Open Access

Data from network meta-analyses can inform clinical practice guidelines and decision-making in diabetes management: perspectives of the taskforce of the guideline workshop



Antonio Ceriello¹, Helena W. Rodbard², Tadei Battelino³, Frank Brosius⁴, Francesco Cosentino⁵, Jennifer Green⁶, Linong Ji⁷, Monika Kellerer⁸, Susan Koob⁹, Mikhail Kosiborod^{10,11}, Nebojsa Lalic¹², Nikolaus Marx¹³, T. Prashant Nedungadi¹⁴, Christopher G. Parkin¹⁵, Lars Rydén¹⁶, Wayne Huey-Herng Sheu¹⁷, Eberhard Standl¹⁸, Per Olav Vandvik¹⁹, Oliver Schnell^{18*} and for the Taskforce of the Guideline Workshop

Abstract

In recent years, several novel agents have become available to treat individuals with type 2 diabetes (T2D), such as sodium-glucose cotransporter-2 inhibitors (SGLT-2i), tirzepatide, which is a dual glucose-dependent insulinotropic polypeptide receptor agonist (GIP RA)/glucagon-like peptide-1 receptor agonist (GLP-1 RA), and finerenone, a nonsteroidal mineralocorticoid receptor antagonist (MRA) that confers significant renal and cardiovascular benefits in individuals with (CKD). New medications have the potential to improve the lives of individuals with diabetes. However, clinicians are challenged to understand the benefits and potential risks associated with these new and emerging treatment options. In this article, we discuss how use of network meta-analyses (NMA) can fill this need.

Keywords Network meta-analysis, Randomized controlled trial, Sodium glucose cotransporter 2 inhibitor, Glucosedependent insulinotropic polypeptide, (GIP RA), Glucagon-like peptide-1 receptor agonist (GLP-1 RA), Tirzepatide, Finerenone

An estimated 537 million people worldwide have diabetes, an alarming number that is projected to reach 643 million by 2030 at an annual cost of over \$1 trillion (USD) [1]. Most of this cost is associated with the acute and chronic complications resulting from overall suboptimal cardiovascular risk management, including insufficient glycaemic control [2]. As reported in recent epidemiological studies, the inability to achieve optimal

*Correspondence: Oliver Schnell oliver.schnell@lrz.uni-muenchen.de Full list of author information is available at the end of the article disease management (glycaemia, lipids, blood pressure) remains problematic for many individuals with diabetes [3–5]. However, as the rate of innovation in the development of new diabetes medications and technologies continues to accelerate, there is a growing and diverse array of treatment options that may facilitate more effective management [6, 7].

In recent years, several novel agents have become available to treat individuals with type 2 diabetes (T2D), such as sodium-glucose cotransporter-2 inhibitors (SGLT-2i). Pharmacologic innovations have also led to new firstin-class medications such as tirzepatide, which is a dual glucose-dependent insulinotropic polypeptide receptor



© The Author(s) 2023. Open Access This article is licensed under a Creative Commons Attribution 4.0 International License, which permits use, sharing, adaptation, distribution and reproduction in any medium or format, as long as you give appropriate credit to the original author(s) and the source, provide a link to the Creative Commons licence, and indicate if changes were made. The images or other third party material in this article are included in the article's Creative Commons licence, unless indicated otherwise in a credit line to the material. If material is not included in the article's Creative Commons licence and your intended use is not permitted by statutory regulation or exceeds the permitted use, you will need to obtain permission directly from the copyright holder. To view a copy of this licence, visit http://creativecommons.org/licenses/by/4.0/. The Creative Commons Public Domain Dedication waiver (http://creativeco mmons.org/publicdomain/zero/1.0/) applies to the data made available in this article, unless otherwise stated in a credit line to the data. agonist (GIP RA)/glucagon-like peptide-1 receptor agonist (GLP-1 RA) that lowers HbA1c with significant reductions in body weight [8–11]. Another new medication class is finerenone, a non-steroidal mineralocorticoid receptor antagonist (MRA) that confers significant renal and cardiovascular benefits in individuals with (CKD) [12].

