

The impact of excluding adverse neonatal outcomes on the creation of gestational weight gain charts among women from low- and middle-income countries with normal and overweight BMI

RANGEL BOUSQUET CARRILHO, Thais, WANG, Dongqing, HUTCHEON, Jennifer A, WANG, Molin, FAWZI, Wafaie W, KAC, Gilberto <<http://orcid.org/0000-0001-8603-9077>>, ACCROMBESSI, Manfred, ADU-AFARWUAH, Seth, ALVES, João Guilherme, LEAL DE ARAÚJO, Carla Adriane, ARIFEEN, Shams, ARTES, Rinaldo, ASHORN, Per, ASHORN, Ulla, ASSEFA, Nega, AYOOLA, Omolola Olukemi, AZIZI, Fereidoun, BAWAH, Ahmed Tijani, BEHBOUDI-GANDEVANI, Samira, BERHANE, Yemane, BERNSTEIN, Robin, BHUTTA, Zulfiqar, BRIAND, Valérie, CALVO, Elvira Beatriz, CARDOSO, Marly Augusto, CHENG, Yue, CHICO-BARBA, Gabriela, CLAYTON, Peter Ellis, COLLINS, Shalean M, COSTELLO, Anthony M, CRUICKSHANK, John Kennedy, DEVAKUMAR, Delanjathan, DEWEY, Kathryn G, DWARKANATH, Pratibha, ESTRADA-GUTIERREZ, Guadalupe, FAIR, Frankie <<http://orcid.org/0000-0001-7613-3393>>, FARIAS, Dayana Rodrigues, FRIIS, Henrik, GHOSH, Shibani, GIRARD, Amy Webb, GOMO, Exnevia, GONDWE, Austrida, HALLAMAA, Lotta, HAMBIDGE, K Michael, HUSSEIN, Hawawu, HUYBREGTS, Lieven, IQBAL, Romaina, KATZ, Joanne, KHATRY, Subarna K, KOLSTEREN, Patrick, KREBS, Nancy F, KULMALA, Teija, KUMAR, Pratap, KURPAD, Anura V, LACHAT, Carl, LARTEY, Anna, LAUER, Jacqueline M, LI, Qian, LIPOETO, Nur Indrawaty, LÓPEZ, Laura Beatriz, LOY, See Ling, MAIYA, G Arun, MALETA, Kenneth, MALTA, Maíra Barreto, MANANDHAR, Dharma S, MANGANI, Charles, MARTÍNEZ-ROJANO, Hugo, MARTIN-PREVEL, Yves, MARTORELL, Reynaldo, MATIAS, Susana L, MCCLURE, Elizabeth M, MELSE-BOONSTRA, Alida, MILLER, Joshua D, MOHAMAD, Marhazlina, JAN MOHAMED, Hamid Jan, MOORE, Sophie, MOSQUERA, Paola Soledad, MRIDHA, Malay Kanti, MUNIM, Shama, MUÑOZ-MANRIQUE, Cinthya, NATAMBA, Barnabas K, OME-KAIUS, Maria, OSRIN, David, PERICHART-PERERA, Otilia, PRENTICE, Andrew M, RAMACHANDRA, Preetha, RAMAKRISHNAN, Usha, RIVERA, Juan, ROBERFROID, Dominique, RODRIGUES, Patricia Lima, RODRÍGUEZ-CANO, Ameyalli, ROGERSON, Stephen J, RONDÓ, Patricia HC, SÁMANO, Reyna, SAVILLE, Naomi M, SHIVALLI, Siddharudha, SHRESTHA, Bhim P, SHRESTHA, Robin, ROBERTO DA SILVA JÚNIOR, José, SOLTANI, Hora <<http://orcid.org/0000-0001-9611-6777>>, SOOFI, Sajid, TEHRANI, Fahimeh Ramezani, THOMAS, Tinku, TIELSCH, James M, UNGER, Holger W, VAZ,

Juliana dos Santos, WORKU, Alemayehu, YANG, Nianhong, YOUNG, Sera L, YUSSIF, Adam Bawa, ZENG, Lingxia, ZHONG, Chunrong and ZHU, Zhonghai

Available from Sheffield Hallam University Research Archive (SHURA) at:

<http://shura.shu.ac.uk/33614/>

This document is the author deposited version. You are advised to consult the publisher's version if you wish to cite from it.

Published version

RANGEL BOUSQUET CARRILHO, Thais, WANG, Dongqing, HUTCHEON, Jennifer A, WANG, Molin, FAWZI, Wafaie W, KAC, Gilberto, ACCROMBESSI, Manfred, ADU-
AFARWUAH, Seth, ALVES, João Guilherme, LEAL DE ARAÚJO, Carla Adriane,
ARIFEEN, Shams, ARTES, Rinaldo, ASHORN, Per, ASHORN, Ulla, ASSEFA, Nega,
AYOOLA, Omolola Olukemi, AZIZI, Fereidoun, BAWAH, Ahmed Tijani, BEHBOUDI-
GANDEVANI, Samira, BERHANE, Yemane, BERNSTEIN, Robin, BHUTTA, Zulfiqar,
BRIAND, Valérie, CALVO, Elvira Beatriz, CARDOSO, Marly Augusto, CHENG, Yue,
CHICO-BARBA, Gabriela, CLAYTON, Peter Ellis, COLLINS, Shalean M,
COSTELLO, Anthony M, CRUICKSHANK, John Kennedy, DEVAKUMAR,
Delanjathan, DEWEY, Kathryn G, DWARKANATH, Pratibha, ESTRADA-
GUTIERREZ, Guadalupe, FAIR, Frankie, FARIAS, Dayana Rodrigues, FRIIS,
Henrik, GHOSH, Shibani, GIRARD, Amy Webb, GOMO, Exnevia, GONDWE,
Austrida, HALLAMAA, Lotta, HAMBIDGE, K Michael, HUSSEIN, Hawawu,
HUYBREGTS, Lieven, IQBAL, Romaina, KATZ, Joanne, KHATRY, Subarna K,
KOLSTEREN, Patrick, KREBS, Nancy F, KULMALA, Teija, KUMAR, Pratap,
KURPAD, Anura V, LACHAT, Carl, LARTEY, Anna, LAUER, Jacqueline M, LI, Qian,
LIPOETO, Nur Indrawaty, LÓPEZ, Laura Beatriz, LOY, See Ling, MAIYA, G Arun,
MALETA, Kenneth, MALTA, Maíra Barreto, MANANDHAR, Dharma S, MANGANI,
Charles, MARTÍNEZ-ROJANO, Hugo, MARTIN-PREVEL, Yves, MARTORELL,
Reynaldo, MATIAS, Susana L, MCCLURE, Elizabeth M, MELSE-BOONSTRA, Alida,
MILLER, Joshua D, MOHAMAD, Marhazlina, JAN MOHAMED, Hamid Jan, MOORE,
Sophie, MOSQUERA, Paola Soledad, MRIDHA, Malay Kanti, MUNIM, Shama,
MUÑOZ-MANRIQUE, Cinthya, NATAMBA, Barnabas K, OME-KAIUS, Maria, OSRIN,
David, PERICHART-PERERA, Otilia, PRENTICE, Andrew M, RAMACHANDRA,
Preetha, RAMAKRISHNAN, Usha, RIVERA, Juan, ROBERFROID, Dominique,
RODRIGUES, Patricia Lima, RODRÍGUEZ-CANO, Ameyalli, ROGERSON, Stephen
J, RONDÓ, Patricia HC, SÁMANO, Reyna, SAVILLE, Naomi M, SHIVALLI,
Siddharudha, SHRESTHA, Bhim P, SHRESTHA, Robin, ROBERTO DA SILVA
JÚNIOR, José, SOLTANI, Hora, SOOFI, Sajid, TEHRANI, Fahimeh Ramezani,
THOMAS, Tinku, TIELSCH, James M, UNGER, Holger W, VAZ, Juliana dos Santos,
WORKU, Alemayehu, YANG, Nianhong, YOUNG, Sera L, YUSSIF, Adam Bawa,
ZENG, Lingxia, ZHONG, Chunrong and ZHU, Zhonghai (2024). The impact of
excluding adverse neonatal outcomes on the creation of gestational weight gain

charts among women from low- and middle-income countries with normal and overweight BMI. The American Journal of Clinical Nutrition, 119 (6), 1465-1474.

