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Agreement between the laboratory- and non-laboratory-based 2019 WHO cardiovascular risk charts in Peru

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3 **Agreement between the laboratory- and non-laboratory-based 2019 WHO cardiovascular**
4 **risk charts in Peru**
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ABSTRACT

Objective: To determine the agreement between the cardiovascular disease (CVD) risk predictions computed with the World Health Organization (WHO) non-laboratory-based model and laboratory-based model in a nationally-representative sample of Peruvian adults.

Design: Cross-sectional analysis of a national health survey.

Methods: Absolute CVD risk was computed with the 2019 WHO laboratory- and non-laboratory-based models. The risk predictions from both models were compared with Bland Altman plots, Lin's concordance coefficient correlation (LCCC) and kappa statistics, stratified by sex, age, body mass index categories, smoking and diabetes status.

Results: 663 people aged 30-59 years were included in the analysis. Overall, there were no substantial differences between the mean CVD risk computed with the laboratory-based model 2.0% (95% CI: 1.8%; 2.2%) and the non-laboratory-based model 2.0% (95% CI: 1.8%; 2.1%). In the Bland Altman plots, the limits of agreement were the widest among people with diabetes (-0.21; 4.37) compared with people without diabetes (-1.17; 0.95). The lowest agreement as per the LCCC was also seen in people with diabetes (0.74 (95% CI: 0.63; 0.82)), the same was observed with the kappa statistic (kappa=0.36). In general, agreement between the scores was appropriate in terms of clinical significance.

Conclusions: The absolute cardiovascular predicted risk was similar between the laboratory-based and non-laboratory-based 2019 WHO cardiovascular risk models. The agreement between these models was less clear in people with cardiovascular risk factors: obesity, smokers and people with diabetes.

Key words: cardiovascular diseases; risk assessment; health metrics

STRENGTHS AND LIMITATIONS OF THIS STUDY

- This analysis provided the first evidence of the agreement between the 2019 WHO CVD risk laboratory-based and non-laboratory-based models.
- We leveraged on the most recent nationally representative survey that included blood biomarkers in Peru.
- Our study population was young, mostly women, and with overall low absolute cardiovascular predicted risk. No one in the study population had an absolute cardiovascular risk $\geq 20\%$.
- We assumed all participants were free of CVD to use the 2019 WHO CVD risk score, as information regarding history of cardiovascular events was not reported.

INTRODUCTION

Cardiovascular diseases (CVD) are the main cause of death globally.¹ In 2019, CVD caused ~18.5 million deaths in adults, representing 36% of all global deaths.^{2,3} Furthermore, CVD impose a huge burden in low- and middle-income countries (LMICs),⁴ where deaths from CVD occur at younger ages compared to high-income countries (HICs).¹ However, CVD can be prevented and managed through a combination of population- and individual-level interventions;⁵ for the latter, the identification of individuals at high cardiovascular risk is a cornerstone in the prevention of CVD. In this line, the World Health Organization (WHO) and the Pan American Health Organization (PAHO),⁶ alongside several clinical guidelines,⁷⁻⁹ recommend CVD risk stratification with CVD risk prediction models to inform evidence-based treatment.

CVD risk prediction models identify people who would benefit the most of preventive interventions (e.g., statin therapy).¹⁰ Although there are several CVD risk prediction models,¹¹ these were mostly developed in HICs limiting their application in LMICs where they would need recalibration to deliver accurate predictions to guide treatment allocation. To overcome this limitation, the WHO convened a global effort to derive, calibrate and validate new CVD risk prediction models for all world regions.¹² Two WHO CVD risk prediction models were developed: a laboratory-based model and a non-laboratory-based model. Because laboratory biomarkers (e.g., total cholesterol)

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3 may not be available in primary health care centres in LMICs,¹³ the non-laboratory-based model
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5 arises as a handy tool for clinicians in LMICs. Nonetheless, evidence regarding the agreement
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7 between the WHO laboratory-based and non-laboratory-based models in countries from Latin
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9 America is missing.¹² Whether the WHO non-laboratory-based model delivers predictions similar
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11 to those of the WHO laboratory-based model remains unknown in Latin America. However,
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13 clinicians need this evidence to inform their choice (non-laboratory-based vs laboratory-based
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15 model), and to interpret the results of the non-laboratory-based model under the assumption that
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17 the laboratory-based model is the gold standard.¹² To provide this evidence for practitioners in
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19 Peru, we determined the agreement between the risk predictions computed with the non-
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21 laboratory-based model and laboratory-based model in a nationally-representative sample of
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23 Peruvian adults.

24 25 26 **METHODS**

27 28 **Data sources**

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30 This is a cross-sectional study of a national survey conducted by the National Centre for Food
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32 and Nutrition (CENAN, for its acronym in Spanish) of Peru. CENAN's survey was conducted
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34 between 2017-2018 on a nationally-representative sample of Peruvian adults aged between 18-
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36 59 years.¹⁴ Of note, this is the most recent nationally-representative survey conducted in Peru
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38 that included blood biomarkers (e.g., lipid profile).

39 40 41 **Study population**

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43 We analysed a complete-case sample regarding all the laboratory- and office-based 2019 ten-
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45 year WHO CVD risk score variables (see Variables section). We studied men and women aged
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47 between 30-59 years. The younger age limit was decided because CVD risk models are not
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49 recommended in younger individuals; the older age limit was decided because of the survey
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51 design. A flowchart of data cleaning is shown in Supplementary figure 1. We did not apply other
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53 selection criteria.

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57 Although the 2019 ten-year WHO CVD risk models were developed for people aged 40-80
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59 years,¹² for people aged <40 years we assumed they had 40 years for the absolute CVD risk
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3 computation. This is consistent with the Package of essential noncommunicable (PEN) disease
4 interventions for primary health care in low-resource settings,¹⁵ and was also done in a recent
5 global work.¹⁶
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10 **Variables**

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12 We calculated the ten-year CVD risk at the individual-level following the laboratory- and non-
13 laboratory-based 2019 ten-year WHO CVD risk model.¹² We used the *whocvdrisk* command in
14 STATA, which was developed by the authors of the 2019 ten-year WHO CVD risk charts.¹⁷ For
15 the laboratory-based model, scores were calculated based on: age (years), current smoking
16 status (yes/no), systolic blood pressure (SBP, mmHg), history of self-reported diabetes diagnosis
17 (yes/no), and total cholesterol (mmol/L).¹² For the non-laboratory-based model, we used age
18 (years), current smoking status (yes/no), systolic BP (mmHg) and body mass index (BMI,
19 kg/m²).¹²
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30 The CENAN's survey collected anthropometrics and three BP measurements which were taken
31 by trained fieldworkers following a standard protocol.¹⁴ We computed BMI using measured weight
32 (kg) divided by the square of height (meters); for descriptive purposes, we classified BMI in three
33 levels: normal weight (BMI <25 kg/m²), overweight (BMI ≥25-29.9 kg/m²), and obesity (BMI ≥30
34 kg/m²). BMI records outside the range 10-80 kg/m² were discarded. As the third BP measurement
35 was only available in few participants (<2% of the initial sample), we used the second SBP
36 measurement only (i.e., the first and third SBP records were discarded in the main analysis). Of
37 note, there were no substantial differences between the first and second SBP records
38 (Supplementary table 1); nonetheless, we performed a sensitivity analysis using the mean SBP
39 of first and second SBP records. We discarded any SBP records outside the range 70-270 mmHg.
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51 For people who self-reported being under antihypertensive treatment, we used the pre-treatment
52 SBP; this is consistent with the PEN protocol and with a previous global work.^{15,16} Pre-treatment
53 SBP was computed as: pre-treatment systolic blood pressure = (current systolic blood pressure-
54 6.3)/0.9.¹⁸ Conversely, for those not taking antihypertensive treatment, we used the recorded SBP
55 as was.
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5 We defined current smoker with one question coded as no versus yes: *do you currently smoke*
6 *any tobacco product such as cigarettes, cigars or pipes?* Self-reported information about a prior
7 history of diabetes was assessed by a question also coded as no versus yes: *have you ever been*
8 *told by a physician or another healthcare worker that you have high blood sugar or diabetes?*
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14 Total cholesterol was obtained via enzymatic colorimetric method.¹⁴ Because CENAN's survey
15 data had total cholesterol in mg/dl, these values were divided by 38.67 to obtain total cholesterol
16 in mmol/L.
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20 21 22 **Statistical analysis** 23

24 We determined the agreement between the absolute CVD risk predicted with the WHO laboratory-
25 and non-laboratory-based models following three methods: Bland Altman plots, Lin's
26 concordance coefficient correlation (LCCC), and kappa statistic. We considered the absolute
27 CVD risk as a continuous variable, and the agreement between both models was examined using
28 Bland Altman plots and the LCCC. Furthermore, we considered the CVD risk as a categorical
29 variable, and divided into three groups: <5%, 5-9%, 10-19%; because there were no observations
30 in the high-risk category (CVD risk $\geq 20\%$), agreement was not examined in this group. For these
31 categories, we evaluated the agreement using the kappa statistic. For the Bland Altman plots,
32 LCCC, and the kappa statistic, results were stratified by sex, 10-year age groups, BMI categories,
33 smoking status, self-reported diabetes diagnosis and urban/rural location.
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45 In the Bland Altman plots, the risk difference between the laboratory- and non-laboratory-based
46 absolute cardiovascular predicted risk was plotted on the vertical axis, and the mean of both
47 scores on the horizontal axis.¹⁹ As the true risk of CVDs at the individual-level is uncertain, the
48 mean of both scores is the best available estimate.^{19,20} The 95% of the limit of agreement was
49 represented by the mean difference of both scores \pm two standard deviations; this limit provides
50 an interval in which 95% of the differences between both scores would be expected to lie.¹⁹ The
51 LCCC between the laboratory- and non-laboratory-based absolute cardiovascular predicted risks
52 was also evaluated. The agreement based on the LCCC ranges between -1 and 1, with 1
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3 suggesting a perfect agreement. The categorical agreement was evaluated using the Kappa
4 statistic. Kappa <0 indicated less than chance agreement, and values 0.01–0.20, 0.21–0.40,
5 0.41–0.60, 0.61–0.80, and 0.81–0.99 represented slight, fair, moderate, substantial, and almost
6 perfect agreement, respectively.²¹
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12 The datasets we used and the analysis code in R (version 4.0.3) and STATA (version 17.0,
13 College Station, Texas 77845, USA) are available as Supplementary Files 1-2. Population
14 characteristics along with their 95% confidence interval (95% CI) were summarized accounting
15 for the complex survey design of the CENAN's survey.¹⁴
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20 21 22 **Ethics**

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24 This study used de-identified nationally-representative survey data that can be requested from
25 the CENAN. We did not seek approval by an Ethics Committee.
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30 31 **Patient and public involvement**

32 No patient involved.
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36 37 **Role of the funding source**

38 The funder had no role in the study design, analysis, interpretation or decision to publish. The
39 authors are collectively responsible for the accuracy of the data. The arguments and opinions in
40 this work are those of the authors alone, and do not represent the position of the institutions to
41 which they belong.
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48 49 **RESULTS**

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51 Our pooled dataset included 663 participants (Supplementary figure 1). The mean age was 44.0
52 years (95% CI: 43.2; 44.7) and the proportion of men was 41.5%. The mean SBP was 108.9
53 mmHg (95% CI: 107.5; 110.3 mmHg) and the mean BMI was 28.8 kg/m² (95% CI: 28.3; 29.2
54 kg/m²). The proportion of people with overweight was 41.0% (95% CI: 36.6; 45.6 %), whereas
55 35.9% (95% CI: (31.6; 40.5%)) of the population were obese. The mean total cholesterol was 4.9
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mmol/L (95% CI: 4.8; 5.0 mmol/L), 11.7% (95% CI: 8.8; 15.3%) of the population were smokers and 7.1% (95% CI: 5.0%; 9.8%) had diabetes (Table 1).

Table 1. Weighted distribution of the predictors in the 2019 WHO CVD risk models, overall and by sex.

	Total	Men	Women
Sample size	663	280	383
Age (mean and 95% CI, years)	44 (43.2-44.7)	44.2 (43-45.4)	43.8 (42.9-44.7)
Proportion of people aged 30-39 years (95% CI, %)	35.8 (31.6-40.3)	37.2 (31-43.9)	34.8 (29.4-40.7)
Proportion of people aged 40-49 years (95% CI, %)	34 (29.8-38.4)	29.5 (24-35.7)	37.1 (31.5-43.1)
Proportion of people aged 50-59 years (95% CI, %)	30.2 (26.3-34.5)	33.3 (27.1-40)	28 (23.5-33.1)
Systolic blood pressure (mean and 95% CI, mmHg)	108.9 (107.5-110.3)	116.1 (113.8-118.4)	103.8 (102.3-105.3)
Diastolic blood pressure (mean and 95% CI, mmHg)	71.9 (71-72.8)	74.6 (73.1-76.1)	70 (69-71)
Body mass index (mean and 95% CI, kg/m²)	28.8 (28.3-29.2)	28.3 (27.6-29)	29.1 (28.5-29.7)
Proportion of people with normal weight (95% CI, %)	23.1 (19.5-27.2)	24.5 (19.5-30.4)	22.1 (17.2-27.9)
Proportion of people with overweight (95% CI, %)	41 (36.6-45.6)	45.3 (38.9-52)	37.9 (31.9-44.3)
Proportion of people with obesity (95% CI, %)	35.9 (31.6-40.5)	30.1 (24.1-36.9)	40 (34-46.2)
Total cholesterol (mean and 95% CI, mmol/L)	4.9 (4.8-5)	4.8 (4.6-5)	4.9 (4.8-5.1)
Proportion of smokers (95% CI, %)	11.7 (8.8-15.3)	21.1 (15.8-27.6)	5 (3-8.3)
Proportion of people with diabetes (95% CI, %)	7.1 (5-9.8)	7.8 (4.7-12.6)	6.5 (4.3-9.9)
Laboratory-based CVD risk score (mean and 95% CI, %)	2 (1.8-2.2)	2.6 (2.3-2.9)	1.6 (1.4-1.7)
Non-laboratory-based CVD risk score (mean and 95% CI, %)	2 (1.8-2.1)	2.7 (2.4-3)	1.5 (1.4-1.6)

Absolute cardiovascular risk according to the 2019 WHO cardiovascular risk models

Overall, there were no substantial differences between the mean absolute cardiovascular risk computed with the laboratory-based and non-laboratory-based models (Table 1). The mean absolute cardiovascular risk was 2.0% (95% CI: 1.8%; 2.2%) according to the laboratory-based model, and 2.0% (95% CI: 1.8%; 2.1%) according to the non-laboratory-based model. In both models, the mean absolute cardiovascular risk was higher in men than women. The sensitivity analysis (using the mean of two SBP records) yielded the same findings in the overall sample: 2.0% (95% CI: 1.8; 2.2%) in the laboratory-based-model and 2.0 (95% CI: 1.8; 2.1%) in the non-laboratory-based model.

Mean difference between risk predictions

Overall, the mean difference between the laboratory-based and the non-laboratory-based models was 0.03% (95% CI: -0.03%; 0.10%). According to sex, the mean difference between models was -0.02% (95% CI: -0.14%; 0.09%) in men, and 0.08% (95% CI: 0.01%; 0.15%) in women. According to age, the mean difference between models was -0.07% (95% CI: -0.12%; 0.04%) in people aged 30-39 years, 0.04% (95% CI: -0.08%; 0.15%) in people aged 40-49 years, and 0.17% (95% CI: 0.02%; 0.32%) in people aged 50-59 years. Stratified by BMI categories, the mean difference was 0.10% (95% CI: 0.02%; 0.18%) in people with normal weight, 0.08% (95% CI: 0.00%; 0.17%) in people with overweight, and -0.07% (95% CI: -0.21%; 0.07%) in people with obesity. The mean difference was 0.05% (95% CI: -0.34%; 0.44%) in smokers and 0.03% (95% CI: -0.02%; 0.09%) in non-smokers. According to self-reported diabetes status, mean difference was 2.08% (95% CI: 1.73%; 2.44%) in people with self-reported diabetes, and -0.11% (95% CI: -0.15%; 0.07%) in people without self-reported diabetes. The sensitivity analysis provided similar results across all variables; for example, the largest mean difference was also observed in people with self-reported diabetes (2.09% (95% CI: 1.73; 2.45%) (Supplementary figure 6).

Bland-Altman plots and limits of agreement.

The limit of agreement was slightly narrower for women (-1.23; 1.39) compared to men (-1.93; 1.89) (Figure 1). The limit of agreement widened with older ages and higher BMI levels; for

example, the limit of agreement was narrower in people aged 30-39 years (-0.72; 0.56) compared to those aged 50-59 years (-1.96; 2.29) (Figure 2), and for people with normal weight (-0.90; 1.09) compared to those with obesity (-2.15; 2.02) (Figure 3). According to smoking and self-reported diabetes status, the limit of agreement was wider in those who had the condition. For example, the limit of agreement in smokers (-3.13; 3.24) was wider compared to non-smokers (-1.25; 1.32) (Figure 4). Similarly, in people with self-reported diabetes (-0.21; 4.37) the limit of agreement was wider compared to people without self-reported diabetes (-1.17; 0.95) (Figure 5). Notably, the sensitivity analysis resulted in similar limits of agreement in all variables (Supplementary figures 2-6).

Agreement by Lin's concordance coefficient correlation

The overall agreement between scores as per the LCCC was 0.87 (95% CI: 0.85; 0.89), and it was virtually the same in men (0.87 (95% CI 0.84; 0.89)) and women (0.85 (95% CI 0.82; 0.87)). Across age groups, the highest agreement was seen in the 30-39 age group (0.87 (95% CI 0.84; 0.9)). Across BMI categories, it was the normal BMI category which had the highest agreement (0.90 (95% CI 0.87; 0.92)). Overall, the lowest agreement values were seen across smokers (0.81 (95% CI 0.72; 0.88)), those aged 40-49 years (0.74 (95% CI 0.67; 0.79)), and those with self-reported diabetes (0.74 (95% CI 0.63; 0.82)) (Table 2). Similar results were seen in the sensitivity analysis (Supplementary table 2).

Table 2. Lin's concordance coefficient correlation showing agreement between laboratory and non-laboratory-based risk models according to the predictors in the 2019 WHO CVD risk models and urban/rural location

Variables	Categories	Lin's concordance coefficient correlation (95% CI)
Sex	Men	0.87 (0.84 - 0.89)
	Women	0.85 (0.82 - 0.87)
Age (years)	30-39	0.87 (0.84 - 0.9)
	40-49	0.74 (0.67 - 0.79)
	50-59	0.83 (0.78 - 0.86)
Body mass index category	Normal	0.9 (0.87 - 0.92)

	Overweight	0.87 (0.85 - 0.9)
	Obese	0.86 (0.82 - 0.89)
Smoking status	Smoker	0.81 (0.72 - 0.88)
	Non-smoker	0.86 (0.84 - 0.88)
Diabetes status	With self-reported diabetes	0.74 (0.63 - 0.82)
	Not with self-reported diabetes	0.91 (0.9 - 0.92)
Urban or rural	Urban	0.88 (0.85 - 0.9)
	Rural	0.85 (0.82 - 0.88)

Categorical agreement

In the overall population, there was a slightly larger number of people categorized as having an absolute CVD risk of 5-9% and 10-19% with the laboratory-based-model compared with the non-laboratory-based model (Supplementary table 3). For example, the laboratory-based model categorized 37 people in the 5-9% CVD risk category, whereas the non-laboratory-based model categorized 26 people. Overall, the agreement between risk categories was substantial ($\kappa=0.62$), and it was better for men ($\kappa=0.70$) compared to women ($\kappa=0.44$) (Supplementary table 4). Of note, the lowest agreement between risk categories was observed among people with self-reported diabetes ($\kappa=0.36$): out of 14 people with self-reported diabetes in the 5-9% CVD risk category following the laboratory-based-model, 4 were placed in the same category following the non-laboratory-based model, as the rest were placed in the 0-5% CVD risk category. The categorical agreement according to all variables is presented in Supplementary tables 3-9.

DISCUSSION

Main findings

In this work, we evaluated the agreement between the CVD risk estimates predicted with the 2019 WHO ten-year laboratory-based and non-laboratory-based models in a nationally-representative sample of Peruvian adults. The mean absolute predicted CVD risk according to both models in the general population was virtually the same. In addition, we found that the limits of agreement between both models increased with a higher CVD risk; for instance, the limits of agreement were

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3 wider in smokers and people with self-reported diabetes. We observed good agreement between
4 the laboratory-based and non-laboratory-based models in terms of clinical significance.^{20,22} These
5 findings suggest that, in a population with a similar profile to that in this study, practitioners could
6 use either the laboratory-based or non-laboratory-based models. Although the difference is very
7 small, careful interpretation could be needed for people with cardiovascular risk factors: obesity,
8 smokers and people with self-reported diabetes, amongst whom the difference was slightly larger
9 than in their peers without these risk factors.
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18 **Public health implications**

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20 We provided insights about the applicability of the 2019 WHO non-laboratory-based model in the
21 Peruvian population. This evidence is relevant in terms of clinical practice and public health in
22 Peru and other similar countries (e.g., Andean Latin America), because it informs whether the
23 predictions based on the laboratory-based and non-laboratory-based CVD risk models are
24 equivalent. If so, either of these models could be used without substantial bias, hence supporting
25 the use of the non-laboratory-based model when blood biomarkers are not available.
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34 Our results suggest that the agreement between the laboratory-based and non-laboratory-based
35 models was appropriate among Peruvians with low CVD risk and younger than 60 years. In other
36 words, our results suggest that the laboratory-based and non-laboratory-based models provide
37 similar predictions and may therefore be used interchangeably as needed, though the profile of
38 our study population ought to be considered when extrating or implementing our findings into
39 clinical practice and public health. Of note, the small number of participants in some variables of
40 interest (e.g., only 44 participants had self-reported diabetes) could have explained the broader
41 limits of agreement in our results. Further studies should include a larger number of participants
42 to further confirm whether limits of agreement are wider according to smoking and diabetes status.
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54 **Research in context**

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56 The study most comparable to ours evaluated the agreement between the Framingham 10-year
57 CVD risk laboratory and non-laboratory models on a population aged 40-75 years in southern
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3 Iran.²² They found the mean CVD risk following the non-laboratory-based model (9.4%) was
4 higher than the laboratory-based model (6.7%).²² Additionally, their limits of agreement between
5 both Framingham models in people <60 years old were wider compared to ours in both men (-
6 1.9-1.9 by our estimates versus -2.5%-8.9% by Rezaei et al.²²) and women (-1.2-1.4 by our
7 estimates versus -2.3%-4.6% by Rezaei et al.²²). This could be explained by the fact that Rezaei
8 et al.²² included an older population, which tend to have a higher absolute CVD risk. As limits of
9 agreement between two models tend to widen with higher CVD risk,^{20,22} our limits of agreement
10 would presumably be wider if we had studied a similar population to that of the work by Rezaei et
11 al.²² The differences between our results could be further explained by the CVD risk score herein
12 used. We used the 2019 WHO CVD risk models,¹² whereas Rezaei et al. used the Framingham
13 risk scores. The Framingham risk score was developed for a more specific population
14 (Caucasians),²³ yet the 2019 WHO CVD risk model was developed for a broader population (i.e.,
15 those living in LMICs).¹²
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30 The agreement between the 2019 WHO laboratory- and non-laboratory-based model was also
31 explored in the global work convened by the WHO.¹² They applied the two models to WHO
32 STEPS surveys, and compared the proportion of people categorized at different levels of
33 predicted CVD risk.¹² Overall, they found moderate agreement between both models, and their
34 discrepancy was attributed to poor performance of the non-laboratory-based model in people with
35 diabetes.¹² This finding is consistent with our results because we found the widest limits of
36 agreement, the lowest LCCC, and the lowest categorical agreement in people with self-reported
37 diabetes. When possible, it would seem reasonable to use the laboratory-based model in those
38 whose have diabetes.
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49 **Strengths and limitations**

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51 To the best of our knowledge, we provided the first evidence of the agreement between the 2019
52 WHO CVD risk laboratory-based and non-laboratory-based models; furthermore, we leveraged
53 on a nationally-representative survey. Nonetheless, this study has also limitations. First, our study
54 population was young (30-59 years), mostly women (58%), and with overall low absolute
55 cardiovascular predicted risk. This led to the observation that no one in the study population had
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3 an absolute cardiovascular risk $\geq 20\%$. Thus, we could only draw conclusions for people within
4 the low and medium CVD risk range, and with a similar demographic and risk factor profile. We
5 acknowledge that further subgroup analysis could be relevant, for example by diabetes status.
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7 However, because of data availability and the reduced number of observations in some groups,
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9 this subgroup analyses would be impossible to conduct. Future work in Peru and Latin America
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11 should verify our results with a larger, older and more diverse population. Second, as CENAN's
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13 survey did not include history of cardiovascular events, we assumed all participants were free of
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15 CVD to use the 2019 WHO CVD risk score. This approach could have led to higher absolute
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17 cardiovascular risk because people who have had a cardiovascular event (e.g., myocardial
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19 infarction) are at higher risk of another cardiovascular event. Third, we only used one blood
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21 pressure record (the second SBP measurement); ideally, we should have used the average of
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23 multiple records having discarded the very first measurement. This approach was not possible
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25 with the available data. We performed a sensitivity analysis using the mean SBP of the first and
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27 second records; this revealed virtually the same results as in our main analysis in which only the
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29 second SBP record was used.
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34 **Conclusions**

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36 The absolute cardiovascular predicted risk was similar between the laboratory-based and non-
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38 laboratory based 2019 WHO cardiovascular risk models. The agreement between these models
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40 was less clear in people with cardiovascular risk factors: obesity, smokers and people with
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42 diabetes. While universal health coverage momentum helps to have laboratory tests in (most)
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44 primary care facilities to use the laboratory-based model, it seems reasonable to use the non-
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46 laboratory-based model for primary prevention of CVD following the risk stratification approach.
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FIGURES

Figure 1. Bland Altman plots showing agreement between laboratory and non-laboratory-based risk scores according to sex.

Figure 2. Bland Altman plots showing agreement between laboratory and non-laboratory-based risk scores according to age groups.

Figure 3. Bland Altman plots showing agreement between laboratory and non-laboratory-based risk scores according to body mass index categories

Figure 4. Bland Altman plots showing agreement between laboratory and non-laboratory-based risk scores according to smoking status.

Figure 5. Bland Altman plots showing agreement between laboratory and non-laboratory-based risk scores according to self-reported diabetes status.

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4

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11 GAQ-V, FFV-R, AB-O and RMC-L. All authors provided relevant scientific contribution. All authors
12 approved the submitted version.
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15

16 **Data sharing statement:** Datasets and analysis code in Supplementary Files 1-2.
17

18 **Word count:** 3,260
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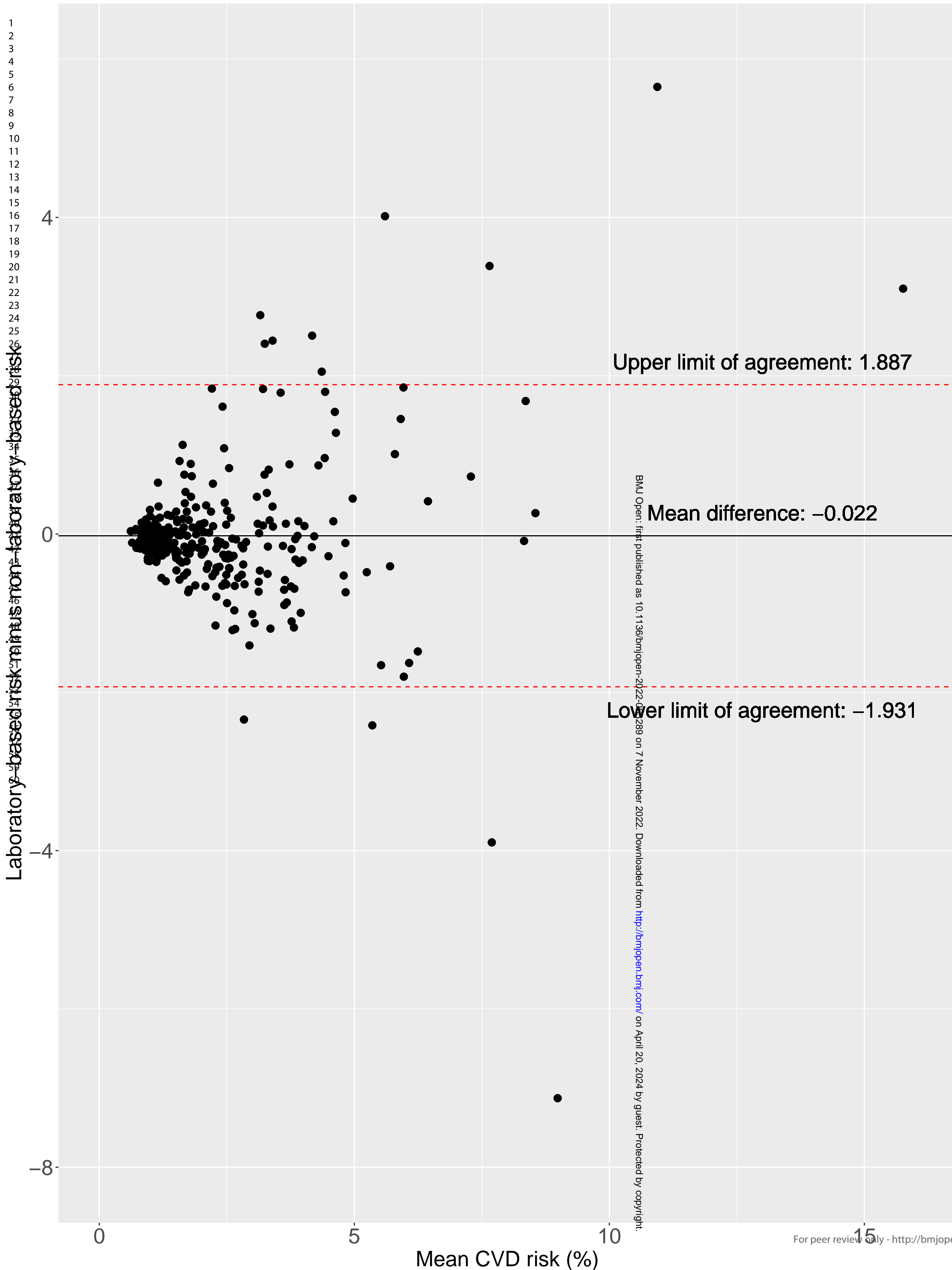
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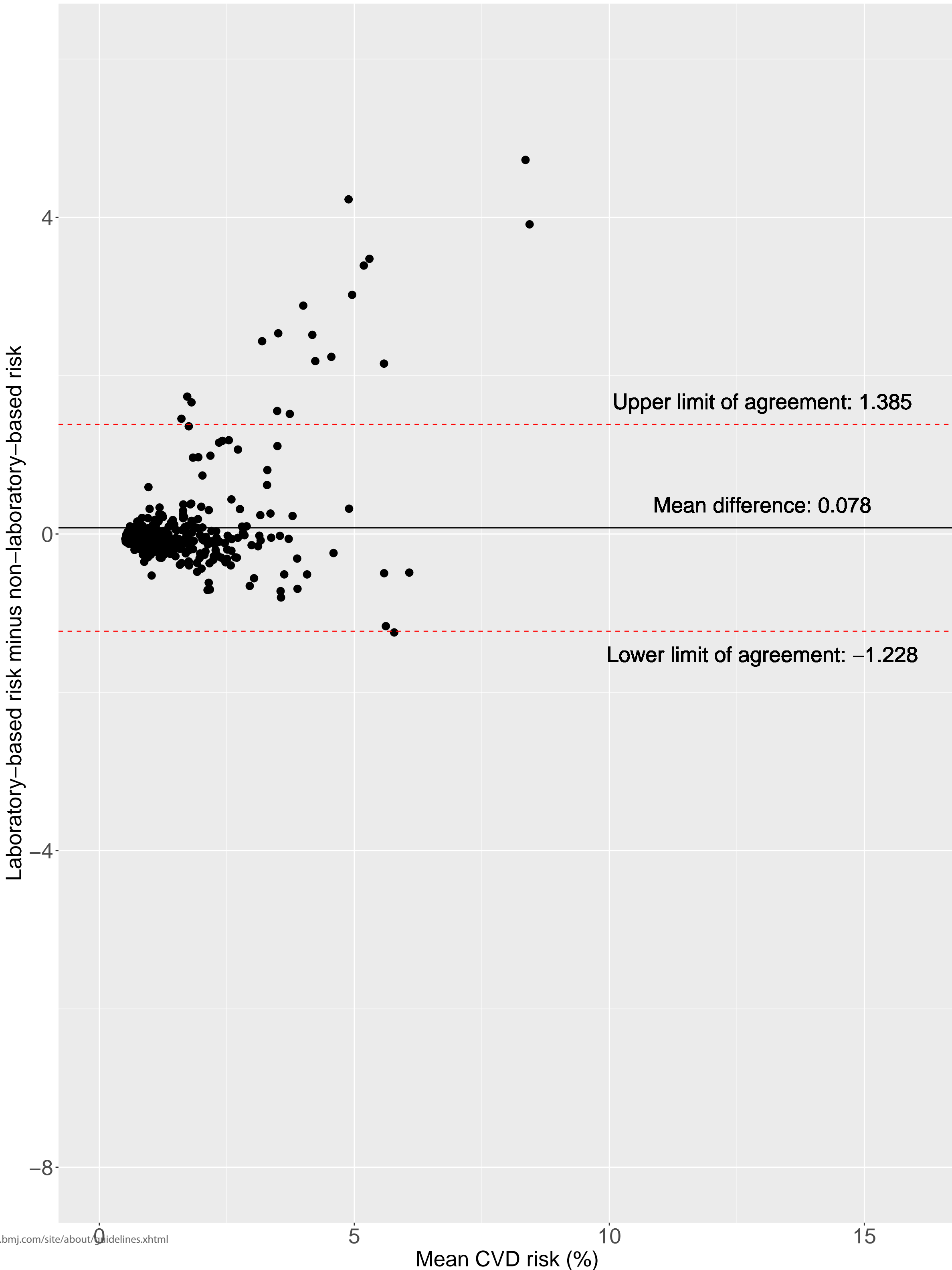
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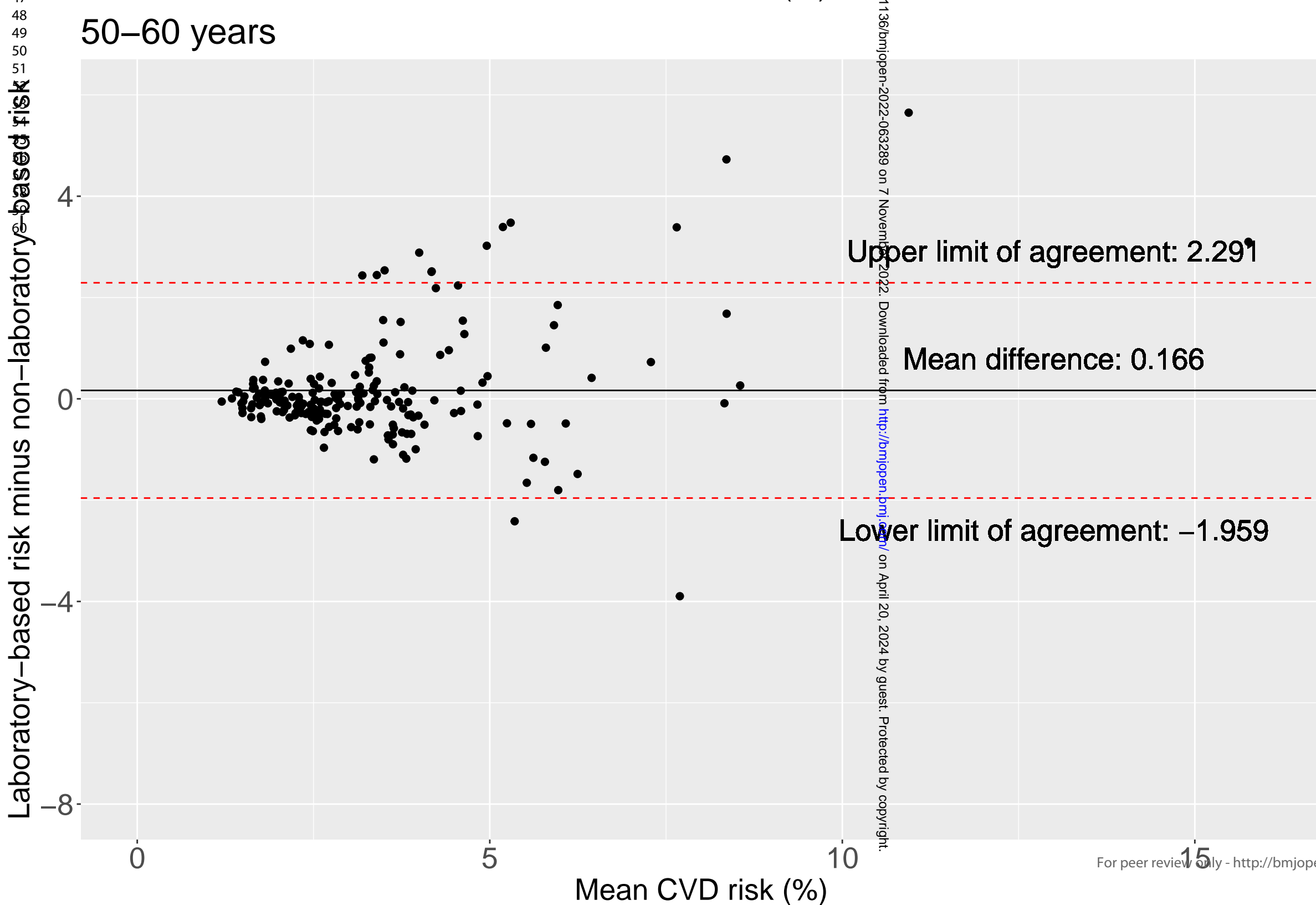
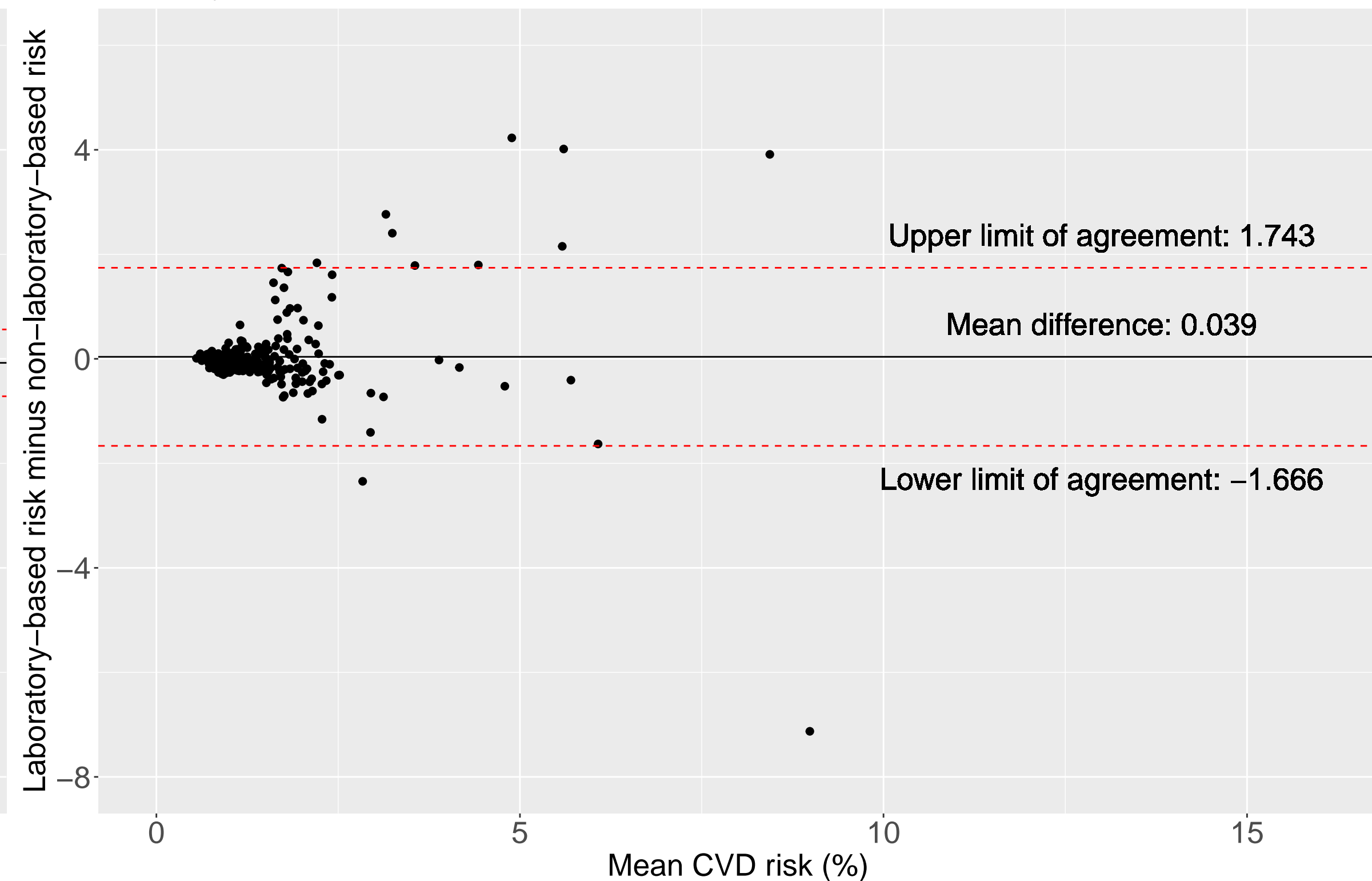
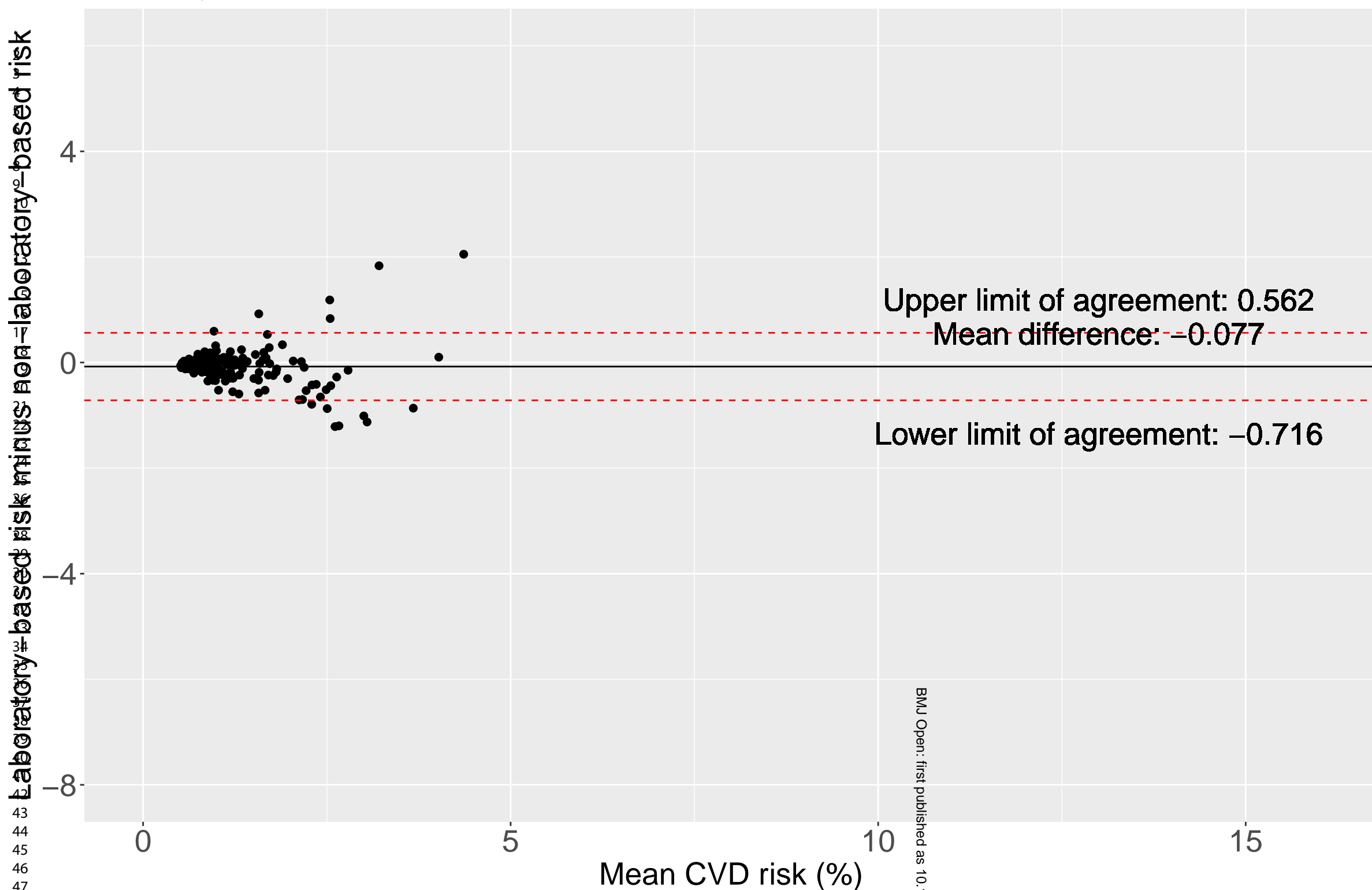
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peer review only

