

Association of insulin resistance with chronic diarrhea: based on the NHANES database 2005-2010

Keywords

chronic diarrhea, insulin resistance, NHANES

Abstract

Introduction

The evidence regarding the association of insulin resistance (IR) with chronic diarrhea (CD) remains scarce, especially without robust epidemiological support from large-scale population cohorts. Therefore, we aimed to explore the association of IR with CD by a cross-sectional study of data drawn from the National Health and Nutrition Examination Survey (NHANES) from 2005 to 2010.

Material and methods

Multiple logistic regression models combined with subgroup analyses were leveraged to explore the association of IR (triglyceride-glucose [TyG] index, metabolic score for IR [METS-IR] score, and Homeostasis Model Assessment of IR [HOMA-IR] score) with the prevalence of CD. Nonlinear correlations were measured by restricted cubic spline (RCS) curves. Finally, the diagnostic efficacy of these indicators was calculated and examined with the receiver operating characteristic (ROC) curve and area under the curve (AUC).

Results

After full adjustment for covariates (e.g., sex, age, race, education, poverty-to-income ratio (PIR), body mass index (BMI), smoking status), the TyG index (OR: 1.35, 95% CI: 1.12-1.63), METS-IR score (OR: 2.00, 95% CI: 1.31-3.06), and HOMA-IR score (OR: 1.02, 95% CI: 1.00-1.03) showed positive associations with the prevalence of CD. The RCS curve analysis showed a non-linear association between HOMA-IR and CD (P -non-linear=0.003), with an inflection point at 2.423. METS-IR and TyG showed a significant positive linear association with CD (both P -non-linear > 0.05). The ROC curve analysis identified METS-IR score as the optimal diagnostic indicator for CD (AUC=0.645, 95% CI: 0.619-0.672).

Conclusions

The present research discovered a significant positive association of IR with CD among a demographically representative cohort of US adults. Elevated IR scores are significantly associated with the prevalence of CD.

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Objective: The evidence regarding the association of insulin resistance (IR) with chronic diarrhea (CD) remains scarce, especially without robust epidemiological support from large-scale population cohorts. Therefore, we aimed to explore the association of IR with CD by a cross-sectional study of data drawn from the National Health and Nutrition Examination Survey (NHANES) from 2005 to 2010.

Methods: Multiple logistic regression models combined with subgroup analyses were leveraged to explore the association of IR (triglyceride-glucose [TyG] index, metabolic score for IR [METS-IR] score, and Homeostasis Model Assessment of IR [HOMA-IR] score) with the prevalence of CD. Nonlinear correlations were measured by restricted cubic spline (RCS) curves. Finally, the diagnostic efficacy of these indicators was calculated and examined with the receiver operating characteristic (ROC) curve and area under the curve (AUC).

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24 **Conclusion:** The present research discovered a significant positive association of IR with CD
25 among a demographically representative cohort of US adults. Elevated IR scores are
26 significantly associated with the prevalence of CD.

27 **Keywords:** chronic diarrhea, insulin resistance, NHANES

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28 1. Introduction

29 Chronic diarrhea (CD) represents a notable public health concern. While this condition does not
30 increase mortality risk, patients with CD exhibit significantly higher prevalence of detected
31 organic pathologies—including colon polyps and ulcerative colitis—relative to the general
32 population [1]. The condition causes physical and mental distress, affects the patients' quality of
33 life and nutrient absorption capacity, and places a greater economic strain on the health care
34 system. CD is usually defined as having more than 3 stools per day, or a total stool volume
35 of >200 g/d, with water content of over 85%, for more than 4 weeks [2]. CD is estimated to be
36 present in 1%-5% of adults [3].

37 CD can be triggered by various diseases, such as functional and organic diseases. The
38 pathophysiological mechanisms of diarrhea caused by different etiologies are different. **Previous**
39 **studies have shown that intestinal function is somewhat affected in patients with type 2 diabetes**
40 **compared with the normal population [4]. It may be related to autonomic neuropathy,**
41 **hyperglycemia-stimulated intestinal microangiopathy and gastrointestinal hormone secretion**
42 **disorders[5, 6], which seriously affect the quality of life and blood glucose control of patients[7].**
43 **Insulin resistance (IR) is not only the main feature of type 2 diabetes and prediabetes, but also a**
44 **systemic metabolic disorder that triggers chronic low-grade inflammation, autonomic**
45 **dysfunction, and changes in hormone levels. These can act directly or indirectly on the**
46 **gastrointestinal tract, thereby causing or aggravating chronic diarrhea. However, currently, most**
47 studies on CD focus on its association with intestinal microecology, inflammatory response, and
48 abnormalities in gastrointestinal mucosal function [8-9]. Studies on the association of IR with
49 CD remain sparse, especially without robust epidemiological evidence from large-scale
50 population cohorts. To quantify IR, there are a number of methods. The

51 hyperinsulinemic-euglycemic clamp (HEGC) test is considered the gold standard for detecting
52 IR [10]. Nevertheless, it is time-consuming, costly, and technically difficult to apply in extensive
53 epidemiological research and clinical practice. Homeostasis Model Assessment of IR
54 (HOMA-IR), computed from fasting insulin (FINS) and fasting plasma glucose (FPG)
55 measurements, is considered a reliable surrogate for HEGC test and is the most extensively
56 utilized index in current practice [11]. Triglyceride-glucose (TyG) index is extensively
57 acknowledged to be a recommended tool for evaluating IR [12]. It is simpler, more economical,
58 and more suitable for clinical studies. Collectively, the TyG index, metabolic score for IR
59 (METS-IR), and HOMA-IR score are widely adopted metrics for assessing IR because of their
60 simplicity, availability, and low cost [13-16]. They have been extensively utilized to study the
61 association of IR with type 2 diabetes, obesity, and cardiovascular disease.