Copyright and re-use policy

See <http://shura.shu.ac.uk/information.html>

The impact of excluding adverse neonatal outcomes on the creation of gestational weight gain charts among women from low- and middle-income countries with normal and overweight BMI

Thais Rangel Bousquet Carrilho^{1,2}, Dongqing Wang³, Jennifer A. Hutcheon², Molin Wang^{5,6}, Wafaie W. Fawzi^{4,5,7}, Gilberto Kac^{*1}, and members of the GWG Pooling Project Consortium.

¹ Nutritional Epidemiology Observatory, Josué de Castro Nutrition Institute, Federal University of Rio de Janeiro, Rio de Janeiro, Brazil.

² Department of Obstetrics and Gynaecology, Faculty of Medicine, University of British Columbia, Vancouver, Canada.

³ Department of Global and Community Health, College of Public Health, George Mason University, Fairfax, Virginia, United States of America.

⁴ Department of Global Health and Population, Harvard T.H. Chan School of Public Health, Harvard University, Boston, Massachusetts, United States of America.

⁵ Department of Epidemiology, Harvard T.H. Chan School of Public Health, Harvard University, Boston, Massachusetts, United States of America.

⁶ Department of Biostatistics, Harvard T.H. Chan School of Public Health, Harvard University, Boston, Massachusetts, United States of America.

⁷ Department of Nutrition, Harvard T.H. Chan School of Public Health, Harvard University, Boston, Massachusetts, United States of America.

Correspondence: Gilberto Kac, Email: gilberto.kac@gmail.com.

* WWF and GK contributed equally as senior authors.

Consortium information

The following are members of the GWG Pooling Project Consortium:

1. Manfred Accrombessi, Faculty of Infectious and Tropical Diseases, Disease Control Department, London School of Hygiene and Tropical Medicine, London, United Kingdom.
2. Seth Adu-Afarwuah, Department of Nutrition and Food Science, University of Ghana, Legon, Ghana.
3. João Guilherme Alves, Instituto de Medicina Integral Prof. Fernando Figueira (IMIP), Recife, Brasil.
4. Carla Adriane Leal de Araújo, Instituto de Medicina Integral Prof. Fernando Figueira (IMIP), Recife, Brasil.
5. Shams Arifeen, Maternal and Child Health Division, International Centre for Diarrheal Disease Research, Dhaka, Bangladesh.
6. Rinaldo Artes, INSPER - Instituto de Ensino e Pesquisa, São Paulo, Brazil.
7. Per Ashorn, Center for Child, Adolescent and Maternal Health Research, Faculty of Medicine and Health Technology, Tampere University and Tampere University Hospital, Tampere, Finland.

8. Ulla Ashorn, Center for Child, Adolescent and Maternal Health Research, Faculty of Medicine and Health Technology, Tampere University and Tampere University Hospital, Tampere, Finland.
9. Nega Assefa, College of Health and Medical Sciences, Haramaya University, Dire Dawa, Ethiopia.
10. Omolola Olukemi Ayoola, Lancashire Teaching Hospitals NHS Foundation Trust, Preston, United Kingdom.
11. Fereidoun Azizi, Endocrine Research Center, Research Institute for Endocrine Sciences, Shahid Beheshti University of Medical Sciences, Tehran, Iran.
12. Ahmed Tijani Bawah, Department of Medical Laboratory Technology, Faculty of Allied Health and Pharmaceutical Sciences, Tamale Technical University, Tamale, Ghana.
13. Samira Behboudi-Gandevani, Faculty of Nursing and Health Sciences, Nord University, Bodø, Norway.
14. Yemane Berhane, Addis Continental Institute of Public Health, Addis Ababa, Ethiopia.
15. Robin Bernstein, Department of Anthropology, University of Colorado, Boulder, Colorado, United States of America.
16. Zulfiqar Bhutta, Centre for Global Child Health, Hospital for Sick Children, Toronto, Ontario, Canada & Institute for Global Health & Development, the Aga Khan University, Karachi, Pakistan.
17. Valérie Briand, National French research institute for sustainable development (IRD), Bordeaux, France.
18. Elvira Beatriz Calvo, Department of Nutrition, Mother & Child Health Direction. Ministry of Health, Argentina.
19. Marly Augusto Cardoso, School of Public Health, University of São Paulo. São Paulo, Brazil.
20. Yue Cheng, Department of Nutrition and Food Safety Research, School of Public Health, Xi'an Jiaotong University Health Science Center, Xi'an, Shaanxi 710061, P.R. China.
21. Gabriela Chico-Barba, Nutrition and Bioprogramming Coordination, Instituto Nacional de Perinatología, Mexico City, Mexico.
22. Peter Ellis Clayton, Faculty of Biology, Medicine & Health, University of Manchester, Manchester, United Kingdom.
23. Shalean M. Collins, Tulane University School of Public Health and Tropical Medicine, New Orleans, United States of America.
24. Anthony M Costello, University College London Institute for Global Health, London, United Kingdom.
25. John Kennedy Cruickshank, St Thomas' & Guy's Hospitals, King's College/ King's Health Partners, London, United Kingdom.
26. Delanjathan Devakumar, University College London Institute for Global Health, London, United Kingdom.
27. Kathryn G Dewey, Department of Nutrition, University of California, Davis, United States of America.
28. Pratibha Dwarkanath, Division of Nutrition, St. John's Research Institute, Bangalore, India.
29. Guadalupe Estrada-Gutierrez, Research Direction, Instituto Nacional de Perinatología, Mexico City, Mexico.
30. Frankie J. Fair, College of Health, Wellbeing and Life Sciences, Sheffield Hallam University, Sheffield, United Kingdom.

