


New Predictive Equations for Estimating Resting Energy Expenditure in Adults With Crohn's Disease

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Abstract

Background: Increased resting energy expenditure (REE) has been hypothesized to be a potential cause of weight loss in individuals with Crohn's disease (CD). This study aimed to develop and validate new predictive equations for estimating REE in adults with CD. **Methods:** Adults, ages 18–65 years, with CD were recruited. Anthropometry, indirect calorimetry, and bioimpedance analysis were performed in all patients. Disease activity was assessed by Crohn's Disease Activity Index. The new predictive equations were generated using different regression models. Prediction accuracy of the new equations was assessed and compared with the most commonly used equations. **Results:** A total of 270 CD patients (159 males, 111 females) were included and randomly assigned to the calibration (n = 180) and validation groups (n = 90). REE was directly correlated with weight and bioimpedance index, whereas the relation with both age and disease activity was inverse. The new equations were suitable for estimating REE at population level (bias: −0.2 and −0.3, respectively). Individual accuracy was good in both models (≥80%, respectively), especially in females; and similar results were shown by some of the selected equations. But, when accuracy was set within ±5%, the new equations gave the highest prediction. **Conclusion:** The new, disease-specific, equations for predicting REE in individuals with CD give a good prediction accuracy as far as those proposed in the literature for the general population. However, the new ones performed better at the individual level. Further studies are needed to verify the reliability and usefulness of these new equations. (*JPEN J Parenter Enteral Nutr.* 2020;44:1021–1028)

Keywords

accuracy; Crohn's disease; dietary advice; energy expenditure; predictive equations

Clinical Relevancy Statement

Variations in resting energy expenditure (REE) can be a potential cause of weight loss in patients with Crohn's disease (CD); hence, an accurate estimation of REE is crucial for assessing their energy needs. Indirect calorimetry is recognized as a criterion method for measuring REE, but it is relatively time-expensive and frequently not available in the clinical setting. Alternatively, REE can be easily and quickly estimated by predictive equations. Thus, this study aims to develop and validate new predictive equations in adult patients with CD, since no specific formula for predicting REE exists in literature.

Introduction

Crohn's disease (CD) is a chronic relapsing, inflammatory condition of the gastrointestinal tract with unpredictable course.^{1–3} In patients with CD, disease symptoms like diarrhea, abdominal pain, anorexia, and malabsorption together with increased resting energy expenditure (REE), possibly related to inflammatory response, can be

assumed as potential causes of weight loss and secondary malnutrition.⁴

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To date, however, inconsistent evidence has emerged on the relationships between REE and disease activity in patients with CD,^{4,5} since comparisons among studies are hampered by differences in data reporting. Sasaki et al⁶ found that REE measured by indirect calorimetry (IC) was higher in CD patients with moderate disease activity compared with healthy controls. Similarly, Kushner et al⁷ observed that measured REE (MREE) increased with increasing disease activity. On the contrary, Stokes et al⁸ showed a mean REE value of 33 kcal/kg/d in patients with CD, which was similar to that observed for healthy subjects even when clinically active patients were included.⁹ Still, adjusting REE per unit of lean mass (REE/kg), it was found to be higher¹⁰ or not different¹¹ in comparison with controls.

The assessment of REE provides basic information for energy requirements in both health and disease statuses.¹² Currently, IC is recognized as a criterion method for measuring REE, but it is relatively time-expensive and frequently not available in the clinical setting. Hence, REE is often estimated by predictive equations, which includes variables that can be easily collected, such as age, stature, and body weight. Among patients with CD, predicted REE (PREE) by equations based on anthropometric variables, for instance the Harris and Benedict (HB) equation, showed lower⁶ or similar results^{4,10} compared with MREE, unrelated to disease activity. However, these results could be affected by small sample size, differences in study population, or inadequate evaluation of disease activity.⁵

From a practical point of view, bioimpedance analysis (BIA) is a commonly used tool for body-composition assessment in the clinical setting¹³ that might be valuable in estimating REE, since the major determinant of REE is fat-free mass (FFM). However, the interpretation of BIA results is highly dependent on the equations used to estimate FFM.^{14,15} Alternatively, raw BIA data, such as bioimpedance index (BI-Index) and phase angle (PhA), might be taken into consideration for evaluating the relationship of FFM with energy expenditure.^{16,17}

To our knowledge, previous studies have not proposed specific predictive equations for estimating REE in individuals with CD. Therefore, the primary aim of the present study was to develop and validate predictive equations for patients with CD, using both anthropometric and raw bioimpedance variables as predictors. As an additional aim, we assessed the accuracy of selected predictive equations of REE (for general population) and compared them with the formulas developed in this paper.