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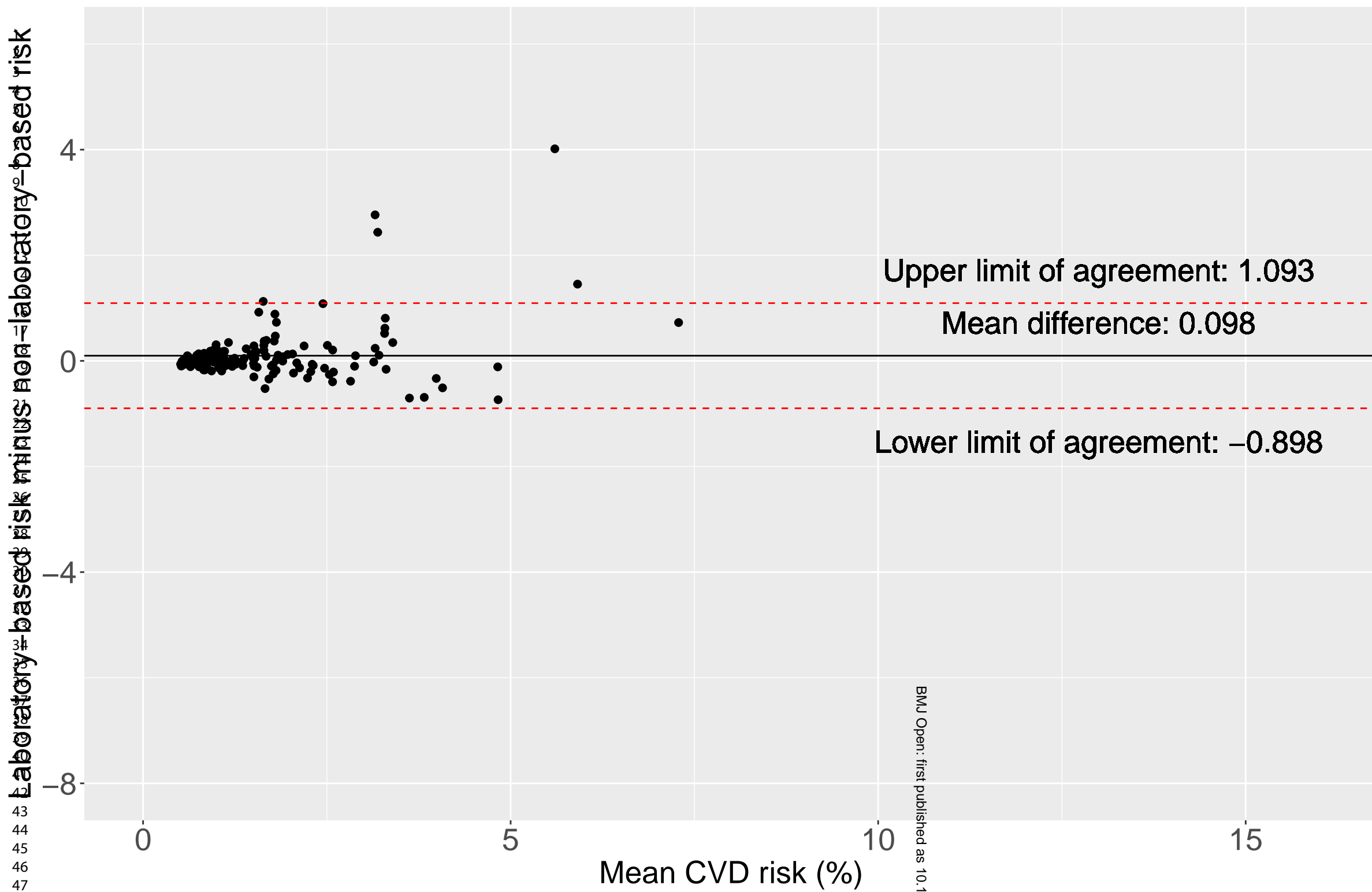
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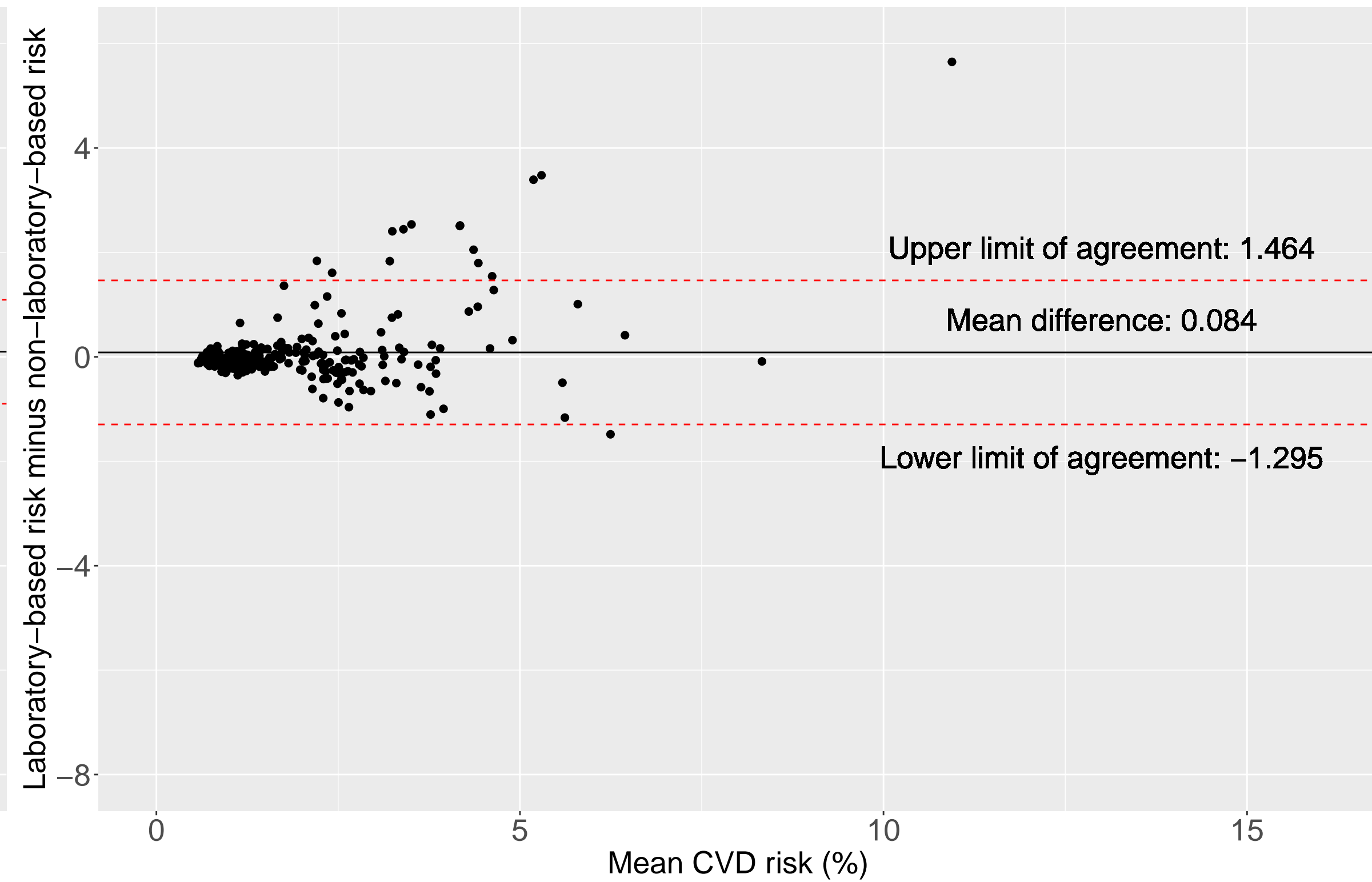


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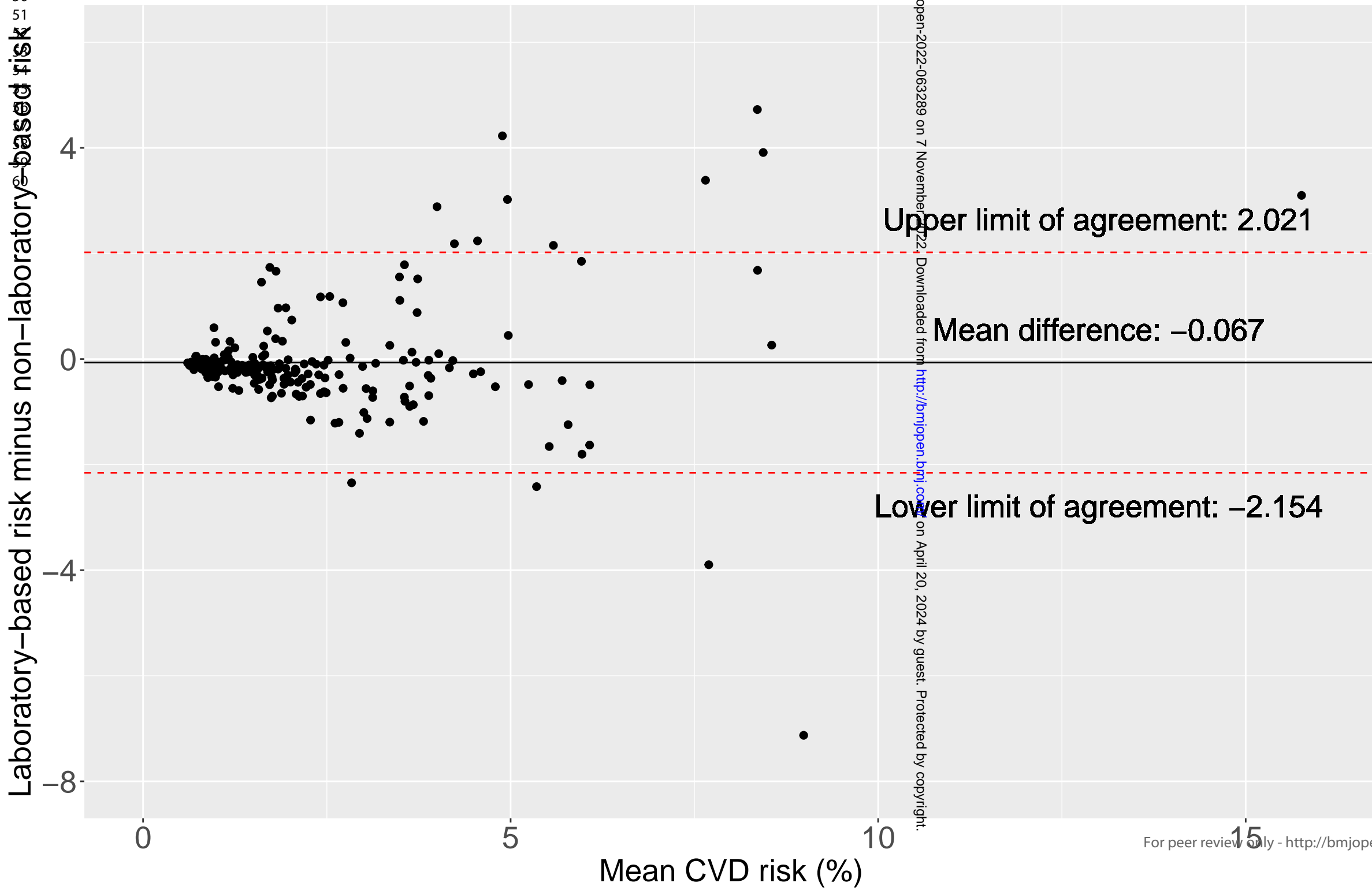
Normal BMI (<25kg/m²)



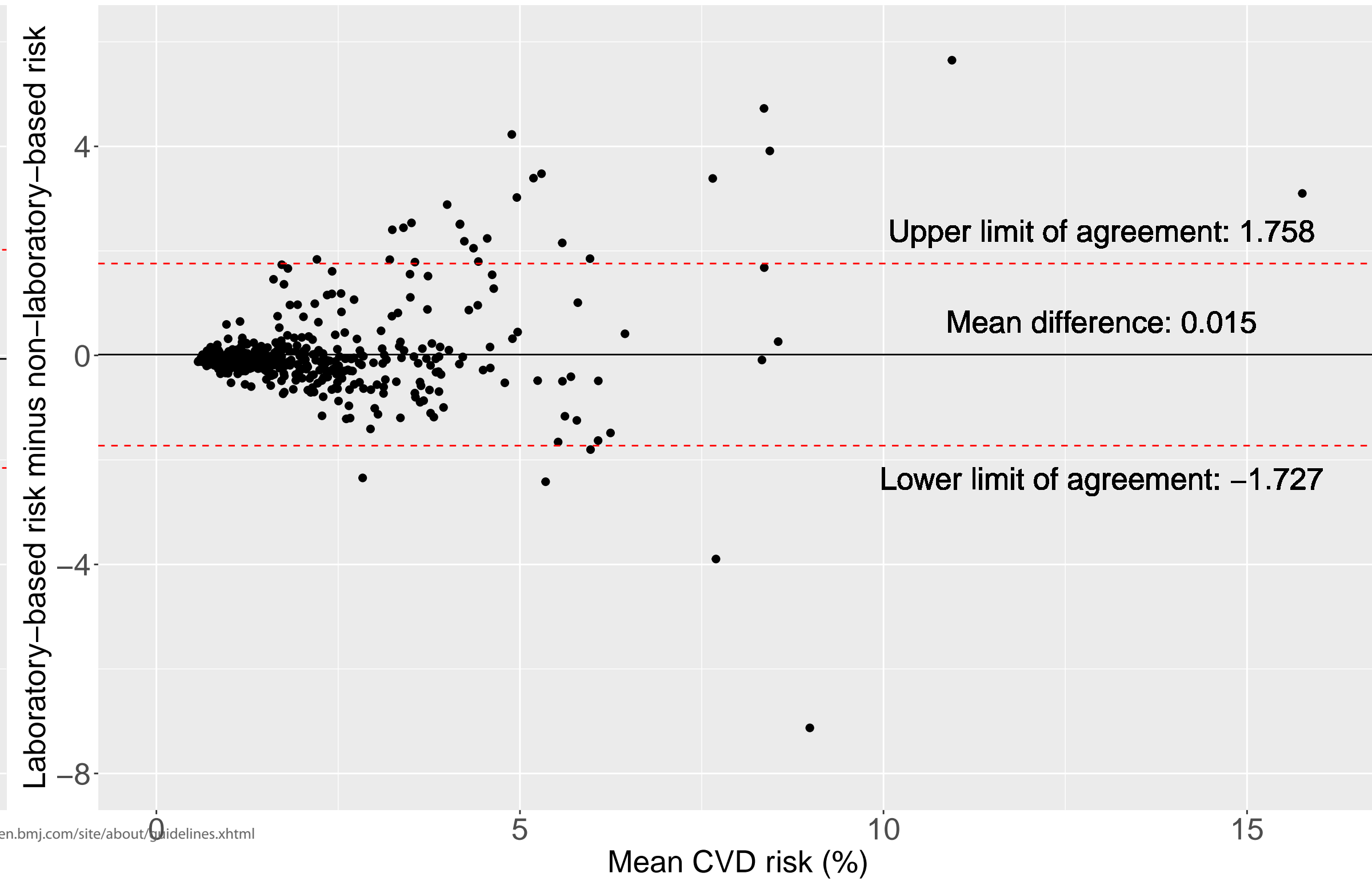
Overweight (25–30 kg/m²)



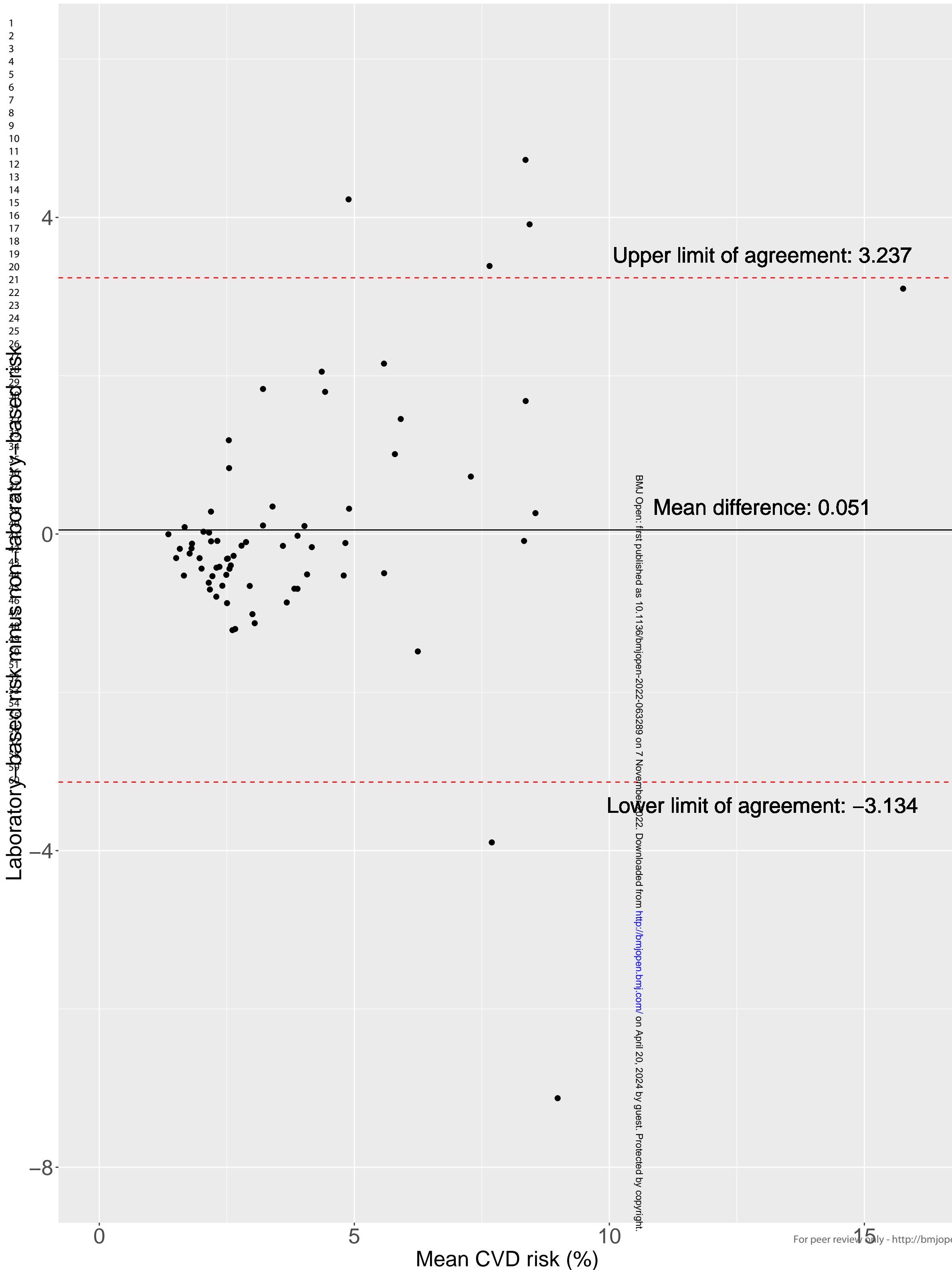
Obese (>30kg/m²)



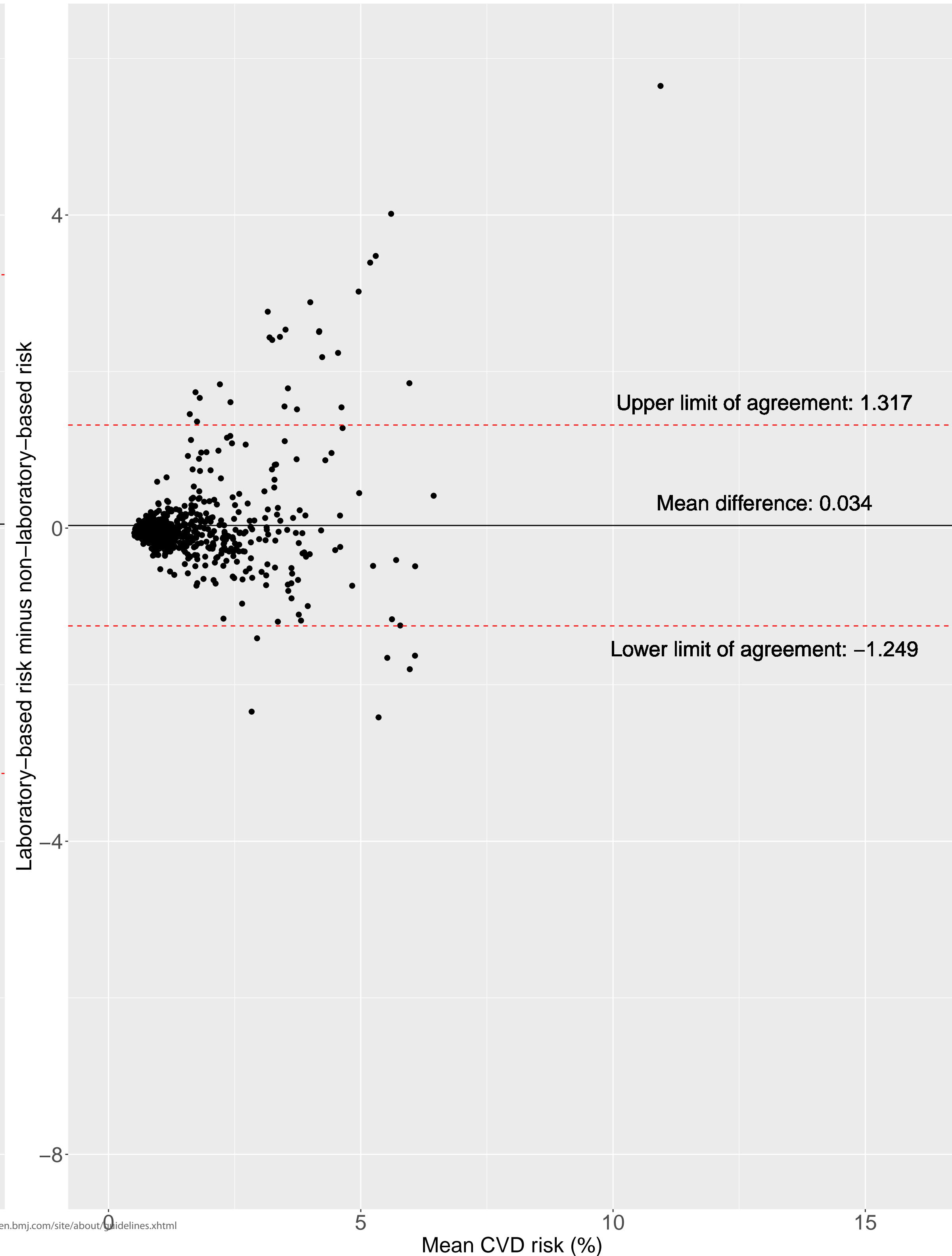
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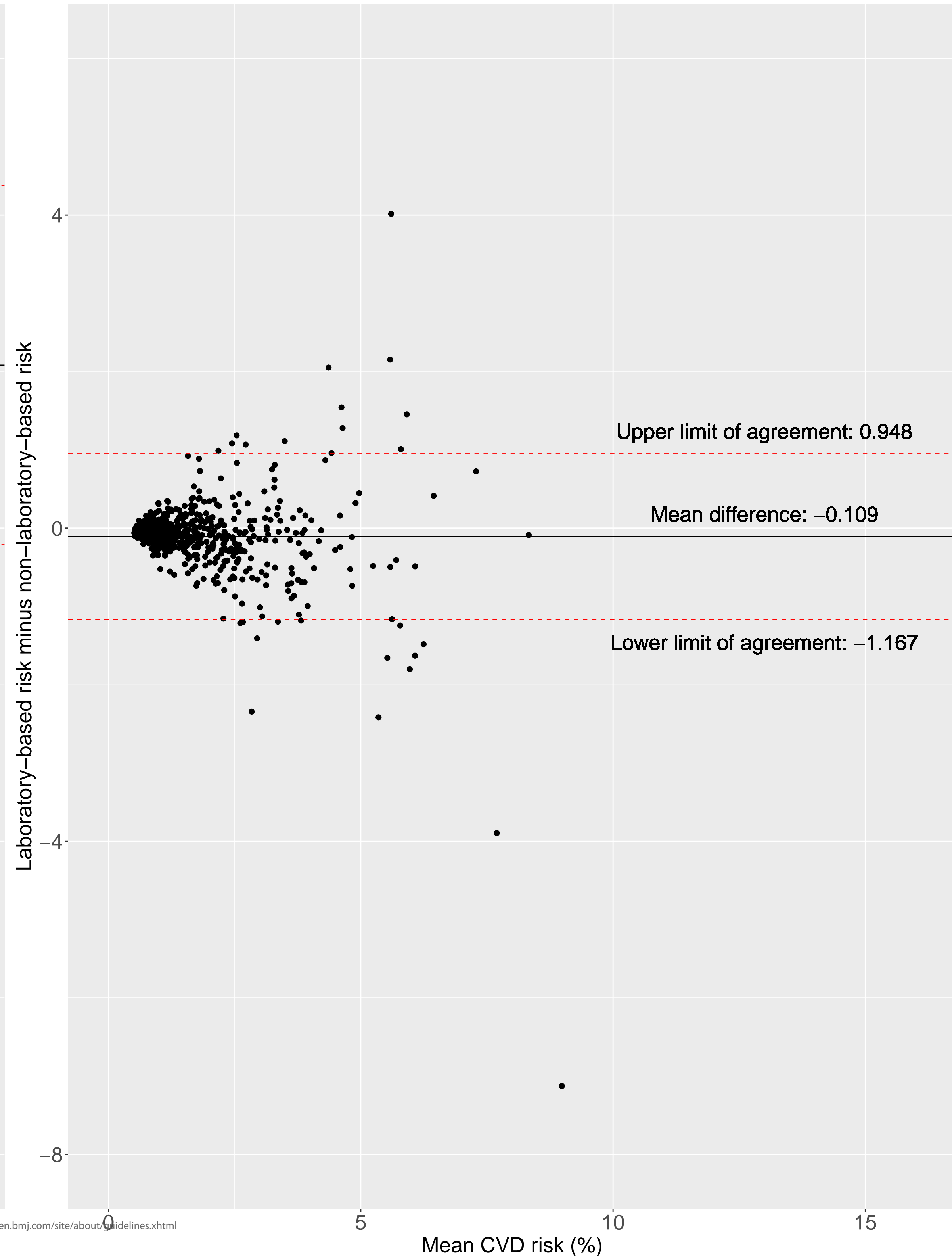
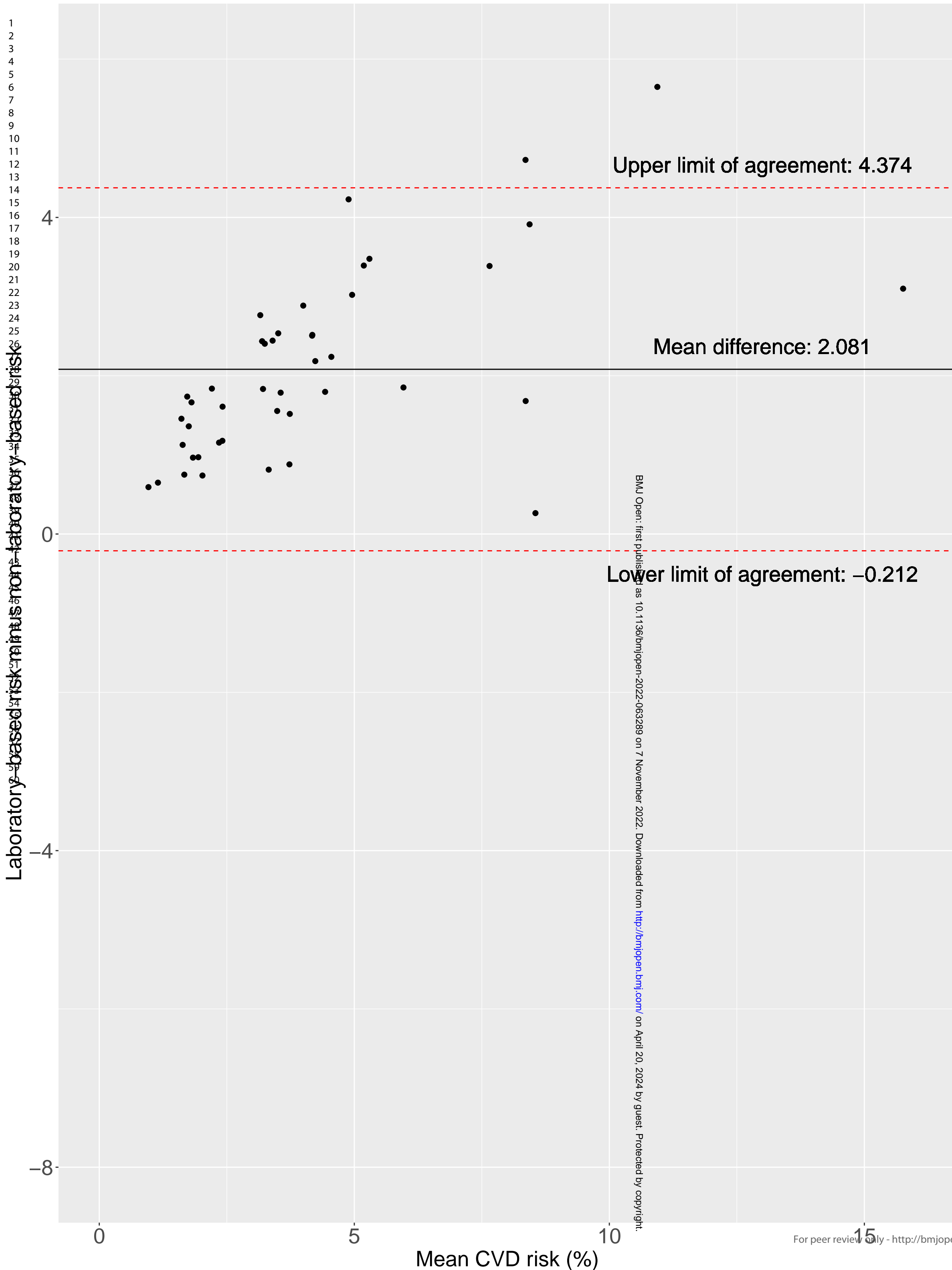


