potential availability bias with regards to the available participant level covariates included in the model, covariate effects will be estimated in IPD studies and compared with AD studies. If considerable differences are found, this will be discussed in the final publication.

Confidence in meta-analytical estimates

To rate the quality of the best available evidence and provide a comprehensive summary, the Grading of Recommendations, Assessment, Development and Evaluation (GRADE) approach will be used. The quality of evidence will be assessed across the domains of risk of bias, consistency, directness, precision, and publication bias. Additional domains may be considered where appropriate.

When evaluating the confidence in the NMA for the primary outcomes, we will use the six domain CINeMA framework[48], which considers within-study bias, across-studies bias, indirectness, imprecision, heterogeneity and incoherence when evaluating the confidence in the NMA. Since there is no current consensus regarding the definition of clinically important effect size for change in depression severity in chronic depression, the clinically important effect size for clinical equivalence will be based on statistical difference.

Patient and public involvement

None.

ETHICS AND DISSEMINATION

As a systematic review of published studies, this IPD-NMA requires no ethical approval. In the event that changes to the protocol occur, changes will be reported in protocol amendments in PROSPERO. All results will be disseminated as peer-reviewed publication in a leading journal in this field and presented at (inter)national scientific conferences.

DISCUSSION

This protocol describes a systematic review and network meta-analysis (NMA) on the efficacy of psychotherapy vs. pharmacotherapy or their combination in the treatment of chronic depression. As the most recent comparable systematic reviews on chronic depression are several years old, do not combine IPD and AD, and do not include various important studies[12,15], an up-to-date overview is urgently needed. To include all available RCTs and maximize statistical power, our NMA approach will allow direct and indirect comparisons based on AD as well as IPD.

In some cases, IPD meta-analysis has demonstrated the potential to produce better quality, more precise, and more reliable results than meta-analysis of only AD[49–52]. However, it is often difficult to access IPD, and subgroup analyses of interest within IPD often will be small[42,44,53]. In order to maximize the use of the available information and attempt to reduce the overall uncertainty, complex methods that allow the inclusion of both IPD and AD can be used to evaluate overall treatment effects, irrespective of subgroups of interest. Then, regression techniques within the combined IPD and AD can enable the examination of the association between treatment effects and potential covariates.

Despite the planned combination of AD and IPD, the expected number of applicable studies fulfilling the defined inclusion and exclusion criteria is small. Thus, any subgroup analysis performed from the identified studies is also expected to be limited by small numbers and therefore to be low powered. The updated NMA will improve our understanding of the therapeutic landscape. However, uncertainty in the estimation of outcomes will likely persist.

It is important to note that, despite attempts to access the individual patient data of all identified studies, the possibility remains that some studies will not be able to provide the IPD. This could lead to an availability bias with regards to the available participant level covariates included in the model. Furthermore, since the NMA is based on existing studies, the analysis can only examine potential predictors or effect modifiers that are measured in the original studies. For aggregate data, this data availability is further limited to the variables that are reported in the original publication. Therefore, it is currently unclear whether all variables of interest defined in the protocol will be incorporated in the NMR model. The use of mean differences would enhance the clinical interpretability of the meta-analysis results if all scales used in the original studies are convertible to the HAM-D. However, if alternative scales are used, it will be necessary to use standardized mean differences to obtain estimates of treatment efficacy, which will limit the clinical interpretability of the results.

This systematic review and meta-analysis will focus on the inclusion of randomized controlled trials (RCTs), since they are considered less susceptible to known and unknown confounders due to the randomization and have well defined criteria, treatments, and endpoints. We plan to complete the systematic literature search, obtain aggregated or individual participant data from the relevant studies, and conduct the statistical analyses by mid-2025. We will aim for the results to be submitted to an international peer-reviewed journal by the end of 2025. The code behind our results will be made accessible through a public GitHub repository. These codes could then potentially be used to create an interactive tool (e.g. an R-Shiny application) that will allow using our prediction model and visualize its results without requiring any programming expertise for users. Ultimately, this systematic review and meta-analysis aims to help elucidate and estimate the impact of treatment alternatives, and associated predictive and effect modifying factors, on chronic depression to enable customized treatment strategies for patients experiencing chronic depression.