While new medications have the potential to improve the lives of individuals with diabetes, clinicians are challenged to understand the benefits and potential risks associated with these new and emerging treatment options. Traditionally, clinicians have relied on clinical practice guidelines based on evidence from cardiovascular outcome trials (CVOTs). To meet standards for trustworthy guidelines, recommendations need to be based on systematic reviews, typically meta-analyses of all available randomized controlled trials (RCTs). However, because a standard pairwise meta-analysis can only compare the efficacy or safety of two medications that have been compared in head-to-head clinical trials, it is impossible to make the same risk-benefit determination when several possible treatments are available to treat patients with the same condition. To provide effective, personalized care to their patients, clinicians need the ability to select the most appropriate treatment among several options.

The use of network meta-analyses (NMA) can fill this need. Also referred to as multiple treatment meta-analyses or mixed treatment comparisons, NMAs combine direct and indirect evidence acquired from one or more common comparators to simultaneously compare multiple treatments in a single pooled analysis [13, 14]. This approach differs from earlier neural node meta-analyses in which compounds are compared to each other for a single measure of efficacy (e.g., HbA1c) vs. current approach in which the common comparator is "standard treatment". Direct evidence is acquired from RCTs that directly compare two medications in head-to-head assessments (e.g., intervention A vs. intervention B), while indirect evidence is acquired from RCTs assessing one or more common comparators. In the absence of a study that reports an A vs. B comparison, it is possible to make this assessment by combining studies with common comparators (e.g., A vs. C and B vs. C). Based on the direct and indirect evidence assessed, a network map is created to graphically depict the number of patients and trials assessed and the network estimate is pooled result of the direct and indirect evidence.

An example of this approach is the recent systematic review and NMA by Shi et al. [15] This NMA is an update of a previous systematic review that informed a clinical practice guideline (BMJ Rapid Recommendations), supported by the MAGIC Evidence Ecosystem Foundation. [16] In the updated NMA, investigators assessed the most current evidence of T2D medication from a larger data set of 821 trials with 471,815 patients [15]. In addition to updated evidence on SGLT-2i and GLP-RA, this NMA included studies of finerenone and tirzepatide, which are new to clinicians. Investigators grouped drug treatments by their treatment class with connections between each drug in all included trials for any outcome. This resulted in 9976 estimates of effect across 13 drugs and 11 outcomes, clearly representing an insurmountable challenge to digest for readers. To ease navigation, interpretation, and use of the evidence in decision-making, the interactive MATCH-IT tool provides user-friendly access to all comparisons and interventions (https://matchit.magicevidence.org/230125dist-diabetes/#!/) [17].

The Taskforce of the Guideline Workshop, an international multidisciplinary team including endocrinologists, cardiologists, and nephrologists, helped formulate the clinical questions and provided input into the study protocol. The aim of the Taskforce is to develop and implement a roadmap for the acceleration and harmonization of clinical guidelines and updates for diabetes, prediabetes, cardiovascular, and kidney diseases. [7, 18] (Fig. 1).

Certainty of the evidence was assessed following Grading of Recommendations Assessment, Development, and Evaluation (GRADE) guidance [19]. This approach focuses on the magnitude of the benefits, harms, and burdens of the interventions and the comparators; the quality of evidence associated with the evidence of benefits, harms, and burdens; and the underlying values and preferences of the population to whom the recommendation applies [20]. Cost, feasibility, and acceptability are also considered [21]. The GRADE approach considers only two types of evidence: randomized trials and observational studies, which are graded as high, moderate, low, and very low. A strong rating identifies recommendations in which the benefits outweigh the harms [22], whereas a weak rating indicates that the recommendation should be considered based on a patient's specific needs and preferences and it should involve shared decision making [7, 23].

To categorise the relative impact of interventions, investigators defined the null effect as the decision threshold and standard treatments as the reference intervention [24, 25]. Standard treatments refer to the control/comparator group included in each study. Treatment options are displayed in rows and outcomes in columns. The cells are colour-coded to indicate the magnitude and certainty of the treatment effect in relation to the reference treatment [25].