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Laboratory-based risk minus non-laboratory-based risk

Laboratory-based risk minus non-laboratory-based risk

STROBE Statement—Checklist of items that should be included in reports of *cross-sectional studies*

	Item No	Recommendation	Page No
Title and abstract	1	(a) Indicate the study's design with a commonly used term in the title or the abstract	3
		(b) Provide in the abstract an informative and balanced summary of what was done and what was found	3
Introduction			
Background/rationale	2	Explain the scientific background and rationale for the investigation being reported	4
Objectives	3	State specific objectives, including any prespecified hypotheses	5
Methods			
Study design	4	Present key elements of study design early in the paper	5
Setting	5	Describe the setting, locations, and relevant dates, including periods of recruitment, exposure, follow-up, and data collection	5-6
Participants	6	(a) Give the eligibility criteria, and the sources and methods of selection of participants	5-6
Variables	7	Clearly define all outcomes, exposures, predictors, potential confounders, and effect modifiers. Give diagnostic criteria, if applicable	6-7
Data sources/ measurement	8*	For each variable of interest, give sources of data and details of methods of assessment (measurement). Describe comparability of assessment methods if there is more than one group	6-7
Bias	9	Describe any efforts to address potential sources of bias	6
Study size	10	Explain how the study size was arrived at	5-6
Quantitative variables	11	Explain how quantitative variables were handled in the analyses. If applicable, describe which groupings were chosen and why	6-7
Statistical methods	12	(a) Describe all statistical methods, including those used to control for confounding	7-8
		(b) Describe any methods used to examine subgroups and interactions	7-8
		(c) Explain how missing data were addressed	5-6
		(d) If applicable, describe analytical methods taking account of sampling strategy	8
		(e) Describe any sensitivity analyses	6
Results			
Participants	13*	(a) Report numbers of individuals at each stage of study—eg numbers potentially eligible, examined for eligibility, confirmed eligible, included in the study, completing follow-up, and analysed	8
		(b) Give reasons for non-participation at each stage	Supp Fig 1
		(c) Consider use of a flow diagram	5
Descriptive data	14*	(a) Give characteristics of study participants (eg demographic, clinical, social) and information on exposures and potential confounders	8
		(b) Indicate number of participants with missing data for each variable of interest	Supp Fig 1
Outcome data	15*	Report numbers of outcome events or summary measures	

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2	Main results	16	(a) Give unadjusted estimates and, if applicable, confounder-adjusted estimates and their precision (eg, 95% confidence interval). Make clear which confounders were adjusted for and why they were included
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6			(b) Report category boundaries when continuous variables were categorized
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9			(c) If relevant, consider translating estimates of relative risk into absolute risk for a meaningful time period
10			-
11	Other analyses	17	Report other analyses done—eg analyses of subgroups and interactions, and sensitivity analyses
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14	Discussion		
15	Key results	18	Summarise key results with reference to study objectives
16			11
17	Limitations	19	Discuss limitations of the study, taking into account sources of potential bias or imprecision. Discuss both direction and magnitude of any potential bias
18			13
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20	Interpretation	20	Give a cautious overall interpretation of results considering objectives, limitations, multiplicity of analyses, results from similar studies, and other relevant evidence
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24	Generalisability	21	Discuss the generalisability (external validity) of the study results
25			11-12
26	Other information		
27	Funding	22	Give the source of funding and the role of the funders for the present study and, if applicable, for the original study on which the present article is based
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*Give information separately for exposed and unexposed groups.

Note: An Explanation and Elaboration article discusses each checklist item and gives methodological background and published examples of transparent reporting. The STROBE checklist is best used in conjunction with this article (freely available on the Web sites of PLoS Medicine at <http://www.plosmedicine.org/>, Annals of Internal Medicine at <http://www.annals.org/>, and Epidemiology at <http://www.epidem.com/>). Information on the STROBE Initiative is available at www.strobe-statement.org.

BMJ Open

Agreement between the laboratory- and non-laboratory-based 2019 WHO cardiovascular risk charts in Peru

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3 **Agreement between the laboratory- and non-laboratory-based 2019 WHO cardiovascular**
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ABSTRACT

Objective: To determine the agreement between the cardiovascular disease (CVD) risk predictions computed with the World Health Organization (WHO) non-laboratory-based model and laboratory-based model in a nationally-representative sample of Peruvian adults.

Design: Cross-sectional analysis of a national health survey.

Methods: Absolute CVD risk was computed with the 2019 WHO laboratory- and non-laboratory-based models. The risk predictions from both models were compared with Bland Altman plots, Lin's concordance coefficient correlation (LCCC) and kappa statistics, stratified by sex, age, body mass index categories, smoking and diabetes status.

Results: 663 people aged 30-59 years were included in the analysis. Overall, there were no substantial differences between the mean CVD risk computed with the laboratory-based model 2.0% (95% CI: 1.8%; 2.2%) and the non-laboratory-based model 2.0% (95% CI: 1.8%; 2.1%). In the Bland Altman plots, the limits of agreement were the widest among people with diabetes (-0.21; 4.37) compared with people without diabetes (-1.17; 0.95). The lowest agreement as per the LCCC was also seen in people with diabetes (0.74 (95% CI: 0.63; 0.82)), the same was observed with the kappa statistic (kappa=0.36). In general, agreement between the scores was appropriate in terms of clinical significance.

Conclusions: The absolute cardiovascular predicted risk was similar between the laboratory-based and non-laboratory-based 2019 WHO cardiovascular risk models. Pending validation from longitudinal studies, the non-laboratory-based model (instead of the laboratory-based) could be used when assessing CVD risk in Peruvian population.

Key words: cardiovascular diseases; risk assessment; health metrics

STRENGTHS AND LIMITATIONS OF THIS STUDY

- This analysis provided the first evidence of the agreement between the 2019 WHO CVD risk laboratory-based and non-laboratory-based models.
- We leveraged on the most recent nationally representative survey that included blood biomarkers in Peru.
- Our study population was young, mostly women, and with overall low absolute cardiovascular predicted risk. No one in the study population had an absolute cardiovascular risk $\geq 20\%$.
- We assumed all participants were free of CVD to use the 2019 WHO CVD risk score, as information regarding history of cardiovascular events was not reported.

INTRODUCTION

Cardiovascular diseases (CVD) are the main cause of death globally.[1] In 2019, CVD caused ~18.5 million deaths in adults, representing 36% of all global deaths.[2,3] Furthermore, CVD impose a huge burden in low- and middle-income countries (LMICs),[4] where deaths from CVD occur at younger ages compared to high-income countries (HICs).[1] However, CVD can be prevented and managed through a combination of population- and individual-level interventions;[5] for the latter, the identification of individuals at high cardiovascular risk is a cornerstone in the prevention of CVD. In this line, the World Health Organization (WHO) and the Pan American Health Organization (PAHO),[6] alongside several clinical guidelines,[7–9] recommend CVD risk stratification with CVD risk prediction models to inform evidence-based treatment.

CVD risk prediction models identify people who would benefit the most of preventive interventions (e.g., statin therapy).[10] Although there are several CVD risk prediction models,[11] these were mostly developed in HICs limiting their application in LMICs where they would need recalibration to deliver accurate predictions to guide treatment allocation. To overcome this limitation, the WHO convened a global effort to derive, calibrate and validate new CVD risk prediction models for all world regions.[12] Two WHO CVD risk prediction models were developed: a laboratory-based

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3 model and a non-laboratory-based model. Because laboratory biomarkers (e.g., total cholesterol)
4 may not be available in primary health care centres in LMICs limiting the use of laboratory-based
5 CVD risk prediction models,[13] the non-laboratory-based model arises as a handy tool for
6 clinicians in LMICs. Similarly, CVD risk prediction models are used to monitor the prevalence of
7 high CVD risk and treatment coverage (i.e., people at high CVD receiving treatment).[14] In this
8 context, countries conducting national or large population-based health surveys without lipid
9 biomarkers, could benefit from the non-laboratory-based models. Peru, for example, does not
10 have regular national health surveys including total cholesterol, but Peru has a yearly national
11 health survey including anthropometrics, blood pressure, and health questionnaires. In Peru, and
12 other similar LMICs, it would not be possible to monitor the burden of high CVD risk with a
13 laboratory-based model and the non-laboratory-based model rises as the only alternative.
14 Nonetheless, evidence regarding the agreement between the WHO laboratory-based and non-
15 laboratory-based models in countries from Latin America is missing.[12] Whether the WHO non-
16 laboratory-based model delivers predictions similar to those of the WHO laboratory-based model
17 remains unknown in Latin America. However, clinicians need this evidence to inform their choice
18 (non-laboratory-based vs laboratory-based model), and to interpret the results of the non-
19 laboratory-based model under the assumption that the laboratory-based model is the gold
20 standard.[12] To provide this evidence for practitioners in Peru, we determined the agreement
21 between the risk predictions computed with the non-laboratory-based model and laboratory-
22 based model in a nationally-representative sample of Peruvian adults.

43 **METHODS**

45 **Data sources**

47 This is a cross-sectional study of a national survey conducted by the National Centre for Food
48 and Nutrition (CENAN, for its acronym in Spanish) of Peru. CENAN's survey was conducted
49 between 2017-2018 on a nationally-representative sample of Peruvian adults aged between 18-
50 59 years.[15] Of note, this is the most recent nationally-representative survey conducted in Peru
51 that included blood biomarkers (e.g., lipid profile). CENAN's survey adhered to ethical guidelines
52 and followed a standardised protocol that has been published elsewhere.[15] Each participant
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3 was informed about all procedures and techniques used in the survey; also, participants could
4 have left the study at any time and their personal information was kept confidential.[15]
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9 The CENAN's survey sample was computed using the formula shown in Supplementary Figure 1
10 and followed a probabilistic sampling design approach with two stages.[15] First, clusters were
11 randomly selected considering three strata: 1) Urban areas except Lima city, 2) Rural areas, and
12 3) Lima city. Then, households (of adults aged 18-59 years living in) were randomly selected
13 within each cluster. To be selected for the survey sample, participants had to fulfil the following
14 inclusion criteria: 1) Adults aged 18-59 years, and 2) Fasting 9-12 hours for blood biomarkers.
15 The following participants were excluded: 1) Pregnant and postpartum women, 2) Adults taking
16 medication that could alter glucose and lipid profiles, 3) Adults with congenital diseases that could
17 limit anthropometrics measurement (e.g., Down syndrome).
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28 **Study population**

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30 We analysed a complete-case sample regarding all the laboratory- and office-based 2019 ten-
31 year WHO CVD risk score variables (see Variables section). We studied men and women aged
32 between 30-59 years. The younger age limit was decided because CVD risk models are not
33 recommended in younger individuals; the older age limit was decided because of the survey
34 design. A flowchart of data cleaning is shown in Supplementary figure 2. We did not apply other
35 selection criteria.
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43 Although the 2019 ten-year WHO CVD risk models were developed for people aged 40-80
44 years,[12] for people aged <40 years we assumed they had 40 years for the absolute CVD risk
45 computation. This is consistent with the Package of essential noncommunicable (PEN) disease
46 interventions for primary health care in low-resource settings,[16] and was also done in a recent
47 global work.[17]
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54 **Variables**

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56 We calculated the ten-year CVD risk at the individual-level following the laboratory- and non-
57 laboratory-based 2019 ten-year WHO CVD risk model.[12] We used the *whocvdrisk* command in
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3 STATA, which was developed by the authors of the 2019 ten-year WHO CVD risk charts.[18] For
4 the laboratory-based model, scores were calculated based on: age (years), current smoking
5 status (yes/no), systolic blood pressure (SBP, mmHg), history of self-reported diabetes diagnosis
6 (yes/no), and total cholesterol (mmol/L).[12] For the non-laboratory-based model, we used age
7 (years), current smoking status (yes/no), systolic BP (mmHg) and body mass index (BMI,
8 kg/m²).[12]
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16 The CENAN's survey collected anthropometrics and three BP measurements which were taken
17 by trained fieldworkers following a standard protocol.[15] We computed BMI using measured
18 weight (kg) divided by the square of height (meters); for descriptive purposes, we classified BMI
19 in three levels: normal weight (BMI <25 kg/m²), overweight (BMI ≥25-29.9 kg/m²), and obesity
20 (BMI ≥30 kg/m²). BMI records outside the range 10-80 kg/m² were discarded. As the third BP
21 measurement was only available in few participants (<2% of the initial sample), we used the
22 second SBP measurement only (i.e., the first and third SBP records were discarded in the main
23 analysis). Of note, there were no substantial differences between the first and second SBP
24 records (Supplementary table 1); nonetheless, we performed a sensitivity analysis using the
25 mean SBP of first and second SBP records. We discarded any SBP records outside the range
26 70-270 mmHg.
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40 For people who self-reported being under antihypertensive treatment, we used the pre-treatment
41 SBP; this is consistent with the PEN protocol and with a previous global work.[16,17]. Pre-
42 treatment SBP was computed as: pre-treatment systolic blood pressure = (current systolic blood
43 pressure-6.3)/0.9.[19] Conversely, for those not taking antihypertensive treatment, we used the
44 recorded SBP as was.
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51 We defined current smoker with one question coded as no versus yes: *do you currently smoke*
52 *any tobacco product such as cigarettes, cigars or pipes?* Self-reported information about a prior
53 history of diabetes was assessed by a question also coded as no versus yes: *have you ever been*
54 *told by a physician or another healthcare worker that you have high blood sugar or diabetes?*
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3 Total cholesterol was obtained via enzymatic colorimetric method.[15] Because CENAN's survey
4 data had total cholesterol in mg/dl, these values were divided by 38.67 to obtain total cholesterol
5 in mmol/L.
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10 **Statistical analysis**

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12 We determined the agreement between the absolute CVD risk predicted with the WHO laboratory-
13 and non-laboratory-based models following three methods: Bland Altman plots, Lin's
14 concordance coefficient correlation (LCCC), and kappa statistic. We considered the absolute
15 CVD risk as a continuous variable, and the agreement between both models was examined using
16 Bland Altman plots and the LCCC. Furthermore, we considered the CVD risk as a categorical
17 variable, and divided into three groups: <5%, 5-9%, 10-19%; because there were no observations
18 in the high-risk category (CVD risk $\geq 20\%$), agreement was not examined in this group. For these
19 categories, we evaluated the agreement using the kappa statistic. For the Bland Altman plots,
20 LCCC, and the kappa statistic, results were stratified by sex, 10-year age groups, BMI categories,
21 smoking status, self-reported diabetes diagnosis and urban/rural location.
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34 In the Bland Altman plots, the risk difference between the laboratory- and non-laboratory-based
35 absolute cardiovascular predicted risk was plotted on the vertical axis, and the mean of both
36 scores on the horizontal axis.[20] As the true risk of CVDs at the individual-level is uncertain, the
37 mean of both scores is the best available estimate.[20,21] The 95% of the limit of agreement was
38 represented by the mean difference of both scores \pm two standard deviations; this limit provides
39 an interval in which 95% of the differences between both scores would be expected to lie.[20] The
40 LCCC between the laboratory- and non-laboratory-based absolute cardiovascular predicted risks
41 was also evaluated. The agreement based on the LCCC ranges between -1 and 1, with 1
42 suggesting a perfect agreement. The categorical agreement was evaluated using the Kappa
43 statistic. Kappa <0 indicated less than chance agreement, and values 0.01–0.20, 0.21–0.40,
44 0.41–0.60, 0.61–0.80, and 0.81–0.99 represented slight, fair, moderate, substantial, and almost
45 perfect agreement, respectively.[22]
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All analyses code were conducted with R (version 4.0.3) and STATA (version 17.0, College Station, Texas 77845, USA). Population characteristics along with their 95% confidence interval (95% CI) were summarized accounting for the complex survey design of the CENAN's survey.[15]

Patient and public involvement

No patient involved.

RESULTS

Our pooled dataset included 663 participants (Supplementary figure 2). The mean age was 44.0 years (95% CI: 43.2; 44.7) and the proportion of men was 41.5%. The mean SBP was 108.9 mmHg (95% CI: 107.5; 110.3 mmHg) and the mean BMI was 28.8 kg/m² (95% CI: 28.3; 29.2 kg/m²). The proportion of people with overweight was 41.0% (95% CI: 36.6; 45.6 %), whereas 35.9% (95% CI: (31.6; 40.5%)) of the population were obese. The mean total cholesterol was 4.9 mmol/L (95% CI: 4.8; 5.0 mmol/L), 11.7% (95% CI: 8.8; 15.3%) of the population were smokers and 7.1% (95% CI: 5.0%; 9.8%) had diabetes (Table 1).

Table 1. Weighted distribution of the predictors in the 2019 WHO CVD risk models, overall and by sex.

	Total	Men	Women
Sample size	663	280	383
Age (mean and 95% CI, years)	44 (43.2-44.7)	44.2 (43-45.4)	43.8 (42.9-44.7)
Proportion of people aged 30-39 years (95% CI, %)	35.8 (31.6-40.3)	37.2 (31-43.9)	34.8 (29.4-40.7)
Proportion of people aged 40-49 years (95% CI, %)	34 (29.8-38.4)	29.5 (24-35.7)	37.1 (31.5-43.1)
Proportion of people aged 50-59 years (95% CI, %)	30.2 (26.3-34.5)	33.3 (27.1-40)	28 (23.5-33.1)
Systolic blood pressure (mean and 95% CI, mmHg)	108.9 (107.5-110.3)	116.1 (113.8-118.4)	103.8 (102.3-105.3)
Diastolic blood pressure (mean and 95% CI, mmHg)	71.9 (71-72.8)	74.6 (73.1-76.1)	70 (69-71)
Body mass index (mean and 95% CI, kg/m²)	28.8 (28.3-29.2)	28.3 (27.6-29)	29.1 (28.5-29.7)
Proportion of people with normal weight (95% CI, %)	23.1 (19.5-27.2)	24.5 (19.5-30.4)	22.1 (17.2-27.9)

Proportion of people with overweight (95% CI, %)	41 (36.6-45.6)	45.3 (38.9-52)	37.9 (31.9-44.3)
Proportion of people with obesity (95% CI, %)	35.9 (31.6-40.5)	30.1 (24.1-36.9)	40 (34-46.2)
Total cholesterol (mean and 95% CI, mmol/L)	4.9 (4.8-5)	4.8 (4.6-5)	4.9 (4.8-5.1)
Proportion of smokers (95% CI, %)	11.7 (8.8-15.3)	21.1 (15.8-27.6)	5 (3-8.3)
Proportion of people with diabetes (95% CI, %)	7.1 (5-9.8)	7.8 (4.7-12.6)	6.5 (4.3-9.9)
Laboratory-based CVD risk score (mean and 95% CI, %)	2 (1.8-2.2)	2.6 (2.3-2.9)	1.6 (1.4-1.7)
Non-laboratory-based CVD risk score (mean and 95% CI, %)	2 (1.8-2.1)	2.7 (2.4-3)	1.5 (1.4-1.6)

Absolute cardiovascular risk according to the 2019 WHO cardiovascular risk models

Overall, there were no substantial differences between the mean absolute cardiovascular risk computed with the laboratory-based and non-laboratory-based models (Table 1). The mean absolute cardiovascular risk was 2.0% (95% CI: 1.8%; 2.2%) according to the laboratory-based model, and 2.0% (95% CI: 1.8%; 2.1%) according to the non-laboratory-based model. In both models, the mean absolute cardiovascular risk was higher in men than women. The sensitivity analysis (using the mean of two SBP records) yielded the same findings in the overall sample: 2.0% (95% CI: 1.8; 2.2%) in the laboratory-based-model and 2.0 (95% CI: 1.8; 2.1%) in the non-laboratory-based model.

Mean difference between risk predictions

Overall, the mean difference between the laboratory-based and the non-laboratory-based models was 0.03% (95% CI: -0.03%; 0.10%). According to sex, the mean difference between models was -0.02% (95% CI: -0.14%; 0.09%) in men, and 0.08% (95% CI: 0.01%; 0.15%) in women. According to age, the mean difference between models was -0.07% (95% CI: -0.12%; 0.04%) in people aged 30-39 years, 0.04% (95% CI: -0.08%; 0.15%) in people aged 40-49 years, and 0.17% (95% CI: 0.02%; 0.32%) in people aged 50-59 years. Stratified by BMI categories, the mean difference was 0.10% (95% CI: 0.02%; 0.18%) in people with normal weight, 0.08% (95% CI: 0.00%; 0.17%) in people with overweight, and -0.07% (95% CI: -0.21%; 0.07%) in people with obesity. The mean difference was 0.05% (95% CI: -0.34%; 0.44%) in smokers and 0.03% (95% CI: -0.02%; 0.09%) in non-smokers. According to self-reported diabetes status, mean difference was 2.08% (95% CI: 1.73%; 2.44%) in people with self-reported diabetes, and -0.11% (95% CI: -0.15%; 0.07%) in people without self-reported diabetes. The sensitivity analysis provided similar results across all variables; for example, the largest mean difference was also observed in people with self-reported diabetes (2.09% (95% CI: 1.73; 2.45%).

Bland-Altman plots and limits of agreement.

The limit of agreement was slightly narrower for women (-1.23; 1.39) compared to men (-1.93; 1.89) (Figure 1). The limit of agreement widened with older ages and higher BMI levels; for

example, the limit of agreement was narrower in people aged 30-39 years (-0.72; 0.56) compared to those aged 50-59 years (-1.96; 2.29) (Figure 2), and for people with normal weight (-0.90; 1.09) compared to those with obesity (-2.15; 2.02) (Figure 3). According to smoking and self-reported diabetes status, the limit of agreement was wider in those who had the condition. For example, the limit of agreement in smokers (-3.13; 3.24) was wider compared to non-smokers (-1.25; 1.32) (Figure 4). Similarly, in people with self-reported diabetes (-0.21; 4.37) the limit of agreement was wider compared to people without self-reported diabetes (-1.17; 0.95) (Figure 5). Notably, the sensitivity analysis resulted in similar limits of agreement in all variables (Supplementary figures 3-7).

Agreement by Lin's concordance coefficient correlation

The overall agreement between scores as per the LCCC was 0.87 (95% CI: 0.85; 0.89), and it was virtually the same in men (0.87 (95% CI 0.84; 0.89)) and women (0.85 (95% CI 0.82; 0.87)). Across age groups, the highest agreement was seen in the 30-39 age group (0.87 (95% CI 0.84; 0.9)). Across BMI categories, it was the normal BMI category which had the highest agreement (0.90 (95% CI 0.87; 0.92)). Overall, the lowest agreement values were seen across smokers (0.81 (95% CI 0.72; 0.88)), those aged 40-49 years (0.74 (95% CI 0.67; 0.79)), and those with self-reported diabetes (0.74 (95% CI 0.63; 0.82)) (Table 2). Similar results were seen in the sensitivity analysis (Supplementary table 2).

Table 2. Lin's concordance coefficient correlation showing agreement between laboratory and non-laboratory-based risk models according to the predictors in the 2019 WHO CVD risk models and urban/rural location

Variables	Categories	Lin's concordance coefficient correlation (95% CI)
Sex	Men	0.87 (0.84 - 0.89)
	Women	0.85 (0.82 - 0.87)
Age (years)	30-39	0.87 (0.84 - 0.9)
	40-49	0.74 (0.67 - 0.79)
	50-59	0.83 (0.78 - 0.86)
Body mass index category	Normal	0.9 (0.87 - 0.92)

	Overweight	0.87 (0.85 - 0.9)
	Obese	0.86 (0.82 - 0.89)
Smoking status	Smoker	0.81 (0.72 - 0.88)
	Non-smoker	0.86 (0.84 - 0.88)
Diabetes status	With self-reported diabetes	0.74 (0.63 - 0.82)
	Not with self-reported diabetes	0.91 (0.9 - 0.92)
Urban or rural	Urban	0.88 (0.85 - 0.9)
	Rural	0.85 (0.82 - 0.88)

Categorical agreement

In the overall population, there was a slightly larger number of people categorized as having an absolute CVD risk of 5-9% and 10-19% with the laboratory-based-model compared with the non-laboratory-based model (Supplementary table 3). For example, the laboratory-based model categorized 37 people in the 5-9% CVD risk category, whereas the non-laboratory-based model categorized 26 people. Overall, the agreement between risk categories was substantial ($\kappa=0.62$), and it was better for men ($\kappa=0.70$) compared to women ($\kappa=0.44$) (Supplementary table 4). Of note, the lowest agreement between risk categories was observed among people with self-reported diabetes ($\kappa=0.36$): out of 14 people with self-reported diabetes in the 5-9% CVD risk category following the laboratory-based-model, 4 were placed in the same category following the non-laboratory-based model, as the rest were placed in the 0-5% CVD risk category. The categorical agreement according to all variables is presented in Supplementary tables 3-9.

DISCUSSION

Main findings

In this work, we evaluated the agreement between the CVD risk estimates predicted with the 2019 WHO ten-year laboratory-based and non-laboratory-based models in a nationally-representative sample of Peruvian adults. The mean absolute predicted CVD risk according to both models in the general population was virtually the same. In addition, we found that the limits of agreement between both models increased with a higher CVD risk; for instance, the limits of agreement were

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3 wider in smokers and people with self-reported diabetes. We observed good agreement between
4 the laboratory-based and non-laboratory-based models in terms of clinical significance.[21,23]
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6 These findings suggest that, in a population with a similar profile to that in this study, practitioners
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8 could use either the laboratory-based or non-laboratory-based models. Although the difference is
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10 very small, careful interpretation could be needed for people with cardiovascular risk factors:
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12 obesity, smokers and people with self-reported diabetes, amongst whom the difference was
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14 slightly larger than in their peers without these risk factors.
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18 **Public health implications**

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20 We provided insights about the applicability of the 2019 WHO non-laboratory-based model in the
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22 Peruvian population. This evidence is relevant in terms of clinical practice and public health in
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24 Peru and other similar countries (e.g., Andean Latin America), because it informs whether the
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26 predictions based on the laboratory-based and non-laboratory-based CVD risk models are
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28 equivalent. If so, either of these models could be used without substantial bias, hence supporting
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30 the use of the non-laboratory-based model when blood biomarkers are not available. The latter is
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32 of special importance in LMICs like Peru, where laboratory measurements are not always
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34 available in primary health centres, which represent >98% of all health care facilities in Peru.[24]
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38 According to the three statistical analyses we implemented (Bland-Altman plots, Lin's
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40 concordance coefficient correlations; and kappa statistics), our results suggest that the
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42 agreement between the laboratory-based and non-laboratory-based models was appropriate
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44 among Peruvians with low CVD risk and younger than 60 years. In other words, our results
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46 suggest that the laboratory-based and non-laboratory-based models provide similar predictions
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48 and may therefore be used interchangeably as needed, though the profile of our study population
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50 ought to be considered when extracting or implementing our findings into clinical practice and
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52 public health. Of note, the small number of participants in some variables of interest (e.g., only 44
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54 participants had self-reported diabetes) could have explained the broader limits of agreement in
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56 our results. Further studies should include a larger number of participants to further confirm
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58 whether limits of agreement are wider according to smoking and diabetes status.
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Research in context

The study most comparable to ours evaluated the agreement between the Framingham 10-year CVD risk laboratory and non-laboratory models on a population aged 40-75 years in southern Iran.[23] They found the mean CVD risk following the non-laboratory-based model (9.4%) was higher than the laboratory-based model (6.7%).[23] Additionally, their limits of agreement between both Framingham models in people <60 years old were wider compared to ours in both men (-1.9-1.9 by our estimates versus -2.5%-8.9% by Rezaei et al.[23]) and women (-1.2-1.4 by our estimates versus -2.3%-4.6% by Rezaei et al.[23]). This could be explained by the fact that Rezaei et al.[23] included an older population, which tend to have higher levels of CVD risk factors and therefore a higher absolute CVD risk. As limits of agreement between two models tend to widen with higher CVD risk,[21,23] our limits of agreement would presumably be wider if we had studied a similar population to that of the work by Rezaei et al.[23] The differences between our results could be further explained by the CVD risk score herein used. We used the 2019 WHO CVD risk models,[12] whereas Rezaei et al. used the Framingham risk scores. The Framingham risk score was developed for a more specific population (Caucasians in the US),[25] yet the 2019 WHO CVD risk model was developed and recalibrated for a global use (e.g., those living in LMICs).[12]

The agreement between the 2019 WHO laboratory- and non-laboratory-based model was also explored in the global work convened by the WHO.[12] They applied the two models to WHO STEPS surveys, and compared the proportion of people categorized at different levels of predicted CVD risk.[12] Overall, they found moderate agreement between both models, and their discrepancy was attributed to poor performance of the non-laboratory-based model in people with diabetes.[12] This finding is consistent with our results because we found the widest limits of agreement, the lowest LCCC, and the lowest categorical agreement in people with self-reported diabetes. When possible, it would seem reasonable to use the laboratory-based model in those whose have diabetes.

Potential explanations

In our study, diabetes status was only included in the laboratory-based model (and not in the non-laboratory-based model) and reasonably, we observed the lowest agreement between both models in those with diabetes. That is, in people with diabetes, the non-laboratory-based model underestimated the absolute CVD risk computed following the laboratory-based model. Probably because people with diabetes are already at high CVD risk and the non-laboratory model, we used without diabetes information would underestimate the absolute risk.

Strengths and limitations

To the best of our knowledge, we provided the first evidence of the agreement between the 2019 WHO CVD risk laboratory-based and non-laboratory-based models; furthermore, we leveraged on a nationally-representative survey conducted in a LMIC.[26] Different CVD risk equations have been created but none has proved to produce reliable estimates for LMICs. The 2019 WHO CVD risk charts were adapted for LMICs using extensive datasets for its derivation, recalibration, and validation, which brings them several advantages over previous risk charts. Nonetheless, this study has also limitations. First, our study population was young (30-59 years), mostly women (58%), and with overall low absolute cardiovascular predicted risk. This led to the observation that no one in the study population had an absolute cardiovascular risk $\geq 20\%$. Thus, we could only draw conclusions for people within the low and medium CVD risk range, and with a similar demographic and risk factor profile. We acknowledge that further subgroup analysis could be relevant, for example by diabetes status. However, because of data availability and the reduced number of observations in some groups, this subgroup analyses would be impossible to conduct. Future work in Peru and Latin America should verify our results with a larger, older and more diverse population. Second, as CENAN's survey did not include history of cardiovascular events, we assumed all participants were free of CVD to use the 2019 WHO CVD risk score. This approach could have led to higher absolute cardiovascular risk because people who have had a cardiovascular event (e.g., myocardial infarction) are at higher risk of another cardiovascular event. Nonetheless, considering that our study population was young and therefore with a low incidence of cardiovascular diseases,[3] the proportion of people with history of CVD excluded from the total sample size would have been small; not excluding potentially a small group may

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3 not have altered the overall results. Third, we only used one blood pressure record (the second
4 SBP out of two measurements). Ideally, and following standard protocols recommended by WHO
5 and other international organizations,[27,28] we should have used the average of multiple records
6 having discarded the very first measurement. This was not possible with the available data
7 because they only measured blood pressure twice. Nonetheless, we performed a sensitivity
8 analysis using the mean SBP of the first and second records and had virtually the same results
9 as in our main analysis in which only the second SBP record was used.
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18 **Conclusions**

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20 The absolute cardiovascular predicted risk was similar between the laboratory-based and non-
21 laboratory based 2019 WHO cardiovascular risk models. Pending validation from longitudinal
22 studies, the non-laboratory-based model (instead of the laboratory-based which requires
23 additional resources) could be used in Peruvian population. Nonetheless, it should be noted that
24 the agreement between these models was less clear in people with cardiovascular risk factors:
25 obesity, smokers and people with diabetes. While universal health coverage momentum helps to
26 have laboratory tests in (most) primary care facilities to use the laboratory-based model, it seems
27 reasonable to use the non-laboratory-based model for primary prevention of CVD following the
28 risk stratification approach.
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FIGURES

Figure 1. Bland Altman plots showing agreement between laboratory and non-laboratory-based risk scores according to sex.

Figure 2. Bland Altman plots showing agreement between laboratory and non-laboratory-based risk scores according to age groups.

Figure 3. Bland Altman plots showing agreement between laboratory and non-laboratory-based risk scores according to body mass index categories

Figure 4. Bland Altman plots showing agreement between laboratory and non-laboratory-based risk scores according to smoking status.

Figure 5. Bland Altman plots showing agreement between laboratory and non-laboratory-based risk scores according to self-reported diabetes status.