Materials and Methods

Patients and Study Design

Inclusion criteria consisted of the following: male and female adult patients, ages 18–65 years, with a diagnosis

of CD. Exclusion criteria were untreated dysthyroidism and type 2 diabetes mellitus; use of corticosteroids in the last 3 months; history of acute or chronic liver or kidney disease; current enteral and parenteral nutrition; presence of fistulae, ileostomy, or colostomy; presence of extensive small-bowel resections (residual small bowel of <2 m); pregnancy or lactation; unstable body weight in the last month; and inability or unwillingness to give informed consent.

Disease activity was clinically defined using the Crohn's Disease Activity Index (CDAI), classifying patients in the active and quiescent phases (≥ 150 and < 150 , respectively).¹⁸

The present study was a retrospective analysis of data collected between 2005 and 2018 from patients undergoing procedures to evaluate nutrition status at the Department of Clinical Medicine and Surgery, Federico II University Hospital, in Naples, Italy. The study was conducted in accordance with the Declaration of Helsinki and received the approval of the Ethical Committee of Federico II University Hospital. Informed consent was obtained from all patients recruited.

All measurements were performed early in the morning after a fasting period of 8–10 hours, according to standardized conditions—abstention from alcohol, smoking, and vigorous physical activity for 24 hours prior to the assessment. As previously reported,¹⁶ smoking was not allowed for occasional and current smokers on the morning of the test until the end of measurements; however, current smokers were asked to maintain their smoking habits on the day before. Data were excluded from analysis if the respiratory quotient (RQ) was outside the expected range (0.71–1.00) and when MREE was ± 3 SDs outside the mean REE.

Anthropometry and Bioimpedance Analysis

Body weight and stature were measured to the nearest 0.1 kg and 0.5 cm, respectively. Measurements were taken while the subject wore light clothes and no shoes, using a platform beam scale with a built-in stadiometer (Seca 709, Seca, Hamburg Germany). Body mass index (BMI) (kg/m^2) was calculated as body weight divided by squared stature.

BIA¹⁵ was performed at 50 kHz (Human Im Plus II, DS Medica, Milan, Italy) at room temperature (22°C–25°C). Measurements were carried out on the nondominant side of the body, in the postabsorptive state, after voiding, and with the participant in the supine position for 20 minutes. The BIA variables considered (data produced by the device) were resistance (R), reactance, and PhA. BI-index was calculated as the ratio $\text{stature}^2/\text{resistance}$ (cm^2/ohm).

Table 1. Characteristics of the Study Sample for the Calibration Group.

| | Males (n = 104) | Females (n = 76) | All (n = 180) |
|------------------------------|--------------------------|------------------|---------------|
| Age, y | 37.7 ± 12.8 | 38.1 ± 14.0 | 37.9 ± 13.3 |
| Weight, kg | 66.6 ± 10.5 ^a | 55.4 ± 9.0 | 61.8 ± 11.3 |
| Stature, cm | 171.9 ± 6.2 ^a | 159.7 ± 6.0 | 166.7 ± 8.6 |
| BMI, kg/m ² | 22.5 ± 3.4 | 21.7 ± 3.4 | 22.2 ± 3.4 |
| BI-Index, cm ² /Ω | 58.5 ± 8.0 ^a | 41.6 ± 6.4 | 51.4 ± 11.1 |
| PhA (degrees) | 6.7 ± 0.9 ^a | 5.6 ± 0.7 | 6.3 ± 1.0 |
| MREE, kcal/d | 1641 ± 195 ^a | 1340 ± 162 | 1514 ± 235 |
| RQ | 0.816 ± 0.076 | 0.812 ± 0.081 | 0.814 ± 0.078 |

Data are expressed as mean ± SD.