Declarations

Competing interests

E. Schramm received modest book royalties and honoraria for workshops and presentations related to the CBASP. T. A. Furukawa reports personal fees from Boehringer-Ingelheim, Daiichi Sankyo, DT Axis, Kyoto University Original, Shionogi, SONY and UpToDate, and a grant from DT Axis and Shionogi, outside the submitted work; In addition, T. A. Furukawa has a patent 7448125 and a pending patent 2022-082495, and intellectual properties for Kokoro-app licensed to Mitsubishi-Tanabe. All other authors have no completing interest to declare.

Funding

This meta-analysis is funded by the Deutsche Forschungsgemeinschaft (DFG, German Research Foundation) – Project-ID 499552394 – SFB 1597. The DFG is not involved in any other aspect of the project, such as the design of the project’s protocol and analysis plan, the collection and analyses. The funder will have no input on the interpretation or publication of the study results.

Contributors

ES is responsible for the overall content as guarantor. ES, ME, JM, NK, TE and AN drafted the manuscript. ES and AN obtained funding. All authors contributed to the development of the selection criteria, the risk of bias assessment strategy and data extraction criteria. ES, ME and JM developed the search strategy. NK, TE, PC, OE, TAF and AN provided statistical expertise. All authors provided feedback and approved the final manuscript.

Acknowledgements

We acknowledge financial support by the Open Access Publication Fund of the University of Freiburg. Furthermore, we thank Sina Wilhelm, Magdalena Hammen, Marion Braun, and Hannah Piosczyk for their valuable contributions to the systematic literature search.

References

- 1 Schramm E, Klein DN, Elsaesser M, *et al.* Review of dysthymia and persistent depressive disorder: history, correlates, and clinical implications. *Lancet Psychiatry*. 2020;7:801–12. doi: 10.1016/S2215-0366(20)30099-7
- 2 Murphy JA, Byrne GJ. Prevalence and correlates of the proposed DSM-5 diagnosis of Chronic Depressive Disorder. *J Affect Disord*. 2012;139:172–80. doi: 10.1016/j.jad.2012.01.033
- 3 Angst J, Gamma A, Rössler W, *et al.* Long-term depression versus episodic major depression: Results from the prospective Zurich study of a community sample. *Journal of Affective Disorders*. 2009;115:112–21. doi: 10.1016/j.jad.2008.09.023
- 4 Köhler S, Wiethoff K, Ricken R, *et al.* Characteristics and differences in treatment outcome of inpatients with chronic vs. episodic major depressive disorders. *J Affect Disord*. 2015;173:126–33. doi: 10.1016/j.jad.2014.10.059
- 5 McIntyre RS, Alsuwaidan M, Baune BT, *et al.* Treatment-resistant depression: definition, prevalence, detection, management, and investigational interventions. *World Psychiatry*. 2023;22:394–412. doi: 10.1002/wps.21120
- 6 McCullough JP. *Treatment for chronic depression. Cognitive behavioral analysis system of psychotherapy*. NY: Guilford Press. 2000.
- 7 McCullough JP. Treatment for chronic depression using Cognitive Behavioral Analysis System of Psychotherapy (CBASP). *J Clin Psychol*. 2003;59:833–46. doi: 10.1002/jclp.10176
- 8 McCullough J. Characteristics of the Optimal Cognitive Behavioral Analysis System of Psychotherapy (CBASP) Therapist Role. *Frontiers in Psychiatry*. 2021;11. doi: 10.3389/fpsy.2020.609954
- 9 Beck AT. *Cognitive Therapy of Depression*. Guilford Press 1979.
- 10 Klerman GL, Weissman MM, Rounsaville BJ, *et al.* *Interpersonal psychotherapy of depression*. NY: Basic Books 1984.
- 11 Jobst A, Brakemeier E-L, Buchheim A, *et al.* European Psychiatric Association Guidance on psychotherapy in chronic depression across Europe. *Eur Psychiatry*. 2016;33:18–36. doi: 10.1016/j.eurpsy.2015.12.003
- 12 Kriston L, von Wolff A, Westphal A, *et al.* Efficacy and acceptability of acute treatments for persistent depressive disorder: a network meta-analysis. *Depress Anxiety*. 2014;31:621–30. doi: 10.1002/da.22236
- 13 Negt P, Brakemeier E-L, Michalak J, *et al.* The treatment of chronic depression with cognitive behavioral analysis system of psychotherapy: a systematic review and meta-analysis of randomized-controlled clinical trials. *Brain Behav*. 2016;6:e00486. doi: 10.1002/brb3.486
- 14 Cuijpers P, Noma H, Karyotaki E, *et al.* A network meta-analysis of the effects of psychotherapies, pharmacotherapies and their combination in the treatment of adult depression. *World Psychiatry*. 2020;19:92–107. doi: 10.1002/wps.20701