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3 **Contributorship:** RMC-L conceived the idea with WCG-V. WCG-V conducted the analysis with
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7
8
9

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14 institutions to which they belong.
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20 **Competing of interests:** None to declare.
21

22 **Ethics:** This study used de-identified nationally-representative survey data that can be requested
23 from the CENAN and is used for independent analyses.[26] Authors had no access to the
24 participants' personal information. Therefore, this work was deemed as of minimal risk and we
25 did not seek approval by an Ethics Committee to conduct the analysis.
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30 **Data sharing:** Datasets available upon request from the CENAN.
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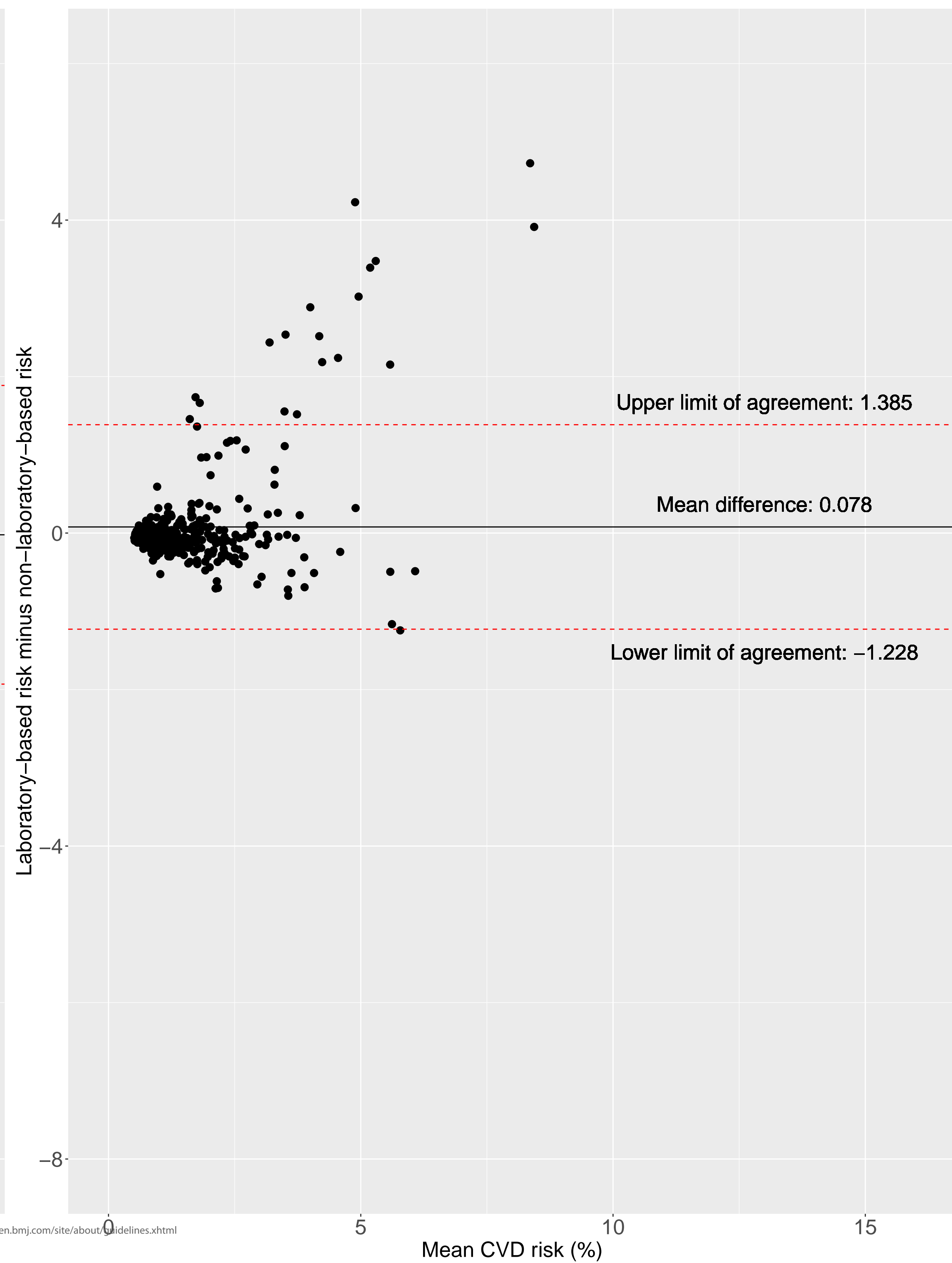
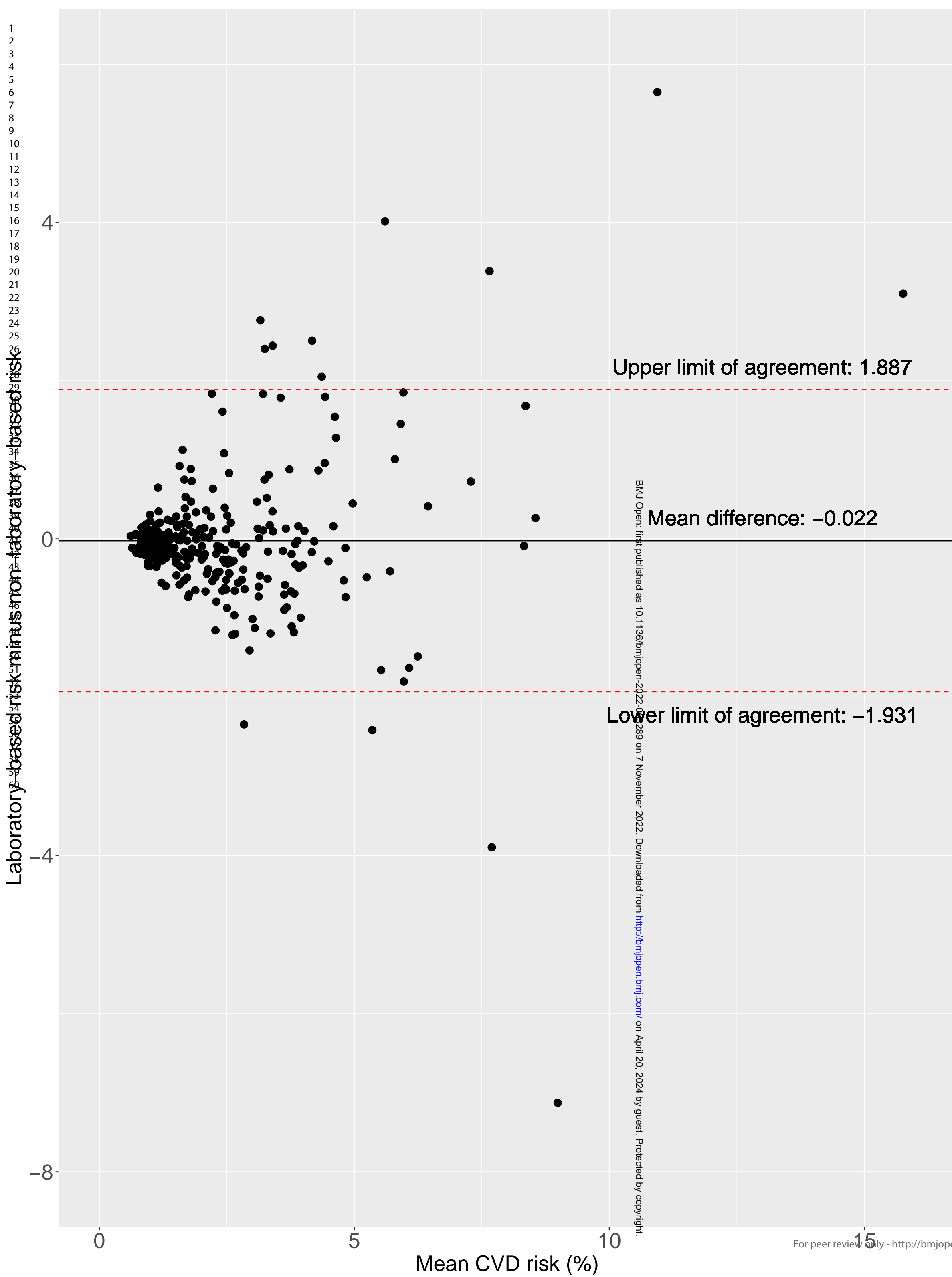
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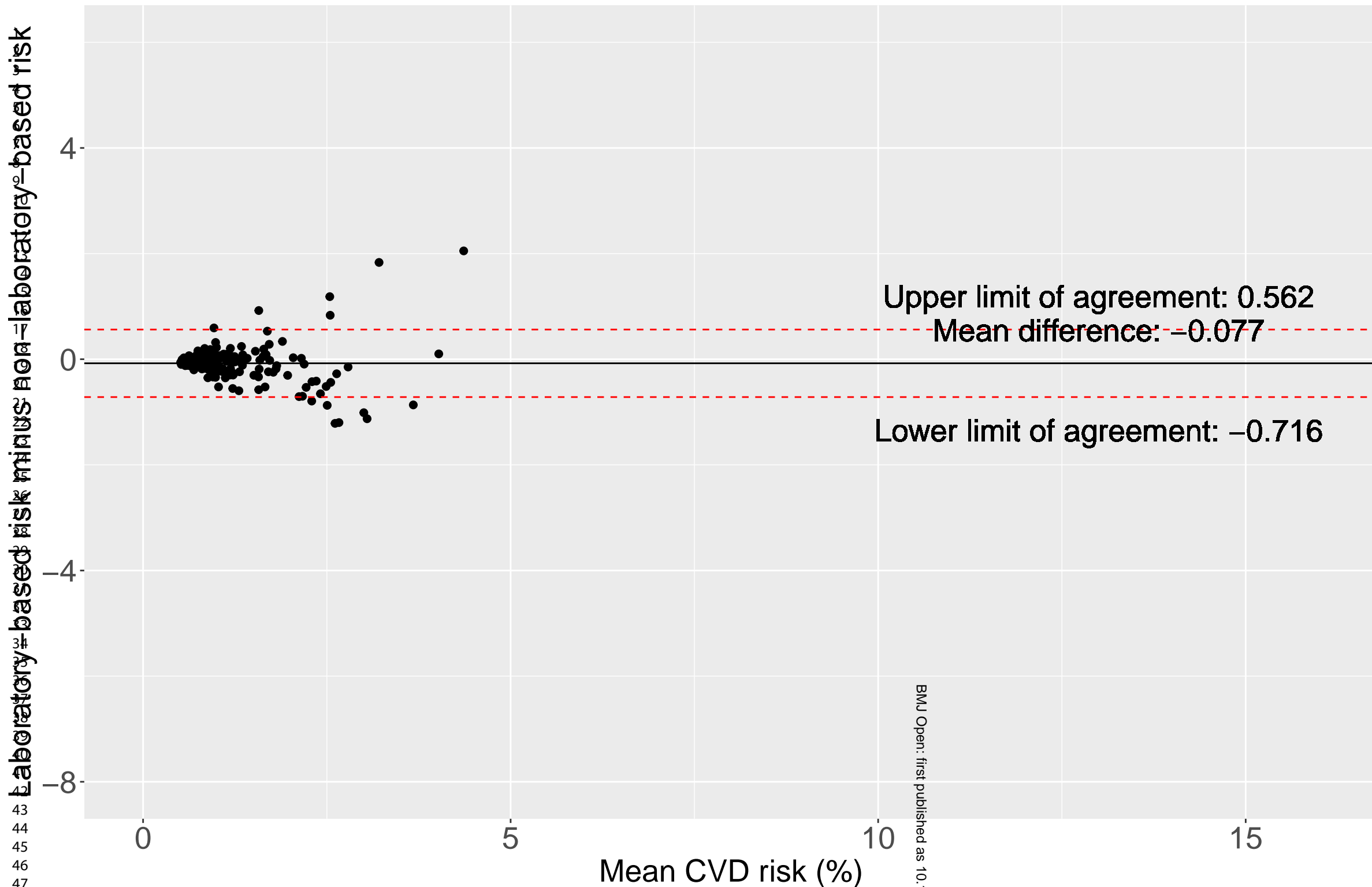
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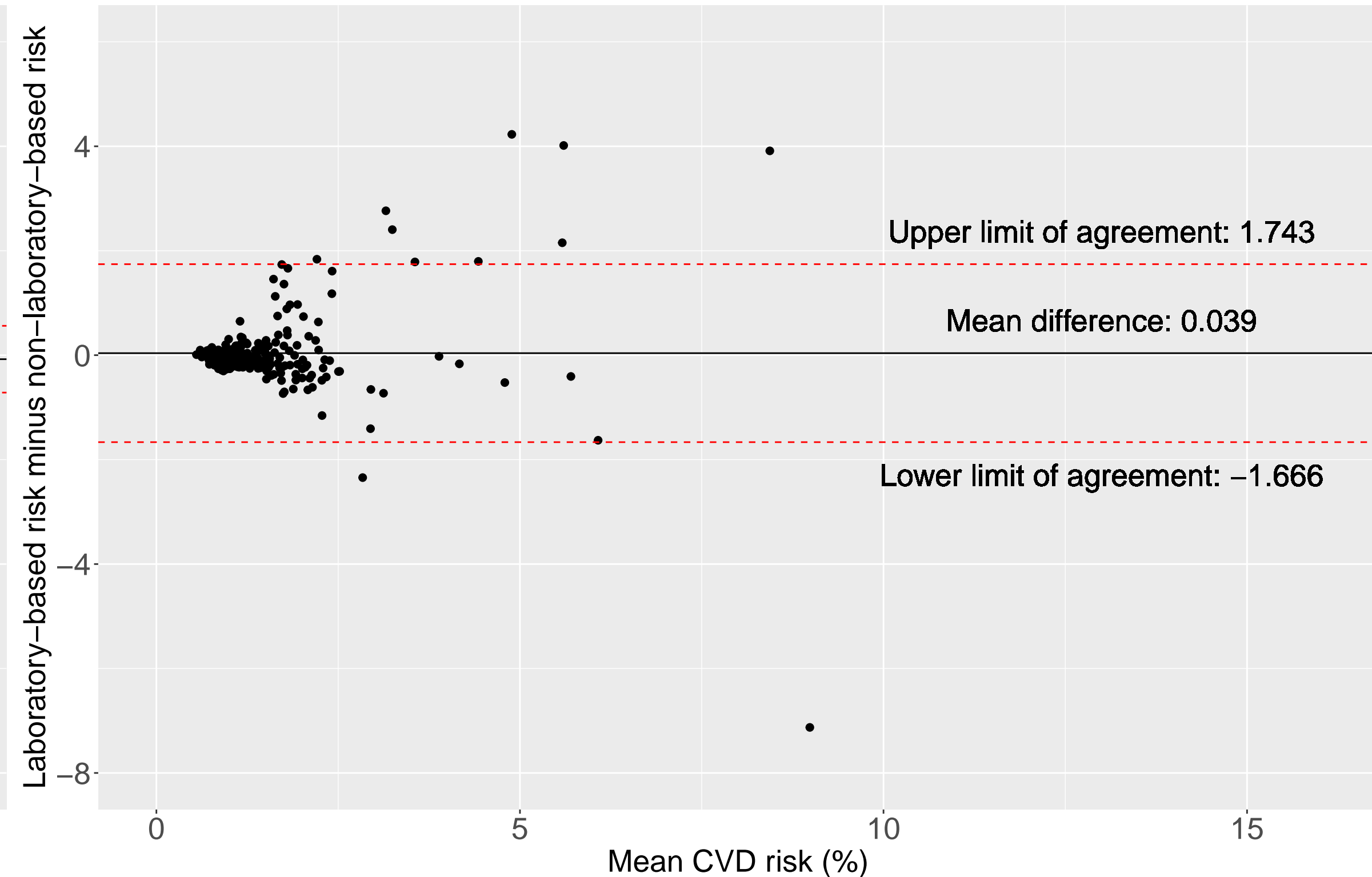
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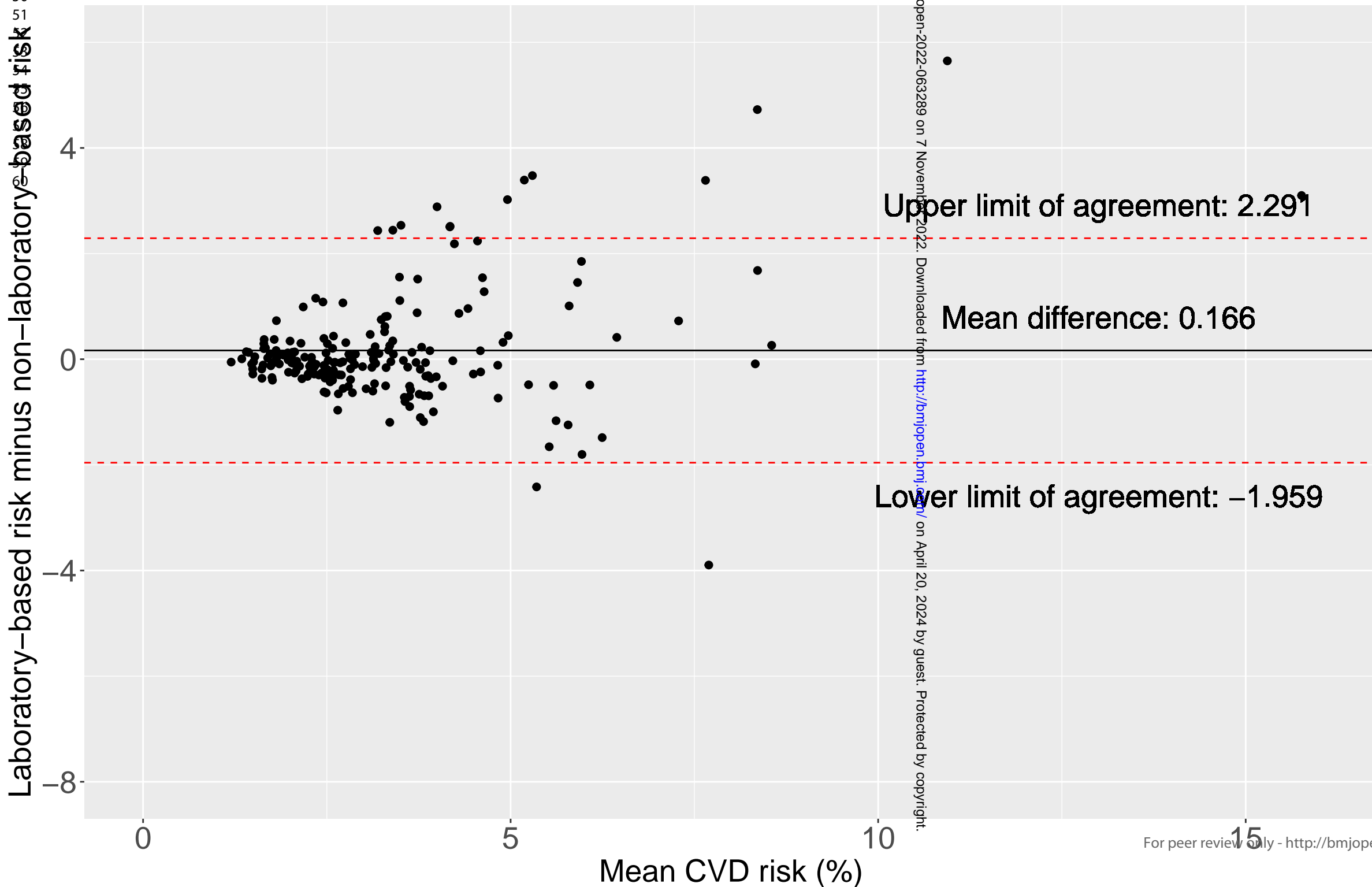
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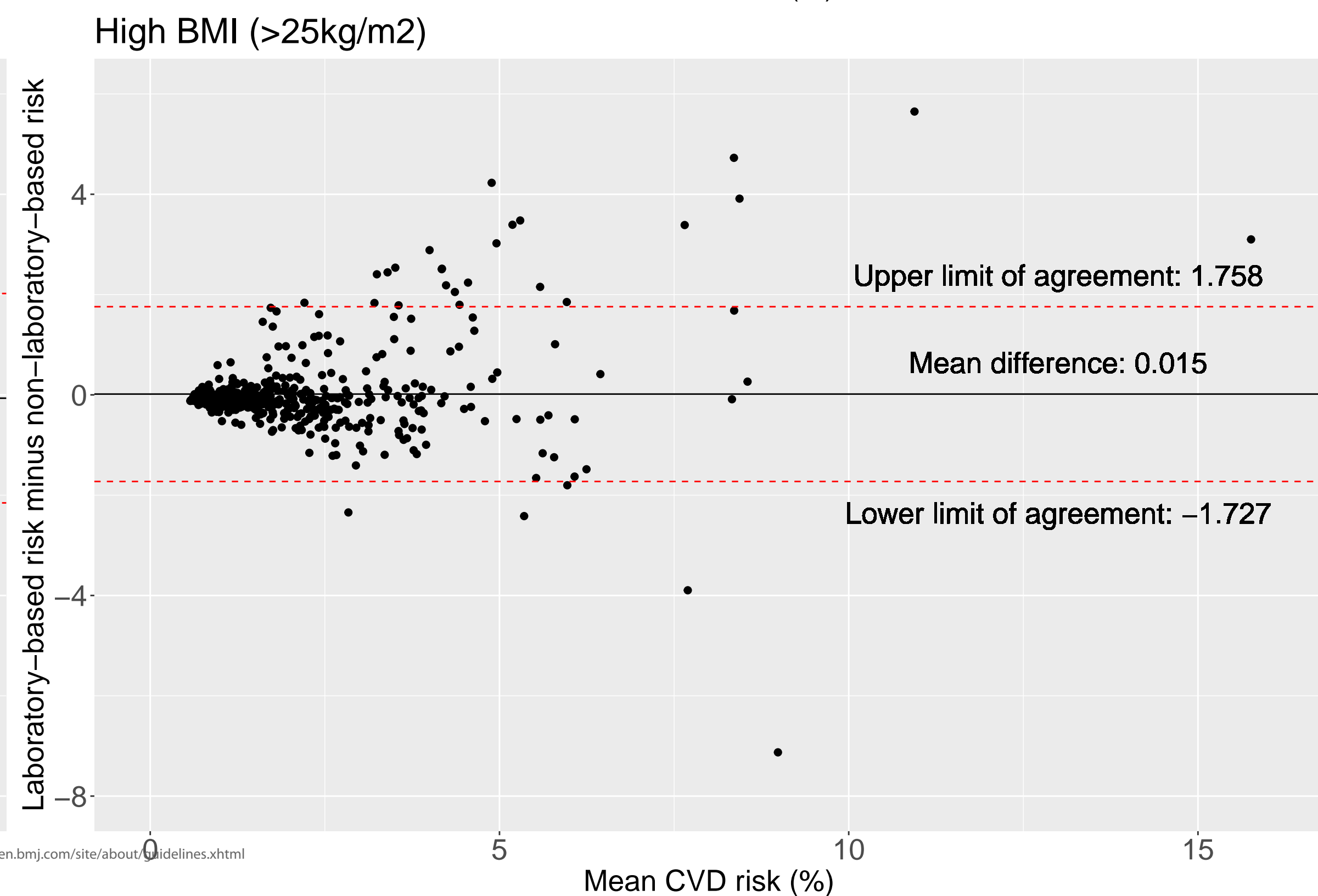
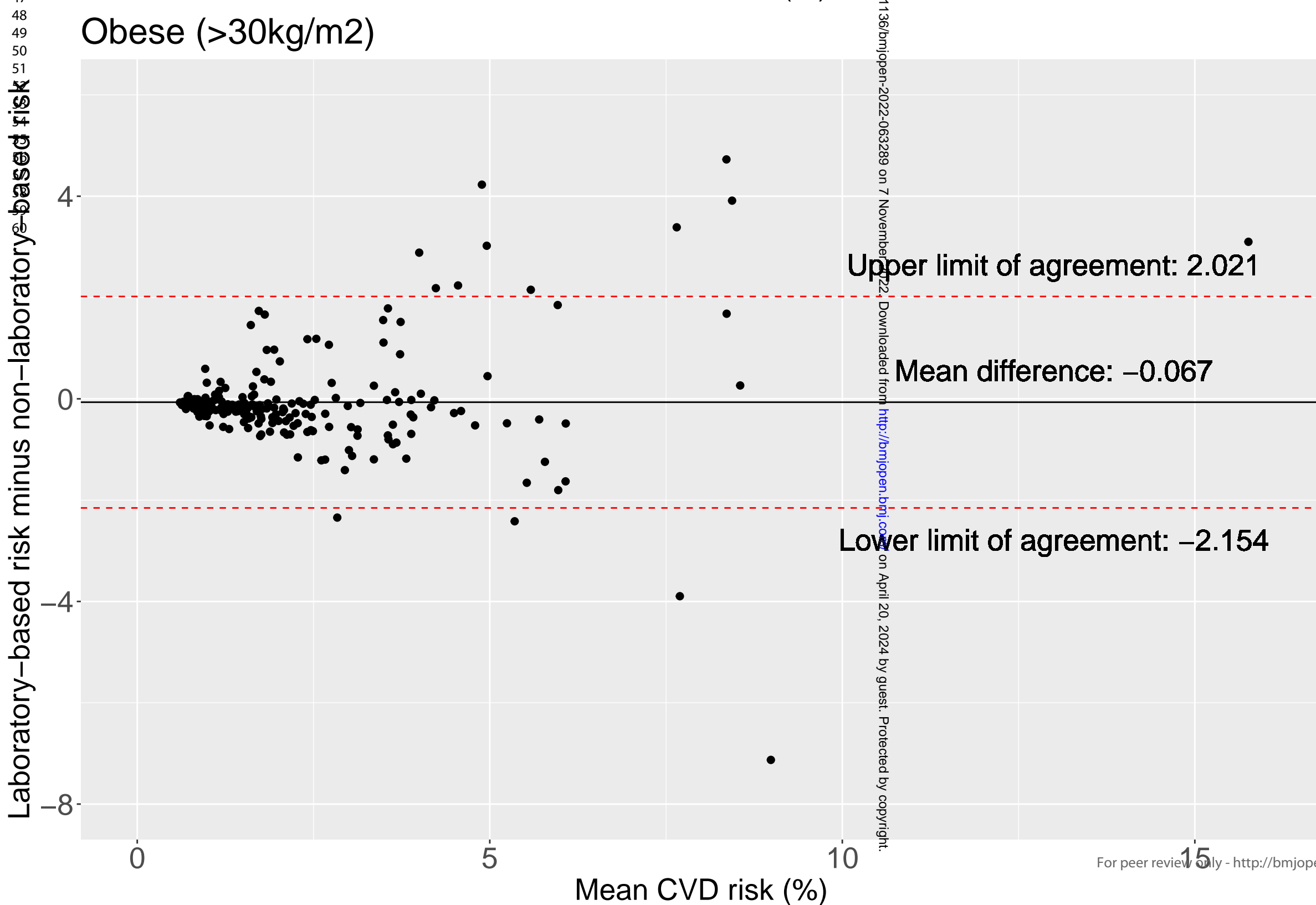
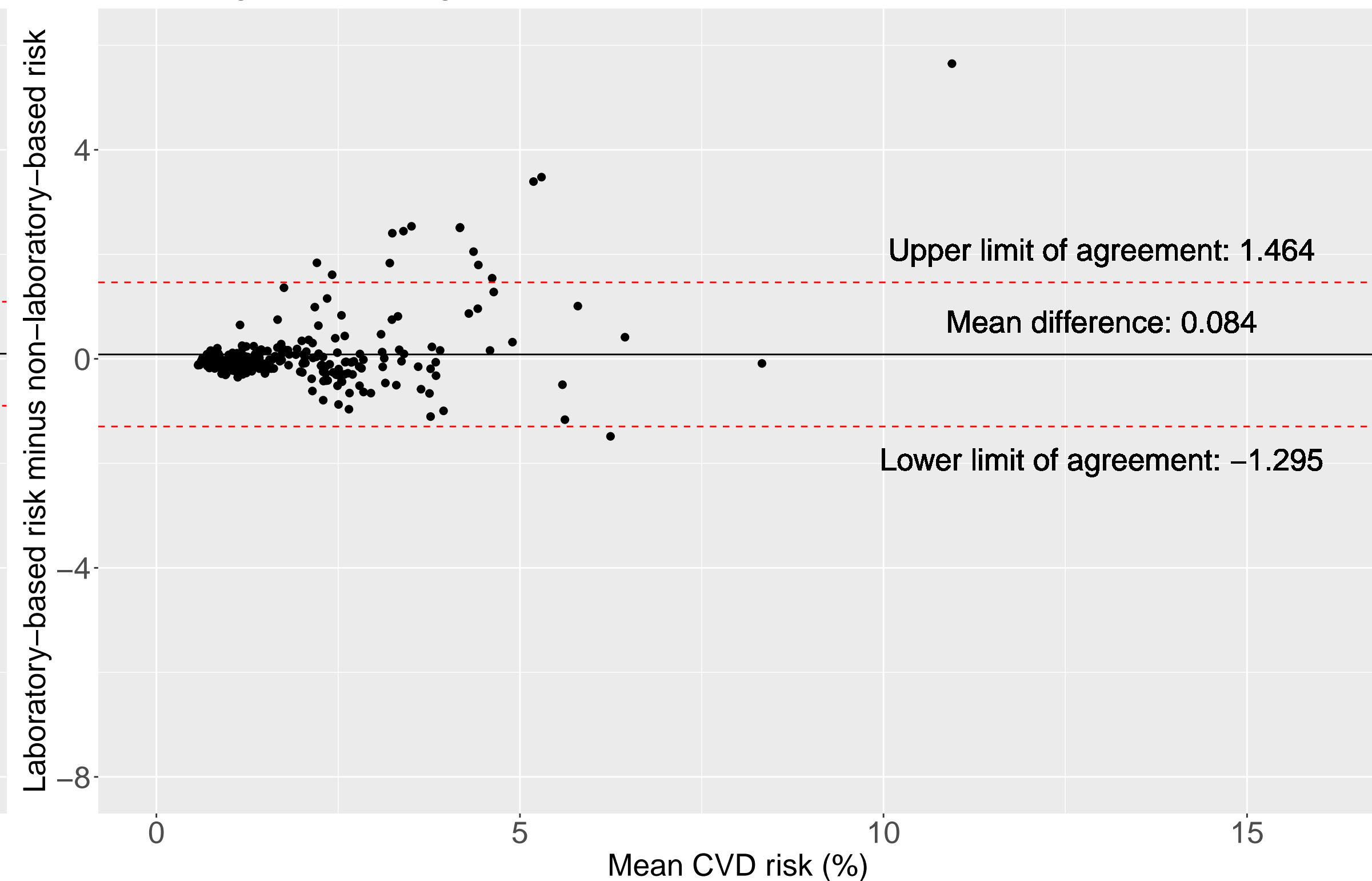
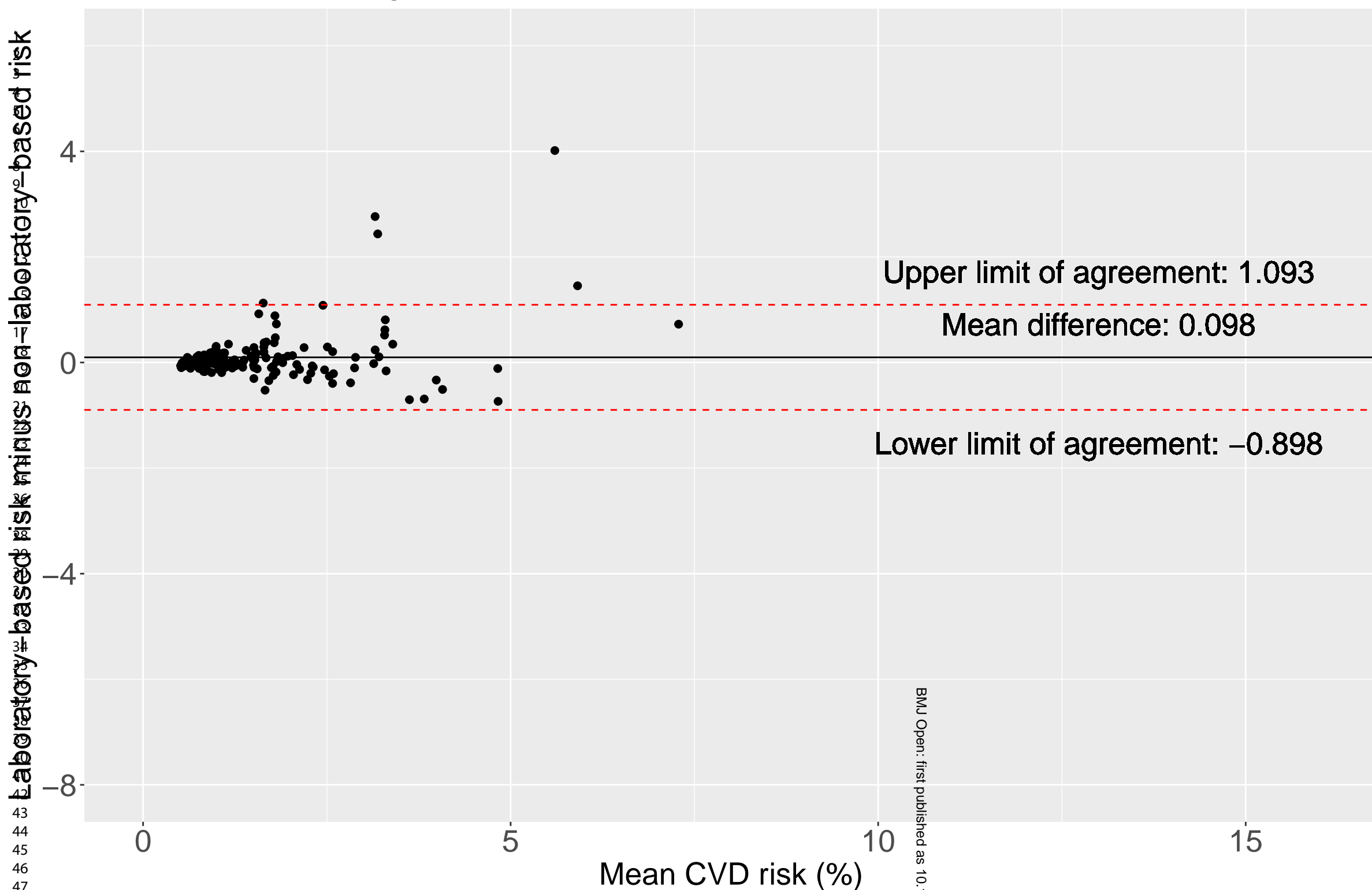