31. Dayana Rodrigues Farias, Nutritional Epidemiology Observatory, Josué de Castro Institute of Nutrition, Federal University of Rio de Janeiro, Rio de Janeiro, Brazil.
32. Henrik Friis, Department of Nutrition, Exercise and Sports, University of Copenhagen, Copenhagen, Denmark.
33. Shibani Ghosh, Friedman School of Nutrition Science and Policy, Tufts University, Boston, United States of America.
34. Amy Webb Girard, Rollins School of Public Health, Emory University, Atlanta, United States of America.
35. Exnevia Gomo, Faculty of Medicine and Health Sciences, University of Zimbabwe, Zimbabwe.
36. Austrida Gondwe, UNC Project, Tidziwe Centre, Kamuzu Central Hospital, Lilongwe, Malawi.
37. Lotta Hallamaa, Center for Child, Adolescent and Maternal Health Research, Faculty of Medicine and Health Technology, Tampere University, Tampere, Finland.
38. K. Michael Hambidge (deceased), University of Colorado School of Medicine, Aurora, Colorado, United States of America.
39. Hawawu Hussein, Department of Medical Laboratory Technology, Faculty of Allied Health and Pharmaceutical Sciences, Tamale Technical University, Tamale, Ghana.
40. Lieven Huybregts, Department of Food Technology, Safety and Health, Ghent University, Gent, Belgium & Nutrition, Diets, and Health Unit, International Food Policy Research Institute, Washington, DC, United States of America.
41. Romaina Iqbal, Department of Community Health Sciences, Aga Khan University, Karachi, Pakistan.
42. Joanne Katz, Department of International Health, Johns Hopkins Bloomberg School of Public Health, Baltimore, MD, United States of America.
43. Subarna K. Khatry, Nepal Nutrition Intervention Project Sarlahi Project, Kathmandu, Nepal.
44. Patrick Kolsteren, Department of Food Technology, Safety and Health, Ghent University, Gent, Belgium.
45. Nancy F. Krebs, University of Colorado School of Medicine, Aurora, Colorado, United States of America.
46. Teija Kulmala, Pihlajalinna Group, Kehräsaari B, Tampere, Finland.
47. Pratap Kumar, Department of Reproductive Medicine and Surgery, Kasturba Medical College, Manipal Academy of Higher Education, Karnataka, India.
48. Anura V. Kurpad, Department of Physiology, St. John's Medical College, Bangalore, India.
49. Carl Lachat, Department of Food Technology, Safety and Health, Ghent University, Gent, Belgium.
50. Anna Lartey, Department of Nutrition and Food Science, University of Ghana, Legon, Ghana.
51. Jacqueline M Lauer, Department of Health Sciences, College of Health & Rehabilitation Sciences: Sargent College, Boston University, Boston, United States of America.
52. Qian Li, Department of Nutrition and Food Hygiene, Hubei Key Laboratory of Food Nutrition and Safety, MOE Key Laboratory of Environment and Health, School of Public Health, Tongji Medical College, Huazhong University of Science & Technology, Wuhan, China.
53. Nur Indrawaty Lipoeto, Department of Nutrition, Andalas University, Padang, Indonesia.

54. Laura Beatriz López, Nutrition School, Faculty of Medicine, University of Buenos Aires, Buenos Aires, Argentina.
55. See Ling Loy, Duke-NUS Medical School, Singapore & Department of Reproductive Medicine, KK Women's and Children's Hospital, Singapore.
56. G. Arun Maiya, Department of Physiotherapy, Manipal College of Health Professions, Manipal Academy of Higher Education, Karnataka, India.
57. Kenneth Maleta, Department of Nutrition and Dietetics, School of Global and Public Health, Kamuzu University of Health Sciences, Blantyre, Malawi.
58. Maíra Barreto Malta, Faculty of Medicine, University of Western São Paulo, Guarujá, Brazil.
59. Dharma S Manandhar, Mother and Infant Research Activities (MIRA), GPO Box 921, Kathmandu, Nepal.
60. Charles Mangani, School of Global and Public Health, Kamuzu University of Health Sciences, Blantyre, Malawi.
61. Hugo Martínez-Rojano, Escuela Superior de Medicina del Instituto Politécnico Nacional, Mexico City, Mexico.
62. Yves Martin-Prevel, MoISA, University of Montpellier, IRD, CIRAD, CIHEAM-IAMM, INRAE, Institut Agro, Montpellier, France.
63. Reynaldo Martorell, Hubert Department of Global Health, Rollins School of Public Health, Emory University, Atlanta, United States of America.
64. Susana L Matias, Department of Nutritional Sciences and Toxicology, University of California, Berkeley, United States of America.
65. Elizabeth M. McClure, RTI International, Durham, North Carolina, United States of America.
66. Alida Melse-Boonstra, Division of Human Nutrition and Health, Wageningen University & Research, the Netherlands.
67. Joshua D. Miller, Department of Nutrition, University of North Carolina at Chapel Hill, North Carolina, United States of America.
68. Marhazlina Mohamad, School of Nutrition and Dietetics, Faculty of Health Sciences, Universiti Sultan Zainal Abidin (UniSZA), Terengganu, Malaysia.
69. Hamid Jan Jan Mohamed, Nutrition Programme, School of Health Sciences, Universiti Sains Malaysia, Malaysia.
70. Sophie Moore, Department of Women and Children's Health, King's College London, St Thomas' Hospital, Westminster Bridge Road, London, United Kingdom & MRC Unit The Gambia at the London School of Hygiene and Tropical Medicine, Fajara, The Gambia.
71. Paola Soledad Mosquera, School of Public Health, University of São Paulo, São Paulo, Brazil.
72. Malay Kanti Mridha, Center for Non-communicable Diseases and Nutrition, BRAC James P Grant School of Public Health, BRAC University, Dhaka, Bangladesh.
73. Shama Munim, Aga Khan University, Karachi, Pakistan.
74. Cinthya Muñoz-Manrique, Nutrition and Bioprogramming Coordination, Instituto Nacional de Perinatología, Mexico City, Mexico.
75. Barnabas K. Natamba (deceased), Department of Research and Development, Ministry of Science Technology and Innovation, Kampala, Republic of Uganda.
76. Maria Ome-Kaius, Papua New Guinea Institute of Medical Research, Papua New Guinea.
77. David Osrin, University College London Institute for Global Health, London, United Kingdom.

78. Otilia Perichart-Perera, Nutrition and Bioprogramming Coordination, Instituto Nacional de Perinatología, Mexico City, Mexico.
79. Andrew M. Prentice, MRC Unit The Gambia at London School of Hygiene & Tropical Medicine, The Gambia.
80. Preetha Ramachandra, Department of Physiotherapy, Manipal College of Health Professions, Manipal Academy of Higher Education, Karnataka, India.
81. Usha Ramakrishnan, Hubert Department of Global Health, Rollins School of Public Health, Emory University, Atlanta, United States of America.
82. Juan Rivera, National Institute of Public Health, Cuernavaca, Morelos, Mexico.
83. Dominique Roberfroid, Faculty of Medicine, University of Namur, Belgium & Belgian Health Care Knowledge Centre (KCE), Brussels, Belgium.
84. Patricia Lima Rodrigues, Instituto de Puericultura e Pediatria Martagão Gesteira, Nutrition Division, Federal University of Rio de Janeiro, Rio de Janeiro, Brazil.
85. Ameyalli Rodríguez-Cano, Nutrition and Bioprogramming Coordination, Instituto Nacional de Perinatología, Mexico City, Mexico.
86. Stephen J Rogerson, Department of Infectious Diseases, Doherty Institute, The University of Melbourne, Melbourne, Australia.
87. Patricia H C Rondó, Nutrition Department, School of Public Health, University of São Paulo, São Paulo, Brazil.
88. Reyna Sámano, Nutrition and Bioprogramming Coordination, Instituto Nacional de Perinatología, Mexico City, Mexico.
89. Naomi M. Saville, University College London Institute for Global Health, London, United Kingdom.
90. Siddharudha Shivalli, London School of Hygiene & Tropical Medicine, London, United Kingdom.
91. Bhim P. Shrestha, Health Research & Develop Forum (HRDF), Kathmandu, Nepal.
92. Robin Shrestha, Friedman School of Nutrition Science and Policy, Tufts University, Boston, United States of America.
93. José Roberto da Silva Júnior, Instituto de Medicina Integral Prof. Fernando Figueira (IMIP), Recife, Brazil.
94. Hora Soltani, College of Health, Wellbeing and Life Sciences, Sheffield Hallam University, Sheffield, United Kingdom.
95. Sajid Soofi, Centre of Excellence in Women and Child Health, the Aga Khan University, Karachi, Pakistan.
96. Fahimeh Ramezani Tehrani, Reproductive Endocrinology Research Center, Research Institute for Endocrine Sciences, Shahid Beheshti University of Medical Sciences, Tehran, Iran.
97. Tinku Thomas, Department of Biostatistics, St. John's Medical College, Bangalore, India.
98. James M. Tielsch, George Washington Milken Institute School of Public Health, Washington, DC, United States of America.
99. Holger W Unger, Menzies School of Health Research, Charles Darwin University, Australia.
100. Juliana dos Santos Vaz, Faculty of Nutrition, Federal University of Pelotas, Pelotas, Brazil.
101. Alemayehu Worku, School of Public Health, Addis Ababa University, Addis Ababa, Ethiopia.
102. Nianhong Yang, Department of Nutrition and Food Hygiene, Hubei Key Laboratory of Food Nutrition and Safety, MOE Key Laboratory of Environment and Health, School

of Public Health, Tongji Medical College, Huazhong University of Science & Technology, Wuhan, China.