- 584 15 Furukawa TA, Efthimiou O, Weitz ES, *et al.* Cognitive-Behavioral Analysis System of
585 Psychotherapy, Drug, or Their Combination for Persistent Depressive Disorder:
586 Personalizing the Treatment Choice Using Individual Participant Data Network
587 Metaregression. *PPS*. 2018;87:140–53. doi: 10.1159/000489227
- 588 16 Schramm E, Kriston L, Zobel I, *et al.* Effect of Disorder-Specific vs Nonspecific
589 Psychotherapy for Chronic Depression: A Randomized Clinical Trial. *JAMA Psychiatry*.
590 2017;74:233–42. doi: 10.1001/jamapsychiatry.2016.3880
- 591 17 Rief W, Bleichhardt G, Dannehl K, *et al.* Comparing the Efficacy of CBASP with Two
592 Versions of CBT for Depression in a Routine Care Center: A Randomized Clinical Trial.
593 *Psychother Psychosom*. 2018;87:164–78. doi: 10.1159/000487893
- 594 18 Schramm E, Kriston L, Elsaesser M, *et al.* Two-Year Follow-Up after Treatment with
595 the Cognitive Behavioral Analysis System of Psychotherapy versus Supportive
596 Psychotherapy for Early-Onset Chronic Depression. *PPS*. 2019;88:154–64. doi:
597 10.1159/000500189
- 598 19 Elsaesser M, Feige B, Kriston L, *et al.* Longitudinal Clusters of Long-Term Trajectories
599 in Patients with Early-Onset Chronic Depression: 2 Years of Naturalistic Follow-Up after
600 Extensive Psychological Treatment. *Psychotherapy and Psychosomatics*. 2023;1–9. doi:
601 10.1159/000535005
- 602 20 Elsaesser M, Herpertz S, Piosczyk H, *et al.* Modular-based psychotherapy (MoBa)
603 versus cognitive-behavioural therapy (CBT) for patients with depression, comorbidities and
604 a history of childhood maltreatment: study protocol for a randomised controlled feasibility
605 trial. *BMJ Open*. 2022;12:e057672. doi: 10.1136/bmjopen-2021-057672
- 606 21 Schramm E, Elsaesser M, Jenkner C, *et al.* Algorithm-based modular psychotherapy
607 vs. cognitive-behavioral therapy for patients with depression, psychiatric comorbidities and
608 early trauma: a proof-of-concept randomized controlled trial. *World Psychiatry*.
609 2024;23:257–66. doi: 10.1002/wps.21204
- 610 22 Elsaesser M, Furukawa TA, Schramm E. Why one answer is not always enough: Reply
611 to “CBASP may not be superior to other treatments for chronic depression.” *J Affect Disord*.
612 2022;312:144–5. doi: 10.1016/j.jad.2022.06.033
- 613 23 Furukawa TA, Schramm E, Weitz ES, *et al.* Cognitive-Behavioural Analysis System of
614 Psychotherapy (CBASP), a drug, or their combination: differential therapeutics for persistent
615 depressive disorder: a study protocol of an individual participant data network meta-
616 analysis. *BMJ Open*. 2016;6:e011769. doi: 10.1136/bmjopen-2016-011769
- 617 24 Efthimiou O, Debray TPA, van Valkenhoef G, *et al.* GetReal in network meta-analysis:
618 a review of the methodology. *Research Synthesis Methods*. 2016;7:236–63. doi:
619 10.1002/jrsm.1195
- 620 25 Montgomery SA, Asberg M. A new depression scale designed to be sensitive to
621 change. *Br J Psychiatry*. 1979;134:382–9. doi: 10.1192/bjp.134.4.382
- 622 26 Rush AJ, Trivedi MH, Ibrahim HM, *et al.* The 16-Item quick inventory of depressive
623 symptomatology (QIDS), clinician rating (QIDS-C), and self-report (QIDS-SR): a
624 psychometric evaluation in patients with chronic major depression. *Biological Psychiatry*.
625 2003;54:573–83. doi: 10.1016/S0006-3223(02)01866-8
- 626 27 Carrozzino D, Patierno C, Fava GA, *et al.* The Hamilton Rating Scales for Depression:
627 A Critical Review of Clinimetric Properties of Different Versions. *PPS*. 2020;89:133–50. doi:
628 10.1159/000506879