The drugs found to be superior or inferior to standard treatments were categorised from the most effective to the most harmful, taking certainty of evidence





Fig. 1 Network map for all included studies [15]

into account. Drugs were further categorized based on the certainty of supporting evidence: "high to moderate certainty" or "low to very low certainty". From these analyses, investigators generated a comprehensive summary of the benefits and harms of the diabetes drugs with estimates that represent the comparative effects of the drugs compared to standard treatments. To address the needs of patient groups with various comorbidities (e.g., T2D with existing CVD), the evidence summary presents the incidence of the pre-defined outcomes to be anticipated with the new treatment approaches compared with standard medical care within the five CVD/CKD risk groups. (e.g., "more" or "fewer" events per 1,000 patients) compared with standard treatments. The clinical outcomes considered in the current network meta-analysis were all-cause death, cardiovascular death, non-fatal myocardial infarction, nonfatal stroke, hospitalization for heart failure, end-stage kidney disease (ESKD), health-related quality of life (HRQoL); severe hypoglycaemia, and drug-specific adverse events. A similar summary of comparative effects was developed for individuals with T2D and CKD.

Investigators found that both the SGLT-2i and GLP-1 RA medications were effective in reducing all-cause death, cardiovascular death, non-fatal myocardial infarction, hospitalization for heart failure, and ESKD. Although only GLP-1 RAs reduced non-fatal stroke, SGLT-2i medications were shown to be superior to other medications in reducing end-stage kidney disease. For patients with T2D and CKD, it was reported that the non-steroidal MRA medication (finerenone) probably reduces hospital admissions for heart failure and end-stage kidney disease and decreases mortality. Tirzepatide appears to facilitate the largest reduction in body weight and increase in health-related quality of life (QoL) in individuals with T2D [15] followed by varying effects of the individual GLP-1 receptor agonists. The key reported harms were largely specific to each medication class; genital infections with SGLT-2 inhibitors, gastrointestinal adverse events with tirzepatide

and GLP-1 receptor agonists, and hyperkalemia, leading to admission to hospital with finerenone.

RCTs remain the gold standard for direct comparison of two interventions. However, when multiple interventions or the same disease or condition are being considered, synthesis of results from RTCs of the various interventions using the NMA model ensures that all relevant direct and indirect evidence is considered. This approach generates more comprehensive and clinically useful estimates of the relative effects of multiple interventions. As demonstrated in the analysis performed by Shi et al. and the accompanying MATCH-IT tool [15, 17], the use of NMAs offers the ability to visualize and interpret a broader picture of the evidence and better understand the relative merits of each intervention when multiple interventions have been used to treat the same disease. Moreover, the NMA model may facilitate creating and updating guidelines more rapidly based on practice-changing evidence. Indeed, the recent NMA on diabetes drugs is now informing an update of the BMJ Rapid Recommendations and in Australia, both in the shape of living guidelines. The CVOT Taskforce recommends that our professional societies to consider use if this NMA to inform their guideline recommendations.

Abbreviations

CKD	Chronic kidney disease
CVOT	Cardiovascular outcome trials
ESKD	End-stage kidney disease
GIP	Glucose-dependent insulinotropic polypeptide receptor agonist
GLP-1RA	Glucagon-like peptide-1 receptor agonist
GRADE	Grading of Recommendations Assessment, Development and
	Evaluation
MRA	Non-steroidal mineralocorticoid receptor antagonist
NMA	Network meta-analyses
QoL	Quality of life
SGLT-2i	Sodium-glucose cotransporter-inhibitors
T2D	Type 2 diabetes
USD	United States dollars

Acknowledgements

Not applicable.

Author contributions

AC, HWR, FB, OS, NM, and CGP wrote the first draft. All other authors provided input on the manuscript and approved the manuscript for publication.

Funding

The author(s) disclosed receipt of the following financial support for the research, authorship, and/or publication of this article: Sponsors of the Guideline Workshop are listed in alphabetical order: Abbott GmbH, AstraZeneca Pharmaceuticals LP, Bayer AG, Boehringer Ingelheim International GmbH, Dexcom, Inc., Eli Lilly and Company, and Novo Nordisk Pharma GmbH. The sponsors had no influence on the content of this manuscript.

Availability of data and materials

Not applicable

Declarations

Ethics approval and consent to participate

Not applicable.

Consent for publication

All authors have provided their consent for publication of this manuscript.

Competing interests

All other authors declared no potential competing interest relevant to the authorship and/or publication of this article.