103. Sera L. Young, Department of Anthropology, Institute for Policy Research, Northwestern University, Evanston, IL, United States of America.

104. Adam Bawa Yussif, Department of Languages and International Relations, Faculty of Applied Arts, Tamale Technical University, Ghana.

105. Lingxia Zeng, Department of Epidemiology and Biostatistics, School of Public Health, Xi'an Jiaotong University Health Science Center, Xi'an, P.R. China.

106. Chunrong Zhong, Department of Nutrition and Food Hygiene, Hubei Key Laboratory of Food Nutrition and Safety, School of Public Health, Tongji Medical College, Huazhong University of Science and Technology, Wuhan, China.

107. Zhonghai Zhu, Department of Epidemiology and Biostatistics, School of Public Health, Xi'an Jiaotong University Health Science Center, Xi'an, P.R. China.

Usha Ramakrishnan and Tinku Thomas are Editorial Board Members for The American Journal of Clinical Nutrition and played no role in the Journal's evaluation of the manuscript.

Nancy Krebs is an Editor for The American Journal of Clinical Nutrition and played no role in the Journal's evaluation of the manuscript.

1 **Abstract**

2 **Background:** Existing gestational weight gain (GWG) charts vary considerably in their
3 choice of exclusion/inclusion criteria, and it is unclear to what extent these criteria
4 create differences in the charts' percentile values. We aimed to establish the impact of
5 including/excluding pregnancies with adverse neonatal outcomes when constructing
6 GWG charts.

7 **Methods:** This is an individual participant data analysis from 31 studies from low- and
8 middle-income countries. We created a dataset that included all participants and a
9 dataset restricted to those with no adverse neonatal outcomes: preterm < 37 weeks,
10 small or large for gestational age – SGA or LGA, low birth weight < 2,500 g, or
11 macrosomia > 4,000 g. Quantile regression models were used to create GWG curves
12 from 9 to 40 weeks, stratified by pre-pregnancy BMI, in each dataset.

13 **Results:** The dataset without the exclusion criteria applied included 14,685 individuals
14 with normal weight and 4,831 with overweight. After removing adverse neonatal
15 outcomes, 10,479 individuals with normal and 3,466 individuals with overweight
16 remained. GWG distributions at 13, 27, and 40 weeks were virtually identical between
17 the datasets with and without the exclusion criteria, except at 40 weeks for normal
18 weight and 27 weeks for overweight. For the 10th and 90th percentiles, the differences
19 between the estimated GWG were larger for overweight (approximately 1.5 kg)
20 compared to normal weight (< 1 kg). Removal of adverse neonatal outcomes had
21 minimal impact on GWG trajectories of normal weight. For overweight, the percentiles
22 estimated in the dataset without the criteria were slightly higher than those in the dataset
23 with the criteria applied. Nevertheless, differences were < 1 kg and virtually nonexistent
24 at the end of pregnancy.

25 **Conclusions:** Removing pregnancies with adverse neonatal outcomes had little or no
26 influence on the GWG trajectories of individuals with normal and overweight.

27

28 **Abbreviations**

29 BMI: Body mass index

30 GA: Gestational age

31 GWG: Gestational weight gain

32 IPD: Individual participant data

33 Ki: Knowledge integration

34 LBW: Low birth weight

35 LGA: Large for gestational age

36 LMICs: low- and middle-income countries

37 SGA: Small for gestational age

38 WHO: World Health Organization

39

40 **Keywords:** gestational weight gain; reference; standards; adverse neonatal outcomes

41

42 **Introduction**

43 Gestational weight gain (GWG) is an important indicator to be monitored during
44 pregnancy because deviations in this indicator are associated with adverse outcomes for
45 both mothers and their infants (1-3). Several GWG curves have been created in the last
46 ten years using national or local data to facilitate the monitoring of GWG during
47 prenatal care (4-11).

48 Although most of the GWG curves have tried to adopt a prescriptive approach – i.e.,
49 charts were constructed using data from a selective, low-risk population to describe how
50 weight gain *ought* to be (12), there is an inconsistency among the criteria adopted in
51 defining the underlying sample to construct the curves. One of these criteria is the

52 removal of individuals who gave birth to children with adverse neonatal outcomes, such
53 as preterm birth, small- (SGA) or large-for-gestational-age (LGA) births, low birth
54 weight (LBW), or macrosomia. Several studies removed individuals who gave birth to
55 preterm infants (4, 5, 7, 9, 10), and some of them also mentioned the removal of
56 individuals based on birth weight (7-10). However, in several studies, including in the
57 INTERGROWTH-21st standards, there is no mention of individuals with those
58 characteristics being removed (6, 11, 13).

59 The decision to include or exclude pregnancies with adverse neonatal outcomes from
60 GWG curves is an important design consideration for ongoing initiatives to create
61 charts, such as the World Health Organization's (WHO) new initiative to develop global
62 standards for GWG (14). Excluding individuals who gave birth to neonates with adverse
63 outcomes has major implications for sample size. If excluding those individuals is
64 necessary, investigators will have to recruit larger samples in early pregnancy to
65 account for those subsequent exclusions. This consideration is especially relevant for
66 researchers from low- and middle-income countries (LMICs), where the prevalence of
67 many adverse neonatal outcomes is higher than in other regions. However, if the
68 exclusion of pregnancies with adverse neonatal outcomes has little practical impact on
69 the chart percentiles, this exclusion may not be necessary. The goal of this study was to
70 identify the impact of removing participants with adverse neonatal outcomes (preterm
71 birth, SGA, LGA, LBW, and macrosomia) when constructing GWG curves, using a
72 pooled dataset from studies conducted in LMICs.

73

74

75

76 **Methods**

77

78 *Study design and selection of study population*

79 This analysis used data from the GWG Pooling Project, a cohort of individual
80 participant data (IPD) from multiple studies that addressed various knowledge gaps of
81 the determinants and consequences of inadequate and excessive GWG in LMICs. The
82 details of the GWG Pooling Project have been described elsewhere (15).

83 In February and March 2019, the GWG Pooling Project team conducted a systematic
84 search using PubMed, Embase, and Web of Science. The search aimed to identify
85 randomized controlled trials and observational cohort studies that collected multiple
86 weight measurements among pregnant individuals (pregnant at enrollment or enrolled
87 before pregnancy and followed up in pregnancy) in low-income, lower-middle-income,
88 or upper-middle-income economies, defined by the World Bank country classification
89 for the 2019 fiscal year. No language restrictions were imposed, but a publication date
90 restriction of 2000 and later was defined, to capture relatively recent studies for
91 generalizability. Studies were excluded if they did not collect two or more measures of
92 gestational weight or only collected self-reported gestational weight. Studies conducted
93 exclusively among women with pre-existing health conditions, such as anemia, human
94 immunodeficiency virus infection, or diabetes were also excluded.

95 Two team members independently screened the titles and abstracts of the identified
96 studies based on the eligibility criteria, with any discrepancies resolved by discussion.