62 Although the National Health and Nutrition Examination Survey (NHANES) lacks direct
63 evidence demonstrating that IR independently contributes to CD, IR may be associated with CD
64 through indirect mechanisms such as metabolic abnormalities, dysbiosis, and inflammation.
65 Therefore, the present study examined the association of IR (TyG index, METS-IR score, and
66 HOMA-IR score) with the prevalence of CD based on NHANES 2005-2010 data. The research
67 aimed to provide new epidemiological evidence for CD and a potential clinical reference for
68 early intervention and management of the disease.

69

70 **2. Materials and Methods**

71 **2.1 Study population**

72 In total, there were 31,034 participants in the NHANES [17] (2005-2010) of the United States of
73 America. Eventually, 5,606 individuals were included in this research. The inclusion criteria

74 were as follows: (i) age ≥ 20 years; (ii) with complete data of FPG, high-density lipoprotein
75 cholesterol (HDL-C), triglycerides (TG), and FINS; (iii) with complete data of Bristol Stool
76 Form Scale (BSFS), and without constipation; (iv) with complete data of age, sex, race,
77 education, PIR, body mass index (BMI), smoking status, alcohol consumption, caffeine intake,
78 depression scale score, and history of diabetes mellitus (DM). The filtering process is
79 demonstrated in Figure 1. All the data were obtained from the official website of NHANES. All
80 the study protocols in the survey were approved by the Ethics Review Committee of the National
81 Center for Health Statistics (NCHS). Informed consent was documented for all participants.
82 Because the data retrieved in this research were publicly available and de-identified, no
83 additional ethical approval was required.

84 **2.2 Exposure variables**

85 The TyG index [18] was computed as $\ln(\text{TG} \times \text{FPG}/2)$ (units: TG and FPG in mg/dL). The
86 METS-IR score [19] was defined by $\ln[\text{BMI} \times (\text{TG} + 2 \times \text{FPG})] / \ln(\text{HDL-C})$ (units: TG, FPG,
87 HDL-C in mg/dL; BMI in kg/m^2). HOMA-IR [20] was determined using $(\text{FPG} \times \text{FINS})/22.5$
88 (FPG in mmol/L; FINS in $\mu\text{U}/\text{mL}$). The TyG index, METS-IR score, and HOMA-IR score were
89 divided into T1, T2, and T3 groups according to tertiles.

90 **2.3 Outcome variables**

91 During the NHANES survey, participants were issued a color card with images and a description
92 card of the BSFS types (1 to 7). They were then asked, "Select the number corresponding to your
93 usual or most common stool type." Participants were confirmed with CD if they identified their
94 usual or most common stool type as BSFS type 6 (fluffy with ragged edges, mushy) or type 7
95 (watery, no solid pieces). They were confirmed with chronic constipation when they chose type 1
96 (separate hard lumps, like nuts) or type 2 (sausage-shaped, frequent use of laxatives) [21, 22].

97 The present study included the normal and diarrhea populations.

98 **2.4 Covariates**

99 The researchers identified potential covariates based on previous studies and clinical expertise.
100 Covariates included age, sex, race, education, PIR, BMI, smoking status, alcohol consumption,
101 DM, depression, and caffeine intake (mg/d). Age was categorized as under 40 years, 40-60 years,
102 and over 60 years. Race was categorized as non-Hispanic white, non-Hispanic black, Mexican
103 American, other Hispanic, and other races. Education was categorized as less than high school,
104 high school and equivalent, and beyond high school. PIR denoted a household's income relative
105 to the federal poverty guidelines, and is categorized as less than 1.3, 1.3-3.5, and more than 3.5.
106 Based on smoking status, participants were divided into former smokers, never smokers, and
107 current smokers. Never smokers were those who have smoked fewer than 100 cigarettes. Former
108 smokers were those who had smoked more than 100 cigarettes, but were non-smokers at the time.
109 Current smokers were those who had smoked more than 100 cigarettes and still smoked on a few
110 days or every day. Alcohol consumption was grouped into drinking and non-drinking. Drinking
111 was characterized by having at least twelve alcoholic drinks a year, based on questionnaires
112 administered via the Computer-Assisted Personal Interview (CAPI) system at the Mobile
113 Examination Center (MEC). History of DM was categorized as either present or absent. A
114 present history of DM was defined as those who had been previously informed by a physician
115 that the respondent had diabetes or those with a glycosylated hemoglobin level $> 6.5\%$.
116 Depression was characterized by a score of >10 in the Patient Health Questionnaire (PHQ).
117 Caffeine intake was examined with a standardized 24-hour dietary recall interview at the dietary
118 assessment center.