- 629 28 Hamilton M. A rating scale for depression. *J Neurol Neurosurg Psychiatry*. 1960;23:56–
630 62.
- 631 29 Hamilton M. Development of a Rating Scale for Primary Depressive Illness. *British*
632 *Journal of Social and Clinical Psychology*. 1967;6:278–96. doi: 10.1111/j.2044-
633 8260.1967.tb00530.x
- 634 30 Veal C, Tomlinson A, Cipriani A, *et al*. Heterogeneity of outcome measures in
635 depression trials and the relevance of the content of outcome measures to patients: a
636 systematic review. *The Lancet Psychiatry*. 2024;11:285–94. doi: 10.1016/S2215-
637 0366(23)00438-8
- 638 31 Leucht S, Fennema H, Engel RR, *et al*. Translating the HAM-D into the MADRS and
639 vice versa with equipercentile linking. *Journal of Affective Disorders*. 2018;226:326–31. doi:
640 10.1016/j.jad.2017.09.042
- 641 32 Beck AT, Steer RA, Brown GK. *Manual for the Beck Depression Inventory-II*. San
642 Antonio, TX: Psychological Corporation 1996.
- 643 33 Rush AJ, Gullion CM, Basco MR, *et al*. The Inventory of Depressive Symptomatology
644 (IDS): psychometric properties. *Psychological Medicine*. 1996;26:477. doi:
645 10.1017/S0033291700035558
- 646 34 Furukawa TA, Reijnders M, Kishimoto S, *et al*. Translating the BDI and BDI-II into the
647 HAMD and vice versa with equipercentile linking. *Epidemiology and Psychiatric Sciences*.
648 2020;29:e24. doi: 10.1017/S2045796019000088
- 649 35 Gameroff MJ, Wickramaratne P, Weissman MM. Testing the Short and Screener
650 versions of the Social Adjustment Scale – Self-report (SAS-SR). *Int J Methods Psychiatr*
651 *Res*. 2011;21:52–65. doi: 10.1002/mpr.358
- 652 36 ANGERMEYER C. WHOQOL-100 und WHOQOL-BREF. *Handbuch für die*
653 *deutschsprachige Version der WHO Instrumente zur Erfassung der Lebensqualität*.
654 Published Online First: 2000.
- 655 37 Ware JE, Sherbourne CD. The MOS 36-item short-form health survey (SF-36). I.
656 Conceptual framework and item selection. *Med Care*. 1992;30:473–83.
- 657 38 McHorney CA, Ware JE, Raczek AE. The MOS 36-Item Short-Form Health Survey (SF-
658 36): II. Psychometric and clinical tests of validity in measuring physical and mental health
659 constructs. *Med Care*. 1993;31:247–63. doi: 10.1097/00005650-199303000-00006
- 660 39 Sterne JAC, Savović J, Page MJ, *et al*. RoB 2: a revised tool for assessing risk of bias
661 in randomised trials. *BMJ*. 2019;366:l4898. doi: 10.1136/bmj.l4898
- 662 40 Higgins J, Thomas J, Chandler J, *et al*. *Cochrane Handbook for Systematic Reviews*
663 *of Interventions version 6.5 (updated August 2024)*. Cochrane 2024.
- 664 41 Quartagno M, Grund S, Carpenter J. jomo: A Flexible Package for Two-level Joint
665 Modelling Multiple Imputation. *The R Journal*. 2019;11:205. doi: 10.32614/RJ-2019-028
- 666 42 Saramago P, Sutton AJ, Cooper NJ, *et al*. Mixed treatment comparisons using
667 aggregate and individual participant level data. *Statistics in Medicine*. 2012;31:3516–36.
668 doi: 10.1002/sim.5442