97 They also reviewed the full texts of the remaining studies to confirm final eligibility,
98 with any discrepancies resolved by discussion. For studies that remained after the full-

99 text screening, individual study investigators were invited to participate in a survey

100 designed to confirm study eligibility and to indicate their willingness to contribute

101 individual-level data. For study teams willing to participate, the Knowledge Integration

102 (Ki) team at the Bill & Melinda Gates Foundation worked with the principal
103 investigators of the studies to pursue data contributor agreements with the respective
104 institutions. As IPD became available, we worked with the Ki team to examine data
105 completeness, map relevant variables, and harmonize the data across studies.
106 Contributed data were harmonized based on a pre-specified variable list and variable
107 definition. We systematically assessed whether the variables included in the analysis
108 were defined consistently across all studies. Among the 337 investigators contacted,
109 approximately 50% responded to the survey, of whom 145 led studies that were eligible
110 for the pooled analysis and were invited to contribute data, out of which 56 studies
111 made IPD available to the project (**Supplementary Figure 1**).

112 For the present study, we included adult pregnant individuals (19 - 45 years old), with
113 singleton pregnancies, without a record of abortion, stillbirth, neonatal, or maternal
114 death in the current pregnancy. We excluded records with implausible gestational age
115 estimates (outside the 1-301 days interval), implausible GWG measurements, and
116 missing initial weights.

117 We identified and removed biologically implausible values (outliers) and assessed the
118 heterogeneity of GWG data across the included studies (**Supplementary Methods**). We
119 removed all measurements flagged as implausible using the selected cross-sectional and
120 longitudinal methods and studies with GWG measurements considered heterogeneous.

121 We also removed GWG measurements obtained < 9 and > 40 weeks due to sample size
122 reasons and because no weight gain related to pregnancy is expected before 9 weeks
123 (16). The dataset retained only a small sample size for the underweight ($n = 3,091$
124 individuals; 7,415 repeated measurements; 12% of the sample) and obesity ($n = 2,106$
125 individuals; 3,630 measurements; 8.2% of the sample) BMI categories after

126 implementing those procedures, which made it uninformative to consider these BMI
127 categories in the analyses.

128 To create the dataset without applying the neonatal exclusion criteria, we selected only
129 individuals classified as normal or overweight according to pre-pregnancy BMI (17).

130 From this dataset, we also removed those with missing data in the variables necessary to
131 construct the neonatal outcomes, i.e., birth weight, sex of the newborn, and gestational
132 age at birth.

133

134 *Creation of the dataset after applying the neonatal exclusion criteria*

135 This dataset was created by excluding all individuals who gave birth to an infant
136 classified as preterm (gestational age at birth < 37 completed weeks) (18), SGA (birth
137 weight < 10th percentile of the INTERGROWTH-21st newborn size standards)(19),
138 LGA (birth weight > 90th percentile of the INTERGROWTH-21st newborn size
139 standards) (19), with LBW (birth weight < 2,500 g), or macrosomia (birth weight >
140 4,000 g).

141

142 *Main variables*

143 GWG was calculated as the difference between the weight in each prenatal care or study
144 visit and an initial weight. This initial weight was chosen using the following hierarchy
145 of available data: 1) measured pre-pregnancy weight, 2) measured first-trimester weight
146 up to 8 weeks of pregnancy, and 3) self-reported pre-pregnancy weight or a pre-
147 pregnancy weight abstracted from medical records.

148 Pre-pregnancy BMI was calculated using the initial weight in kilograms, and the first
149 registered height, in meters squared. BMI (kg/m^2) was then classified according to the
150 World Health Organization (WHO) cut-offs as underweight (< 18.5 kg/m^2), normal

151 weight (≥ 18.5 and < 25.0 kg/m²), overweight (≥ 25.0 and < 30.0 kg/m²) and obesity (\geq
152 30.0 kg/m²) (17).

153 Gestational age (GA) was ascertained through various methods across the contributing
154 studies. When multiple methods for ascertaining GA were used in a single study,
155 preferences were given in the following order (from the most prioritized to the least
156 prioritized): crown-rump length, fetal biometry (biparietal diameter and femur length),
157 best obstetric estimate, GA as reported by the ultrasound machine, last menstrual
158 period, new Ballard score, GA as provided in the raw data, and the estimated date of
159 delivery.

160

161 *Ethics*

162 This is a secondary analysis of existing data. All studies included in this analysis were
163 approved by their respective ethics committees.

164

165 *Statistical analyses*

166 Characteristics of the study population were described by calculating the 25th, 50th
167 (median), and 75th percentiles for continuous variables and counts with percentages for
168 categorical ones. The analyses were stratified by normal weight and overweight pre-
169 pregnancy BMI categories only.

170 To account for the non-linear relationship between GWG and GA, we identified the best
171 fitting powers for the GWG curves using second-degree fractional polynomials (20),
172 which were further modeled in quantile regressions with robust errors. Quantile
173 regressions with robust errors were used to create GWG curves from 9 to 40 weeks for
174 the 10th, 25th, 50th, 75th, and 90th percentiles. The predicted GWG values with
175 accompanying 95% confidence intervals for the 25th, 50th, and 75th percentiles for pre-

176 pregnancy normal and overweight at 13, 27, and 40 weeks were compared. Differences
177 between the percentiles were considered clinically meaningful if they were > 1 kg.
178 Density plots were created to compare the distributions of GWG in each dataset at 13,
179 27, and 40 weeks for normal and overweight pregnant individuals. The predicted 10th
180 and 90th percentiles from the quantile regressions were added to the graphs. The
181 analyses were performed in R, version 4.2.0, and Stata, version 15.

182

183 **Results**

184 The initial pooled dataset comprised 56 studies and 109,567 individuals with 689,196
185 measurements. Initial data cleaning and application of eligibility criteria reduced the
186 dataset to 28,176 individuals and 66,151 measurements (**Figure 1**).

187 The cross-sectional methods applied identified 1,359 individuals and 4,970 weight
188 measurements and 17 pre-pregnancy BMI measurements as outliers (data not shown).
189 Subsequently, the identification of longitudinal outliers flagged 23 individuals and 161
190 measurements as implausible that were removed from the dataset (**Supplementary**
191 **Figure 2**).

192 Several studies in both BMI categories had heterogenous GWG distribution
193 (**Supplementary Figure 3**). Therefore, 669 individuals and 5,584 GWG measurements
194 were removed from studies or visits conducted in intervals in which the study was
195 considered heterogeneous. GWG measurements taken outside the 9-40 weeks range
196 were removed, resulting in a dataset with 25,532 individuals and 53,631 measurements.
197 Finally, removing individuals with pre-pregnancy underweight and obesity and with
198 missing data on birth variables resulted in the dataset without the neonatal exclusion
199 criteria applied, with 14,685 individuals with normal weight and 4,831 with overweight,
200 with 32,130 and 8,825 weight gain measurements, respectively (**Figure 1**). Of these,

201 10,479 individuals and 22,118 GWG measurements for normal weight and 3,466
202 individuals and 6,232 GWG measurements for overweight remained after application of
203 the neonatal exclusion criteria.

204 The dataset for normal weight comprised data from 16 countries; for overweight, 13
205 countries were included. Among normal weight participants, the median (IQR) maternal
206 age was 27.0 years (24.0 – 30.0), while for overweight, the median (IQR) age was 30.0
207 (26.0 – 33.0) years. The prevalence of LGA was 9.8% for normal weight and 14.4% for
208 overweight. The prevalence of preterm birth (8.3 vs. 6.7%) and macrosomia (4.3 vs.
209 3.1%) was also higher among overweight individuals (**Table 1**). We did not observe any
210 differences in the distribution of key variables (BMI, GWG, maternal age) in the final
211 cohort (**Table 1**) and the data before removing missing data on birth variables
212 (**Supplementary Table 1**). GWG distribution at the end of each trimester (13, 27, and
213 40 weeks) was virtually identical for both BMI categories, except at 40 weeks for
214 normal and 27 weeks for overweight. For normal weight, there was a shift to the right in
215 the GWG distribution at 40 weeks. For overweight, at 27 weeks, a small shift to the left
216 was observed. When we compared the 10th and 90th percentiles, the differences between
217 the curves created in the dataset with and without the exclusion criteria applied were
218 larger for overweight (approx. 1.5 kg) compared to normal weight individuals (< 1 kg)
219 (**Figures 2 and 3**).