119 **2.5 Statistical methods**

120 All continuous variables conforming to a normal distribution were expressed as mean (SD), and
121 inter-group comparisons were examined with the independent samples t-test. Non-normally
122 distributed continuous variables were expressed as [M(P25, P75)], and inter-group comparisons
123 were conducted with the Mann-Whitney U-test. Categorical variables were expressed as
124 percentages(%), and inter-group disparities were compared with the X^2 test. Variance inflation
125 factor (VIF) was leveraged to examine potential multicollinearity between covariates. All
126 included covariates showed a VIF value of less than 5, denoting no significant multicollinearity
127 in the current study. Logistic regression models were leveraged to ascertain the association of
128 TyG index, METS-IR score, and HOMA-IR score with CD, respectively. Model 1 was
129 unadjusted; model 2 was adjusted for age, sex, and race; and model 3 was further adjusted for
130 education, PIR, BMI, smoking status, alcohol consumption, caffeine intake, history of depression,
131 and history of DM, based on model 2. METS-IR, HOMA-IR, and TyG were included in
132 regression models both in continuous form and in tertile groups, with OR and 95% CI calculated.
133 Stratified analyses and interaction analyses were implemented by sex, age, race, PIR, education,
134 smoking status, alcohol consumption, caffeine intake, history of depression, and history of DM.
135 Also, restricted cubic spline (RCS) plots were leveraged to further examine the nonlinear
136 relationships of TyG index, METS-IR score, and HOMA-IR score with CD. Additionally,
137 receiver operating characteristic (ROC) curves and area under the curve (AUC) analyses were
138 employed to appraise the prognostic performance of IR indices in the full cohort. All data
139 analyses were executed with the R 4.4.3 package (<http://www.R-project.org>), and differences
140 were deemed to be statistically significant at $P < 0.05$ (two-sided).

141

142 3. Results

143 **3.1 Baseline characteristics of the participants**

144 In total, 5,606 participants were included, of whom 49.92% were males and 50.08% were
145 females, and 511 (9.12%) had CD. The prevalence of CD was considerably higher in females,
146 current smokers, individuals aged 40-60 years old, individuals with DM, depression, PIR less
147 than 3.5, education less than high school education. A statistically significant difference was
148 noted ($P<0.05$). The IR scores in individuals with CD [TyG: 8.78 (8.41,9.24); METS-IR: 2.14
149 (1.95,2.34); HOMA-IR: 2.79 (1.62,4.97)] were all higher than those of the individuals without
150 CD [TyG: 8.62(8.24,9.04), METS-IR: 2.26 (1.37,4.01), HOMA-IR: 2.06 (1.89,2.25)], with
151 statistically significant differences ($P<0.001$). The details are delineated in Table 1.

152 **3.2 Association of TyG index, METS-IR score, and HOMA-IR score with CD**

153 As continuous variables, IR (TyG index, METS-IR score, and HOMA-IR score) were positively
154 associated with the prevalence of CD. The unadjusted logistic regression analyses yielded the
155 following ORs with 95% CIs: METS-IR (OR: 2.13, 95% CI: 1.52-2.98), HOMA-IR (OR: 1.03,
156 95% CI: 1.01-1.04), and TyG index (OR: 1.47, 95% CI: 1.26-1.73), as demonstrated in Table 2.
157 After fully adjusting for covariates, the IR remained positively associated with CD **prevalence**.
158 Specifically, per unit increase in METS-IR (OR: 2.00, 95% CI: 1.31-3.06) was associated with
159 approximately a 1-fold increase in the prevalence of CD. Significant positive associations were
160 also observed for HOMA-IR (OR: 1.02; 95% CI: 1.00-1.03) and TyG index (OR: 1.35; 95% CI:
161 1.12-1.63). Further analyses supported this finding. With the tertile T1 as a reference, the third
162 quartile groups of METS-IR, HOMA-IR, and TyG were associated with increased prevalence of
163 CD, regardless of model adjustment ($P<0.05$), as shown in Table 2. To better understand the
164 nonlinear association of IR with CD **prevalence**, we plotted RCS curves. The results showed that
165 HOMA-IR exhibited a nonlinear positive association with the prevalence of CD

166 (P-non-linear=0.003) with an inflection point at 2.423, and METS-IR and TyG demonstrated
167 significant positive linear associations with CD (P-non-linear = 0.056 and 0.567, respectively)
168 (Figure 2A-C).

169 **3.3 Subgroup analyses of the association of METS-IR score, TyG index, and HOMA-IR** 170 **score with CD**

171 The results of the subgroup analyses indicated inconsistent associations of the METS-IR score,
172 TyG index, and HOMA-IR score with CD (Table 3).

173 In subgroup analyses, the positive association of METS-IR with CD was consistent in the
174 subgroups of females, non-Hispanic whites, non-Hispanic blacks, other Hispanics, aged < 40
175 years, PIR < 3.5, beyond high school education, less than high school education, current smokers,
176 non-drinking, with history of DM, no history of DM, and no depression (P<0.05). The
177 interaction test yielded that sex and alcohol consumption could influence the positive association
178 between METS-IR score and CD (P for interaction<0.05), and no interaction was observed for
179 the remaining components (Table 3).

180 The positive association of TyG index with CD was consistent across the subgroups of females,
181 non-Hispanic whites, non-Hispanic blacks, and other Hispanics, aged <40 years, PIR <3.5,
182 beyond high school education, less than high school education, current smokers, no depression,
183 drinking, and non-drinking (P<0.05). The interaction test yielded that education could impact the
184 positive association of TyG index with CD (P for interaction<0.05), and no interaction was
185 observed for the remaining components (Table 3).