Author details

¹ IRCCS MultiMedica. Via Milanese 300, 20099 Sesto San Giovanni. Ml. Italv. ²Endocrine and Metabolic Consultants, 3200 Tower Oaks Blvd., Suite 250, Rockville, MD 20852, USA. ³University Medical Center Ljubljana, and Faculty of Medicine, University of Ljubljana, Ljubljana, Slovenia.⁴University of Arizona College of Medicine, 1501 N. Campbell Ave, Tucson, AZ 85724-5022, USA. ⁵Cardiology Unit, Department of Medicine, Karolinska Institute and Karolinska University Hospital, Solna, Stockholm, Sweden. ⁶Duke University Medical Center, Duke Clinical Research Institute, 200 Morris St, DUMC Box 3850, Durham, NC 27715, USA. ⁷Peking University People's Hospital, 11 Xizhimen S St, Xicheng District, Beijing, China.⁸Marienhospital Stuttgart, Böheimstraße 37, 70199 Stuttgart, Germany. ⁹PCNA National Office, 613 Williamson Street, Suite 200, Madison, WI 53703, USA. ¹⁰Saint Luke's Mid America Heart Institute and University of Missouri-Kansas City, 4401 Wornall Rd, Kansas City, MO 64111, USA.¹¹The George Institute for Global Health and University of New South Wales, Sydney, Australia, ¹²University Clinical Center of Serbia, University of Belgrade, Pasterova 2, 11000 Belgrade, Serbia. ¹³Department of Internal Medicine I, University Hospital Aachen, RWTH Aachen University, Pauwelsstraße 30, 52074 Aachen, Germany.¹⁴American Heart Association, 7272 Greenville Avenue, Dallas, TX 75082, USA. ¹⁵CGParkin Communications, Inc., 2675 Windmill Pkwy, Suite 2721, Henderson, NV 89074, USA. ¹⁶Department of Medicine K2, Karolinska Institute, Stockholm, Sweden.¹⁷Institute of Molecular and Genomic Medicine, National Research Health Institutes, Zhunan, Miaoli 350, Taiwan.¹⁸Forschergruppe Diabetes E. V, Ingolstaedter Landstraße 1, Neuherberg, 85764 Munich, Germany. ¹⁹Institute of Health and Society, University of Oslo, Oslo, Norway.

Received: 17 June 2023 Accepted: 14 September 2023 Published online: 13 October 2023

References

- International Diabetes Federation. IDF Diabetes Atlas, 10th ed. Brussels, International Diabetes Federation. 2021. https://diabetesatlas.org/atlas/ tenth-edition/. Accessed 31 8 Jan 2023.
- Ferrannini G, de Bacquer D, De Backer G, on behalf of the EUROASPIRE V team, et al. Screening for glucose perturbations and risk factor management in dysglycaemic patients with coronary artery disease - a persistent challenge in need of substantial improvement. A report from EUROASPIRE V. Diabetes Care. 2020;43:726–33.
- Khunti K, Ceriello A, Cos X, De Block C. Achievement of guideline targets for blood pressure, lipid, and glycaemic control in type 2 diabetes: a meta-analysis. Diabetes Res Clin Pract. 2018;137:137–48. https://doi.org/ 10.1016/j.diabres.2017.12.004.
- Prigge R, McKnight JA, Wild SH, et al. International comparison of glycaemic control in people with type 1 diabetes: an update and extension. Diabet Med. 2022;39(5):e14766. https://doi.org/10.1111/dme.14766.
- Bin Rakhis Sr SA, AlDuwayhis NM, Aleid N, AlBarrak AN, Aloraini AA. Glycemic control for type 2 diabetes mellitus patients: a systematic review. Cureus. 2022;14(6):e26180. https://doi.org/10.7759/cureus.26180.
- Schnell O, Battelino T, Bergenstal RM, Summit CVOT, et al. Report: new cardiovascular, kidney, and glycemic outcomes. Cardiovasc Diabetol. 2022;2023:22. https://doi.org/10.1186/s12933-023-01788-6.
- Marx N, Ryden L, Brosius F, et al. Towards living guidelines on cardiorenal outcomes in diabetes: a pilot project of the Taskforce of the Guideline Workshop 2020. Diabetes Res Clin Pract. 2021;177:108870. https://doi. org/10.1016/j.diabres.2021.108870.