220 The population-level GWG percentiles of normal-weight individuals were very similar
221 when charts with and without removing the adverse neonatal outcomes were compared.
222 The GWG values at each of the estimated percentiles tended to be higher when those
223 individuals were removed, but the differences were always < 1 kg and virtually
224 nonexistent at the end of pregnancy (**Figure 4A**). For overweight, the GWG values at
225 each of the estimated percentiles tended to be lower when those individuals were

226 removed. The differences between the curves for overweight were also larger when
227 compared to normal weight, especially for the 75th and 90th percentiles. However, at 40
228 weeks, all differences were < 1 kg (**Figure 4B**).

229 In general, the GWG values at each of the estimated percentiles of the curves obtained
230 in the dataset without neonatal adverse outcomes were modestly lower than those
231 estimated in the dataset with those outcomes, but differences were all < 1 kg for both
232 BMI categories when the most central percentiles of the distribution (25th, 50th, and 75th)
233 were examined, and confidence intervals were overlapping (**Table 2**). For normal
234 weight women, the predicted GWG 75th percentile at 40 weeks was 9.6 kg in the curve
235 modeled in the dataset without the application of the neonatal exclusion criteria, and
236 10.1 kg in the curve modeled after removal of those outcomes, representing the largest
237 difference of 0.5 kg. For overweight, the largest difference (0.8 kg) was also observed
238 for the GWG at the 75th percentile at 27 weeks: 8.0 kg in the curves created in the
239 dataset without the criteria applied v. 7.2 kg in those created in the dataset after
240 application of the criteria.

241

242 **Discussion**

243 In this analysis of IPD from multiple LMIC cohorts, the exclusion of individuals who
244 gave birth to neonates who were preterm, classified as SGA or LGA, or with LBW or
245 macrosomia had little or no influence on GWG trajectories and the estimated GWG
246 percentiles throughout pregnancy of mothers with pre-pregnancy normal or overweight.
247 For normal weight, GWG trajectories and the predicted values of GWG at selected
248 percentiles and gestational ages were virtually identical in the datasets with and without
249 the exclusion criteria applied. For overweight, the estimated percentiles for the dataset
250 without excluding the neonatal outcomes were slightly higher than those estimated in

251 the data after removing those, but the differences were always < 1 kg and virtually
252 nonexistent at the end of pregnancy.

253 There are two possible explanations for the relatively small impact of removing
254 individuals with adverse neonatal outcomes on GWG trajectories. First, individuals with
255 adverse neonatal outcomes constituted only a small fraction of the cohort (< 15%);
256 exclusion of this relatively small number of records may not have been sufficient to
257 influence the GWG trajectories. The exclusion of adverse outcomes with higher
258 prevalence may have a greater influence on GWG trajectories. However, the prevalence
259 of the adverse neonatal outcomes observed in this dataset is similar to those reported in
260 other studies using data from LMICs (21-23), supporting the generalizability of our
261 findings.

262 A second possible explanation is that the magnitude of the association between GWG
263 and the selected adverse outcomes is modest. Although insufficient and excessive GWG
264 are consistently reported to be associated with those outcomes, the results of meta-
265 analyses show odds ratios between 1.1 – 2.5 (1-3). Farias et al. (24) also showed that the
266 ability of GWG classified according to several curves to predict SGA and LGA in a
267 sample of Brazilian women had low sensitivity and high specificity for both outcomes.
268 The modest magnitude of association and the low ability of GWG to predict those
269 outcomes could help explain the small or no effect that the removal of individuals with
270 adverse outcomes had on the GWG curves.

271 To the best of our knowledge, this is the first study to analyze the role of excluding
272 adverse neonatal outcomes on GWG charts. The comparison of the values of GWG
273 observed for this set of normal-weight individuals from LMICs with the standards
274 proposed by the INTERGROWTH-21st, created based on data of a highly-selected
275 sample (6) showed that the differences between the predicted GWG at 40 weeks for the

276 25th, 50th, and 75th were mostly < 1 kg. For example, the 50th percentile of
277 INTERGROWTH-21st standards corresponds to a weight gain of 13.7 kg, while in the
278 dataset of the LMICs data without excluding the adverse neonatal outcomes, the
279 predicted GWG is 12.9 kg.

280 The practical difference between references and standards has been previously
281 evaluated by Hutcheon and Liauw (25) in the context of fetal growth. These authors
282 observed that the distribution of estimated fetal weight obtained from a reference versus
283 standard population (which included 30% of the referent cohort) from Canada was
284 similar, suggesting that exclusion criteria played no role on the fetal growth charts
285 created from those datasets. This small difference aligns with the results of the current
286 study for GWG.

287 In many situations, excluding individuals with selected characteristics from GWG
288 charts is the standard procedure adopted without further reflection on the real impact of
289 those factors in the curves. This removal poses a critical constraint on the available
290 sample size for constructing those curves and should be carefully evaluated. In addition,
291 it is important to mention that other relevant factors that should be considered in
292 creating GWG standards, such as excess postpartum weight retention, child obesity,
293 gestational diabetes, and hypertensive disorders of pregnancy, were not evaluated in the
294 current study. Therefore, future studies aiming to develop GWG standards should
295 consider the true influence of removing individuals with those and other characteristics
296 related to GWG in constructing the curves.

297 The strengths of this study include the use of a large dataset from multiple settings and
298 the application of rigorous procedures to flag and remove implausible values and assess
299 the data's heterogeneity and construct the curves. Limitations include the insufficient
300 data for individuals with underweight and obesity, and the lack of a measured pre-

301 pregnancy weight. Many studies rely on a first-trimester measurement as a proxy of the
302 weight at conception, which may not be accurate due to expected weight changes in this
303 period. Thus, we decided to create a priority “first weight” variable considering a
304 hierarchy based on agreement analyses performed in the same dataset (data not shown).
305 The low number of individuals with underweight and obesity is of particular concern,
306 especially the lack of data for underweight individuals in LMICs. Since SGA and LBW
307 are more prevalent among individuals with underweight, it is possible that the exclusion
308 of these pregnancies could have a larger impact on weight gain distributions than among
309 individuals with normal weight. Repeating our analyses in cohorts where undernutrition
310 is more common and the prevalence of averse neonatal outcomes associated with GWG
311 is higher, such as South Asia and Sub-Saharan Africa, would be valuable.

312 The exclusion of individuals who gave birth to neonates classified as preterm or with
313 SGA, LGA, LBW, or macrosomia had little or no influence on the GWG percentiles of
314 individuals with normal- and overweight. Thus, excluding those individuals from GWG
315 curves may have little practical impact. However, an essential step is to repeat these
316 analyses in cohorts with higher prevalence of underweight and obesity. Combining
317 these findings with results from the present study will allow more definitive conclusions
318 on the need to exclude pregnancies complicated by adverse neonatal outcomes from
319 GWG curves. In addition, future studies considering GWG determinants and other
320 GWG-related outcomes, such as excess postpartum weight retention and child obesity,
321 are needed to clarify the appropriate criteria to be adopted to construct GWG standards.