186 The positive association of HOMA-IR scores with CD was consistent in the subgroups of
187 females, non-Hispanic whites, beyond high school education, current smokers, and depression
188 (P<0.05). Interaction tests showed that no interaction was noted for all components (P for

189 interaction >0.05), as shown in Table 3.

190 3.4 ROC curve analysis of METS-IR score, TyG index, and HOMA-IR score with CD

191 The ROC curves were adopted (Figure 3A-C) to further examine the predictive utility of IR for
192 the prevalence of CD. The AUC of METS-IR was 0.645 (95% CI: 0.619-0.672), which was
193 higher than that of the TyG index (AUC=0.644,95% CI:0.617-0.671) and the HOMA-IR score
194 (AUC=0.641,95% CI:0.614- 0.668), as shown in Figure 3A-C.

195

196 4. Discussions

197 Using cross-sectional data from U.S. adults (≥ 20 years) in NHANES (2005-2010), we
198 discovered that HOMA-IR exhibited a nonlinear positive correlation with CD. Separately, both
199 METS-IR and TyG demonstrated significant positive linear correlations with CD after
200 adjustment for key demographic and lifestyle confounders. Furthermore, ROC curve analysis
201 showed that METS-IR score had the strongest diagnostic performance for CD. This indicates that
202 high scores of IR (METS-IR, HOMA-IR, and TyG) are highly associated with the prevalence of
203 CD, suggesting the therapeutic potential of targeting IR in CD patients. Previous studies have
204 also shown that the incidence of gastrointestinal symptoms in patients with diabetes is higher
205 than that in the general population. The fasting blood glucose level of patients with
206 gastrointestinal symptoms is significantly higher than that of patients without gastrointestinal
207 symptoms, and diabetes patients with gastrointestinal symptoms have a worse quality of life [23,
208 24]. Patients with high IR index scores should be given appropriate guidance as early as possible
209 to control them at a lower level. This may be beneficial for lowering the prevalence of CD and
210 optimizing the quality of survival for patients.

211 Studies of recent years have shown that gut microbiota disorders may be involved in IR by

212 affecting host metabolism, immunity, and intestinal barrier integrity [25, 26]. The butyric
213 acid-producing bacteria *Faecalibacterium prausnitzii* and *Clostridium leptum* were both
214 negatively correlated with HOMA-IR [27]. Meanwhile, the abundance of *Faecalibacterium*
215 *prausnitzii* and *Clostridium leptum* was significantly lower in inflammatory bowel disease,
216 which is a common disorder characterized by CD [28-30]. This indicates that butyric
217 acid-producing bacteria are negatively correlated with both CD and HOMA-IR. Additional
218 studies denote that short-chain fatty acids, encompassing acetic acid and butyric acid, may
219 exacerbate both CD and IR through the mediation of 5-hydroxytryptamine (5-HT) [31, 32]. The
220 gut microbiota and its other product, bile acids, form the bile acid-gut microbiota axis.
221 Disruption at either end of this axis may promote the onset of IR [33, 34]. Furthermore, gut
222 microbiota dysbiosis may elevate the permeability of the intestinal mucosa, leading to increased
223 production of inflammatory mediators, including IL-6, which exacerbates IR [35]. Inflammatory
224 mediators like IL-6 and IL-1B affect peristalsis, secretion and reabsorption and may be involved
225 in the pathogenesis of CD [36]. **Supplementation with *Bifidobacterium* preparations can reduce**
226 **gastrointestinal symptoms, alleviate low-level inflammation in individuals with type 2 diabetes,**
227 **and improve glucose metabolism and IR [37, 38].** From the above, it can be seen that gut
228 microbiota is a critical node in the interaction between CD and IR. The mediators involved in
229 this process include not only short-chain fatty acids, 5-HT, bile acids, and inflammatory
230 mediators, but also numerous other unmentioned factors. Therefore, whether gut microbiota
231 plays a promoting or suppressing role in CD and IR depends on the types, proportions, and
232 relative activity of mediating molecules. In our research, a positive linear correlation between IR
233 and CD is observed. Hence, it can be speculated that IR is primarily related to the gut microbiota
234 and its metabolite derivatives that promote the development of CD.

235 In addition, the present research revealed sex-specific disparity in the association of METS-IR
236 with the prevalence of CD, with a stronger association observed in women. This may be related
237 to the differences in the synthesis of estrogens and serotonin in the brain, which also involves the
238 functions of gut microbiota and inflammatory mediators [34, 35]. Education partly reflects the
239 socioeconomic status of an individual. Lower education levels indicate limited healthcare access
240 and suboptimal hygiene and health literacy, which may result in delayed medical consultation
241 and reduced compliance. These factors thereby exacerbate IR and increase susceptibility to CD,
242 ultimately leading to variations in the prevalence of CD across education levels.

243 This study has the following strengths. First, it is the first study to identify a direct association of
244 IR with CD. The study has a clear way of sample acquisition, sufficient sample size, and reliable
245 data sources, which contribute to the reliability of the results. Second, the TyG index, METS-IR,
246 and HOMA-IR are accessible and widely used non-invasive methods. Their relationship with the
247 prevalence of diarrhea is plain to see. Elevated levels of these indices predict a higher prevalence
248 of diarrhea, suggesting that patients with metabolic abnormalities require IR screening and
249 interventions, such as dietary modifications and glucagon-like peptide-1 (GLP-1) agonists. These
250 approaches can guide further diagnostics, therapies, and preventive measures, thereby providing
251 references for clinical practice.