BI-Index, bioimpedance index; BMI, body mass index; MREE, measured resting energy expenditure; PhA, phase angle; RQ, respiratory quotient.

^a*P* < .05 between sexes.

Table 2. Characteristics of the Study Sample for the Validation Group.

| | Males (n = 55) | Females (n = 35) | All (n = 90) |
|------------------------------|--------------------------|------------------|---------------|
| Age, y | 37.9 ± 12.5 | 39.5 ± 15.1 | 38.5 ± 13.5 |
| Weight, kg | 68.8 ± 11.4 ^a | 55.9 ± 10.6 | 63.8 ± 12.7 |
| Stature, cm | 172.0 ± 6.1 ^a | 159.4 ± 5.6 | 167.1 ± 8.5 |
| BMI, kg/m ² | 23.2 ± 3.7 | 21.9 ± 3.7 | 22.7 ± 3.7 |
| BI-Index, cm ² /Ω | 60.0 ± 7.7 ^a | 41.8 ± 6.9 | 53.0 ± 11.6 |
| PhA (degrees) | 6.7 ± 0.9 ^a | 5.6 ± 0.6 | 6.3 ± 1.0 |
| MREE, kcal/d | 1679 ± 182 ^a | 1360 ± 151 | 1555 ± 231.1 |
| RQ | 0.810 ± 0.075 | 0.790 ± 0.075 | 0.802 ± 0.075 |

Data are expressed as mean ± SD.

BI-Index, bioimpedance index; BMI, body mass index; MREE, measured resting energy expenditure; PhA, phase angle; RQ, respiratory quotient.

^a*P* < .05 between sexes.

Measurements of Resting Energy Expenditure

REE was measured by IC¹⁹ using a canopy system, V max29 (Sensor Medics, Anaheim, CA, USA). The instrument was routinely checked by burning ethanol, whereas oxygen and carbon dioxide analyzers were calibrated on the test day using nitrogen and standardized gases (mixtures of nitrogen, carbon dioxide, and oxygen).

Measurement conditions for IC were defined following the suggestions made by Compher et al²⁰ and Fullmer et al.²¹ REE was assessed at an ambient temperature of 22°C–25°C and, in fertile women, during the follicular phase to avoid any potential effects of the menstrual cycle. Participants lay down on a bed in a quiet environment for a 15-minute adaptation period. Afterward, oxygen consumption and carbon dioxide production were measured for 45 minutes, discarding the first 5 minutes. Energy expenditure was calculated using the abbreviated Weir formula, neglecting protein oxidation.²²

Predictive Equations

REE was also estimated using the following predictive equations: HB,²³ Food and Agriculture Organization (FAO),²⁴ Schofield,²⁵ Owen,^{26,27} Muller,²⁸ and De Lorenzo.²⁹ Thus,

accuracy of the new predictive equations at the population and individual levels was calculated and then compared with those equations.

Statistical Analysis

Statistical analyses were performed using IBM SPSS (version 24, Chicago, IL, USA). All data are presented as mean ± SD, unless otherwise specified, and significance was defined as *P* < .05. As highlighted in Tables 1 and 2, participants were randomly assigned to a calibration or a validation subset in a way that the ratio between them remained constant.¹⁶ To examine whether variables were normally distributed, the Kolmogorov-Smirnov test and the Shapiro-Wilk test were used.

Data were compared between genders using unpaired *t*-tests, whereas linear correlation was applied for evaluating associations between variables. Multivariate linear regression analysis was performed to develop the new predictive equations, with REE measured by IC as the dependent variable. We generated models as follows: in Model 1, age, sex, weight, stature, and CDAI were set as predictors, whereas in Model 2, we added the raw BIA variables (BI-Index and PhA). Coefficient of determination (*R*²) and standard error of the estimate (SEE) were considered for

assessing the predictive power of formulas. The regression equations derived from the calibration subset were applied to the validation subset.

