# PROBABILISTIC ASSESSMENT OF SYMPTOMLESS INFLAMMATION IN CROHN'S DISEASE PATIENTS

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#### 1. INTRODUCTION

The new goal in treating Crohn's Disease (CD) patients is not only the improvement of symptoms and the induction of clinical remission but also the achievement of endoscopic and histological remission, with a reduced need of hospitalization and CD-related surgery (Pariente *et al.*, 2011).

Faecal calprotectin (FC) is a non-invasive biomarker currently used for monitoring inflammatory bowel disease (IBD) activity (Røseth *et al.*, 1999). Levels of FC have been shown to correlate well with both endoscopic and histological parameters of the disease activity in patients with CD (Sipponen *et al.*, 2008).

In particular a FC level  $\geq 150 \mu g/g$  intercepts endoscopically and histologically active disease (i.e. evident mucosal inflammation) and identifies patients in clinical remission but with active disease and prone to clinical relapse (Gisbert *et al.*, 2009).

The question of scientific interest in this study is to propose a method for identifying patients at high risk of symptomless but ongoing inflammation: the ability to discriminate subjects in this way would allow clinicians to improve the management of CD patients.

In this framework a key role is played by the Crohn's Disease Activity Index (CDAI) (Best *et al.*, 1976) which is the gold standard for monitoring clinical activity in CD patients (Yoshida, 1999).

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In clinical practice, CDAI scores < 150 are associated with quiescent (asymptomatic) or clinically non-active disease. CDAI scores  $\geq$  150 are indicative of clinically active disease, while scores over 450 indicate extremely severe disease.

The issue concerning the potential usefulness of Crohn's Disease Activity Index for assessing the endoscopic activity of the disease has been largely debated without achieving a definite conclusion (Brignola *et al.*, 1986; Cellier *et al.*, 1994; Jones *et al.*, 2008; af Björkesten *et al.*, 2012).

The aim of the present work is to set out the problem from a statistical point of view involving: a multivariate approach, which enables us to consider the joint role of CDAI and other clinical covariates using a single model; a logistic regression model, which allows us to quantify the risk of  $FC \geq 150 \mu g/g$  (i.e. an endoscopic active disease).

The results indicate that the proposed model provides a clear picture of the relationship among the selected predictors and the outcome of interest and has a very appreciable ability for identifying patients at high risk of symptomless but persistent inflammation.

The paper is organized as follows. The following Section introduces the data and describes materials and methods. Section 3 contains the estimation of the multivariate regression model and the and the assessment of its performance. Section 4 reports results and discussion. Finally, Section 5 contains our concluding remarks.

#### 2. Data, materials and methods

The data set consists of 193 consecutive Crohn's Disease patients attending an outpatient clinic, between January 2013 to December 2013, according to the following inclusion and exclusion criteria.

Inclusion criteria: age from 16 to 89 years; Crohn's Disease diagnosed at least 6 months previously. Exclusion criteria: patients with overlapping infectious, enterocolitis from Salmonella, Shighella, Campylobacter, Clostridium difficile or Cytomegalovirus; patients with a pouch; patients with a colostomy; patients with known neoplasia.

During the visit a blood sample and a sample of faeces were collected. The faecal sample was stored in a refrigerator at  $2^{\circ}C-8^{\circ}C$  for a maximum of 48 hours before being sent to the central laboratory of Sant'Orsola Hospital, Bologna, where the calprotectin assay was performed using a quantitative, enzyme-linked immunosorbent assay (ELISA). The blood sample was also analysed at the Sant'Orsola Hospital central laboratory to determine hematocrit values.

During the same visit, clinical parameters were recorded, using validated questionnaires, which, together with the hematocrit value, were used to determine the Crohn's Disease Activity Index. A summary of the clinical characteristics of our population study is shown in Table 1.

#### 3. The model

Logistic regression is well suited for studying the relation between a categorical or qualitative outcome variable (Y) and one or more predictors.