252 There are certain limitations in this research. First, the data were drawn from NHANES, and
253 most of the participants were of Mexican, Caucasian, and Black ethnicity. Their physical
254 conditions and lifestyles differed greatly from those of Asians. Therefore, the findings need to be
255 corroborated in studies with Asian cohorts. Second, despite adjustment for some confounders,
256 the potential influence of residual confounders cannot be excluded, such as bile acid
257 malabsorption. Third, this is a cross-sectional investigation, which assesses chronic diarrhea and

258 IR at the same time. Hence, the time series between CD and IR could not be determined, and it
259 does not present any causality. Future prospective cohort studies could integrate bile acid
260 metabolism and the characteristics of gut microbiota to provide robust evidence for a causal
261 association between IR and CD.

262

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363 **Figure legends**

364 **Figure 1** Flow chart of the selection of participants based on NHANES 2005-2010

365 TG: triglycerides; FPG: fasting plasma glucose; HDL-C: high-density lipoprotein cholesterol;

366 FINS: fasting insulin; PIR: poverty income ratio; BMI: body mass index

367 **Figure 2** RCS analysis (A) The link of HOMA-IR with CD; (B) The link of METS-IR with CD;

368 (C)The link of TyG with CD

369 **Figure 3** (A) The ROC curve of METS-IR scores predicting CD; (B) The ROC curve of TyG

370 index predicting CD; (C) The ROC curve of HOMA-IR scores predicting CD

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371 **Table 1** Demographic information of the participants

Characteristic	N	Overall	Non-diarrhea	CD	p-value
Sex	5,606				0.027
Female		2,750(50.08%)	2,497(49.55%)	253(56.63%)	
Male		2,856(49.92%)	2,655(50.45%)	201(43.37%)	
Age	5,606				0.200
<40		2,003(36.94%)	1,875(37.44%)	128(30.79%)	
>60		1,730(23.24%)	1,573(23.19%)	157(23.77%)	
40-60		1,873(39.82%)	1,704(39.37%)	169(45.44%)	
Race	5,606				0.400
Mexican American		993(7.13%)	892(6.99%)	101(8.87%)	
Non-Hispanic Black		1,048(10.14%)	960(10.02%)	88(11.64%)	
Non-Hispanic White		2,902(73.87%)	2,695(74.17%)	207(70.23%)	
Other Hispanic		430(3.81%)	389(3.74%)	41(4.74%)	

Other Race		233(5.04%)	216(5.08%)	17(4.52%)	
Education	5,606				0.004
<High school diploma		1,488(17.19%)	1,308(16.54%)	180(25.20%)	
>High school diploma		2,781(58.63%)	2,608(59.26%)	173(50.84%)	
High school diploma/equivalent		1,337(24.19%)	1,236(24.21%)	101(23.95%)	
Family-PIR	5,606				0.027
<1.3		1,033(12.19%)	910(11.82%)	123(16.74%)	
>3.5		2,404(52.34%)	2,244(53.00%)	160(44.17%)	
1.3-3.5		2,169(35.48%)	1,998(35.18%)	171(39.09%)	
Alcohol consumption	5,606				0.200
No		1,528(22.97%)	1,388(22.65%)	140(26.97%)	
Yes		4,078(77.03%)	3,764(77.35%)	314(73.03%)	
Smoking status	5,606				0.032

Before		1,497(26.02%)	1,380(26.15%)	117(24.40%)	
Never		2,886(50.97%)	2,670(51.42%)	216(45.36%)	
Now		1,223(23.01%)	1,102(22.43%)	121(30.24%)	
Caffeine intake	5,606	128.00(29.00,274.00)	128.00(29.00,274.00)	130.00(36.00,296.00)	0.400
DM	5,606				<0.001
No		4,825(89.94%)	4,470(90.54%)	355(82.64%)	
Yes		781(10.06%)	682(9.46%)	99(17.36%)	
Depression	5,606				<0.001
No		5,181(93.56%)	4,803(94.22%)	378(85.43%)	
Ye		425(6.44%)	349(5.78%)	76(14.57%)	
TyG	5,606	8.63(8.25,9.06)	8.62(8.24,9.04)	8.78(8.41,9.24)	<0.001
HOMA-IR	5,606	2.29(1.38,4.06)	2.26(1.37,4.01)	2.79(1.62,4.97)	<0.001
METS-IR	5,606	2.07(1.89,2.26)	2.06(1.89,2.25)	2.14(1.95,2.34)	<0.001
Fasting plasma glucose (mmol/L)	5,606	5.50(5.11,6.00)	5.50(5.11,5.94)	5.61(5.22,6.27)	0.002

Fasting insulin (uU/mL)	5,606	9.24(5.77,15.42)	9.13(5.71,15.26)	10.41(6.66,18.17)	0.002
Triglyceride (mg/dL)	5,606	112.00(78.00,162.00)	111.00(78.00,160.00)	125.00(88.00,187.00)	<0.001
Direct HDL-C(mg/dL)	5,606	52.00(42.00,64.00)	52.00(42.00,64.00)	49.00(41.00,60.00)	0.009
Body mass index (kg/m ²)	5,606	27.64(24.10,32.19)	27.48(24.05,31.93)	29.29(24.94,34.89)	<0.001

372 DM: diabetes mellitus; PIR: poverty to income ratio; HOMA-IR: Homeostasis Model Assessment of insulin resistance; METS-IR:
373 metabolic score for insulin resistance; HDL-C: high-density lipoprotein cholesterol.