Mean difference between PREE and MREE, as well as bias (that is, the average percent difference), were both used as a measure of accuracy at the group level. Bias was found acceptable if within $\pm 5\%$.^{30,31} Concurrently, the percentage of patients with a PREE within 90%–110% and the percentage of patients with an MREE within 95%–105% were both used as measures of accuracy at the individual level. According to the range, values lower than 90% and 95% were classified as underprediction, whereas values higher than 105% and 110% were classified as overprediction. The root mean squared error (RMSE) was used to define the predictions obtained with these models. Finally, comparisons of PREE-MREE differences vs mean PREE-MREE values were performed by Bland-Altman plots to estimate the limits of agreement.³²

Results

Two hundred eighty-four patients with CD were selected for this study. Ten were excluded for not fulfilling the inclusion criteria, and 4 were excluded for taking corticosteroids. Therefore, a total of 270 patients with CD (159 males and 111 females) were included in the analysis (Supplementary material). Anthropometric, raw BIA variables, and MREE data are summarized for the calibration and validation groups in Tables 1 and 2, respectively. Age, BMI, and RQ did not vary between genders, whereas body weight, stature, MREE, and raw BIA variables significantly differed. Thirty-six percent of patients were in active phase (CDAI 225 ± 58 ; median = 215, min-max = 152–472), whereas 64% were in quiescent phase (CDAI 71 ± 41 ; median = 77, min-max = 2–140). The proportion of patients in the active phase was similar in males and females.

Pearson linear correlation showed that in patients with CD, MREE was directly correlated with individual characteristics and raw BIA variables except for age and CDAI, which displayed an inverse relation. Overall, BI-Index had the strongest correlation with MREE ($r = .762$, $P < .001$), followed by body weight ($r = 0.733$, $P < .001$), stature ($r = 0.681$, $P < .001$), and PhA ($r = 0.464$, $P < .001$). On the contrary, neither age ($r = -0.102$, $P = .175$) nor CDAI ($r = -0.113$, $P = .165$) was significantly correlated with MREE.

Next, multiple regression analysis was performed to assess the relationship between MREE and various combinations of potential predictors. Age, basic anthropometric measures (stature, weight, and BMI), and CDAI were considered first (Model 1) to create the following equation:

$$\begin{aligned} \text{REE (kcal/d)} &= 10.8 \times \text{Weight} + 6.42 \\ &\times \text{Stature} - 1.85 \times \text{Age} - 211 \quad (+102 \text{ if male}) \\ &\text{(unstandardized regression coefficients,} \\ &R^2 = 0.687; \text{ SEE} = 133 \text{ kcal/d)} \end{aligned} \quad (1)$$

When raw BIA variables were added to the model (Model 2), BI-Index and PhA were both included in the equation:

$$\begin{aligned} \text{REE (kcal/d)} &= 7.33 \times \text{Weight} + 5.02 \times \text{BI} - \text{Index} \\ &+ 7.59 \times \text{Stature} + 34.1 \times \text{PhA} - 678 \\ &\text{(unstandardized regression coefficients,} \\ &R^2 = 0.699; \text{ SEE} = 130 \text{ kcal/d)} \end{aligned} \quad (2)$$

Validation of the New Predictive Equations

To evaluate the accuracy of the new predictive equations, 90 individuals with CD (55 males and 35 females) were randomly assigned to the validation group, using the statistical software. Prediction accuracy at the group level (assessed by the difference between PREE and MREE, percent bias, and the RMSE in kcal/d) was reported for the new equations and selected equations from the literature in the Table 3. We found that the newly developed predictive equations were accurate at the group level in both sexes, since mean bias was $<1\%$ (Equation (1): -0.2% and Equation (2): -0.3% in males; Equation (1): -1% and Equation (2): -0.7% in females). When REE was predicted using selected equations from the literature, bias was acceptable for the HB (-3.0%), FAO (-0.1%), Schofield (0.2%), and De Lorenzo equations (-2.0%) in males and for the HB (-2.7%), FAO (-2.8%), and Schofield equations (-4.4%) in females.