Acknowledgements

TRBC, GK, and JAH planned the data analyses. TRBC analyzed the data and wrote the manuscript with input from DQ, JAH, MW, WWF, and GK. WWF is the coordinator of the GWG pooling project. WWF and GK contributed equally as senior authors. TRBC, JAH, and GK had primary responsibility for the final content. All authors read and approved the final manuscript.

Data Availability

The data that support the findings of this study are available from the Knowledge integration (Ki) initiative (Bill and Melinda Gates Foundation), but restrictions apply to the availability of these data, which were used under license for the current study, and so are not publicly available. Data are however available upon reasonable request and with signature of data access agreements with the Ki.

Funding

This study was supported by funding from the Bill and Melinda Gates Foundation (INV-016436). TRBC is a Michael Smith Health Research BC Research Trainee. JAH holds a Canada Research Chair in Perinatal Population Health. The funders had no role in study design, data collection and analysis, decision to publish or preparation of the manuscript.

Author Disclosures

The authors declare that they have no conflicts to disclose.

References

1. Goldstein RF, Abell SK, Ranasinha S, Misso ML, Boyle JA, Harrison CL, et al. Gestational weight gain across continents and ethnicity: systematic review and meta-analysis of maternal and infant outcomes in more than one million women. *BMC Medicine*. 2018;16(1):153.
2. Goldstein RF, Abell SK, Ranasinha S, Misso M, Boyle JA, Black MH, et al. Association of gestational weight gain with maternal and infant outcomes: a systematic review and meta-analysis. *JAMA*. 2017;317(21):2207-25.
3. Rogozinska E, Zamora J, Marlin N, Betran AP, Astrup A, Bogaerts A, et al. Gestational weight gain outside the Institute of Medicine recommendations and adverse pregnancy outcomes: analysis using individual participant data from randomised trials. *BMC Pregnancy Childbirth*. 2019;19(1):322.
4. Hutcheon JA, Platt RW, Abrams B, Himes KP, Simhan HN, Bodnar LM. A weight-gain-for-gestational-age z score chart for the assessment of maternal weight gain in pregnancy. *Am J Clin Nutr*. 2013;97(5):1062-7.
5. Hutcheon JA, Platt RW, Abrams B, Himes KP, Simhan HN, Bodnar LM. Pregnancy weight gain charts for obese and overweight women. *Obesity*. 2015;23(3):532-5.
6. Cheikh Ismail L, Bishop DC, Pang R, Ohuma EO, Kac G, Abrams B, et al. Gestational weight gain standards based on women enrolled in the Fetal Growth Longitudinal Study of the INTERGROWTH-21st Project: a prospective longitudinal cohort study. *BMJ*. 2016;352:i555.
7. Santos S, Eekhout I, Voerman E, Gaillard R, Barros H, Charles MA, et al. Gestational weight gain charts for different body mass index groups for women in Europe, North America, and Oceania. *BMC Medicine*. 2018;16(1):201.

8. Huang A, Xiao Y, Hu H, Zhao W, Yang Q, Ma W, et al. Gestational weight gain charts by gestational age and body mass index for Chinese women: A population-based follow-up study. *J Epidemiol.* 2019.
9. Kac G, Carrilho TRB, Rasmussen KM, Reichenheim ME, Farias DR, Hutcheon JA. Gestational weight gain charts: results from the Brazilian Maternal and Child Nutrition Consortium. *Am J Clin Nutr.* 2021;113(5):1351-60.
10. Xu J, Luntamo M, Kulmala T, Ashorn P, Cheung YB. A longitudinal study of weight gain in pregnancy in Malawi: unconditional and conditional standards. *Am J Clin Nutr.* 2014;99(2):296-301.
11. Thiruvengadam R, Desiraju BK, Natchu UCM, Wadhwa N, Sachdeva K, Misra S, et al. Gestational weight gain trajectories in GARBH-Ini pregnancy cohort in North India and a comparative analysis with global references. *Eur J Clin Nutr.* 2022;76(6):855-62.
12. Ananth CV, Brandt JS, Vintzileos AM. Standard vs population reference curves in obstetrics: which one should we use? *Am J Obstet Gynecol.* 2019;220(4):293-6.
13. Johansson K, Hutcheon JA, Stephansson O, Cnattingius S. Pregnancy weight gain by gestational age and BMI in Sweden: a population-based cohort study. *Am J Clin Nutr.* 2016;103(5):1278-84.
14. World Health Organization, First global call for data on gestational weight gain [Internet] [cited February 28, 2024]. Available from: <https://www.who.int/news-room/articles-detail/first-global-call-for-data-on-gestational-weight-gain>.
15. Liu E, Wang D, Darling AM, Perumal N, Wang M, Ahmed T, et al. Effects of prenatal nutritional supplements on gestational weight gain in low- and middle-income countries: a meta-analysis of individual participant data. *Am J Clin Nutr.* 2022;116(6):1864-76.
16. Pitkin RM. Nutritional support in obstetrics and gynecology. *Clinical obstetrics and gynecology.* 1976;19(3):489-513.

17. WHO Expert Committee on Physical Status. Physical status: the use and interpretation of anthropometry. Geneva: World Health Organization; 1995. x, 452 p. p.
18. World Health Organization. ICD-10 version: 2010. International Statistical Classification of Diseases and Related Health Problems, 10th Revision. Geneva: WHO; 2010.
19. Villar J, Cheikh Ismail L, Victora CG, Ohuma EO, Bertino E, Altman DG, et al. International standards for newborn weight, length, and head circumference by gestational age and sex: the Newborn Cross-Sectional Study of the INTERGROWTH-21st Project. *Lancet*. 2014;384(9946):857-68.
20. Royston P, Altman DG. Regression using fractional polynomials of continuous covariates: parsimonious parametric modelling. *J R Stat Soc, C: Appl Stat*. 1994;43(3):429-67.
21. Falcao IR, Ribeiro-Silva RC, de Almeida MF, Fiaccone RL, Silva NJ, Paixao ES, et al. Factors associated with small- and large-for-gestational-age in socioeconomically vulnerable individuals in the 100 Million Brazilian Cohort. *Am J Clin Nutr*. 2021;114(1):109-16.
22. Pusdekar YV, Patel AB, Kurhe KG, Bhargav SR, Thorsten V, Garces A, et al. Rates and risk factors for preterm birth and low birthweight in the global network sites in six low- and low middle-income countries. *Reproductive Health*. 2020;17(3):187.
23. Lee AC, Katz J, Blencowe H, Cousens S, Kozuki N, Vogel JP, et al. National and regional estimates of term and preterm babies born small for gestational age in 138 low-income and middle-income countries in 2010. *The Lancet Global Health*. 2013;1(1):e26-36.

24. Farias DR, Carrilho TRB, Rasmussen KM, Hutcheon JA, Reichenheim ME, Barros DC, et al. Comparison between the Brazilian and three international gestational weight gain charts. *Am J Clin Nutr.* 2022.
25. Hutcheon JA, Liauw J. Should fetal growth charts be references or standards? *Epidemiology.* 2021;32(1):14-7.

Tables

Table 1. Maternal and neonatal characteristics of the cohort before applying the neonatal exclusion criteria according to pre/early-pregnancy body mass index (BMI) category.