Table 2 Weighted logistic regression analysis of the link of TyG index, the METS-IR score, and the HOMA-IR score with CD

Participants	Model 1		Model 2		Model 3	
	OR (95%CI)	<i>p</i> -value	OR (95%CI)	<i>p</i> -value	OR (95%CI)	<i>p</i> -value
METS-IR						
Continuous	2.13(1.52, 2.98)	<0.001	2.57(1.74, 3.81)	<0.001	2.00(1.31, 3.06)	0.002
Tertiles						
T1	Ref		Ref		Ref	
T2	1.26(0.86, 1.83)	0.200	1.32(0.91, 1.94)	0.140	1.25(0.84, 1.86)	0.300
T3	1.91(1.43, 2.55)	<0.001	2.18(1.57, 3.03)	<0.001	1.89(1.34, 2.68)	<0.001
TyG						
Continuous	1.47(1.26, 1.73)	<0.001	1.51(1.26, 1.80)	<0.001	1.35(1.12, 1.63)	0.003
Tertiles						
T1	Ref		Ref		Ref	
T2	1.38(0.95, 2.00)	0.087	1.44(0.95, 2.00)	0.065	1.38(0.94, 2.02)	0.100
T3	1.84(1.31, 2.59)	<0.001	1.90(1.31, 2.77)	0.001	1.64(1.11, 2.43)	0.014
HOMA-IR						
Continuous	1.03(1.01, 1.04)	0.006	1.02(1.01, 1.04)	0.010	1.02(1.00, 1.03)	0.036
Tertiles						
T1	Ref		Ref		Ref	
T2	1.36(0.97, 1.90)	0.072	1.34(0.95, 1.87)	0.090	1.35(0.98, 1.86)	0.064
T3	1.67(1.27, 2.20)	<0.001	1.62(1.23, 2.13)	0.001	1.44(1.07, 1.94)	0.017

Note: Model 1 was not adjusted; model 2 was adjusted for age, sex, and race; model 3 was further adjusted for education, poverty-to-income ratio, body mass index, smoking status, alcohol consumption, caffeine intake, history of depression, and history of diabetes mellitus, on the basis of model 2.

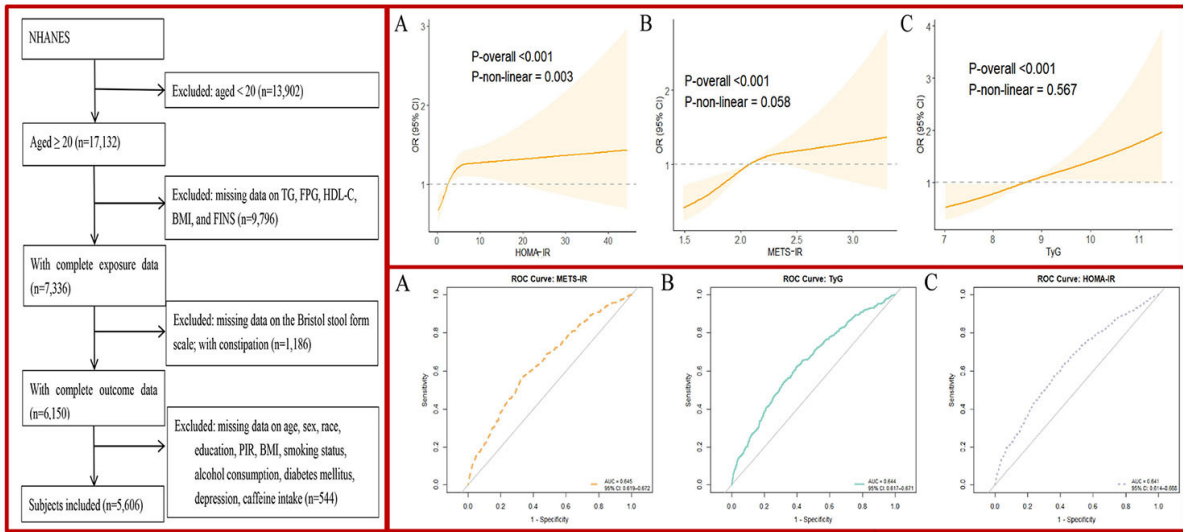
Table 3 Subgroup analyses of the link of METS-IR score, TyG index, and HOMA-IR score with CD