As far as the accuracy at the individual level is concerned, the percentage of participants with a PREE within $\pm 10\%$ of MREE is presented for the new and other predictive equations in Figure 1. The new equations gave the highest accuracy in males (Equation (1): 80%; Equation (2): 82%) and females (Equation (1): 80%; Equation (2): 83%). The HB, FAO, Schofield, and De Lorenzo equations were also accurate ($\approx 80\%$) in male patients, whereas the HB (74%) and FAO (80%) equations performed well in female patients.

Conversely, by setting the accuracy range within $\pm 5\%$ of MREE, all equations considered showed low prediction accuracy ($<50\%$) at the individual level in both genders, as presented in Figure 2. Specifically, most of them tended to underpredict REE, except for Muller equation, which overpredicted REE. On the contrary, the new equations provided the best accuracy in males (Equation (1): 58%; Equation (2): 64%), whereas the one including raw BIA variables gave the highest prediction in females (Equation (2): 63%).

Bland-Altman Plots of PREE-MREE Differences

Finally, the Bland-Altman plots of PREE-MREE differences vs mean PREE-MREE values were shown in Figure 3 for the new equations, since those selected plots highlight the best agreement.

Table 3. Evaluation of New and Selected Predictive Equations in Crohn's Disease Patients, According to Sex.

| REE predictive equations | Males (n = 55) | | | Females (n = 35) | | |
|---------------------------|------------------------------|-------------------|--------------|------------------------------|-------------------|--------------|
| | Difference PREE-MREE, kcal/d | Bias ^a | RMSE, kcal/d | Difference PREE-MREE, kcal/d | Bias ^a | RMSE, kcal/d |
| | Mean (SD) | % | | Mean (SD) | % | |
| Equation (1) | -12 (123) | -0.2 | 93 | -19 (98) | -1 | 80 |
| Equation (2) ^b | -13 (117) | -0.3 | 88 | -15 (92) | -0.7 | 69 |
| HB | -58 (129) | -3.0 | 111 | -43 (109) | -2.7 | 96 |
| FAO | -11 (127) | -0.1 | 101 | -47 (113) | -2.8 | 99 |
| De Lorenzo | -41 (128) | -2.0 | 104 | -74 (109) | -5.1 | 109 |
| Schofield | -5 (128) | 0.2 | 102 | -60 (115) | -3.8 | 105 |
| Muller | 170 (145) | 11 | 189 | 150 (117) | 12 | 160 |
| Owen | -98 (133) | -5.2 | 135 | -164 (110) | -11.4 | 166 |

HB, Harris and Benedict; FAO, Food and Agriculture Organization; MREE, measured resting energy expenditure; PREE, predicted resting energy expenditure; REE, resting energy expenditure; RMSE, root mean square error.

^aMean percentage error between predicted and measured REE.

^bIncluding bioimpedance index and phase angle.