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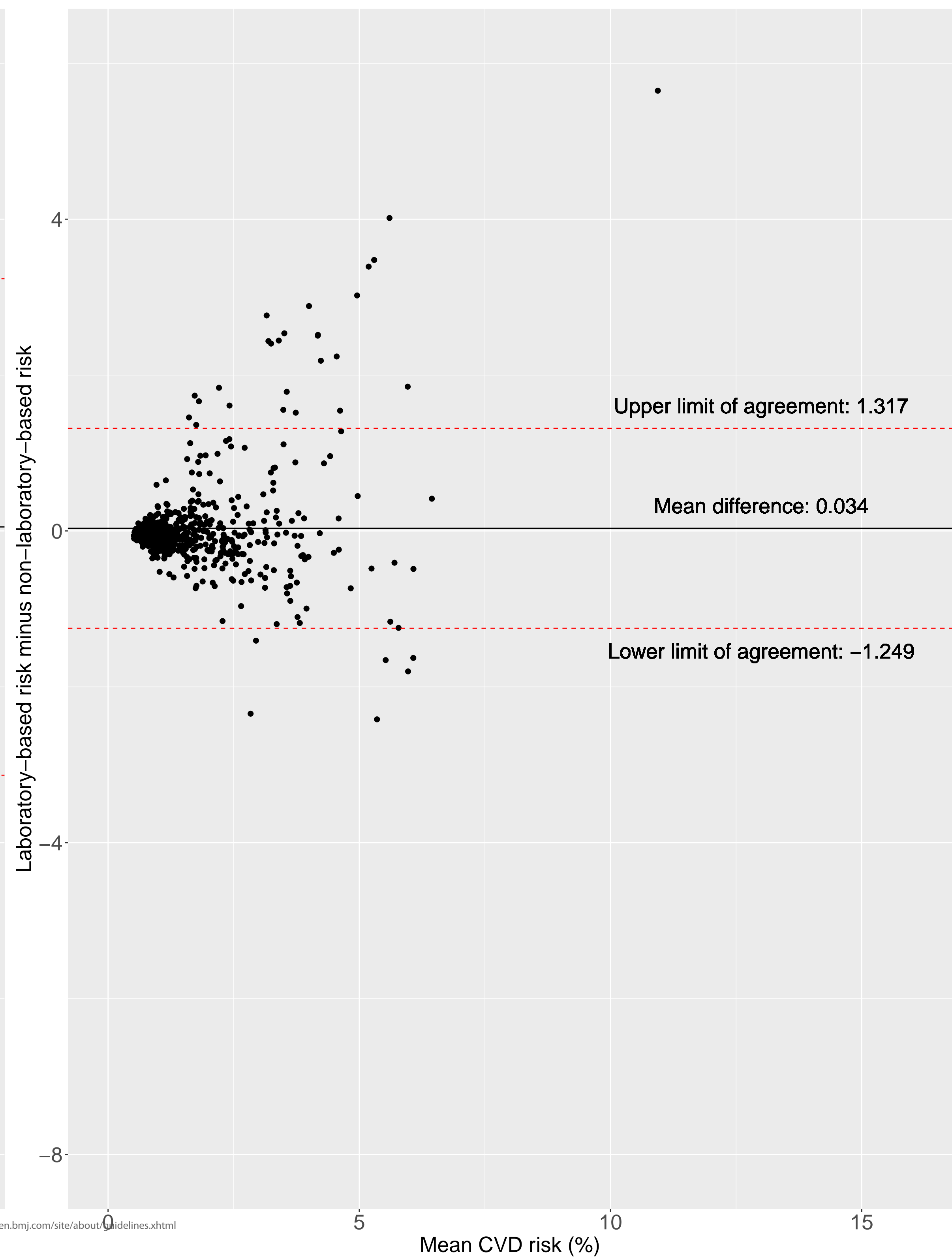
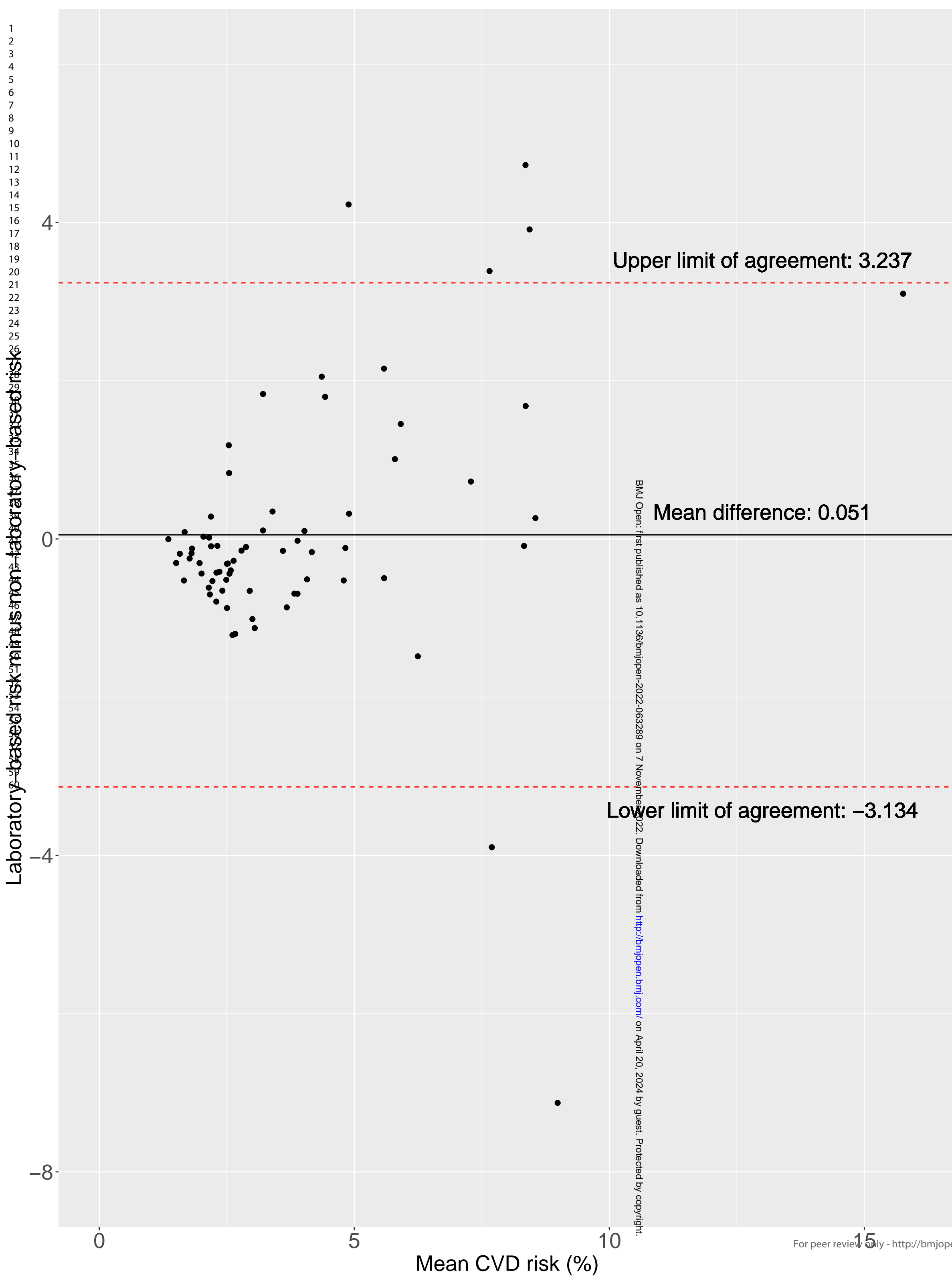


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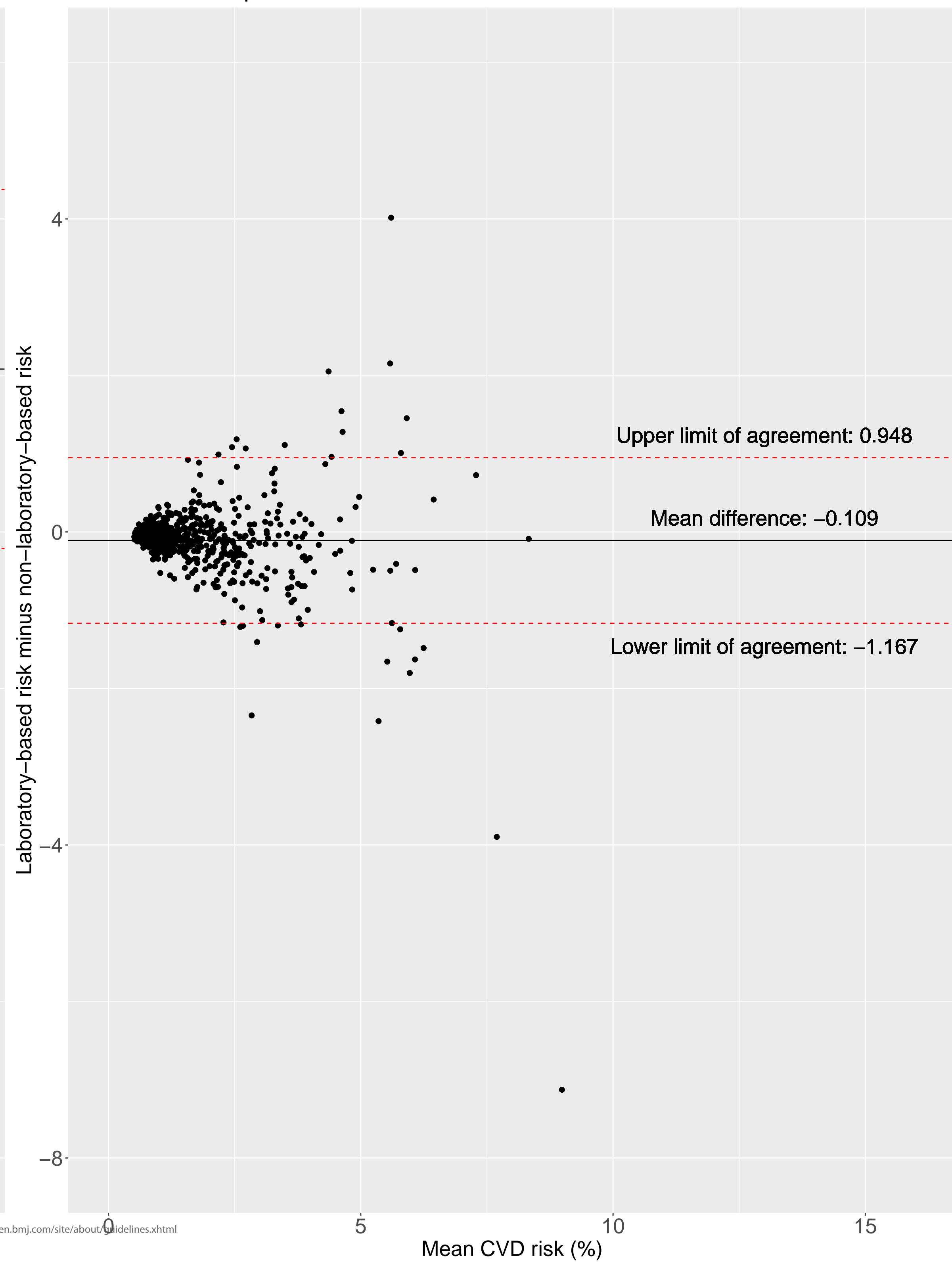
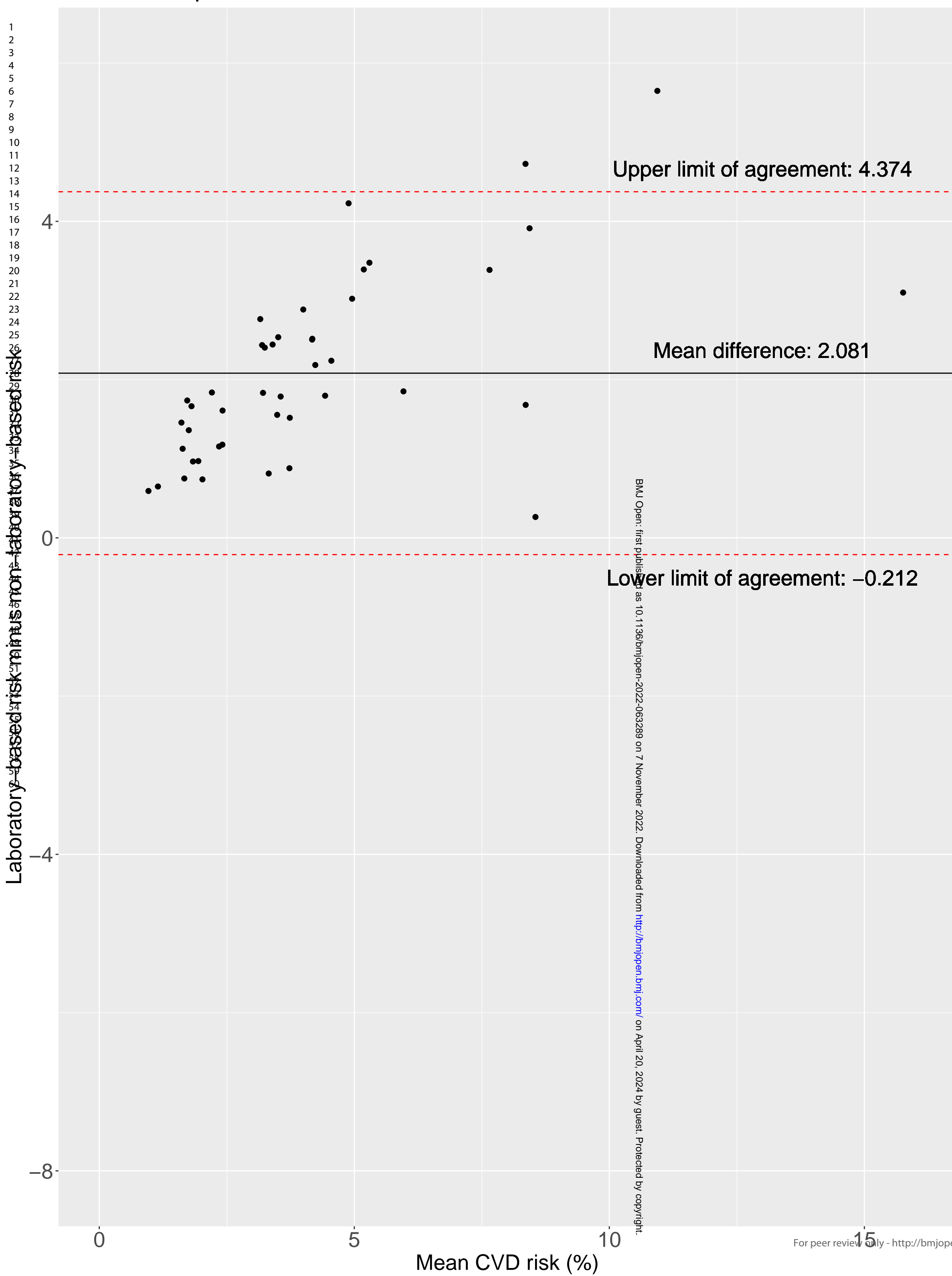
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3 **Agreement between the laboratory- and non-laboratory-based 2019 WHO cardiovascular risk**
4 **charts in Peru**
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Supplementary Figure 1. Formula to compute the survey sample size

$$n_h = \frac{N_h Z^2 P_h Q_h}{(N_h - 1) d^2 + Z^2 P_h Q_h} * TNR$$

Where:

N_h : number of people within an age group in the “h” conglomerate

n_h : number of people in the sample within an age group in the “h” conglomerate

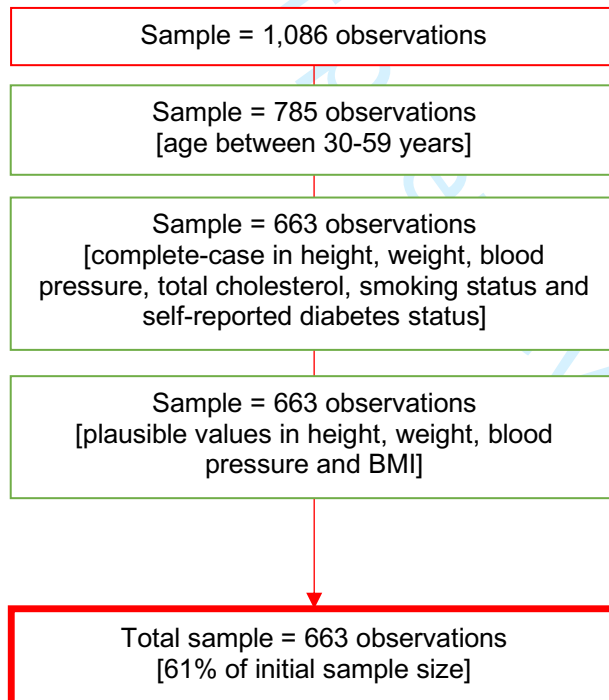
d : error margin assumed in the estimation of P_h

Z : 95% confidence level

TNR (*Tasa de No Respuesta* in Spanish): Expected refusal rate

P_h : prevalence of overweight in adults in the “h” conglomerate

Supplementary Figure 2. Flowchart of data cleaning and inclusion criteria



Supplementary Table 1. Summary statistics of the first and second systolic blood pressure (SBP) records in the overall sample

	First SBP record	Second SBP record
Minimum value	78	75
1st quartile	101	99
Median	109	108
Mean	110.33	109.54
3rd quartile	118	118
Maximum value	203	198

Supplementary Table 2. Sensitivity Analysis: Lin's concordance coefficient correlation showing agreement between laboratory and non-laboratory-based risk models according to the predictors in the 2019 WHO CVD risk models and urban/rural location

Variables	Categories	Lin's concordance coefficient correlation (95% CI)
Sex	Men	0.87 (0.84 - 0.89)
Sex	Women	0.85 (0.83 - 0.87)
Age (years)	30-39	0.87 (0.84 - 0.9)
Age (years)	40-49	0.74 (0.68 - 0.79)
Age (years)	50-59	0.83 (0.78 - 0.87)
Body mass index category	Normal	0.9 (0.87 - 0.92)
Body mass index category	Overweight	0.87 (0.85 - 0.9)
Body mass index category	Obese	0.86 (0.82 - 0.89)
Smoking status	Smoker	0.82 (0.73 - 0.88)
Smoking status	Non-smoker	0.86 (0.84 - 0.88)
Diabetes status	With self-reported diabetes	0.74 (0.63 - 0.82)
Diabetes status	Not with self-reported diabetes	0.91 (0.9 - 0.92)
Urban or rural	Urban	0.88 (0.86 - 0.9)
Urban or rural	Rural	0.86 (0.82 - 0.88)

Supplementary Table 3. Kappa statistics showing agreement between laboratory and non-laboratory-based risk scores in the overall sample

Laboratory-based-risk category	Non-laboratory-based-risk category			kappa
	0-5	5-9	10-19	
0-5	618	4	0	0.62
5-9	17	19	1	
10-19	0	3	1	

Supplementary Table 4. Kappa statistics showing agreement between laboratory and non-laboratory-based risk scores by sex

Laboratory-based-risk category	Sex	Non-laboratory-based-risk category			kappa
		0-5	5-9	10-19	
0-5	Men	251	4	0	0.7
5-9	Men	7	15	1	
10-19	Men	0	1	1	
0-5	Women	367	0	0	0.44
5-9	Women	10	4	0	
10-19	Women	0	2	0	

Supplementary Table 5. Kappa statistics showing agreement between laboratory and non-laboratory-based risk scores by age groups

Laboratory-based-risk category	Age group	Non-laboratory-based-risk category			kappa
		0-5	5-9	10-19	
0-5	30-39	236	0	0	0
5-9	30-39	1	0	0	
10-19	30-39	0	0	0	
0-5	40-49	216	1	0	0.45
5-9	40-49	4	2	1	
10-19	40-49	0	1	0	
0-5	50-59	166	3	0	0.65
5-9	50-59	12	17	0	
10-19	50-59	0	2	1	

Supplementary Table 6. Kappa statistics showing agreement between laboratory and non-laboratory-based risk scores by body mass index categories

Laboratory-based-risk category	Body mass index category	Non-laboratory-based-risk category			kappa
		0-5	5-9	10-19	
0-5	Normal	161	1	0	0.66
5-9	Normal	1	2	0	
10-19	Normal	0	0	0	
0-5	Obese	202	3	0	0.65
5-9	Obese	7	11	1	
10-19	Obese	0	2	1	
0-5	Overweight	255	0	0	0.55
5-9	Overweight	9	6	0	
10-19	Overweight	0	1	0	

Supplementary Table 7. Kappa statistics showing agreement between laboratory and non-laboratory-based risk scores by smoking status

Laboratory-based-risk category	Smoking status	Non-laboratory-based-risk category			kappa
		0-5	5-9	10-19	
0-5	Non-smoker	568	3	0	0.53
5-9	Non-smoker	12	9	0	
10-19	Non-smoker	0	1	0	
0-5	Smoker	50	1	0	0.67
5-9	Smoker	5	10	1	
10-19	Smoker	0	2	1	

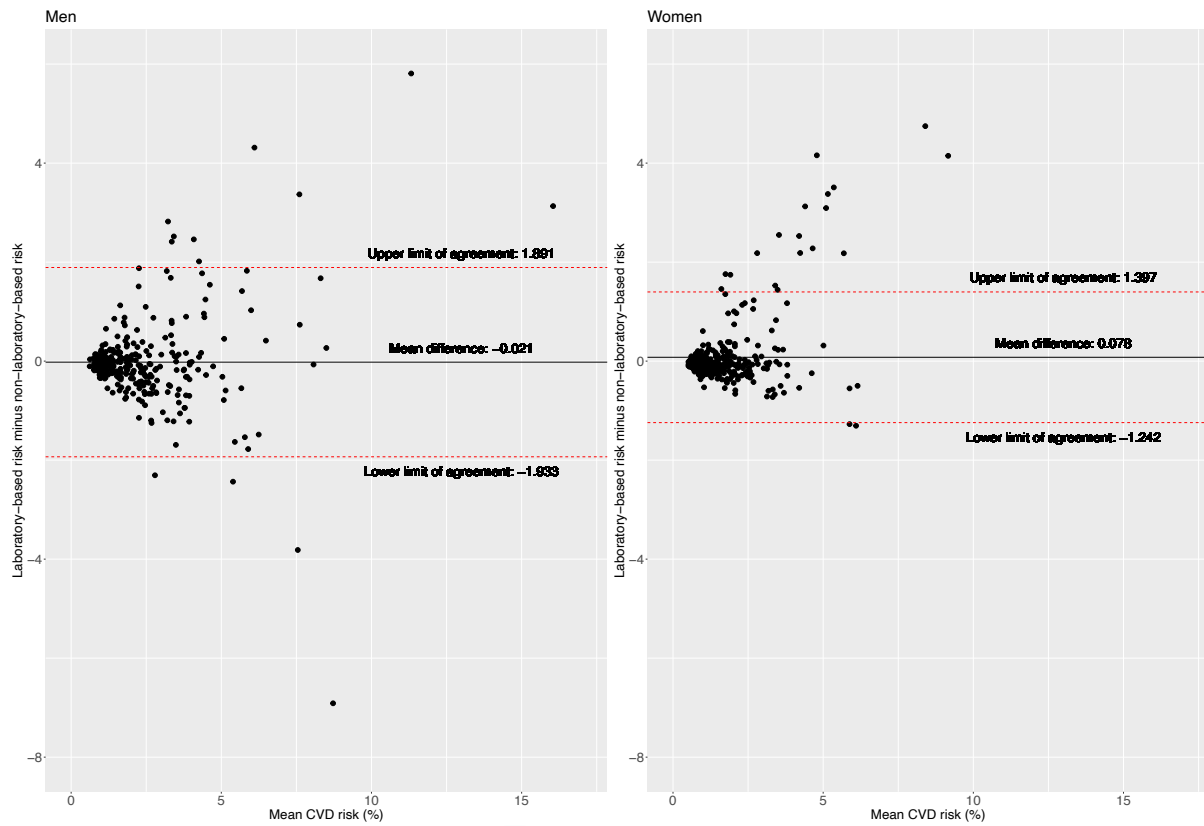
Supplementary Table 8. Kappa statistics showing agreement between laboratory and non-laboratory-based risk scores by diabetes status

Laboratory-based-risk category	Diabetes status	Non-laboratory-based-risk category			kappa
		0-5	5-9	10-19	
0-5	Not with self-reported diabetes	592	4	0	0.71
5-9	Not with self-reported diabetes	7	15	1	
10-19	Not with self-reported diabetes	0	0	0	
0-5	With self-reported diabetes	26	0	0	0.36
5-9	With self-reported diabetes	10	4	0	
10-19	With self-reported diabetes	0	3	1	

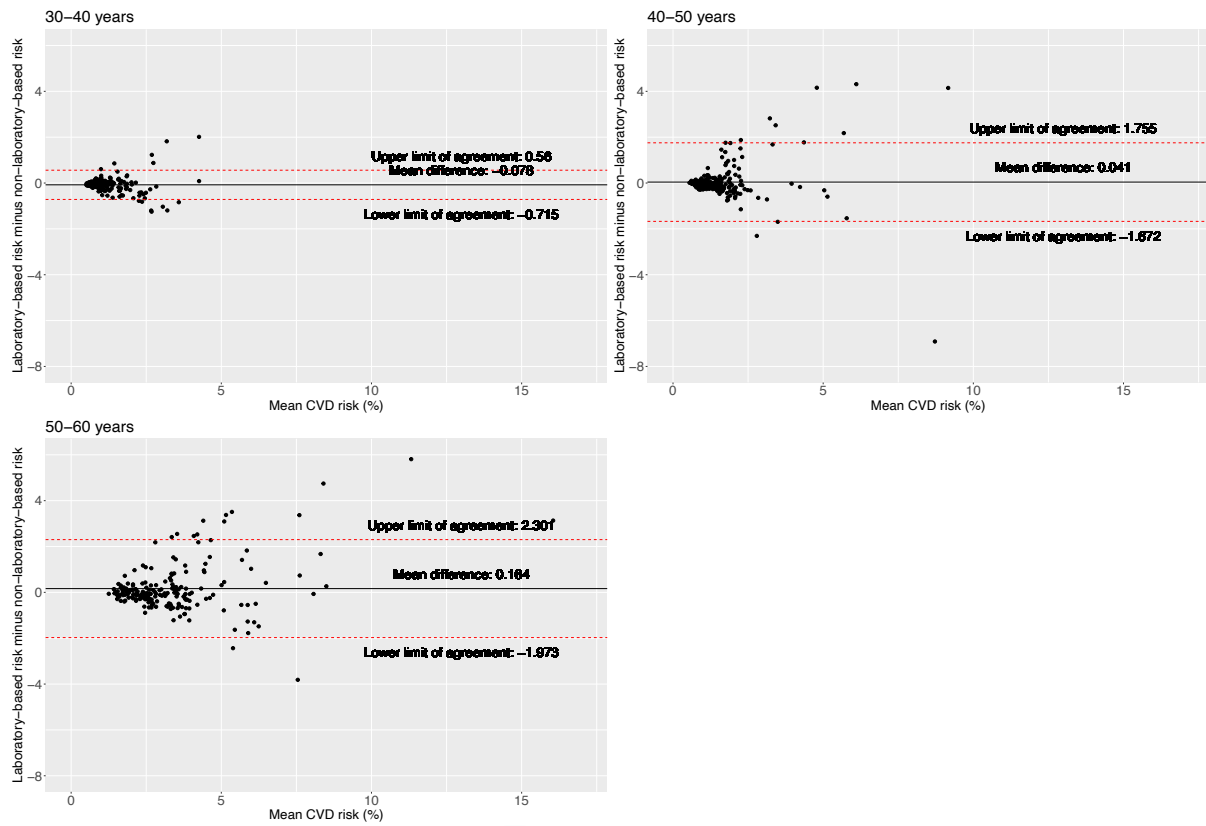
Supplementary Table 9. Kappa statistics showing agreement between laboratory and non-laboratory-based risk scores by urban/rural location

Laboratory-based-risk category	Urban/Rural	Non-laboratory-based-risk category			kappa
		0-5	5-9	10-19	
0-5	Rural	224	1	0	0.53
5-9	Rural	4	3	0	
10-19	Rural	0	1	0	
0-5	Urban	394	3	0	0.64
5-9	Urban	13	16	1	
10-19	Urban	0	2	1	

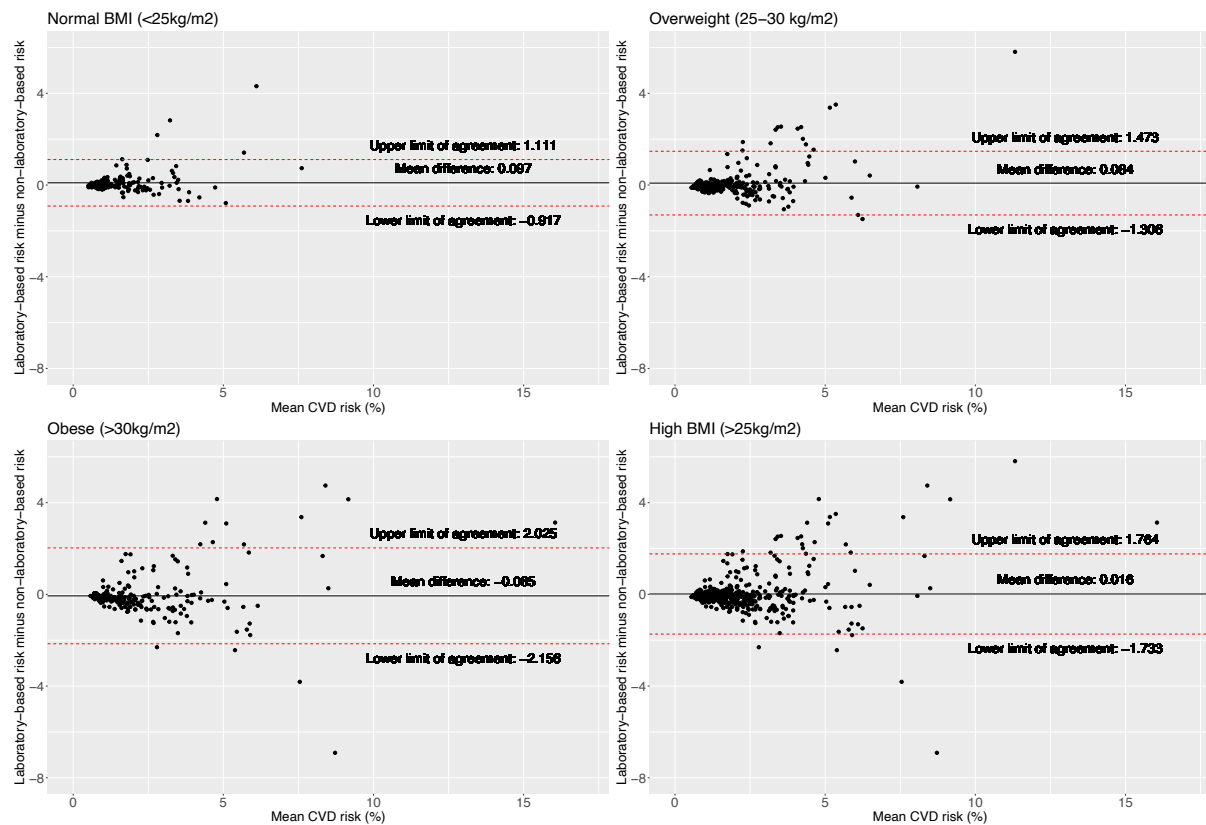
Supplementary Figure 3. Sensitivity Analysis: Bland Altman plots showing agreement between laboratory and non-laboratory-based risk scores according to sex.



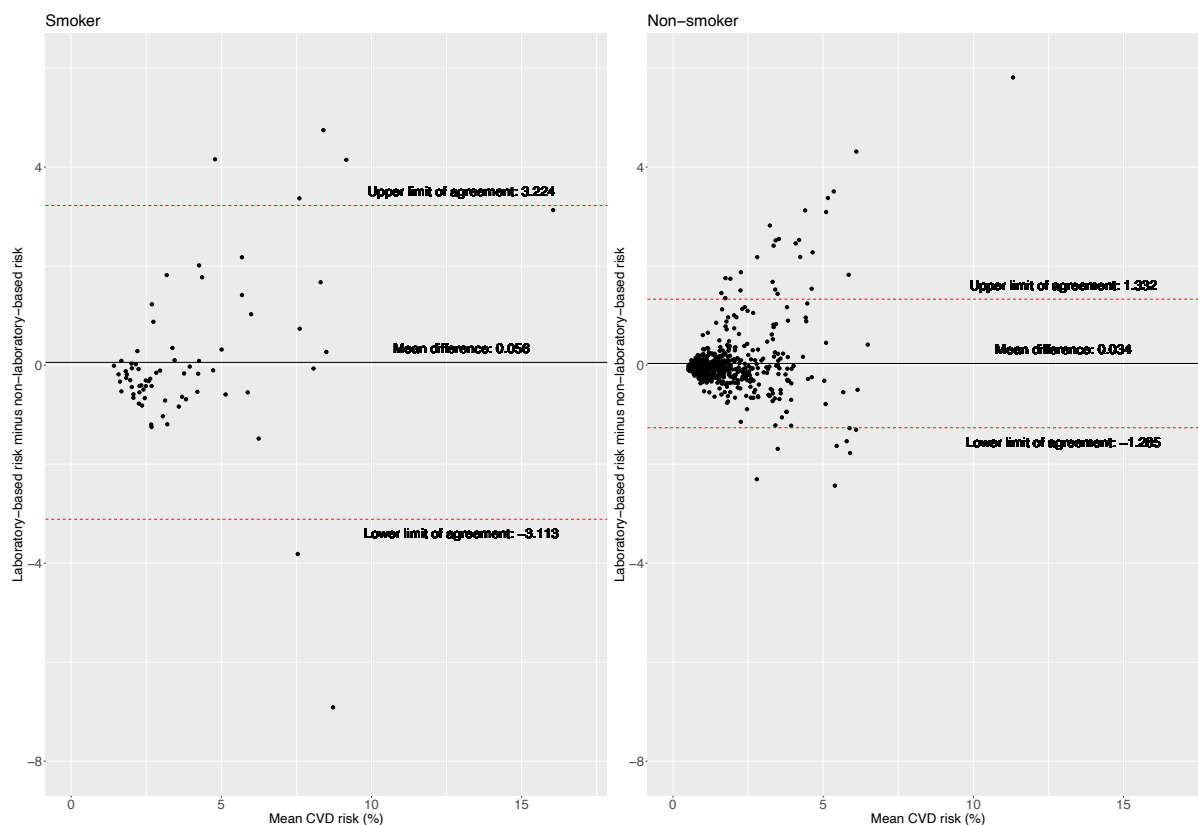
Supplementary Figure 4. Sensitivity Analysis: Bland Altman plots showing agreement between laboratory and non-laboratory-based risk scores according to age groups.



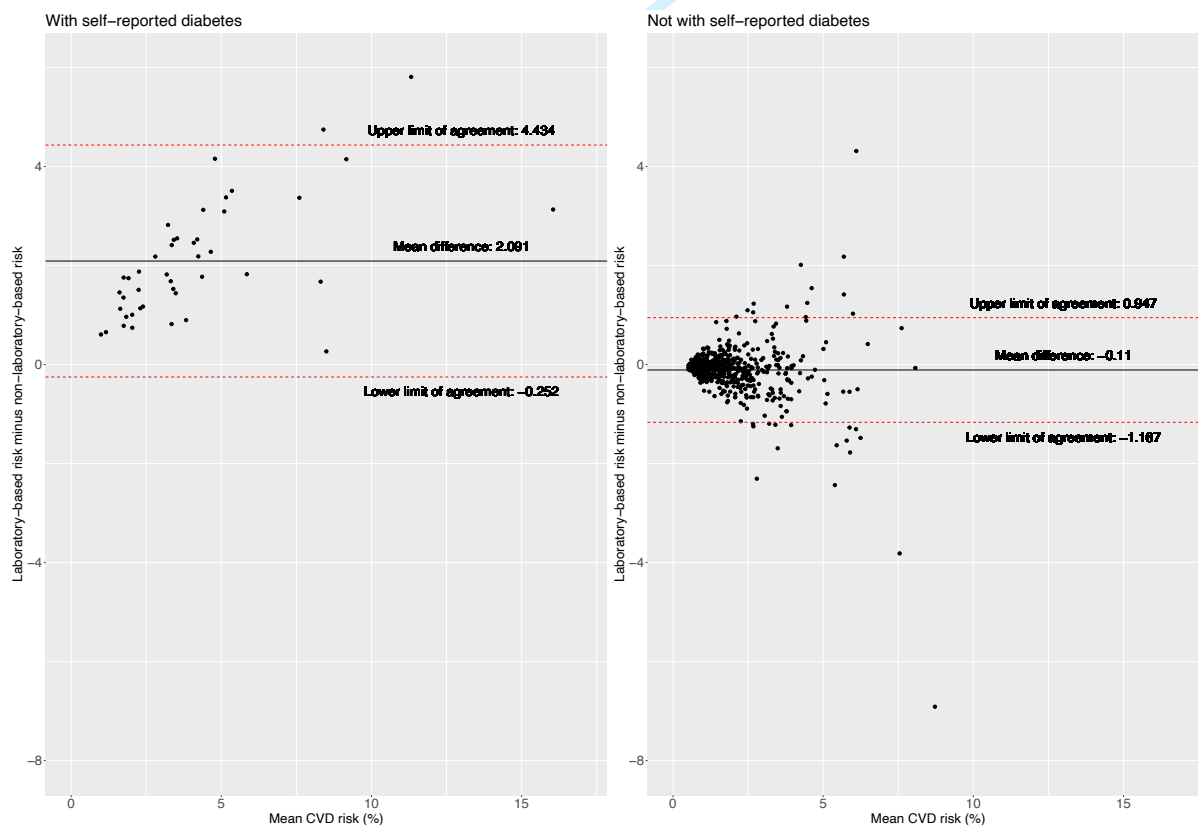
Supplementary Figure 5. Sensitivity Analysis: Bland Altman plots showing agreement between laboratory and non-laboratory-based risk scores according to body mass index categories.



Supplementary Figure 6. Sensitivity Analysis: Bland Altman plots showing agreement between laboratory and non-laboratory-based risk scores according to smoking status.



Supplementary Figure 7. Sensitivity Analysis: Bland Altman plots showing agreement between laboratory and non-laboratory-based risk scores according to self-reported diabetes status.