Subgroup	METS-TR			TyG			HOMA-IR		
	OR (95%CI)	<i>p</i> -value	<i>p</i> for interaction	OR (95%CI)	<i>p</i> -value	<i>p</i> for interaction	OR (95%CI)	<i>p</i> -value	<i>p</i> for interaction
Sex			0.016			0.233			0.053
Female	3.63(1.82, 7.27)	<0.001		1.60(1.16, 2.21)	0.006		1.04(1.01,1.08)	0.023	
Male	1.13(0.62,2.08)	0.700		1.16(0.90, 1.49)	0.200		1.00(0.98,1.02)	0.900	
Race			0.341			0.081			0.489
Mexican American	0.62(0.19,2.03)	0.400		0.84(0.48, 1.47)	0.500		1.02(0.961,0.7)	0.500	
Other Hispanic	9.24(2.63,32.42)	0.002		4.36(2.08, 9.13)	<0.001		0.97(0.96,1.01)	0.300	
Non-Hispanic White	2.01(1.14,3.54)	0.018		1.32(1.01, 1.71)	0.041		1.03(1.00,1.05)	0.036	
Non-Hispanic Black	4.33(1.60,11.8)	0.010		1.62(1.13, 2.34)	0.010		1.03(0.99,1.07)	0.200	
Other race	1.65(0.13,21.4)	0.700		1.10(0.36, 3.34)	0.900		1.09(0.91,1.31)	0.300	
Age			0.362			0.237			0.291
<40	2.84(1.30,6.20)	0.010		1.88(1.31, 2.70)	0.001		1.00(0.95,1.05)	>0.900	
40-60	1.63(0.78,3.37)	0.200		1.15(0.83, 1.60)	0.400		1.01(1.00,1.03)	0.093	
>60	2.03(0.92,4.47)	0.080		1.32(0.87, 2.01)	0.200		1.03(0.99,1.08)	0.130	
Poverty to income ratio			0.396			0.242			0.462
<1.3	3.33(1.61,6.90)	0.002		1.48(1.03, 2.12)	0.034		1.01(0.99,1.03)	0.300	
1.3-3.5	1.95(1.19,3.18)	0.009		1.53(1.12, 2.10)	0.009		1.03(1.00,1.07)	0.072	
>3.5	1.48(0.72,3.04)	0.300		1.11(0.84, 1.45)	0.400		1.00(0.97,1.04)	0.800	
Education			0.065			0.030			0.286
<High school diploma	2.12(1.07,4.19)	0.032		1.45(1.05, 2.01)	0.027		1.00(0.98,1.02)	0.700	
High school diploma/equivalent	0.84(0.27,2.60)	0.800		0.85(0.49, 1.47)	0.600		1.02(0.98,1.06)	0.400	
>High school diploma	2.88(1.55,5.37)	0.002		1.55(1.15, 2.10)	0.006		1.04(1.01,1.08)	0.019	
Smoking status			0.692			0.847			0.169
Never	1.78(0.96,3.28)	0.066		1.17(0.84, 1.64)	0.300		1.01(0.98,1.04)	0.500	
Former	2.3(0.96,5.50)	0.061		1.41(0.97, 2.06)	0.072		1.01(0.99,1.029)	0.500	
Current	2.04(1.16,3.58)	0.015		1.51(1.11, 2.05)	0.010		1.06(1.03,1.09)	<0.001	

Alcohol consumption			0.004			0.089		0.902
No	4.47(1.95,10.24)	<0.001		1.65(1.09, 2.50)	0.019		1.01(0.98,1.03)	0.700
Yes	1.62(0.99,2.65)	0.056		1.24(1.00, 1.55)	0.049		1.02(1.00,1.05)	0.079
Diabetes Mellitus			0.065			0.613		0.503
No	1.81(1.076,3.04)	0.027		1.28(0.98, 1.68)	0.066		1.03(0.99,1.06)	0.120
Yes	2.61(1.13,6.04)	0.026		1.39(1.00, 1.93)	0.053		1.00(0.98,1.03)	0.800
Depression			0.409			0.410		0.057
No	2.17(1.35,3.49)	0.002		1.43(1.16, 1.76)	0.002		1.01(1.00,1.03)	0.130
Yes	1.20(0.39,3.70)	0.700		0.86(0.50, 1.49)	0.600		1.06(1.01,1.11)	0.030