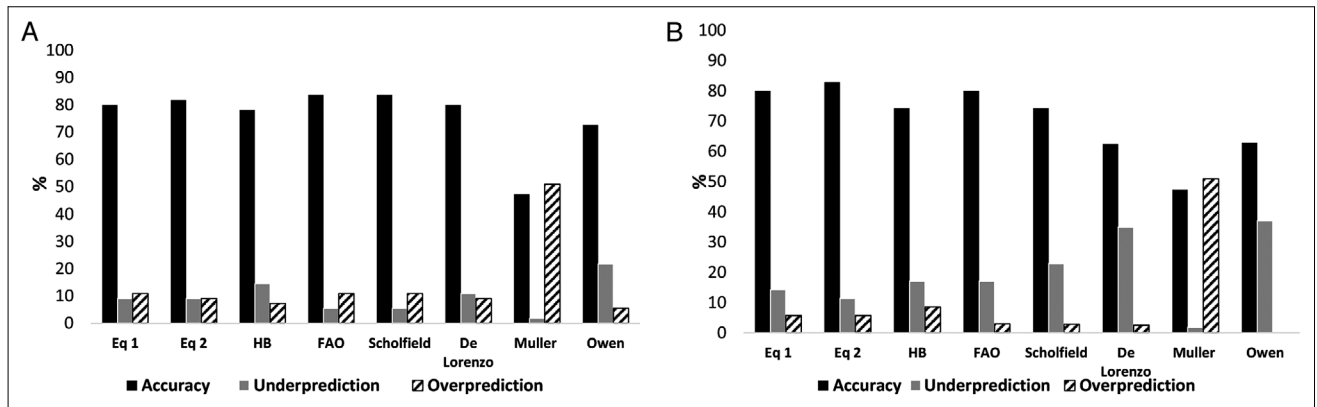


Figure 1. Prediction accuracy within $\pm 10\%$. Accuracy of prediction equations for measurements of resting energy expenditure within $\pm 10\%$ using each equation in 55 males and 35 females with Crohn's disease (A and B), respectively. HB, Harris and Benedict; FAO, Food and Agriculture Organization; Eq 1, equation generated by Model 1; Eq 2, equation generated by Model 2.

Discussion

This study aimed to develop and cross-validate, in individuals with CD, specific predictive equations for estimating REE and to explore the relationships of REE with age, disease activity, and main anthropometric and raw BIA variables. Our results showed that REE is largely predicted by BI-Index and body weight. The new equations have a good prediction power, providing by far the best results when accuracy was set within $\pm 5\%$.

The role of increased REE in the worsening of nutrition status in patients with CD⁵ is still unclear. So far, only a few studies have evaluated the use of predictive equations of REE in participants with CD, showing contrasting results. Three papers have given results on the HB equation, whereas no data are available on other predictive equations proposed in the literature for the general population. Barot et al³³

did not find any significant difference between MREE and PREE in 9 CD patients undergoing nutrition supplementation. Similarly, Chan et al⁴ showed that REE measured by IC in 54 CD patients did not significantly differ from PREE. On the contrary, Stokes and Hill⁸ found that MREE was 14% higher than that predicted. The great variability among those studies can be explained by different inclusion criteria, small sample size, and several methodological shortfalls.^{8,33}

Theoretically, disease-specific equations could ensure a better predictive accuracy. Hence, in the present study, we developed new predictive equations based on age and basic anthropometric parameters (age, weight, stature, and BMI) (Model 1) or also including raw BIA variables in the model (Model 2). Our results showed that weight and BI-Index were the best predictors of REE, whereas (not surprisingly) we found an inverse association between REE and age. In

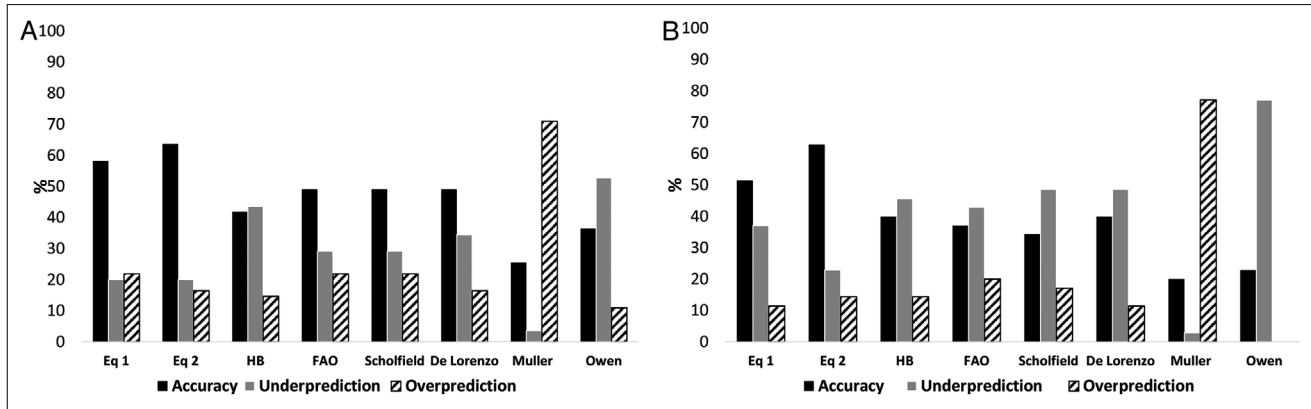


Figure 2. Prediction accuracy within $\pm 5\%$. Accuracy of prediction equations for measurements of resting energy expenditure within $\pm 5\%$ using each equation in 55 males and 35 females with Crohn's disease (A and B), respectively. HB, Harris and Benedict; FAO, Food and Agriculture Organization; Eq 1, equation generated by Model 1; Eq 2, equation generated by Model 2.