	Normal weight (BMI \geq 18.5 and $<$ 25.0 kg/m ² , 14,685 individuals and 32,130 measurements)	Overweight (BMI \geq 25.0 and $<$ 30.0 kg/m ² , 4,831 individuals and 8,825 measurements)
<i>Continuous</i>		
	Median (IQR)	
Maternal age	27.0 (24.0 – 30.0)	30.0 (26.0 – 33.0)
Maternal pre-/early-pregnancy BMI	21.3 (19.9 – 22.9)	27.1 (25.9 – 28.3)
Gestational weight gain (kg)		
First trimester	0.7 (-0.6 – 2.3)	0.2 (-1.1 – 2.0)
Second trimester	2.5 (0.7 – 4.6)	2.1 (-0.2 – 5.2)
Third trimester	8.1 (4.9 – 12.0)	9.0 (6.0 – 12.5)
Birth weight (g)	3,180 (2,880 – 3,500)	3,250 (2,960 – 3,510)
Gestational age at birth (weeks)	39.4 (38.4 – 40.3)	39.1 (38.1 – 40.0)
<i>Categorical</i>		
	Number of individuals (%)	
Country		
Argentina	658 (4.5)	224 (4.6)
Bangladesh	484 (3.3)	47 (1.0)
Benin	133 (0.9)	29 (0.6)
Brazil	776 (5.3)	350 (7.2)

Burkina Faso	165 (1.1)	4 (0.08)
China	5,788 (39.4)	520 (10.8)
Ghana	154 (1.0)	105 (2.2)
Guatemala	167 (1.1)	110 (2.3)
India	230 (1.6)	48 (1.0)
Islamic Republic of Iran	3,615 (24.6)	3,162 (65.4)
Mexico	311 (2.1)	184 (3.8)
Nepal	1,731 (11.8)	-
Nigeria	34 (0.2)	-
Pakistan	51 (0.3)	9 (0.2)
Papua New Guinea	8 (0.05)	-
The Gambia	380 (2.6)	39 (0.8)
Neonatal outcomes		
Small for gestational age	2,043 (13.9)	386 (8.0)
Large for gestational age	1,444 (9.8)	698 (14.4)
Preterm birth	986 (6.7)	400 (8.3)
Low birth weight	1,043 (7.1)	264 (5.5)
Macrosomia	458 (3.1)	207 (4.3)

Notes: BMI: body mass index; IQR: interquartile range.

Table 2. Smoothed gestational weight gain values and 95% confidence intervals (in kg) for the selected percentiles at the end of each pregnancy trimester.

Percentiles		Normal weight (BMI ≥ 18.5 and < 25.0 kg/m ²)			Overweight (BMI ≥ 25.0 and < 30.0 kg/m ²)		
		13 weeks	27 weeks	40 weeks	13 weeks	27 weeks	40 weeks
		(1 st trimester)	(2 nd trimester)	(3 rd trimester)	(1 st trimester)	(2 nd trimester)	(3 rd trimester)
<i>25th percentile</i>							
Dataset without the							
exclusion criteria		-0.3 (-0.4 – -0.3)	2.7 (2.7 – 2.8)	9.6 (9.4 – 9.8)	-1.3 (-1.6 – -1.1)	2.2 (2.0 – 2.5)	8.5 (8.2 – 8.8)
applied							
Dataset with the							
exclusion criteria		-0.3 (-0.4 – -0.3)	3.1 (3.0 – 3.2)	10.1 (9.9 – 10.3)	-1.8 (-2.0 – -1.5)	2.2 (2.0 – 3.3)	8.3 (8.1 – 8.6)
applied							
<i>50th percentile</i>							

Dataset without the exclusion criteria applied	1.1 (1.0 – 1.1)	4.8 (4.7 – 4.9)	12.9 (12.7 – 13.0)	0.5 (0.3 – 0.6)	5.1 (4.9 – 5.3)	11.1 (10.9 – 11.3)
Dataset with the exclusion criteria applied	1.1 (1.0 – 1.1)	5.2 (5.1 – 5.3)	13.1 (12.9 – 13.2)	0.4 (0.3 – 0.6)	4.7 (4.6 – 4.8)	11.3 (11.1 – 11.5)
 <i>75th percentile</i>						
Dataset without the exclusion criteria applied	2.7 (2.6 – 2.8)	7.1 (7.0 – 7.2)	16.2 (16.0 – 16.4)	2.2 (2.0 – 2.4)	8.0 (7.7 – 8.2)	14.2 (13.8 – 14.5)
Dataset with the exclusion criteria applied	2.7 (2.6 – 2.8)	7.5 (7.4 – 7.6)	16.3 (16.1 – 16.5)	2.4 (2.2 – 2.5)	7.2 (7.1 – 7.3)	14.7 (14.1 – 15.0)

Notes: BMI: body mass index.

Figure legends

Figure 1: Flowchart for the construction of the database.

Notes: **Initial weight refers to a pre-pregnancy weight measured, self-reported or abstracted from medical records or a weight measured at the beginning of pregnancy (up to 8 weeks). Body mass index (BMI) was calculated based on a pre-pregnancy weight or in early pregnancy (up to 8 weeks). Normal weight: $\text{BMI} \geq 18.5$ and $< 25.0 \text{ kg/m}^2$; Overweight: $\text{BMI} \geq 25.0$ and $< 30.0 \text{ kg/m}^2$.

Abbreviations: GWG: gestational weight gain; SD: standard deviation; SSD: standardized site difference.

Figure 2: Distribution of gestational weight gain of the dataset without v. dataset with the exclusion criteria applied for individuals with normal weight: **A:** at 13 weeks; **B:** at 27 weeks; **C:** at 40 weeks.

Notes: Body mass index (BMI) was calculated based on a pre-pregnancy weight or in early pregnancy (up to 8 weeks). Normal weight: $\text{BMI} \geq 18.5$ and $< 25.0 \text{ kg/m}^2$. Vertical lines refer to the predicted 10th and 90th percentiles of the curves generated using quantile regression: lighter lines are the values for the dataset without the application of the criteria and darker lines are for the dataset after the criteria was applied. The latter was created from the larger dataset (n=14,685 individuals and 32,130 GWG measurements), by removing individuals who gave birth to an infant classified as preterm, small-for-gestational-age or large-for-gestational-age, and low birth weight or macrosomia (n = 10,479 individuals and 22,118 GWG measurements).

Figure 3: Distribution of gestational weight gain of the dataset without v. dataset with the exclusion criteria applied for individuals with overweight: **A:** at 13 weeks; **B:** at 27 weeks; **C:** at 40 weeks.

Notes: Body mass index (BMI) was calculated based on a pre-pregnancy weight or in early pregnancy (up to 8 weeks). Overweight: $\text{BMI} \geq 25.0$ and $< 30.0 \text{ kg/m}^2$. Vertical lines refer to the predicted 10th and 90th percentiles of the curves generated using quantile regression: lighter lines are the values for the dataset without the application of the criteria and darker lines are for the dataset after the criteria was applied. The latter was created from the larger dataset (n=4,831 individuals and 8,825 GWG measurements), by removing individuals who gave birth to an infant classified as preterm, small-for-gestational-age or large-for-gestational-age, and low birth weight or macrosomia (n = 3,466 individuals and 6,232 GWG measurements).

Figure 4: Comparison of the gestational weight gain trajectories of the dataset without v. dataset with the exclusion criteria applied: **A:** Normal weight; **B:** Overweight.

Notes: Body mass index (BMI) was calculated based on a pre-pregnancy weight or in early pregnancy (up to 8 weeks). Normal weight: $\text{BMI} \geq 18.5$ and $< 25.0 \text{ kg/m}^2$; Overweight: $\text{BMI} \geq 25.0$ and $< 30.0 \text{ kg/m}^2$. The dataset with the exclusion criteria applied was created from the larger dataset (normal weight: n=14,685 individuals and 32,130 GWG measurements; overweight: n=4,831 individuals and 8,825 GWG measurements), by removing individuals who gave birth to an infant classified as preterm, small-for-gestational-age or large-for-gestational-age, and low birth weight or macrosomia (normal weight: n = 10,479 individuals and 22,118 GWG measurements; overweight: n = 3,466 individuals and 6,232 GWG measurements).