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STROBE Statement—Checklist of items that should be included in reports of *cross-sectional studies*

	Item No	Recommendation	Page No
Title and abstract	1	(a) Indicate the study's design with a commonly used term in the title or the abstract	3
		(b) Provide in the abstract an informative and balanced summary of what was done and what was found	3
Introduction			
Background/rationale	2	Explain the scientific background and rationale for the investigation being reported	4
Objectives	3	State specific objectives, including any prespecified hypotheses	5
Methods			
Study design	4	Present key elements of study design early in the paper	5
Setting	5	Describe the setting, locations, and relevant dates, including periods of recruitment, exposure, follow-up, and data collection	5-6
Participants	6	(a) Give the eligibility criteria, and the sources and methods of selection of participants	5-6
Variables	7	Clearly define all outcomes, exposures, predictors, potential confounders, and effect modifiers. Give diagnostic criteria, if applicable	6-7
Data sources/ measurement	8*	For each variable of interest, give sources of data and details of methods of assessment (measurement). Describe comparability of assessment methods if there is more than one group	6-7
Bias	9	Describe any efforts to address potential sources of bias	6
Study size	10	Explain how the study size was arrived at	5-6
Quantitative variables	11	Explain how quantitative variables were handled in the analyses. If applicable, describe which groupings were chosen and why	6-7
Statistical methods	12	(a) Describe all statistical methods, including those used to control for confounding	7-8
		(b) Describe any methods used to examine subgroups and interactions	7-8
		(c) Explain how missing data were addressed	5-6
		(d) If applicable, describe analytical methods taking account of sampling strategy	8
		(e) Describe any sensitivity analyses	6
Results			
Participants	13*	(a) Report numbers of individuals at each stage of study—eg numbers potentially eligible, examined for eligibility, confirmed eligible, included in the study, completing follow-up, and analysed	8
		(b) Give reasons for non-participation at each stage	Supp Fig 1
		(c) Consider use of a flow diagram	5
Descriptive data	14*	(a) Give characteristics of study participants (eg demographic, clinical, social) and information on exposures and potential confounders	8
		(b) Indicate number of participants with missing data for each variable of interest	Supp Fig 1
Outcome data	15*	Report numbers of outcome events or summary measures	

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Main results	16	(a) Give unadjusted estimates and, if applicable, confounder-adjusted estimates and their precision (eg, 95% confidence interval). Make clear which confounders were adjusted for and why they were included	8-10
		(b) Report category boundaries when continuous variables were categorized	8-10
		(c) If relevant, consider translating estimates of relative risk into absolute risk for a meaningful time period	-
Other analyses	17	Report other analyses done—eg analyses of subgroups and interactions, and sensitivity analyses	10
Discussion			
Key results	18	Summarise key results with reference to study objectives	11
Limitations	19	Discuss limitations of the study, taking into account sources of potential bias or imprecision. Discuss both direction and magnitude of any potential bias	13
Interpretation	20	Give a cautious overall interpretation of results considering objectives, limitations, multiplicity of analyses, results from similar studies, and other relevant evidence	11-12
Generalisability	21	Discuss the generalisability (external validity) of the study results	11-12
Other information			
Funding	22	Give the source of funding and the role of the funders for the present study and, if applicable, for the original study on which the present article is based	2

*Give information separately for exposed and unexposed groups.

Note: An Explanation and Elaboration article discusses each checklist item and gives methodological background and published examples of transparent reporting. The STROBE checklist is best used in conjunction with this article (freely available on the Web sites of PLoS Medicine at <http://www.plosmedicine.org/>, Annals of Internal Medicine at <http://www.annals.org/>, and Epidemiology at <http://www.epidem.com/>). Information on the STROBE Initiative is available at www.strobe-statement.org.

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Agreement between the laboratory- and non-laboratory-based WHO cardiovascular risk charts: a cross-sectional analysis of a national health survey in Peru

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3 **Agreement between the laboratory- and non-laboratory-based WHO cardiovascular risk**
4 **charts: a cross-sectional analysis of a national health survey in Peru**
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ABSTRACT

Objective: To determine the agreement between the cardiovascular disease (CVD) risk predictions computed with the World Health Organization (WHO) non-laboratory-based model and laboratory-based model in a nationally-representative sample of Peruvian adults.

Design: Cross-sectional analysis of a national health survey.

Methods: Absolute CVD risk was computed with the 2019 WHO laboratory- and non-laboratory-based models. The risk predictions from both models were compared with Bland Altman plots, Lin's concordance coefficient correlation (LCCC) and kappa statistics, stratified by sex, age, body mass index categories, smoking and diabetes status.

Results: 663 people aged 30-59 years were included in the analysis. Overall, there were no substantial differences between the mean CVD risk computed with the laboratory-based model 2.0% (95% CI: 1.8%; 2.2%) and the non-laboratory-based model 2.0% (95% CI: 1.8%; 2.1%). In the Bland Altman plots, the limits of agreement were the widest among people with diabetes (-0.21; 4.37) compared with people without diabetes (-1.17; 0.95). The lowest agreement as per the LCCC was also seen in people with diabetes (0.74 (95% CI: 0.63; 0.82)), the same was observed with the kappa statistic (kappa=0.36). In general, agreement between the scores was appropriate in terms of clinical significance.

Conclusions: The absolute cardiovascular predicted risk was similar between the laboratory-based and non-laboratory-based 2019 WHO cardiovascular risk models. Pending validation from longitudinal studies, the non-laboratory-based model (instead of the laboratory-based) could be used when assessing CVD risk in Peruvian population.

Key words: cardiovascular diseases; risk assessment; health metrics

STRENGTHS AND LIMITATIONS OF THIS STUDY

- This analysis provided the first evidence of the agreement between the 2019 WHO CVD risk laboratory-based and non-laboratory-based models.
- We leveraged on the most recent nationally representative survey that included blood biomarkers in Peru.
- Our study population was young, mostly women, and with overall low absolute cardiovascular predicted risk. No one in the study population had an absolute cardiovascular risk $\geq 20\%$.
- We assumed all participants were free of CVD to use the 2019 WHO CVD risk score, as information regarding history of cardiovascular events was not reported.

INTRODUCTION

Cardiovascular diseases (CVD) are the main cause of death globally.[1] In 2019, CVD caused ~18.5 million deaths in adults, representing 36% of all global deaths.[2,3] Furthermore, CVD impose a huge burden in low- and middle-income countries (LMICs),[4] where deaths from CVD occur at younger ages compared to high-income countries (HICs).[1] However, CVD can be prevented and managed through a combination of population- and individual-level interventions;[5] for the latter, the identification of individuals at high cardiovascular risk is a cornerstone in the prevention of CVD. In this line, the World Health Organization (WHO) and the Pan American Health Organization (PAHO),[6] alongside several clinical guidelines,[7–9] recommend CVD risk stratification with CVD risk prediction models to inform evidence-based treatment.

CVD risk prediction models identify people who would benefit the most of preventive interventions (e.g., statin therapy).[10] Although there are several CVD risk prediction models,[11] these were mostly developed in HICs limiting their application in LMICs where they would need recalibration to deliver accurate predictions to guide treatment allocation. To overcome this limitation, the WHO convened a global effort to derive, calibrate and validate new CVD risk prediction models for all world regions.[12] Two WHO CVD risk prediction models were developed: a laboratory-based

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3 model and a non-laboratory-based model. Because laboratory biomarkers (e.g., total cholesterol)
4 may not be available in primary health care centres in LMICs limiting the use of laboratory-based
5 CVD risk prediction models,[13] the non-laboratory-based model arises as a handy tool for
6 clinicians in LMICs. Similarly, CVD risk prediction models are used to monitor the prevalence of
7 high CVD risk and treatment coverage (i.e., people at high CVD receiving treatment).[14] In this
8 context, countries conducting national or large population-based health surveys without lipid
9 biomarkers, could benefit from the non-laboratory-based models. Peru, for example, does not
10 have regular national health surveys including total cholesterol, but Peru has a yearly national
11 health survey including anthropometrics, blood pressure, and health questionnaires. In Peru, and
12 other similar LMICs, it would not be possible to monitor the burden of high CVD risk with a
13 laboratory-based model and the non-laboratory-based model rises as the only alternative.
14 Nonetheless, evidence regarding the agreement between the WHO laboratory-based and non-
15 laboratory-based models in countries from Latin America is missing.[12] Whether the WHO non-
16 laboratory-based model delivers predictions similar to those of the WHO laboratory-based model
17 remains unknown in Latin America. However, clinicians need this evidence to inform their choice
18 (non-laboratory-based vs laboratory-based model), and to interpret the results of the non-
19 laboratory-based model under the assumption that the laboratory-based model is the gold
20 standard.[12] To provide this evidence for practitioners in Peru, we determined the agreement
21 between the risk predictions computed with the non-laboratory-based model and laboratory-
22 based model in a nationally-representative sample of Peruvian adults.

43 **METHODS**

45 **Data sources**

47 This is a cross-sectional study of a national survey conducted by the National Centre for Food
48 and Nutrition (CENAN, for its acronym in Spanish) of Peru. CENAN's survey was conducted
49 between 2017-2018 on a nationally-representative sample of Peruvian adults aged between 18-
50 59 years.[15] Of note, this is the most recent nationally-representative survey conducted in Peru
51 that included blood biomarkers (e.g., lipid profile). CENAN's survey adhered to ethical guidelines
52 and followed a standardised protocol that has been published elsewhere.[15] Each participant
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3 was informed about all procedures and techniques used in the survey; also, participants could
4 have left the study at any time and their personal information was kept confidential.[15]
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9 The CENAN's survey sample was computed using the formula shown in Supplementary Figure 1
10 and followed a probabilistic sampling design approach with two stages.[15] First, clusters were
11 randomly selected considering three strata: 1) Urban areas except Lima city, 2) Rural areas, and
12 3) Lima city. Then, households (of adults aged 18-59 years living in) were randomly selected
13 within each cluster. To be selected for the survey sample, participants had to fulfil the following
14 inclusion criteria: 1) Adults aged 18-59 years, and 2) Fasting 9-12 hours for blood biomarkers.
15 The following participants were excluded: 1) Pregnant and postpartum women, 2) Adults taking
16 medication that could alter glucose and lipid profiles, 3) Adults with congenital diseases that could
17 limit anthropometrics measurement (e.g., Down syndrome).
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28 **Study population**

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30 We analysed a complete-case sample regarding all the laboratory- and office-based 2019 ten-
31 year WHO CVD risk score variables (see Variables section). We studied men and women aged
32 between 30-59 years. The younger age limit was decided because CVD risk models are not
33 recommended in younger individuals; the older age limit was decided because of the survey
34 design. A flowchart of data cleaning is shown in Supplementary figure 2. We did not apply other
35 selection criteria.
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43 Although the 2019 ten-year WHO CVD risk models were developed for people aged 40-80
44 years,[12] for people aged <40 years we assumed they had 40 years for the absolute CVD risk
45 computation. This is consistent with the Package of essential noncommunicable (PEN) disease
46 interventions for primary health care in low-resource settings,[16] and was also done in a recent
47 global work.[17]
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54 **Variables**

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56 We calculated the ten-year CVD risk at the individual-level following the laboratory- and non-
57 laboratory-based 2019 ten-year WHO CVD risk model.[12] We used the *whocvdrisk* command in
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3 STATA, which was developed by the authors of the 2019 ten-year WHO CVD risk charts.[18] For
4 the laboratory-based model, scores were calculated based on: age (years), current smoking
5 status (yes/no), systolic blood pressure (SBP, mmHg), history of self-reported diabetes diagnosis
6 (yes/no), and total cholesterol (mmol/L).[12] For the non-laboratory-based model, we used age
7 (years), current smoking status (yes/no), systolic BP (mmHg) and body mass index (BMI,
8 kg/m²).[12]
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16 The CENAN's survey collected anthropometrics and three BP measurements which were taken
17 by trained fieldworkers following a standard protocol.[15] We computed BMI using measured
18 weight (kg) divided by the square of height (meters); for descriptive purposes, we classified BMI
19 in three levels: normal weight (BMI <25 kg/m²), overweight (BMI ≥25-29.9 kg/m²), and obesity
20 (BMI ≥30 kg/m²). BMI records outside the range 10-80 kg/m² were discarded. As the third BP
21 measurement was only available in few participants (<2% of the initial sample), we used the
22 second SBP measurement only (i.e., the first and third SBP records were discarded in the main
23 analysis). Of note, there were no substantial differences between the first and second SBP
24 records (Supplementary table 1); nonetheless, we performed a sensitivity analysis using the
25 mean SBP of first and second SBP records. We discarded any SBP records outside the range
26 70-270 mmHg.
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40 For people who self-reported being under antihypertensive treatment, we used the pre-treatment
41 SBP; this is consistent with the PEN protocol and with a previous global work.[16,17]. Pre-
42 treatment SBP was computed as: pre-treatment systolic blood pressure = (current systolic blood
43 pressure-6.3)/0.9.[19] Conversely, for those not taking antihypertensive treatment, we used the
44 recorded SBP as was.
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51 We defined current smoker with one question coded as no versus yes: *do you currently smoke*
52 *any tobacco product such as cigarettes, cigars or pipes?* Self-reported information about a prior
53 history of diabetes was assessed by a question also coded as no versus yes: *have you ever been*
54 *told by a physician or another healthcare worker that you have high blood sugar or diabetes?*
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3 Total cholesterol was obtained via enzymatic colorimetric method.[15] Because CENAN's survey
4 data had total cholesterol in mg/dl, these values were divided by 38.67 to obtain total cholesterol
5 in mmol/L.
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10 **Statistical analysis**

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12 We determined the agreement between the absolute CVD risk predicted with the WHO laboratory-
13 and non-laboratory-based models following three methods: Bland Altman plots, Lin's
14 concordance coefficient correlation (LCCC), and kappa statistic. We considered the absolute
15 CVD risk as a continuous variable, and the agreement between both models was examined using
16 Bland Altman plots and the LCCC. Furthermore, we considered the CVD risk as a categorical
17 variable, and divided into three groups: <5%, 5-9%, 10-19%; because there were no observations
18 in the high-risk category (CVD risk $\geq 20\%$), agreement was not examined in this group. For these
19 categories, we evaluated the agreement using the kappa statistic. For the Bland Altman plots,
20 LCCC, and the kappa statistic, results were stratified by sex, 10-year age groups, BMI categories,
21 smoking status, self-reported diabetes diagnosis and urban/rural location.
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34 In the Bland Altman plots, the risk difference between the laboratory- and non-laboratory-based
35 absolute cardiovascular predicted risk was plotted on the vertical axis, and the mean of both
36 scores on the horizontal axis.[20] As the true risk of CVDs at the individual-level is uncertain, the
37 mean of both scores is the best available estimate.[20,21] The 95% of the limit of agreement was
38 represented by the mean difference of both scores \pm two standard deviations; this limit provides
39 an interval in which 95% of the differences between both scores would be expected to lie.[20] The
40 LCCC between the laboratory- and non-laboratory-based absolute cardiovascular predicted risks
41 was also evaluated. The agreement based on the LCCC ranges between -1 and 1, with 1
42 suggesting a perfect agreement. The categorical agreement was evaluated using the Kappa
43 statistic. Kappa <0 indicated less than chance agreement, and values 0.01–0.20, 0.21–0.40,
44 0.41–0.60, 0.61–0.80, and 0.81–0.99 represented slight, fair, moderate, substantial, and almost
45 perfect agreement, respectively.[22]
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All analyses code were conducted with R (version 4.0.3) and STATA (version 17.0, College Station, Texas 77845, USA). Population characteristics along with their 95% confidence interval (95% CI) were summarized accounting for the complex survey design of the CENAN's survey.[15]

Patient and public involvement

No patient involved.

RESULTS

Our pooled dataset included 663 participants (Supplementary figure 2). The mean age was 44.0 years (95% CI: 43.2; 44.7) and the proportion of men was 41.5%. The mean SBP was 108.9 mmHg (95% CI: 107.5; 110.3 mmHg) and the mean BMI was 28.8 kg/m² (95% CI: 28.3; 29.2 kg/m²). The proportion of people with overweight was 41.0% (95% CI: 36.6; 45.6 %), whereas 35.9% (95% CI: (31.6; 40.5%)) of the population were obese. The mean total cholesterol was 4.9 mmol/L (95% CI: 4.8; 5.0 mmol/L), 11.7% (95% CI: 8.8; 15.3%) of the population were smokers and 7.1% (95% CI: 5.0%; 9.8%) had diabetes (Table 1).

Table 1. Weighted distribution of the predictors in the 2019 WHO CVD risk models, overall and by sex.

	Total	Men	Women
Sample size	663	280	383
Age (mean and 95% CI, years)	44 (43.2-44.7)	44.2 (43-45.4)	43.8 (42.9-44.7)
Proportion of people aged 30-39 years (95% CI, %)	35.8 (31.6-40.3)	37.2 (31-43.9)	34.8 (29.4-40.7)
Proportion of people aged 40-49 years (95% CI, %)	34 (29.8-38.4)	29.5 (24-35.7)	37.1 (31.5-43.1)
Proportion of people aged 50-59 years (95% CI, %)	30.2 (26.3-34.5)	33.3 (27.1-40)	28 (23.5-33.1)
Systolic blood pressure (mean and 95% CI, mmHg)	108.9 (107.5-110.3)	116.1 (113.8-118.4)	103.8 (102.3-105.3)
Diastolic blood pressure (mean and 95% CI, mmHg)	71.9 (71-72.8)	74.6 (73.1-76.1)	70 (69-71)
Body mass index (mean and 95% CI, kg/m²)	28.8 (28.3-29.2)	28.3 (27.6-29)	29.1 (28.5-29.7)
Proportion of people with normal weight (95% CI, %)	23.1 (19.5-27.2)	24.5 (19.5-30.4)	22.1 (17.2-27.9)

Proportion of people with overweight (95% CI, %)	41 (36.6-45.6)	45.3 (38.9-52)	37.9 (31.9-44.3)
Proportion of people with obesity (95% CI, %)	35.9 (31.6-40.5)	30.1 (24.1-36.9)	40 (34-46.2)
Total cholesterol (mean and 95% CI, mmol/L)	4.9 (4.8-5)	4.8 (4.6-5)	4.9 (4.8-5.1)
Proportion of smokers (95% CI, %)	11.7 (8.8-15.3)	21.1 (15.8-27.6)	5 (3-8.3)
Proportion of people with diabetes (95% CI, %)	7.1 (5-9.8)	7.8 (4.7-12.6)	6.5 (4.3-9.9)
Laboratory-based CVD risk score (mean and 95% CI, %)	2 (1.8-2.2)	2.6 (2.3-2.9)	1.6 (1.4-1.7)
Non-laboratory-based CVD risk score (mean and 95% CI, %)	2 (1.8-2.1)	2.7 (2.4-3)	1.5 (1.4-1.6)

Absolute cardiovascular risk according to the 2019 WHO cardiovascular risk models

Overall, there were no substantial differences between the mean absolute cardiovascular risk computed with the laboratory-based and non-laboratory-based models (Table 1). The mean absolute cardiovascular risk was 2.0% (95% CI: 1.8%; 2.2%) according to the laboratory-based model, and 2.0% (95% CI: 1.8%; 2.1%) according to the non-laboratory-based model. In both models, the mean absolute cardiovascular risk was higher in men than women. The sensitivity analysis (using the mean of two SBP records) yielded the same findings in the overall sample: 2.0% (95% CI: 1.8; 2.2%) in the laboratory-based-model and 2.0 (95% CI: 1.8; 2.1%) in the non-laboratory-based model.

Mean difference between risk predictions

Overall, the mean difference between the laboratory-based and the non-laboratory-based models was 0.03% (95% CI: -0.03%; 0.10%). According to sex, the mean difference between models was -0.02% (95% CI: -0.14%; 0.09%) in men, and 0.08% (95% CI: 0.01%; 0.15%) in women. According to age, the mean difference between models was -0.07% (95% CI: -0.12%; 0.04%) in people aged 30-39 years, 0.04% (95% CI: -0.08%; 0.15%) in people aged 40-49 years, and 0.17% (95% CI: 0.02%; 0.32%) in people aged 50-59 years. Stratified by BMI categories, the mean difference was 0.10% (95% CI: 0.02%; 0.18%) in people with normal weight, 0.08% (95% CI: 0.00%; 0.17%) in people with overweight, and -0.07% (95% CI: -0.21%; 0.07%) in people with obesity. The mean difference was 0.05% (95% CI: -0.34%; 0.44%) in smokers and 0.03% (95% CI: -0.02%; 0.09%) in non-smokers. According to self-reported diabetes status, mean difference was 2.08% (95% CI: 1.73%; 2.44%) in people with self-reported diabetes, and -0.11% (95% CI: -0.15%; 0.07%) in people without self-reported diabetes. The sensitivity analysis provided similar results across all variables; for example, the largest mean difference was also observed in people with self-reported diabetes (2.09% (95% CI: 1.73; 2.45%).

Bland-Altman plots and limits of agreement.

The limit of agreement was slightly narrower for women (-1.23; 1.39) compared to men (-1.93; 1.89) (Figure 1). The limit of agreement widened with older ages and higher BMI levels; for

example, the limit of agreement was narrower in people aged 30-39 years (-0.72; 0.56) compared to those aged 50-59 years (-1.96; 2.29) (Figure 2), and for people with normal weight (-0.90; 1.09) compared to those with obesity (-2.15; 2.02) (Figure 3). According to smoking and self-reported diabetes status, the limit of agreement was wider in those who had the condition. For example, the limit of agreement in smokers (-3.13; 3.24) was wider compared to non-smokers (-1.25; 1.32) (Figure 4). Similarly, in people with self-reported diabetes (-0.21; 4.37) the limit of agreement was wider compared to people without self-reported diabetes (-1.17; 0.95) (Figure 5). Notably, the sensitivity analysis resulted in similar limits of agreement in all variables (Supplementary figures 3-7).

Agreement by Lin's concordance coefficient correlation

The overall agreement between scores as per the LCCC was 0.87 (95% CI: 0.85; 0.89), and it was virtually the same in men (0.87 (95% CI 0.84; 0.89)) and women (0.85 (95% CI 0.82; 0.87)). Across age groups, the highest agreement was seen in the 30-39 age group (0.87 (95% CI 0.84; 0.9)). Across BMI categories, it was the normal BMI category which had the highest agreement (0.90 (95% CI 0.87; 0.92)). Overall, the lowest agreement values were seen across smokers (0.81 (95% CI 0.72; 0.88)), those aged 40-49 years (0.74 (95% CI 0.67; 0.79)), and those with self-reported diabetes (0.74 (95% CI 0.63; 0.82)) (Table 2). Similar results were seen in the sensitivity analysis (Supplementary table 2).

Table 2. Lin's concordance coefficient correlation showing agreement between laboratory and non-laboratory-based risk models according to the predictors in the 2019 WHO CVD risk models and urban/rural location

Variables	Categories	Lin's concordance coefficient correlation (95% CI)
Sex	Men	0.87 (0.84 - 0.89)
	Women	0.85 (0.82 - 0.87)
Age (years)	30-39	0.87 (0.84 - 0.9)
	40-49	0.74 (0.67 - 0.79)
	50-59	0.83 (0.78 - 0.86)
Body mass index category	Normal	0.9 (0.87 - 0.92)

	Overweight	0.87 (0.85 - 0.9)
	Obese	0.86 (0.82 - 0.89)
Smoking status	Smoker	0.81 (0.72 - 0.88)
	Non-smoker	0.86 (0.84 - 0.88)
Diabetes status	With self-reported diabetes	0.74 (0.63 - 0.82)
	Not with self-reported diabetes	0.91 (0.9 - 0.92)
Urban or rural	Urban	0.88 (0.85 - 0.9)
	Rural	0.85 (0.82 - 0.88)

Categorical agreement

In the overall population, there was a slightly larger number of people categorized as having an absolute CVD risk of 5-9% and 10-19% with the laboratory-based-model compared with the non-laboratory-based model (Supplementary table 3). For example, the laboratory-based model categorized 37 people in the 5-9% CVD risk category, whereas the non-laboratory-based model categorized 26 people. Overall, the agreement between risk categories was substantial ($\kappa=0.62$), and it was better for men ($\kappa=0.70$) compared to women ($\kappa=0.44$) (Supplementary table 4). Of note, the lowest agreement between risk categories was observed among people with self-reported diabetes ($\kappa=0.36$): out of 14 people with self-reported diabetes in the 5-9% CVD risk category following the laboratory-based-model, 4 were placed in the same category following the non-laboratory-based model, as the rest were placed in the 0-5% CVD risk category. The categorical agreement according to all variables is presented in Supplementary tables 3-9.

DISCUSSION

Main findings

In this work, we evaluated the agreement between the CVD risk estimates predicted with the 2019 WHO ten-year laboratory-based and non-laboratory-based models in a nationally-representative sample of Peruvian adults. The mean absolute predicted CVD risk according to both models in the general population was virtually the same. In addition, we found that the limits of agreement between both models increased with a higher CVD risk; for instance, the limits of agreement were

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3 wider in smokers and people with self-reported diabetes. We observed good agreement between
4 the laboratory-based and non-laboratory-based models in terms of clinical significance.[21,23]
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6 These findings suggest that, in a population with a similar profile to that in this study, practitioners
7 could use either the laboratory-based or non-laboratory-based models. Although the difference is
8 very small, careful interpretation could be needed for people with cardiovascular risk factors:
9 obesity, smokers and people with self-reported diabetes, amongst whom the difference was
10 slightly larger than in their peers without these risk factors.
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18 **Public health implications**

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20 We provided insights about the applicability of the 2019 WHO non-laboratory-based model in the
21 Peruvian population. This evidence is relevant in terms of clinical practice and public health in
22 Peru and other similar countries (e.g., Andean Latin America), because it informs whether the
23 predictions based on the laboratory-based and non-laboratory-based CVD risk models are
24 equivalent. If so, either of these models could be used without substantial bias, hence supporting
25 the use of the non-laboratory-based model when blood biomarkers are not available. The latter is
26 of special importance in LMICs like Peru, where laboratory measurements are not always
27 available in primary health centres, which represent >98% of all health care facilities in Peru.[24]
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38 According to the three statistical analyses we implemented (Bland-Altman plots, Lin's
39 concordance coefficient correlations; and kappa statistics), our results suggest that the
40 agreement between the laboratory-based and non-laboratory-based models was appropriate
41 among Peruvians with low CVD risk and younger than 60 years. In other words, our results
42 suggest that the laboratory-based and non-laboratory-based models provide similar predictions
43 and may therefore be used interchangeably as needed, though the profile of our study population
44 ought to be considered when extracting or implementing our findings into clinical practice and
45 public health. Of note, the small number of participants in some variables of interest (e.g., only 44
46 participants had self-reported diabetes) could have explained the broader limits of agreement in
47 our results. Further studies should include a larger number of participants to further confirm
48 whether limits of agreement are wider according to smoking and diabetes status.
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Research in context

The study most comparable to ours evaluated the agreement between the Framingham 10-year CVD risk laboratory and non-laboratory models on a population aged 40-75 years in southern Iran.[23] They found the mean CVD risk following the non-laboratory-based model (9.4%) was higher than the laboratory-based model (6.7%).[23] Additionally, their limits of agreement between both Framingham models in people <60 years old were wider compared to ours in both men (-1.9-1.9 by our estimates versus -2.5%-8.9% by Rezaei et al.[23]) and women (-1.2-1.4 by our estimates versus -2.3%-4.6% by Rezaei et al.[23]). This could be explained by the fact that Rezaei et al.[23] included an older population, which tend to have higher levels of CVD risk factors and therefore a higher absolute CVD risk. As limits of agreement between two models tend to widen with higher CVD risk,[21,23] our limits of agreement would presumably be wider if we had studied a similar population to that of the work by Rezaei et al.[23] The differences between our results could be further explained by the CVD risk score herein used. We used the 2019 WHO CVD risk models,[12] whereas Rezaei et al. used the Framingham risk scores. The Framingham risk score was developed for a more specific population (Caucasians in the US),[25] yet the 2019 WHO CVD risk model was developed and recalibrated for a global use (e.g., those living in LMICs).[12]

The agreement between the 2019 WHO laboratory- and non-laboratory-based model was also explored in the global work convened by the WHO.[12] They applied the two models to WHO STEPS surveys, and compared the proportion of people categorized at different levels of predicted CVD risk.[12] Overall, they found moderate agreement between both models, and their discrepancy was attributed to poor performance of the non-laboratory-based model in people with diabetes.[12] This finding is consistent with our results because we found the widest limits of agreement, the lowest LCCC, and the lowest categorical agreement in people with self-reported diabetes. When possible, it would seem reasonable to use the laboratory-based model in those whose have diabetes.

Potential explanations

In our study, diabetes status was only included in the laboratory-based model (and not in the non-laboratory-based model) and reasonably, we observed the lowest agreement between both models in those with diabetes. That is, in people with diabetes, the non-laboratory-based model underestimated the absolute CVD risk computed following the laboratory-based model. Probably because people with diabetes are already at high CVD risk and the non-laboratory model, we used without diabetes information would underestimate the absolute risk.

Strengths and limitations

To the best of our knowledge, we provided the first evidence of the agreement between the 2019 WHO CVD risk laboratory-based and non-laboratory-based models; furthermore, we leveraged on a nationally-representative survey conducted in a LMIC.[26] Different CVD risk equations have been created but none has proved to produce reliable estimates for LMICs. The 2019 WHO CVD risk charts were adapted for LMICs using extensive datasets for its derivation, recalibration, and validation, which brings them several advantages over previous risk charts. Nonetheless, this study has also limitations. First, our study population was young (30-59 years), mostly women (58%), and with overall low absolute cardiovascular predicted risk. This led to the observation that no one in the study population had an absolute cardiovascular risk $\geq 20\%$. Thus, we could only draw conclusions for people within the low and medium CVD risk range, and with a similar demographic and risk factor profile. We acknowledge that further subgroup analysis could be relevant, for example by diabetes status. However, because of data availability and the reduced number of observations in some groups, this subgroup analyses would be impossible to conduct. Future work in Peru and Latin America should verify our results with a larger, older and more diverse population. Second, as CENAN's survey did not include history of cardiovascular events, we assumed all participants were free of CVD to use the 2019 WHO CVD risk score. This approach could have led to higher absolute cardiovascular risk because people who have had a cardiovascular event (e.g., myocardial infarction) are at higher risk of another cardiovascular event. Nonetheless, considering that our study population was young and therefore with a low incidence of cardiovascular diseases,[3] the proportion of people with history of CVD excluded from the total sample size would have been small; not excluding potentially a small group may

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3 not have altered the overall results. Third, we only used one blood pressure record (the second
4 SBP out of two measurements). Ideally, and following standard protocols recommended by WHO
5 and other international organizations,[27,28] we should have used the average of multiple records
6 having discarded the very first measurement. This was not possible with the available data
7 because they only measured blood pressure twice. Nonetheless, we performed a sensitivity
8 analysis using the mean SBP of the first and second records and had virtually the same results
9 as in our main analysis in which only the second SBP record was used.
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18 **Conclusions**

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20 The absolute cardiovascular predicted risk was similar between the laboratory-based and non-
21 laboratory based 2019 WHO cardiovascular risk models. Pending validation from longitudinal
22 studies, the non-laboratory-based model (instead of the laboratory-based which requires
23 additional resources) could be used in Peruvian population. Nonetheless, it should be noted that
24 the agreement between these models was less clear in people with cardiovascular risk factors:
25 obesity, smokers and people with diabetes. While universal health coverage momentum helps to
26 have laboratory tests in (most) primary care facilities to use the laboratory-based model, it seems
27 reasonable to use the non-laboratory-based model for primary prevention of CVD following the
28 risk stratification approach.
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FIGURES

Figure 1. Bland Altman plots showing agreement between laboratory and non-laboratory-based risk scores according to sex.

Figure 2. Bland Altman plots showing agreement between laboratory and non-laboratory-based risk scores according to age groups.

Figure 3. Bland Altman plots showing agreement between laboratory and non-laboratory-based risk scores according to body mass index categories

Figure 4. Bland Altman plots showing agreement between laboratory and non-laboratory-based risk scores according to smoking status.

Figure 5. Bland Altman plots showing agreement between laboratory and non-laboratory-based risk scores according to self-reported diabetes status.

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3 **Contributorship:** RMC-L conceived the idea with WCG-V. WCG-V conducted the analysis with
4 support from GAQ-V and FFV-R. WCG-V wrote the first draft of the manuscript with support from
5 GAQ-V, FFV-R, AB-O and RMC-L. All authors provided relevant scientific contribution. All authors
6 approved the submitted version.
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9

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12 to publish. The authors are collectively responsible for the accuracy of the data. The arguments
13 and opinions in this work are those of the authors alone, and do not represent the position of the
14 institutions to which they belong.
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20 **Competing of interests:** None to declare.
21