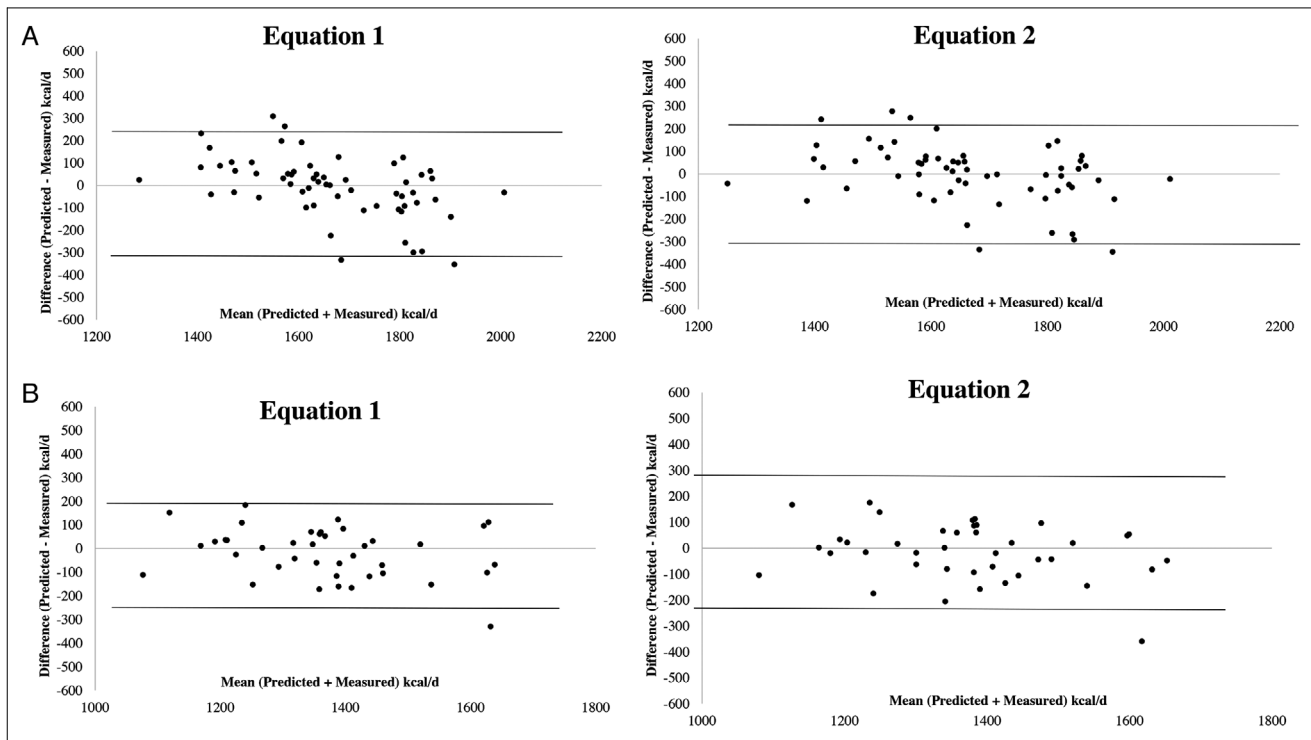


Figure 3. Bland-Altman plots. Bland-Altman plots between differences and mean predicted-measured resting energy expenditure using the new predictive equations (Equations (1) and (2)) in 55 males (A) and in females (B) with Crohn's disease. The dotted lines represent 2 SDs from the mean (limits of agreement).

the calibration groups, the new formula based on individual parameters (Model 1) led to similar SEEs in the 2 genders.