43 Salanti G, Ades AE, Ioannidis JPA. Graphical methods and numerical summaries for presenting results from multiple-treatment meta-analysis: an overview and tutorial. *Journal of Clinical Epidemiology*. 2011;64:163–71. doi: 10.1016/j.jclinepi.2010.03.016

44 Jansen JP. Network meta-analysis of individual and aggregate level data. *Research Synthesis Methods*. 2012;3:177–90. doi: 10.1002/jrsm.1048

45 van Klaveren D, Steyerberg EW, Serruys PW, *et al*. The proposed ‘concordance-statistic for benefit’ provided a useful metric when modeling heterogeneous treatment effects. *Journal of Clinical Epidemiology*. 2018;94:59–68. doi: 10.1016/j.jclinepi.2017.10.021

46 Efthimiou O, Hoogland J, Debray TPA, *et al*. Measuring the performance of prediction models to personalize treatment choice. *Stat Med*. 2023;42:1188–206. doi: 10.1002/sim.9665

47 Hoogland J, Efthimiou O, Nguyen TL, *et al*. Evaluating individualized treatment effect predictions: a model-based perspective on discrimination and calibration assessment. 2023.

48 Nikolakopoulou A, Higgins JPT, Papakonstantinou T, *et al*. CINeMA: An approach for assessing confidence in the results of a network meta-analysis. *PLOS Medicine*. 2020;17:e1003082. doi: 10.1371/journal.pmed.1003082

49 Balduzzi S, Rücker G, Schwarzer G. How to perform a meta-analysis with R: a practical tutorial. *BMJ Ment Health*. 2019;22:153–60. doi: 10.1136/ebmental-2019-300117

50 Kanters S, Karim ME, Thorlund K, *et al*. When does the use of individual patient data in network meta-analysis make a difference? A simulation study. *BMC Medical Research Methodology*. 2021;21:21. doi: 10.1186/s12874-020-01198-2

51 Lambert PC, Sutton AJ, Abrams KR, *et al*. A comparison of summary patient-level covariates in meta-regression with individual patient data meta-analysis. *Journal of Clinical Epidemiology*. 2002;55:86–94. doi: 10.1016/S0895-4356(01)00414-0

52 Riley RD, Dias S, Donegan S, *et al*. Using individual participant data to improve network meta-analysis projects. *BMJ Evidence-Based Medicine*. 2023;28:197–203. doi: 10.1136/bmjebm-2022-111931

53 Sutton AJ, Kendrick D, Coupland C a. C. Meta-analysis of individual- and aggregate-level data. *Statistics in Medicine*. 2008;27:651–69. doi: 10.1002/sim.2916