22 **Ethics:** This study used de-identified nationally-representative survey data that can be requested
23 from the CENAN and is used for independent analyses.[26] Authors had no access to the
24 participants' personal information. Therefore, this work was deemed as of minimal risk and we
25 did not seek approval by an Ethics Committee to conduct the analysis.
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30 **Data sharing:** Datasets available upon request from the CENAN.
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32 **Word count:** 4,220
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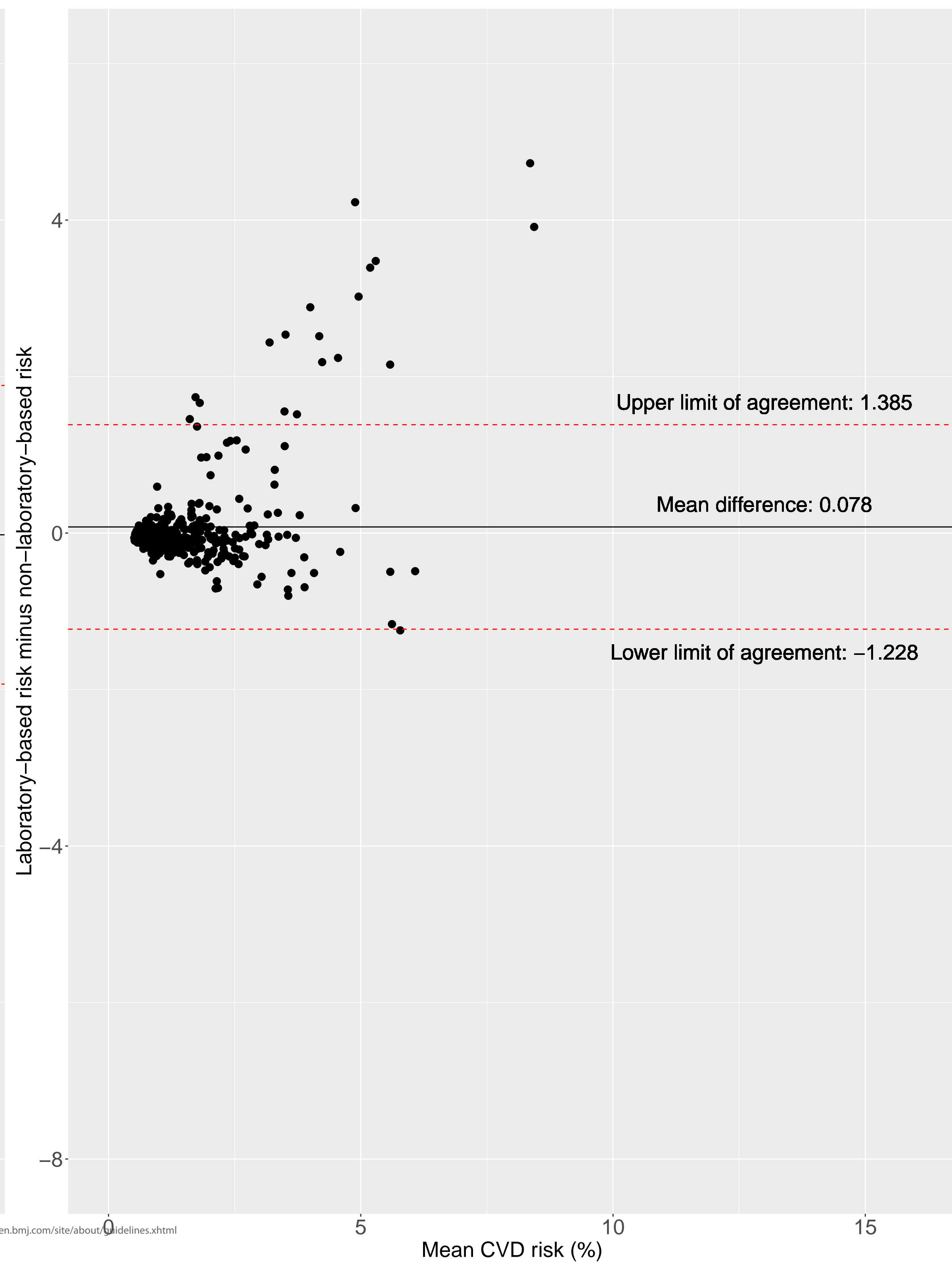
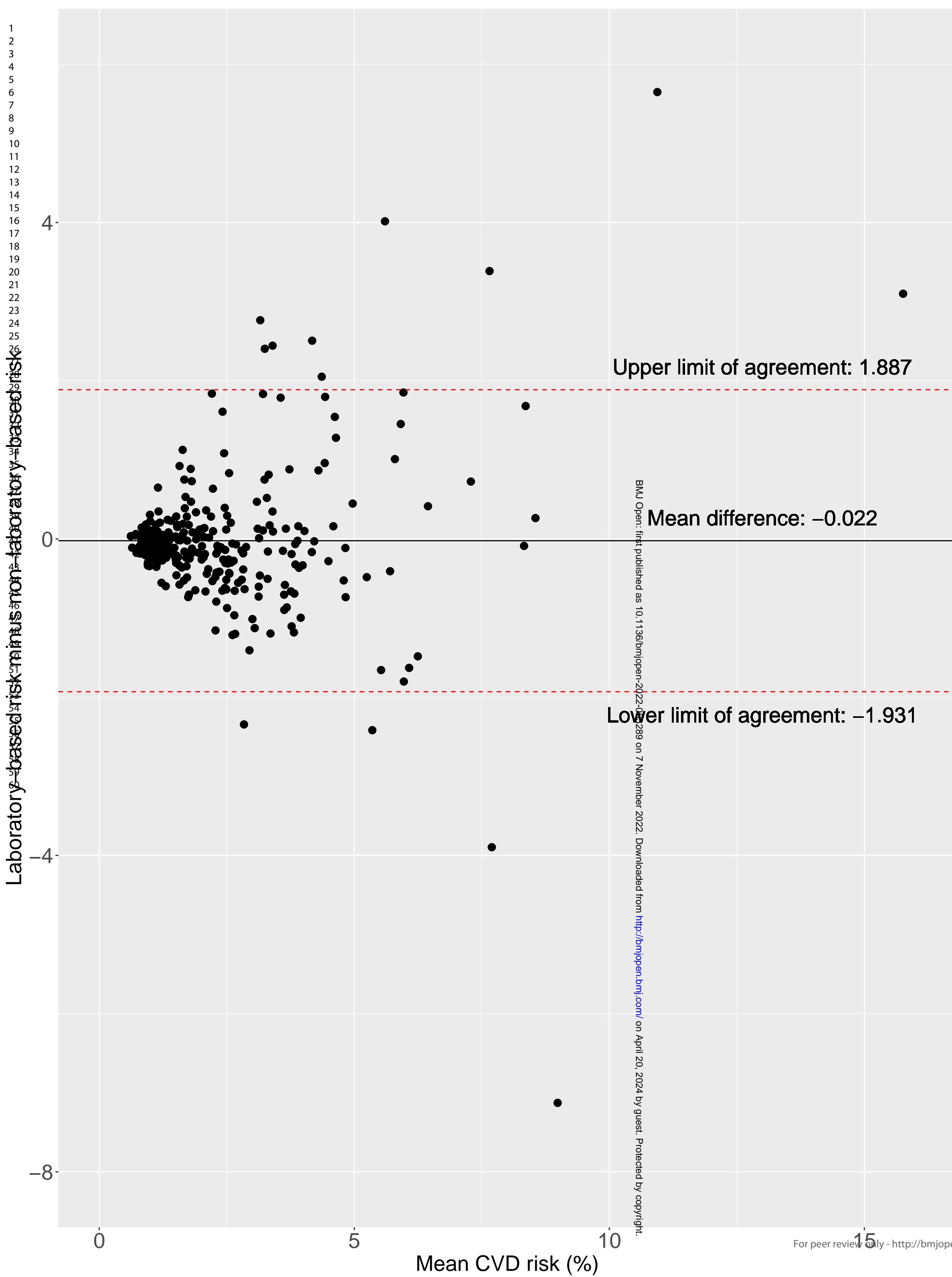
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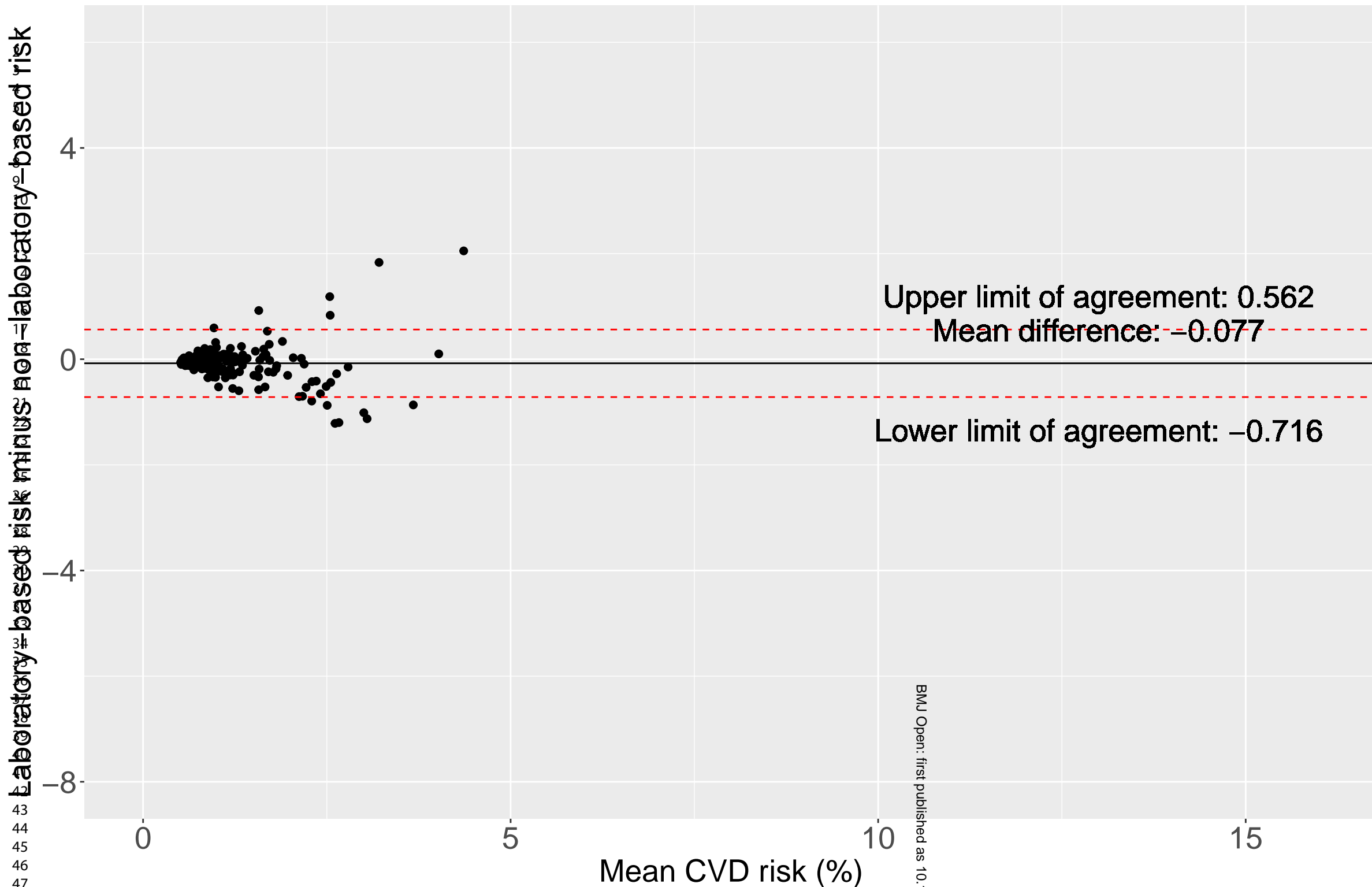
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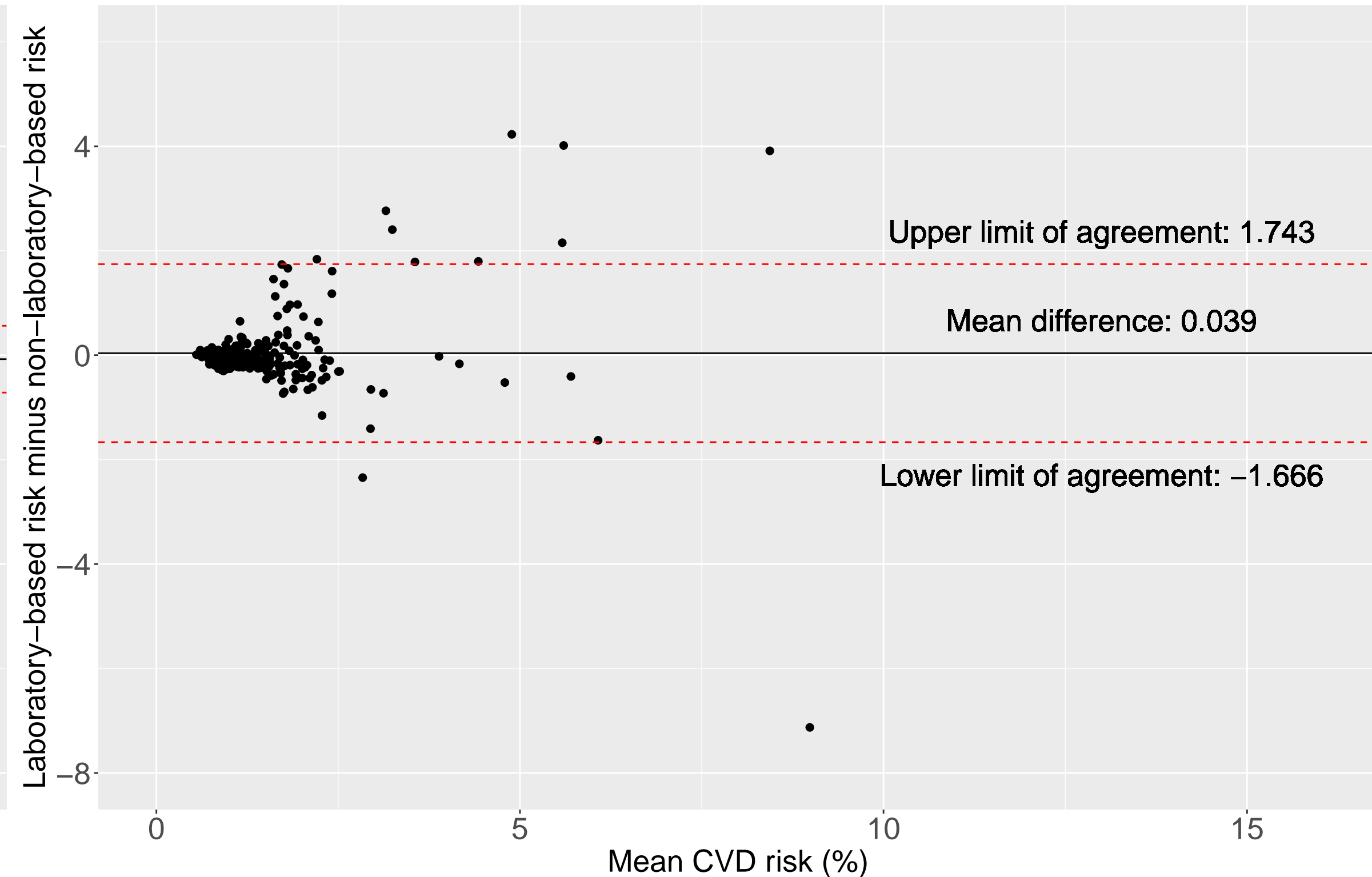
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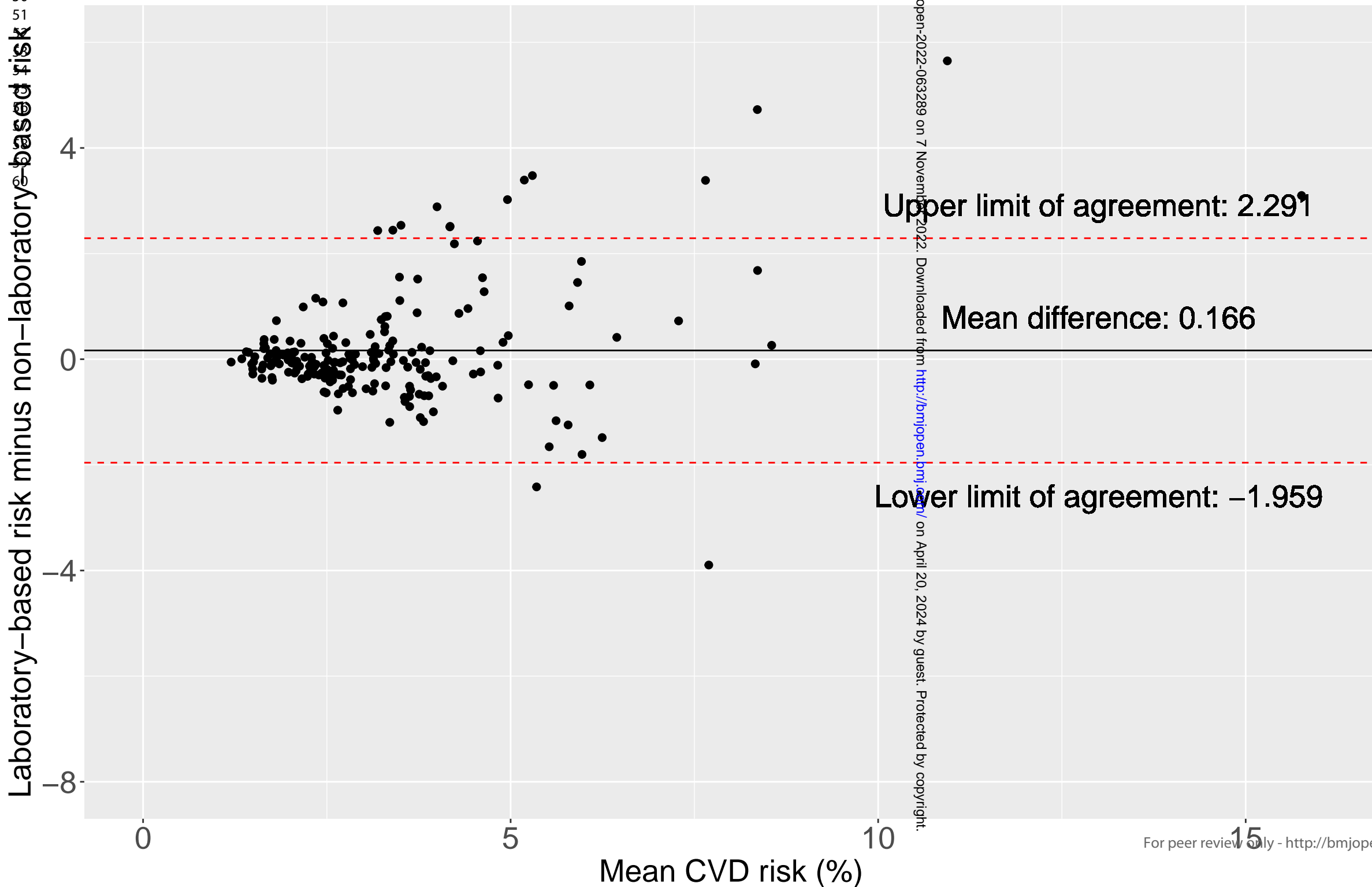
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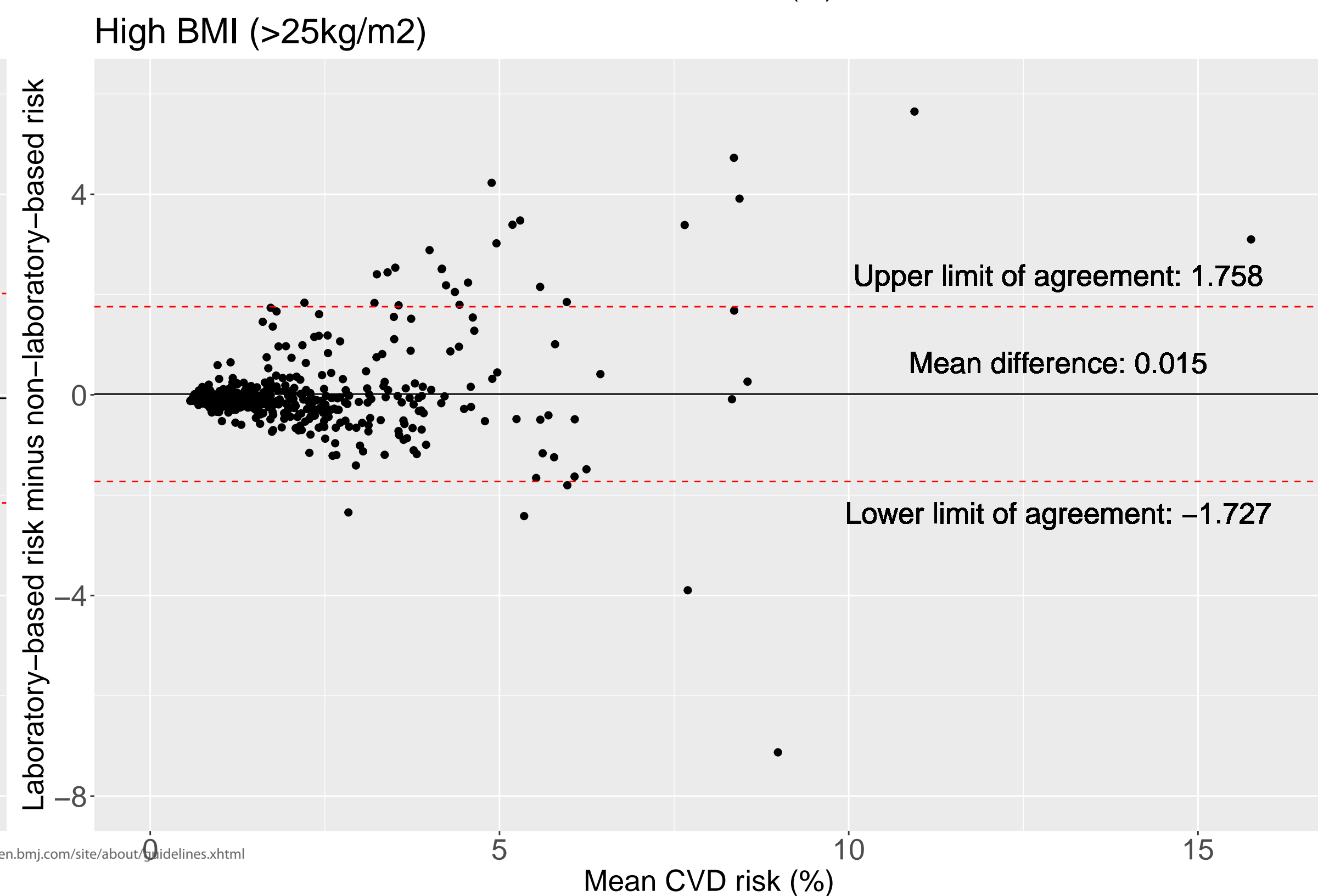
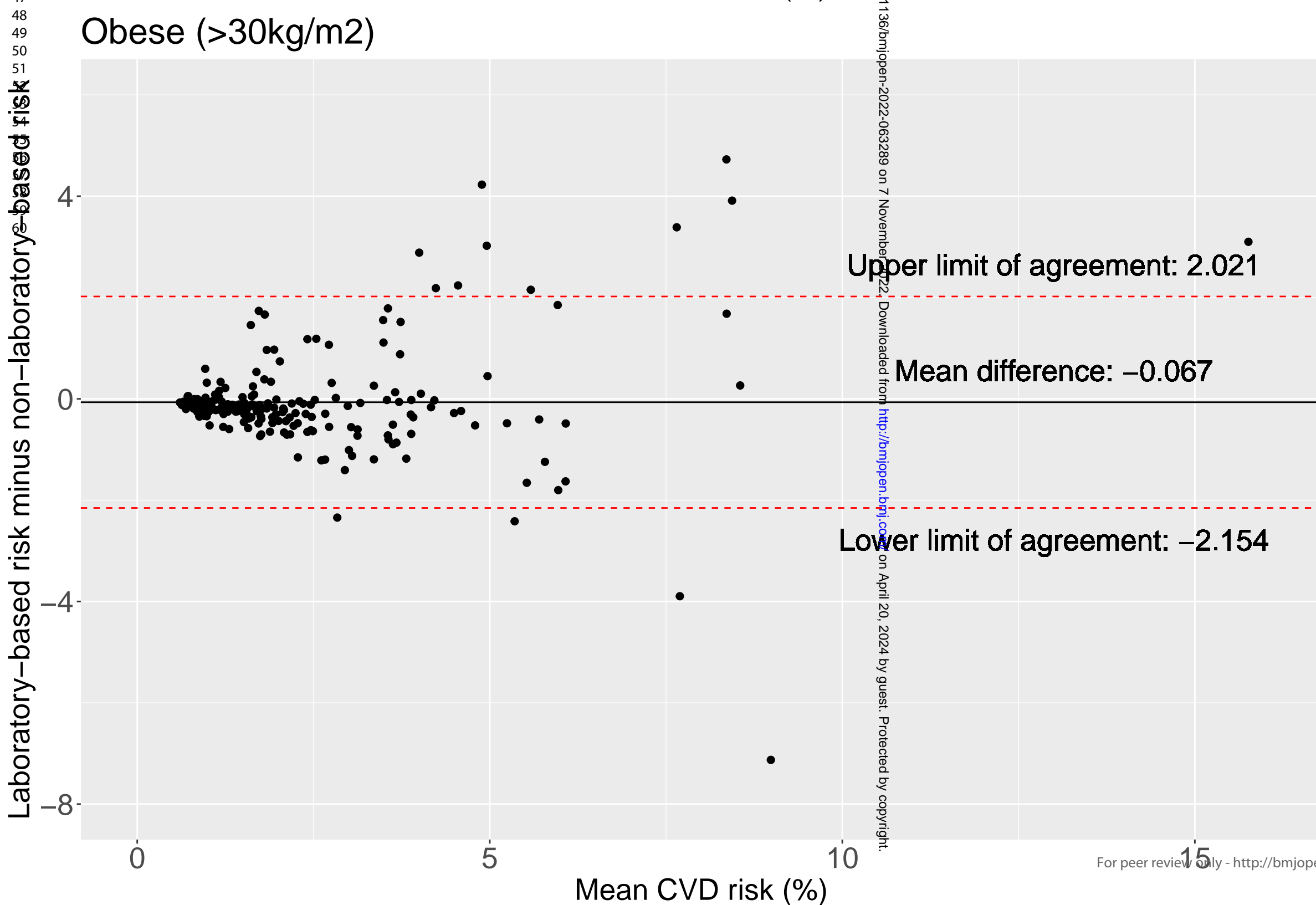
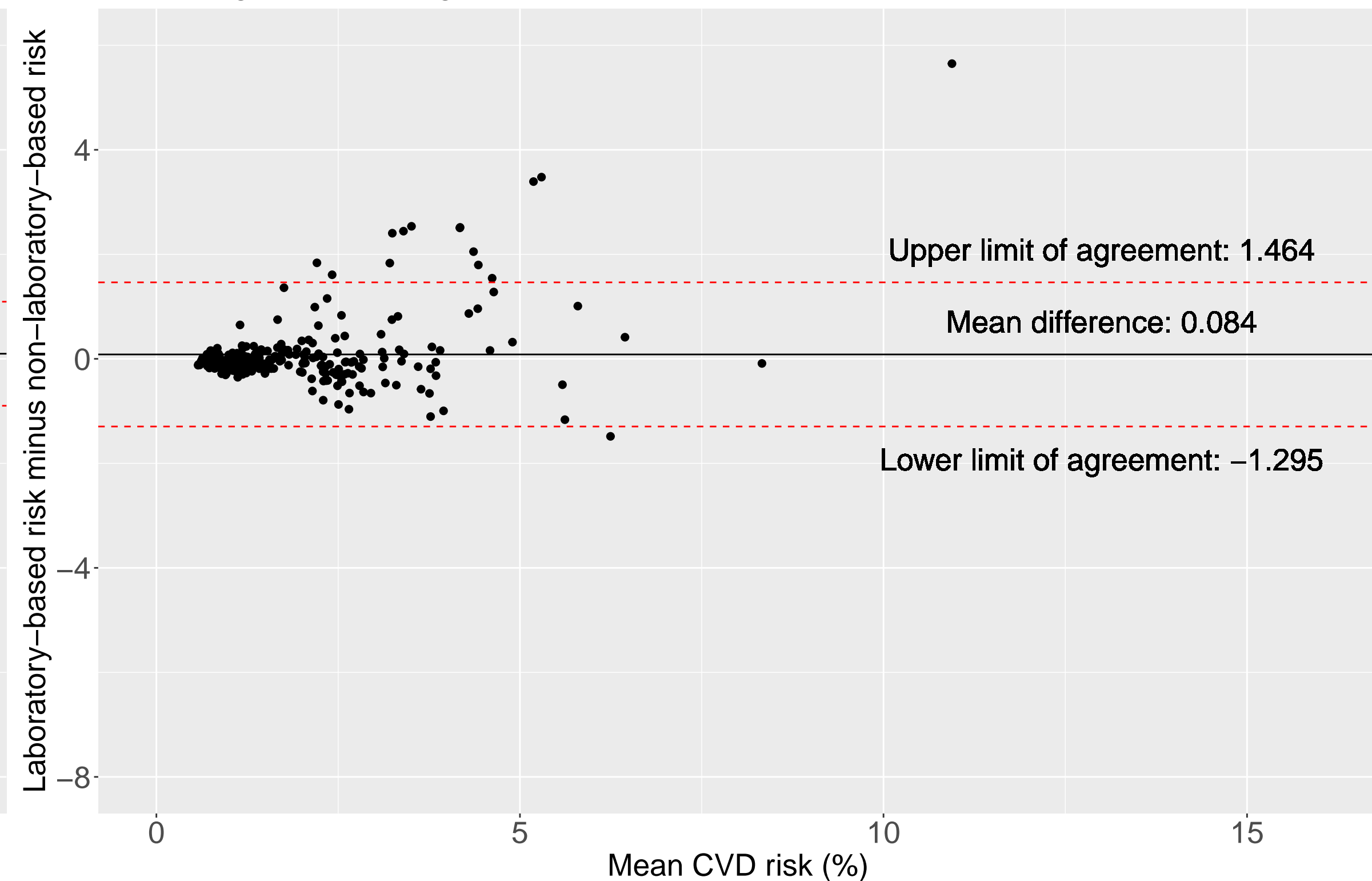
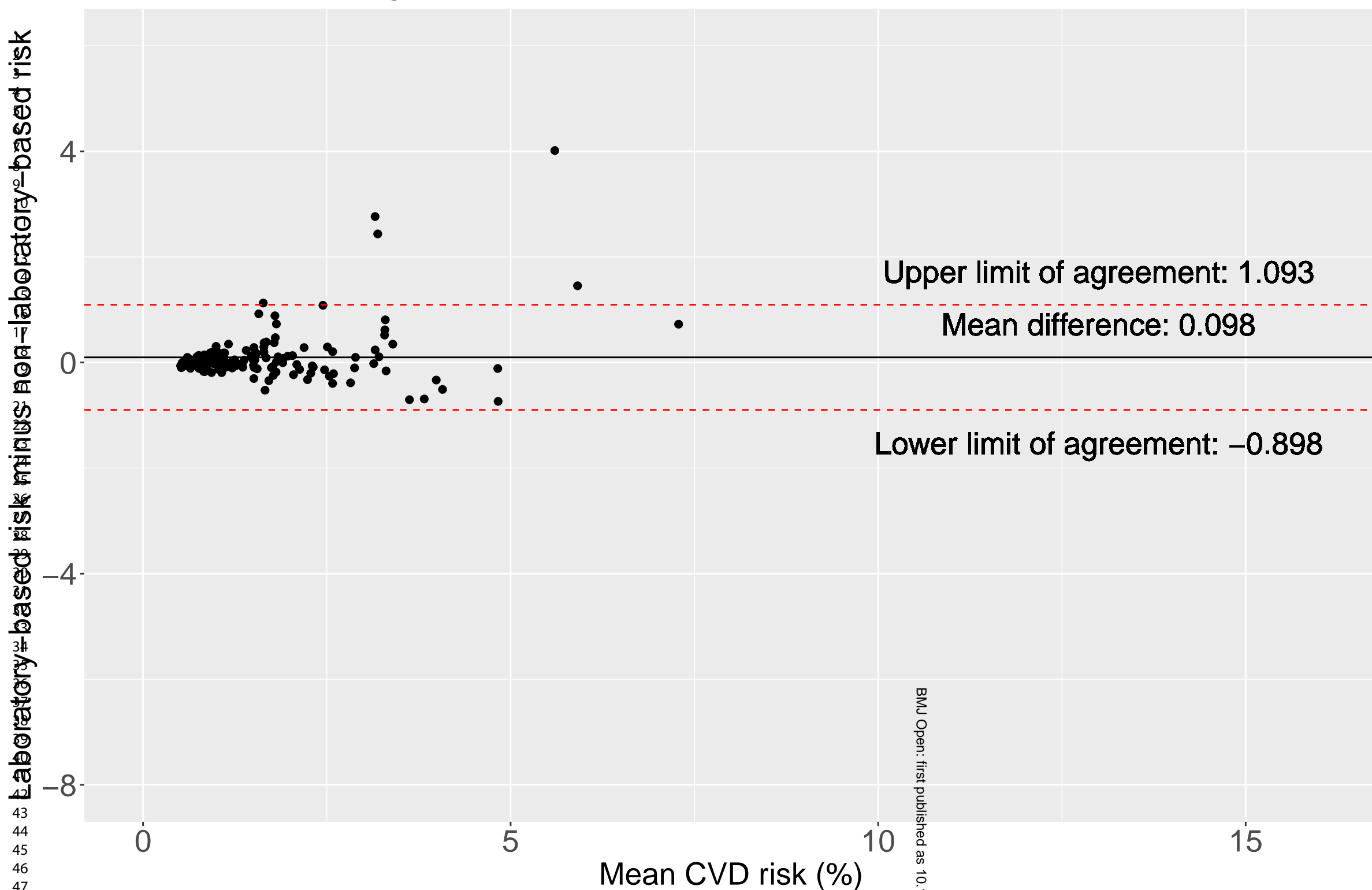


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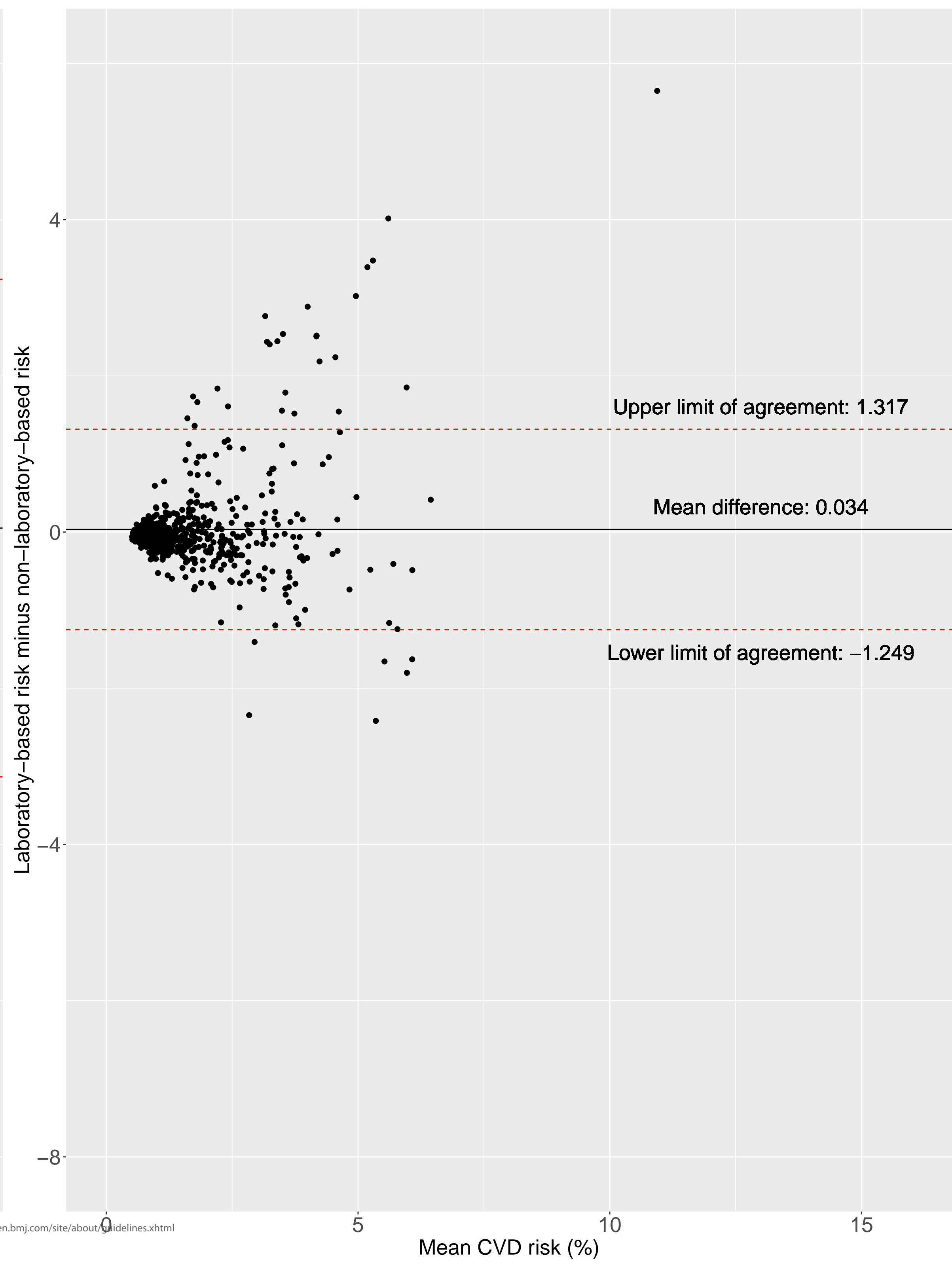
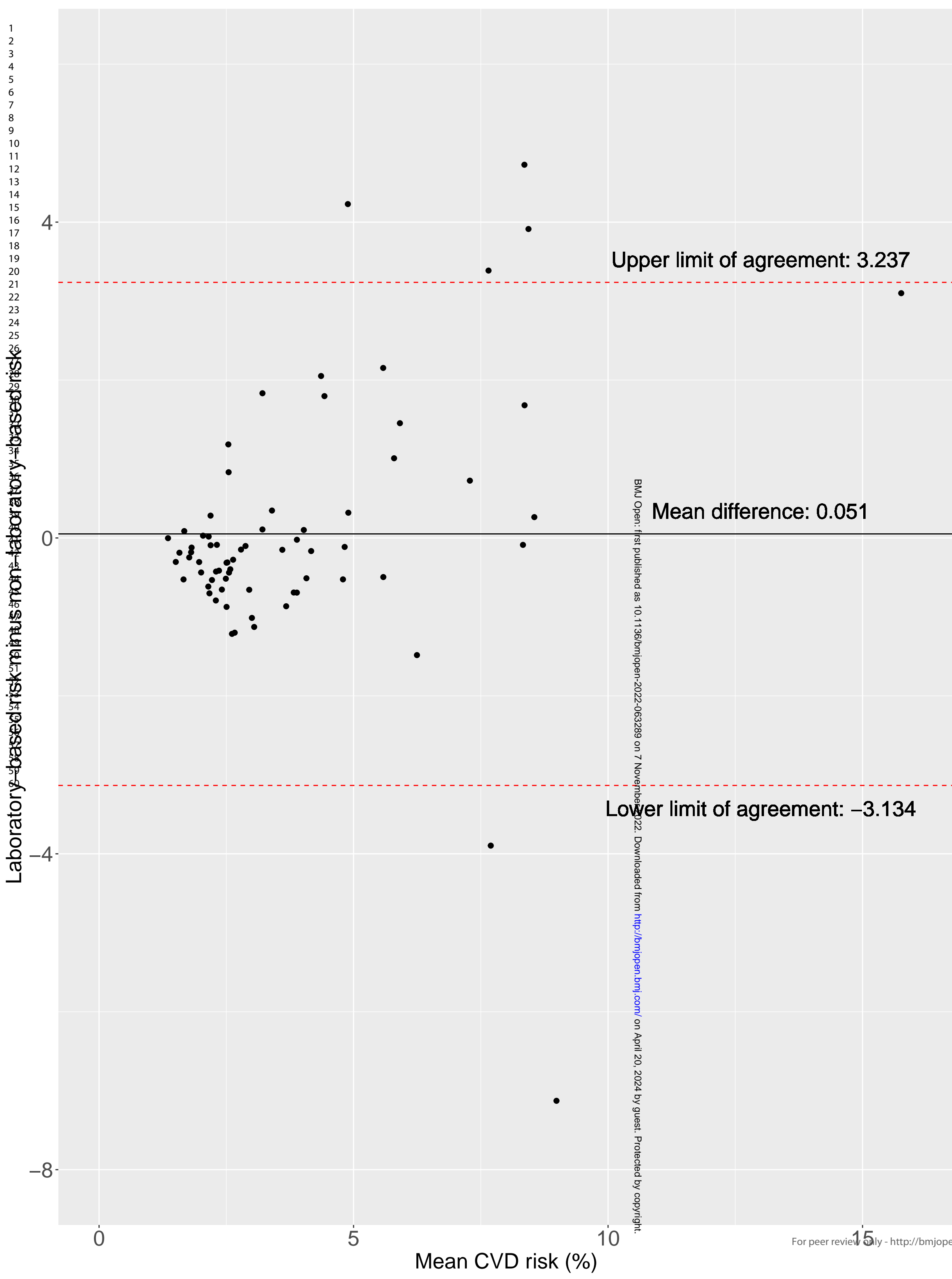