We have opted for including raw BIA variables in the regression model. Previously, we found that both BI-Index and PhA can estimate REE in individuals with obesity¹⁶ and anorexia nervosa¹⁷; and so they are expected to be potential predictors of REE in patients with CD as well. Moreover, we recently found that BIA-derived PhA is a

valid indicator of nutrition status in these patients, and its values were impaired with increasing disease activity.³⁴ First, we evaluated predictive equations of REE including raw BIA variables alone ($R^2 = 0.614$; SEE: 147 kcal/d) or in combination with age ($R^2 = 0.624$, SEE: 145 kcal/d), but the prediction power was lower compared with those including age and basic anthropometric variables. Secondly, we included age and basic anthropometric variables plus

raw BIA variables in the model, finding an increase of R^2 and a decrease in SEE values. In such a case, both sex and age were not identified as significant predictors, as occurred in both models also for disease activity (assessed by CDAI). However, 64% of patients recruited were in clinical remission, whereas those clinically active showed from mild to moderate disease activity, likely without having any influence on REE prediction. Nowadays, although CDAI is easy to get, simple, and applicable to large populations, it is no longer the “gold standard” tool for assessing disease severity among patients with CD. Alternatively, the use of the endoscopy-based scores, which address mucosal healing, might be more suitable for classifying disease severity, with potential effects on REE prediction, but this option needs to be further investigated.

In the validation group, the accuracy at the population level was very good for our equations, since it ranged within $\pm 1\%$, and the accuracy at individual level (within $\pm 10\%$) was also reasonable with equivalent figures in the 2 models ($\approx 80\%$). When we set the accuracy range within $\pm 5\%$, we found that the new equations gave by far the highest accuracy (Equation (1): $\approx 60\%$ in males and $\approx 51\%$ in females; Equation (2): $\approx 64\%$ in males and $\approx 63\%$ in females).

On the contrary, the bias at the population level was similarly within $\pm 5\%$ for the HB, FAO, and Schofield equations in both genders, confirming previous results on the use of the HB equation in CD.^{4,33} At the individual level, accuracy within $\pm 10\%$ was good and close to 80% for different equations, whereas, by setting the range within $\pm 5\%$, prediction accuracy sharply decreased to $< 50\%$ in both genders. The choice of analyzing and reporting prediction accuracy within $\pm 5\%$ will be useful for providing specific equations that are able to enhance REE prediction in the clinical setting. Surprisingly, we noted that the inclusion of raw BIA variables (Equation (2)) has slightly improved REE prediction compared with the equation based on anthropometric variables (Equation (1)); but, unfortunately, BIA is often not used or available in clinical settings. However, in the absence of BIA, we can reasonably opt for Equation (1), since its accuracy achieved almost 60% in males, which is good, and gave the best result ($> 51\%$) among females, being higher than other available equations based on anthropometric variables.

To the best of our knowledge, this is the first study that develops and cross-validates disease-specific equations to predict REE in participants with CD, also considering raw BIA variables as potential predictors. Overall, we performed a cross-sectional study in a reasonable sample of individuals, using known and documented methods and in line with previous studies that derived predictive equations for REE in healthy as well as ill participants. However, some limitations need to be considered. Firstly, as this is a single-center study including adult patients with CD

showing from mild to moderate disease activity, our findings need to be substantiated in other clinical subgroups or in different clinical settings. Secondly, although conventionally accepted, the use of CDAI might not be the best choice for defining disease activity because it is based on subjective criteria symptoms, resulting in a measure of severity of illness rather than of mucosal inflammatory activity.

In conclusion, the new, disease-specific equations proposed here to predict REE in individuals with CD give a good prediction accuracy at population level. Raw BIA variables are significant predictors of REE, but their inclusion in the model improves the prediction power by only a small extent. The new, disease-specific equations ensure a good accuracy also at the individual level and perform much better than the equations proposed in the literature for the general population. Further studies are needed to verify the reliability and usefulness of these new equations and to explore their role in estimating REE in clinically active patients.

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Statement of Authorship

M. Marra and I. Cioffi equally contributed to the conception and design of the research; F. Pasanisi contributed to the design of the research; O. Di Vincenzo, D. Morlino, I. Cioffi, N. Imperatore, M. C. Pagano, L. Alfonsi, L. Santarpia, and F. Castiglione contributed to the acquisition of the data; M. Marra contributed to the analysis and interpretation of the data; I. Cioffi and L. Scalfi drafted the manuscript. All authors critically revised the manuscript, agree to be fully accountable for ensuring the integrity and accuracy of the work, and read and approved the final manuscript.

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