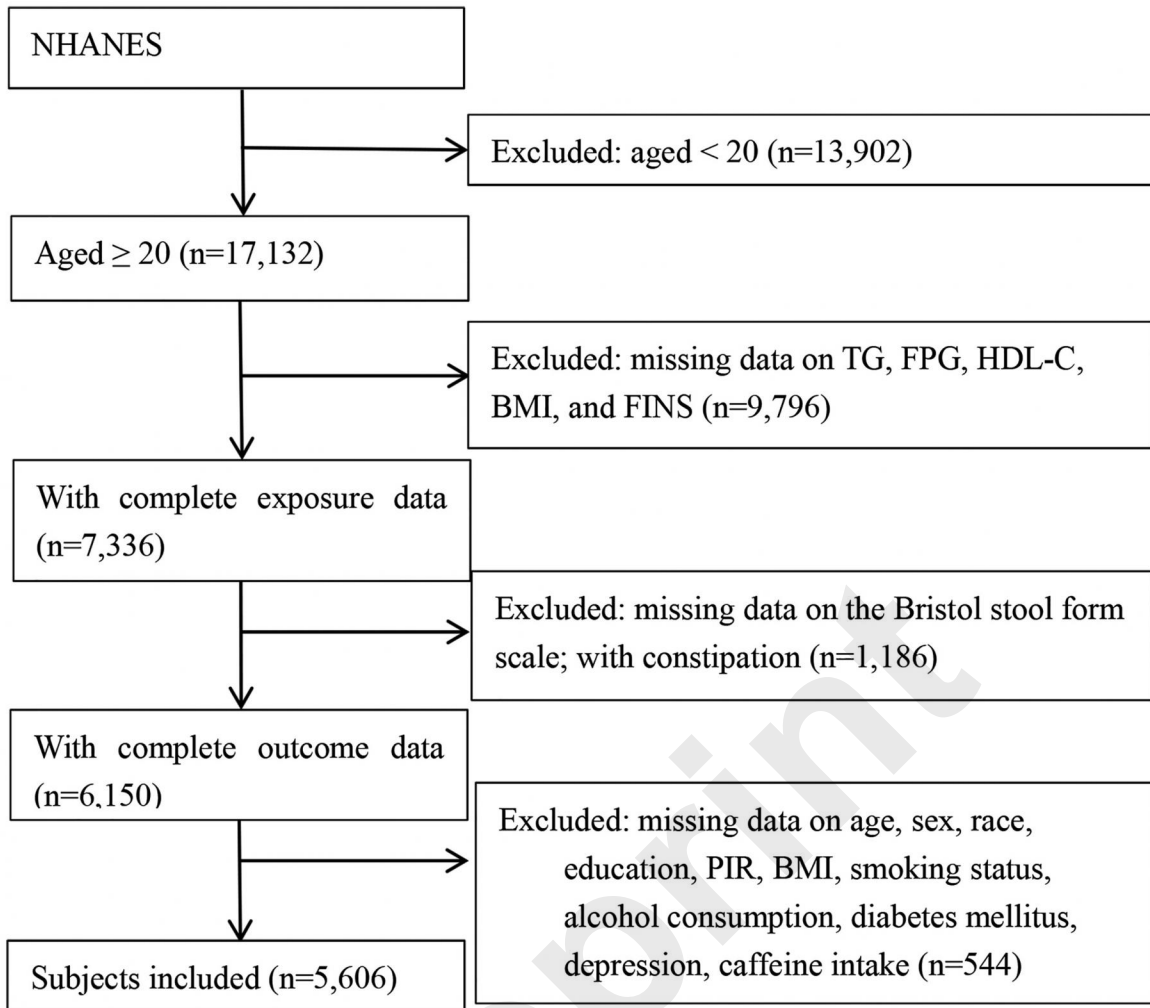
Note: Subgroup analyses were adjusted for age, sex, race, education, poverty-to-income ratio, body mass index, smoking status, alcohol consumption, caffeine intake, depression, and diabetes mellitus.

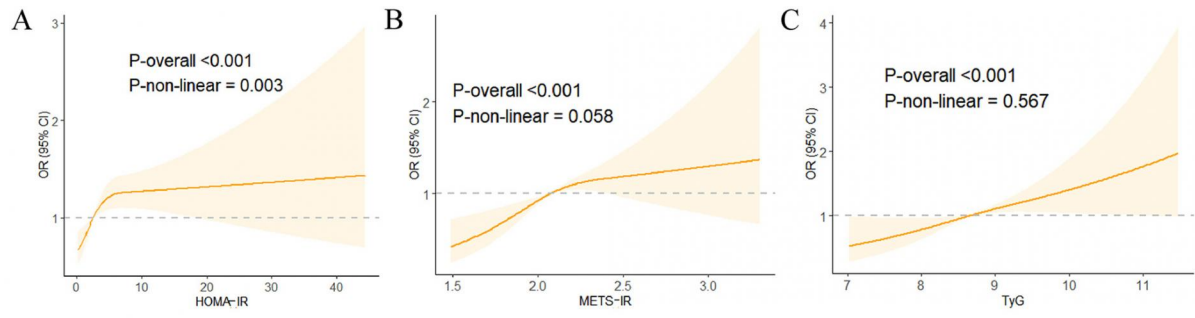
Association of IR with CD



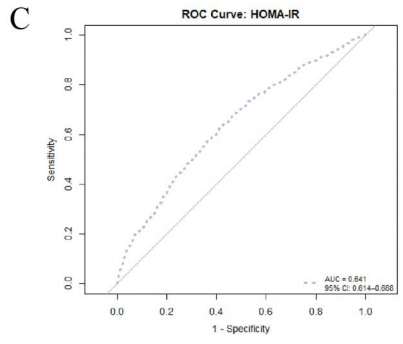
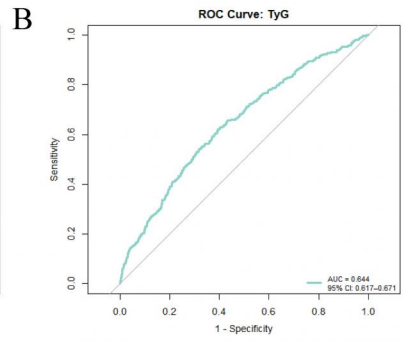
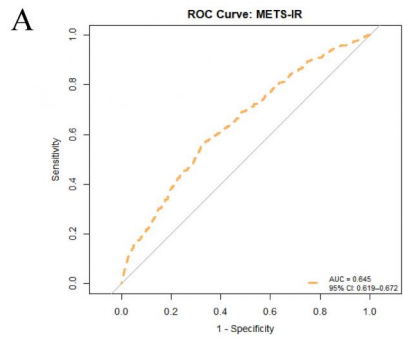
In this cross-sectional investigation, we discovered that HOMA-IR exhibited a nonlinear positive correlation with CD. Separately, both METS-IR and TyG demonstrated significant positive linear correlations with CD after adjustment for key demographic and lifestyle confounders. Furthermore, ROC curve analysis showed that METS-IR score had the strongest diagnostic performance for CD.

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