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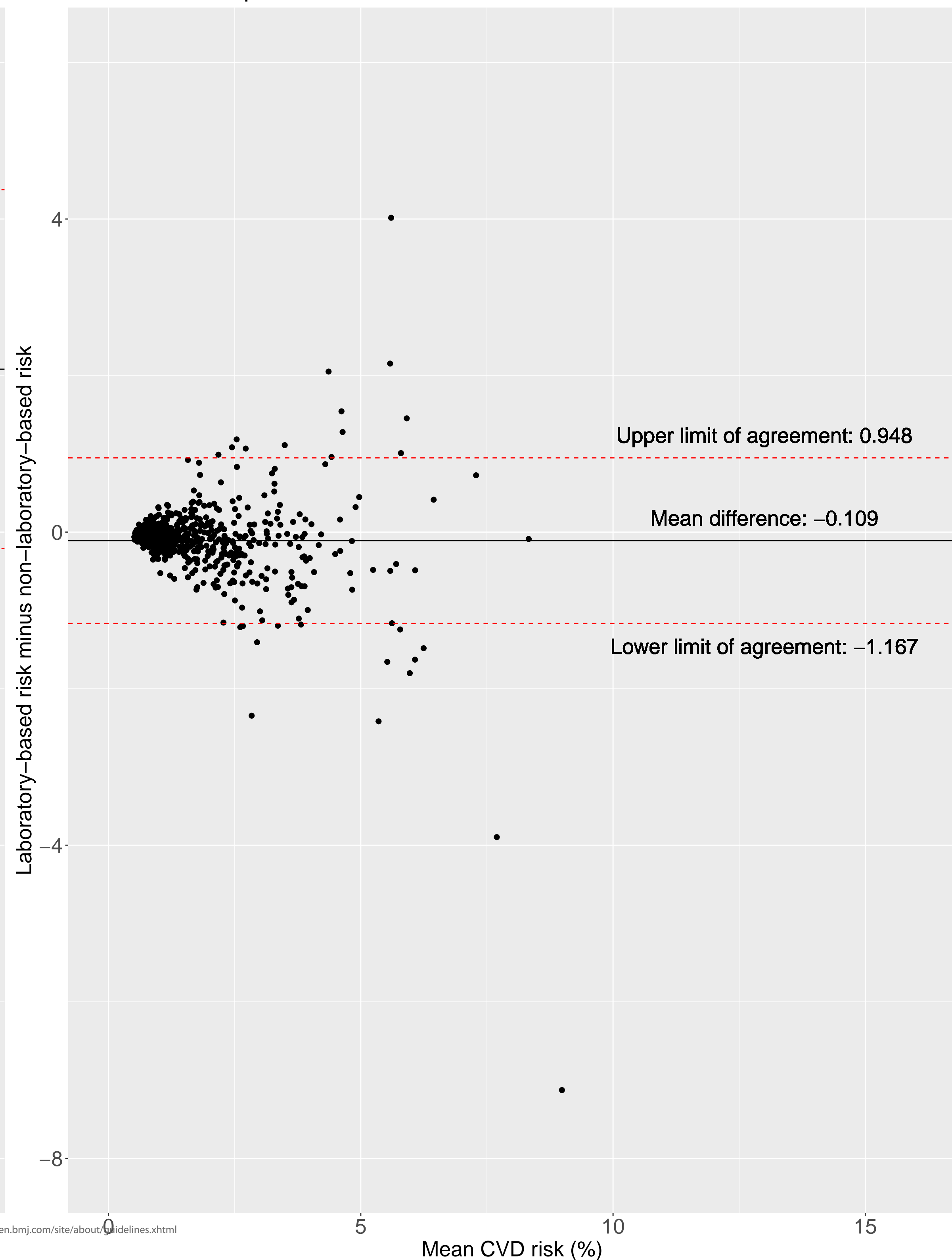
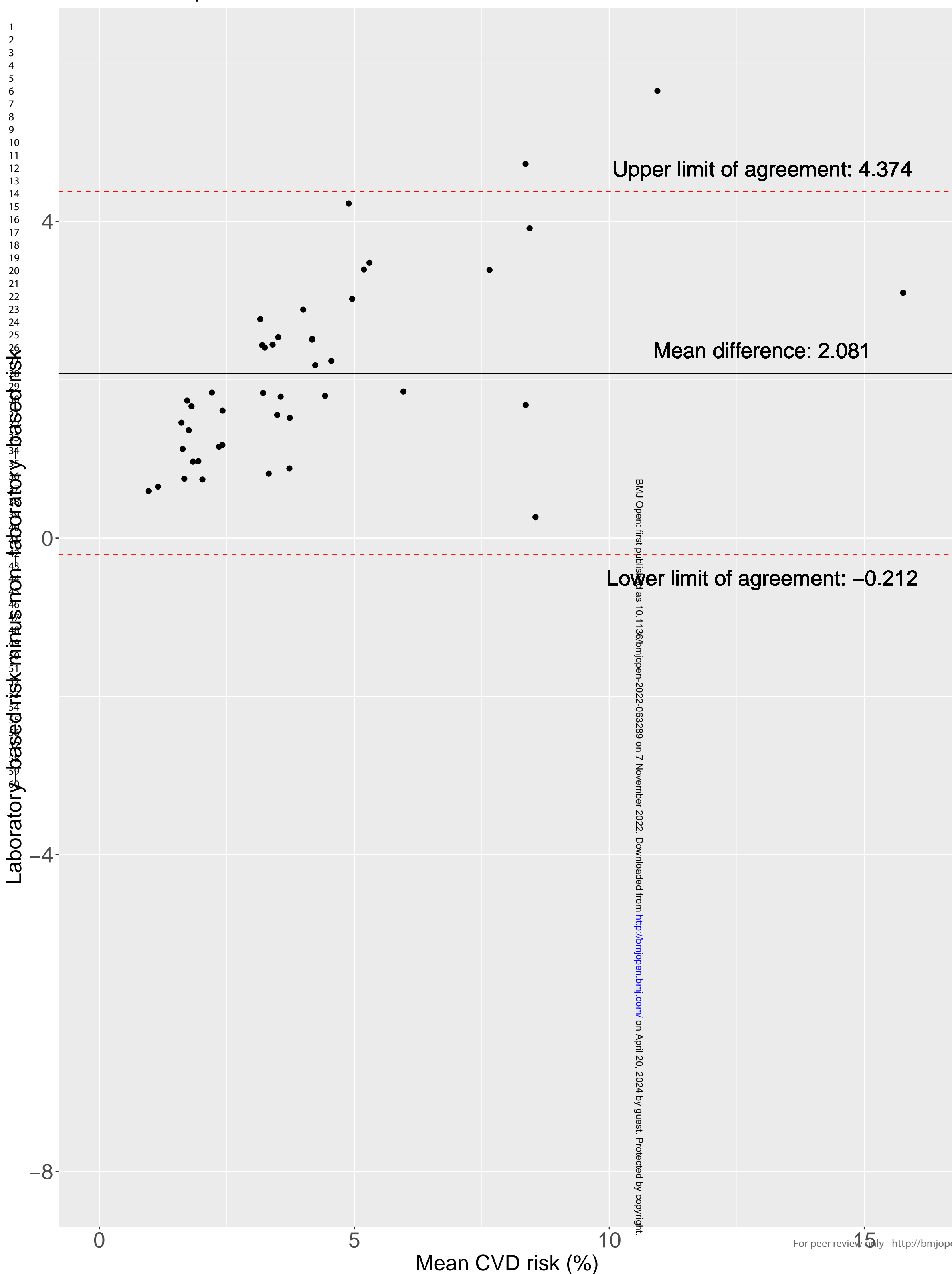
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3 **Agreement between the laboratory- and non-laboratory-based WHO cardiovascular risk charts:**
4 **a cross-sectional analysis of a national health survey in Peru**
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31	<i>and non-laboratory-based risk scores by smoking status</i>	6
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33	<i>Supplementary Table 8. Kappa statistics showing agreement between laboratory</i>	
34	<i>and non-laboratory-based risk scores by diabetes status</i>	6
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36	<i>Supplementary Table 9. Kappa statistics showing agreement between laboratory</i>	
37	<i>and non-laboratory-based risk scores by urban/rural location</i>	7
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39	<i>Supplementary Figure 3. Sensitivity Analysis: Bland Altman plots showing</i>	
40	<i>agreement between laboratory and non-laboratory-based risk scores according to</i>	
41	<i>sex.</i>	7
42		
43	<i>Supplementary Figure 4. Sensitivity Analysis: Bland Altman plots showing</i>	
44	<i>agreement between laboratory and non-laboratory-based risk scores according to</i>	
45	<i>age groups</i>	8
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47	<i>Supplementary Figure 5. Sensitivity Analysis: Bland Altman plots showing</i>	
48	<i>agreement between laboratory and non-laboratory-based risk scores according to</i>	
49	<i>body mass index categories.</i>	9
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51	<i>Supplementary Figure 6. Sensitivity Analysis: Bland Altman plots showing</i>	
52	<i>agreement between laboratory and non-laboratory-based risk scores according to</i>	
53	<i>smoking status.</i>	10
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55	<i>Supplementary Figure 7. Sensitivity Analysis: Bland Altman plots showing</i>	
56	<i>agreement between laboratory and non-laboratory-based risk scores according to</i>	
57	<i>self-reported diabetes status</i>	11
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Supplementary Figure 1. Formula to compute the survey sample size

$$n_h = \frac{N_h Z^2 P_h Q_h}{(N_h - 1) d^2 + Z^2 P_h Q_h} * TNR$$

Where:

N_h : number of people within an age group in the “h” conglomerate

n_h : number of people in the sample within an age group in the “h” conglomerate

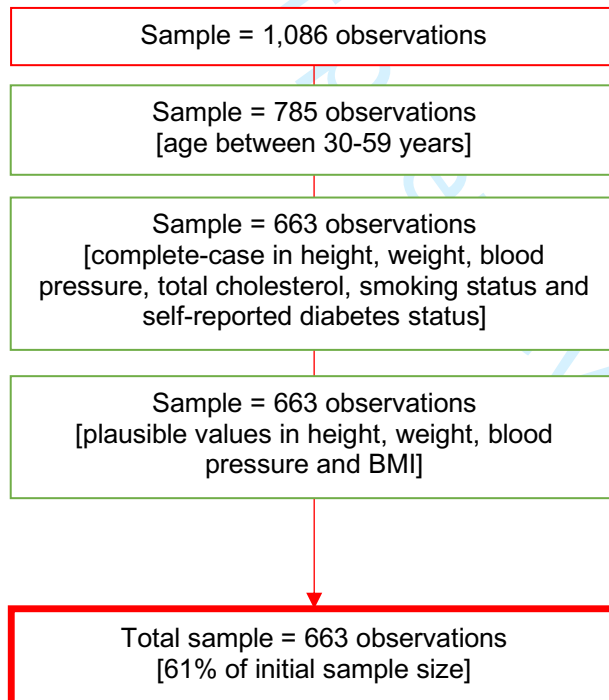
d: error margin assumed in the estimation of P_h

Z: 95% confidence level

TNR (*Tasa de No Respuesta* in Spanish): Expected refusal rate

P_h : prevalence of overweight in adults in the “h” conglomerate

Supplementary Figure 2. Flowchart of data cleaning and inclusion criteria



Supplementary Table 1. Summary statistics of the first and second systolic blood pressure (SBP) records in the overall sample

	First SBP record	Second SBP record
Minimum value	78	75
1st quartile	101	99
Median	109	108
Mean	110.33	109.54
3rd quartile	118	118
Maximum value	203	198

Supplementary Table 2. Sensitivity Analysis: Lin's concordance coefficient correlation showing agreement between laboratory and non-laboratory-based risk models according to the predictors in the 2019 WHO CVD risk models and urban/rural location

Variables	Categories	Lin's concordance coefficient correlation (95% CI)
Sex	Men	0.87 (0.84 - 0.89)
Sex	Women	0.85 (0.83 - 0.87)
Age (years)	30-39	0.87 (0.84 - 0.9)
Age (years)	40-49	0.74 (0.68 - 0.79)
Age (years)	50-59	0.83 (0.78 - 0.87)
Body mass index category	Normal	0.9 (0.87 - 0.92)
Body mass index category	Overweight	0.87 (0.85 - 0.9)
Body mass index category	Obese	0.86 (0.82 - 0.89)
Smoking status	Smoker	0.82 (0.73 - 0.88)
Smoking status	Non-smoker	0.86 (0.84 - 0.88)
Diabetes status	With self-reported diabetes	0.74 (0.63 - 0.82)
Diabetes status	Not with self-reported diabetes	0.91 (0.9 - 0.92)
Urban or rural	Urban	0.88 (0.86 - 0.9)
Urban or rural	Rural	0.86 (0.82 - 0.88)

Supplementary Table 3. Kappa statistics showing agreement between laboratory and non-laboratory-based risk scores in the overall sample

Laboratory-based-risk category	Non-laboratory-based-risk category			kappa
	0-5	5-9	10-19	
0-5	618	4	0	0.62
5-9	17	19	1	
10-19	0	3	1	

Supplementary Table 4. Kappa statistics showing agreement between laboratory and non-laboratory-based risk scores by sex

Laboratory-based-risk category	Sex	Non-laboratory-based-risk category			kappa
		0-5	5-9	10-19	
0-5	Men	251	4	0	0.7
5-9	Men	7	15	1	
10-19	Men	0	1	1	
0-5	Women	367	0	0	0.44
5-9	Women	10	4	0	
10-19	Women	0	2	0	

Supplementary Table 5. Kappa statistics showing agreement between laboratory and non-laboratory-based risk scores by age groups

Laboratory-based-risk category	Age group	Non-laboratory-based-risk category			kappa
		0-5	5-9	10-19	
0-5	30-39	236	0	0	0
5-9	30-39	1	0	0	
10-19	30-39	0	0	0	
0-5	40-49	216	1	0	0.45
5-9	40-49	4	2	1	
10-19	40-49	0	1	0	
0-5	50-59	166	3	0	0.65
5-9	50-59	12	17	0	
10-19	50-59	0	2	1	

Supplementary Table 6. Kappa statistics showing agreement between laboratory and non-laboratory-based risk scores by body mass index categories

Laboratory-based-risk category	Body mass index category	Non-laboratory-based-risk category			kappa
		0-5	5-9	10-19	
0-5	Normal	161	1	0	0.66
5-9	Normal	1	2	0	
10-19	Normal	0	0	0	
0-5	Obese	202	3	0	0.65
5-9	Obese	7	11	1	
10-19	Obese	0	2	1	
0-5	Overweight	255	0	0	0.55
5-9	Overweight	9	6	0	
10-19	Overweight	0	1	0	

Supplementary Table 7. Kappa statistics showing agreement between laboratory and non-laboratory-based risk scores by smoking status

Laboratory-based-risk category	Smoking status	Non-laboratory-based-risk category			kappa
		0-5	5-9	10-19	
0-5	Non-smoker	568	3	0	0.53
5-9	Non-smoker	12	9	0	
10-19	Non-smoker	0	1	0	
0-5	Smoker	50	1	0	0.67
5-9	Smoker	5	10	1	
10-19	Smoker	0	2	1	

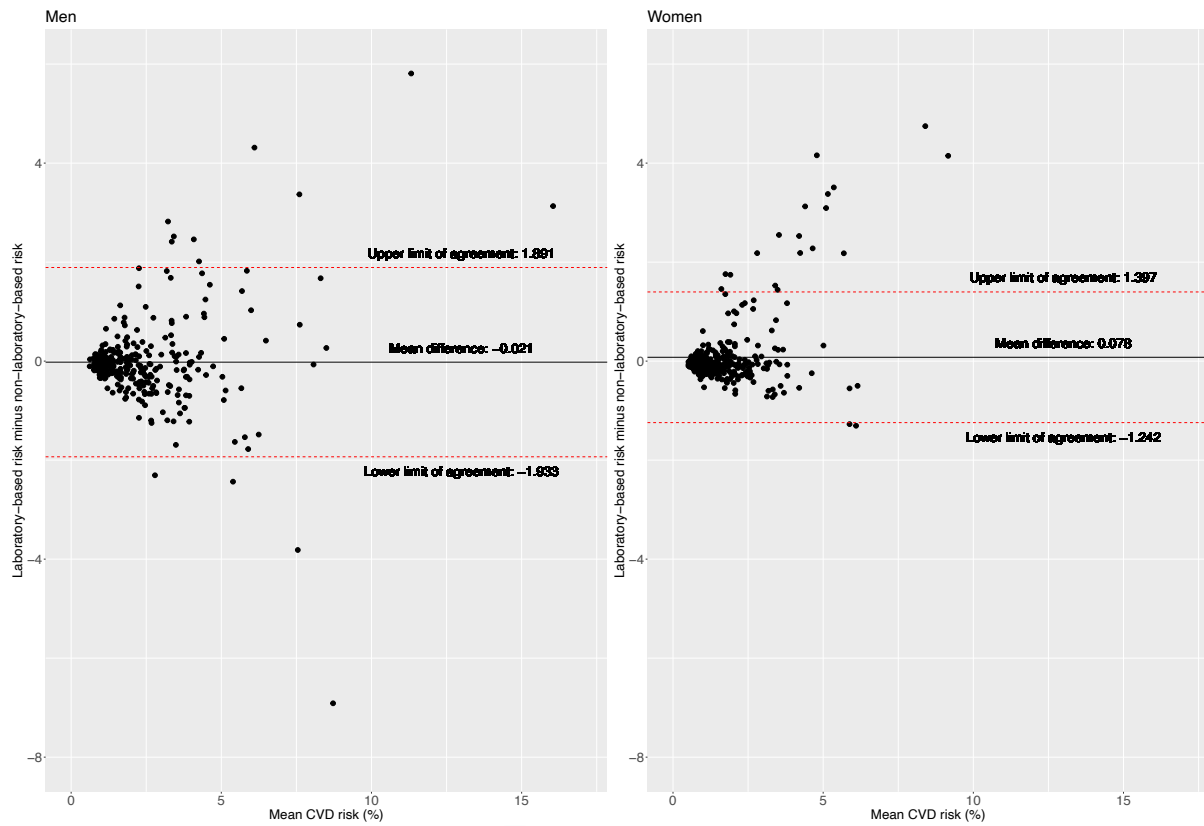
Supplementary Table 8. Kappa statistics showing agreement between laboratory and non-laboratory-based risk scores by diabetes status

Laboratory-based-risk category	Diabetes status	Non-laboratory-based-risk category			kappa
		0-5	5-9	10-19	
0-5	Not with self-reported diabetes	592	4	0	0.71
5-9	Not with self-reported diabetes	7	15	1	
10-19	Not with self-reported diabetes	0	0	0	
0-5	With self-reported diabetes	26	0	0	0.36
5-9	With self-reported diabetes	10	4	0	
10-19	With self-reported diabetes	0	3	1	

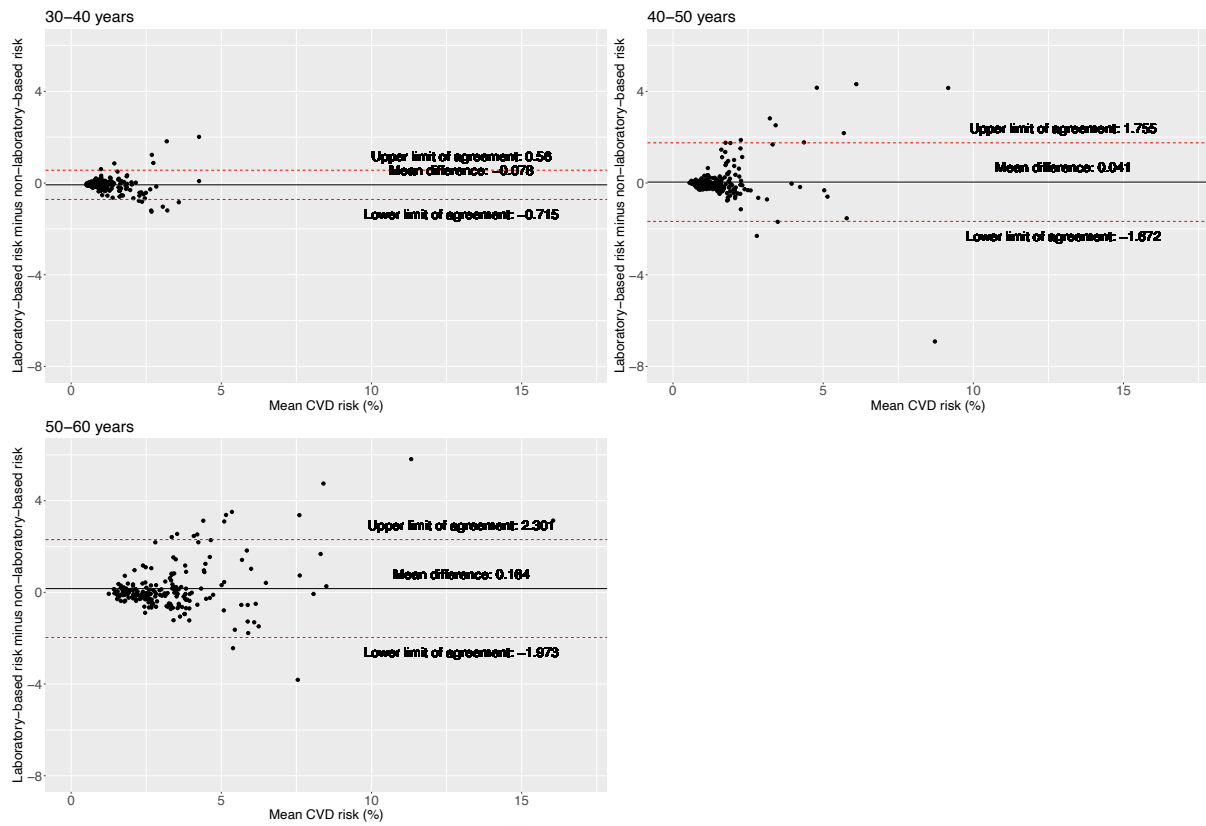
Supplementary Table 9. Kappa statistics showing agreement between laboratory and non-laboratory-based risk scores by urban/rural location

Laboratory-based-risk category	Urban/Rural	Non-laboratory-based-risk category			kappa
		0-5	5-9	10-19	
0-5	Rural	224	1	0	0.53
5-9	Rural	4	3	0	
10-19	Rural	0	1	0	
0-5	Urban	394	3	0	0.64
5-9	Urban	13	16	1	
10-19	Urban	0	2	1	

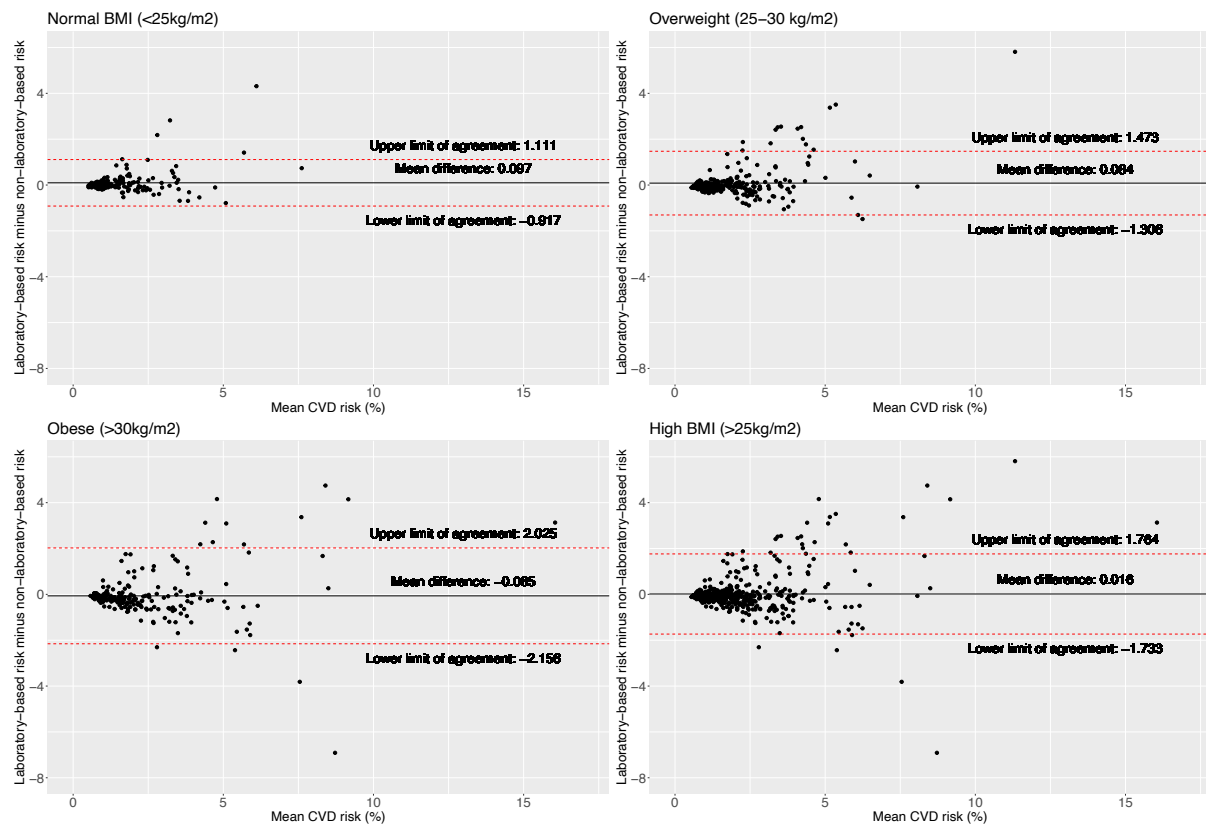
Supplementary Figure 3. Sensitivity Analysis: Bland Altman plots showing agreement between laboratory and non-laboratory-based risk scores according to sex.



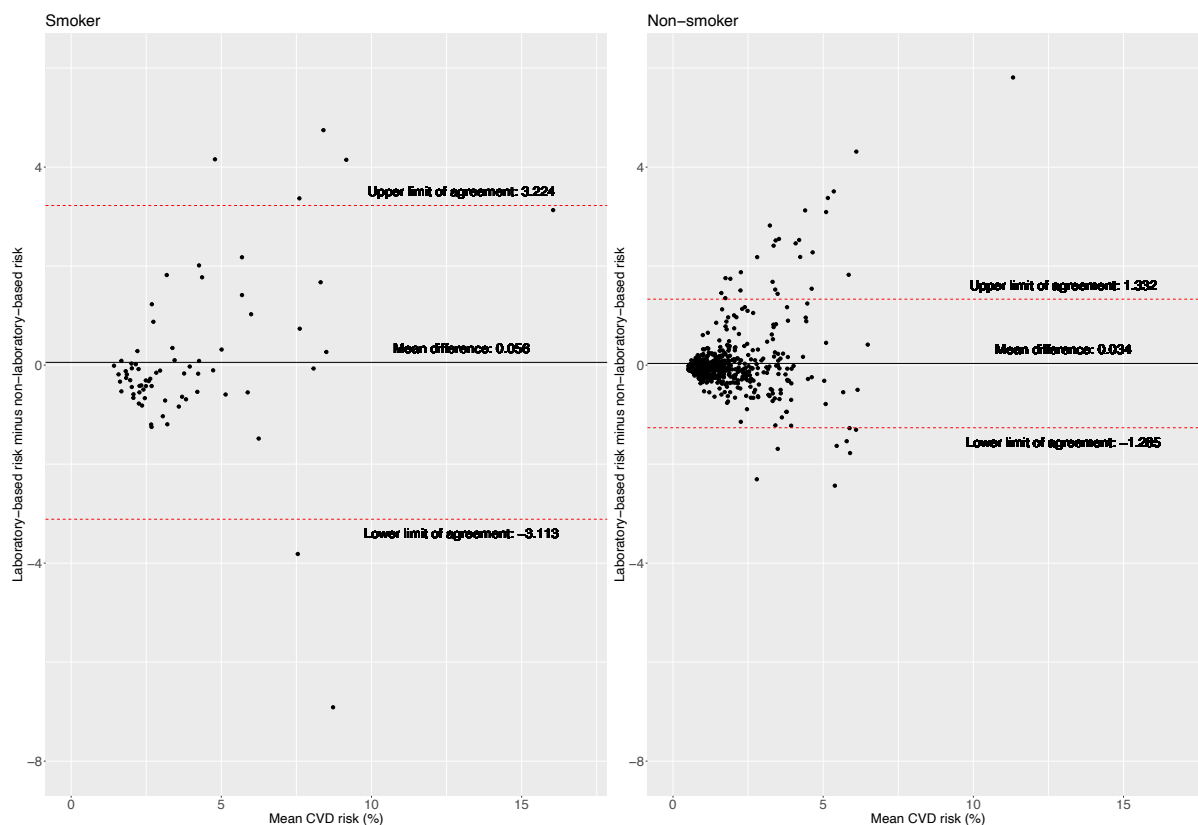
Supplementary Figure 4. Sensitivity Analysis: Bland Altman plots showing agreement between laboratory and non-laboratory-based risk scores according to age groups.



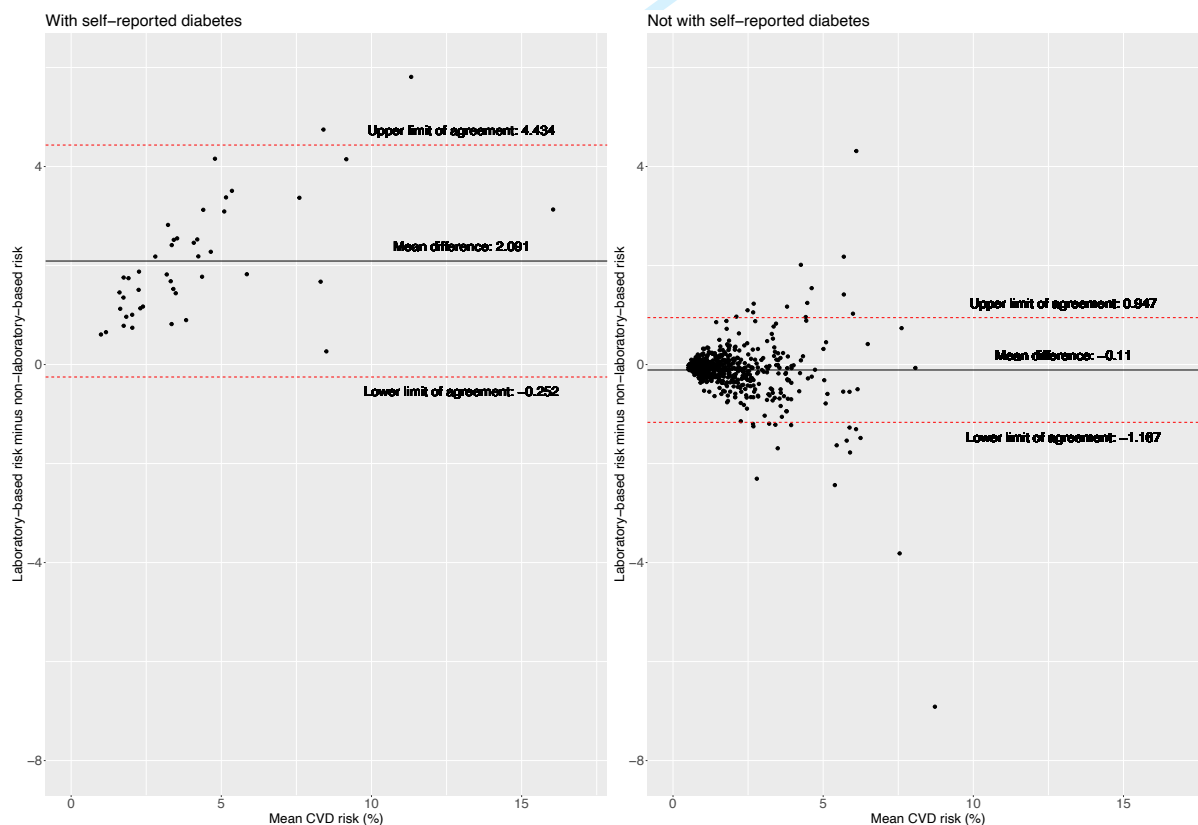
Supplementary Figure 5. Sensitivity Analysis: Bland Altman plots showing agreement between laboratory and non-laboratory-based risk scores according to body mass index categories.



Supplementary Figure 6. Sensitivity Analysis: Bland Altman plots showing agreement between laboratory and non-laboratory-based risk scores according to smoking status.



Supplementary Figure 7. Sensitivity Analysis: Bland Altman plots showing agreement between laboratory and non-laboratory-based risk scores according to self-reported diabetes status.



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For peer review only

STROBE Statement—Checklist of items that should be included in reports of *cross-sectional studies*

	Item No	Recommendation	Page No
Title and abstract	1	(a) Indicate the study's design with a commonly used term in the title or the abstract	3
		(b) Provide in the abstract an informative and balanced summary of what was done and what was found	3
Introduction			
Background/rationale	2	Explain the scientific background and rationale for the investigation being reported	4
Objectives	3	State specific objectives, including any prespecified hypotheses	5
Methods			
Study design	4	Present key elements of study design early in the paper	5
Setting	5	Describe the setting, locations, and relevant dates, including periods of recruitment, exposure, follow-up, and data collection	5-6
Participants	6	(a) Give the eligibility criteria, and the sources and methods of selection of participants	5-6
Variables	7	Clearly define all outcomes, exposures, predictors, potential confounders, and effect modifiers. Give diagnostic criteria, if applicable	6-7
Data sources/ measurement	8*	For each variable of interest, give sources of data and details of methods of assessment (measurement). Describe comparability of assessment methods if there is more than one group	6-7
Bias	9	Describe any efforts to address potential sources of bias	6
Study size	10	Explain how the study size was arrived at	5-6
Quantitative variables	11	Explain how quantitative variables were handled in the analyses. If applicable, describe which groupings were chosen and why	6-7
Statistical methods	12	(a) Describe all statistical methods, including those used to control for confounding	7-8
		(b) Describe any methods used to examine subgroups and interactions	7-8
		(c) Explain how missing data were addressed	5-6
		(d) If applicable, describe analytical methods taking account of sampling strategy	8
		(e) Describe any sensitivity analyses	6
Results			
Participants	13*	(a) Report numbers of individuals at each stage of study—eg numbers potentially eligible, examined for eligibility, confirmed eligible, included in the study, completing follow-up, and analysed	8
		(b) Give reasons for non-participation at each stage	Supp Fig 1
		(c) Consider use of a flow diagram	5
Descriptive data	14*	(a) Give characteristics of study participants (eg demographic, clinical, social) and information on exposures and potential confounders	8
		(b) Indicate number of participants with missing data for each variable of interest	Supp Fig 1
Outcome data	15*	Report numbers of outcome events or summary measures	

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Main results	16	(a) Give unadjusted estimates and, if applicable, confounder-adjusted estimates and their precision (eg, 95% confidence interval). Make clear which confounders were adjusted for and why they were included	8-10
		(b) Report category boundaries when continuous variables were categorized	8-10
		(c) If relevant, consider translating estimates of relative risk into absolute risk for a meaningful time period	-
Other analyses	17	Report other analyses done—eg analyses of subgroups and interactions, and sensitivity analyses	10
Discussion			
Key results	18	Summarise key results with reference to study objectives	11
Limitations	19	Discuss limitations of the study, taking into account sources of potential bias or imprecision. Discuss both direction and magnitude of any potential bias	13
Interpretation	20	Give a cautious overall interpretation of results considering objectives, limitations, multiplicity of analyses, results from similar studies, and other relevant evidence	11-12
Generalisability	21	Discuss the generalisability (external validity) of the study results	11-12
Other information			
Funding	22	Give the source of funding and the role of the funders for the present study and, if applicable, for the original study on which the present article is based	2

*Give information separately for exposed and unexposed groups.

Note: An Explanation and Elaboration article discusses each checklist item and gives methodological background and published examples of transparent reporting. The STROBE checklist is best used in conjunction with this article (freely available on the Web sites of PLoS Medicine at <http://www.plosmedicine.org/>, Annals of Internal Medicine at <http://www.annals.org/>, and Epidemiology at <http://www.epidem.com/>). Information on the STROBE Initiative is available at www.strobe-statement.org.