- Jastreboff AM, Aronne LJ, Ahmad NN, et al. Tirzepatide once weekly for the treatment of obesity. N Engl J Med. 2022;387:205–16. https://doi.org/ 10.1056/NEJMoa2206038.
- Frías JP, Davies MJ, Rosenstock J, et al. Tirzepatide versus semaglutide once weekly in patients with type 2 diabetes. N Engl J Med. 2021;385:503–15. https://doi.org/10.1056/NEJMoa2107519.
- Chavda VP, Ajabiya J, Teli D, et al. Tirzepatide, a new era of dualtargeted treatment for diabetes and obesity: a mini-review. Molecules. 2022;27(13):4315. https://doi.org/10.3390/molecules27134315.
- Dahl D, Onishi Y, Norwood P, et al. Effect of subcutaneous tirzepatide vs placebo added to titrated insulin glargine on glycemic control in patients with type 2 diabetes: the SURPASS-5 randomized clinical trial. JAMA. 2022;327(6):534–45. https://doi.org/10.1001/jama.2022.0078.
- Agarwal R, Filippatos G, Pitt B, et al. Cardiovascular and kidney outcomes with finerenone in patients with type 2 diabetes and chronic kidney disease: the FIDELITY pooled analysis. Eur Heart J. 2022;43(6):474–84. https:// doi.org/10.1093/eurheartj/ehab777.
- Rouse B, Chaimani A, Li T. Network meta-analysis: an introduction for clinicians. Intern Emerg Med. 2017;12(1):103–11. https://doi.org/10.1007/ s11739-016-1583-7.
- Phillips MR, Steel DH, Wykoff CC, et al. A clinician's guide to network meta-analysis. Eye. 2022;36:1523–6. https://doi.org/10.1038/ s41433-022-01943-5.
- Shi Q, Nong K, Vandvik PO, et al. Benefits and harms of pharmacotherapy for type 2 diabetes: a systematic review and network meta-analysis of randomised controlled trials. BMJ. 2023;381:e074068.
- MAGIC Evidence Ecosystem Foundation. https://magicevidence.org. Accessed 1 Feb 2023.
- MAGIC Evidence Ecosystem Foundation. MATCH-IT tool. https://matchit. magicevidence.org/230125dist-diabetes/#I/. Accessed 1 Feb 2023.
- Marx N, Rydén L, Brosius F, et al. Proceedings of the Guideline Workshop 2019 - Strategies for the optimization of guideline processes in diabetes, cardiovascular diseases and kidney diseases. Diabetes Res Clin Pract. 2020;162:108092. https://doi.org/10.1016/j.diabres.2020.108092.
- Atkins D, Best D, Briss PA, Eccles M, Falck-Ytter Y, Flottorp S, GRADE Working Group, et al. Grading quality of evidence and strength of recommendations. BMC. 2004;328(7454):1490.
- Andrews JC, Schunemann HJ, Oxman AD, et al. GRADE guidelines: 15. Going from evidence to recommendation-determinants of a recommendation's direction and strength. J Clin Epidemiol. 2013;66(7):726–35.
- Alonso-Coello P, Oxman AD, Moberg J, et al. GRADE Evidence to Decision (EtD) frameworks: a systematic and transparent approach to making well informed healthcare choices. 2: Clinical practice guidelines. BMJ. 2016;353:i2089.
- 22. Guyatt GH, Oxman AD, Kunz R, et al. Going from evidence to recommendations. BMJ. 2008;336(7652):1049–51.
- Andrews J, Guyatt G, Oxman AD, et al. GRADE guidelines: 14. Going from evidence to recommendations: the significance and presentation of recommendations. J Clin Epidemiol. 2013;66(7):719–25.
- Brignardello-Petersen R, Florez ID, Izcovich A, et al. GRADE approach to drawing conclusions from a network meta-analysis using a minimally contextualised framework. BMJ. 2020;371:m3900. https://doi.org/10. 1136/bmj.m3900.
- Phillips MR, Sadeghirad B, Busse JW, et al. Development and design validation of a novel network meta-analysis presentation tool for multiple outcomes: a qualitative descriptive study. BMJ Open. 2022;12:e056400. https://doi.org/10.1136/bmjopen-2021-056400.

Publisher's Note

Springer Nature remains neutral with regard to jurisdictional claims in published maps and institutional affiliations.

Ready to submit your research? Choose BMC and benefit from:

- fast, convenient online submission
- thorough peer review by experienced researchers in your field
- rapid publication on acceptance
- support for research data, including large and complex data types
- gold Open Access which fosters wider collaboration and increased citations
- maximum visibility for your research: over 100M website views per year

At BMC, research is always in progress.

Learn more biomedcentral.com/submissions