Consider a collection of p independent variables denoted by the vector  $\mathbf{x}' = (x_1, x_2, ..., x_p)$  and let the conditional probability that the outcome is present be

TABLE 1					
Clinical characteristics of CD patients. (PUFAs, polyunsaturated fatty acid	ds; CDAI,				
Crohn's disease activity index; TNF- $\alpha$ , tumor necrosis factor- $\alpha$ )					

Patient characteristics	Continuous Variables (median;min-max;mean $\pm$ SD ) Categorical Variables n (%)	
Age (years)	$36.5; 18 - 75; 41 \pm 15.05$	
Duration of disease (years)	$4; 1 - 36; 7.1 \pm 8.1$	
CDAI	$84.8; 32 - 401.6; 93.3 \pm 50.8$	
Faecal Calprotectin $(\mu g/g)$	$112; 4 - 1231; 199.4 \pm 218.86$	
Gender		
Male	87 (45.1%)	
Female	106 (54.9%)	
Disease Location	· · · · · · · · · · · · · · · · · · ·	
Ileum	83 (43%)	
Colon	27 (14%)	
Ileocolon	83 (43%)	
Prior Surgery		
Yes	87 (45.1%)	
No	106 (54.9%)	
Smoker	· · · · · · · · · · · · · · · · · · ·	
Yes	58(30.1%)	
No	135(69.9%)	
Disease behaviuor	· · · · ·	
Inflammatory	88(45.6%)	
Stenosing	59 (30.6%)	
Penetrating	46 (23.8%)	
Ongoing Treatement	· · · · · · · · · · · · · · · · · · ·	
no medication	31 (16.1%)	
5-aminosalicylic acid	57 (29.5%)	
n-3 PUFAs	2 (1.0%)	
corticosteroids	29 (15.0%)	
thiopurine	34 (17.6%)	
methotrexate	3 (1.6%)	
TNF- $\alpha$ antagonist	37 (19.32%)	

Predictors	Estimate	Std. Err.	z-value	p-value
Intercept	-1.6667	0.5594	-2.9791	0.0029
DIS_LOC	-2.5826	0.9803	-2.6345	0.0084
CDAI	0.0147	0.0057	2.5834	0.0098
$DIS\_LOC \times CDAI$	0.0257	0.0095	2.7017	0.0069

TABLE 2Estimation results for the FC model

denoted by  $\pi(\mathbf{x}) = Pr(Y = 1|\mathbf{x}).$ 

The multivariate logistic regression model has the general form

$$\pi\left(\mathbf{x}\right) = \frac{1}{1 + \exp\{-g\left(\mathbf{x}\right)\}}\tag{1}$$

where  $g(\mathbf{x}) = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \dots + \beta_p x_p$  is the logit of the model,  $\beta_0$  is the intercept and  $\beta_j$   $(j = 1, 2, \dots, p)$  are the slope parameters.

For our purposes, using the FC threshold usually adopted in clinical practice  $(150\mu g/g)$ , we define as dichotomous outcome the variable Y "dichotomised calprotectin" which assumes the following values:  $y_j = 0$  if subject j has  $FC < 150\mu g/g$ ;  $y_j = 1$  if subject j has  $FC \ge 150\mu g/g$  (j = 1, 2, ..., n).

Crohn's Disease is a complex disease influenced by several factors thus, with the aim of finding the best fitting and most parsimonious clinically interpretable model, we initially fit a multivariate logistic model including all clinically and intuitively relevant covariates.

The results showed that: Age and Duration of disease, considered as continuous covariates, did not result statistically significant; Disease behavior, Ongoing treatment, Prior surgery and Smoking considered as a polychotomous covariates and included in the model using design variables (two design variables for Disease behavior, six design variables for Ongoing treatment, one design variable for Prior surgery and one design variable for Smoking) did not result statistically significant; all interaction effects did not result statistically significant (Hosmer *et al.*, 2013).

The logistic regression analysis showed that two clinical covariates and their interaction effect are significant in estimating the probability of  $FC \ge 150 \mu g/g$  (i.e. the probability of an endoscopically and/or histologically active disease):

- 1. Disease location with two levels, Ileum and Colon or Ileocolon, coded in the model with the design variable *DIS\_LOC*: *DIS\_LOC* = 0 for Ileum; *DIS\_LOC* = 1 for Colon or Ileocolon;
- 2. CDAI score as a continuous covariate.

The estimated logistic regression coefficients with the corresponding standard errors are shown in Table 2.

Summarizing the estimated logit (or risk score) is

$$\widehat{g}(\mathbf{x}) = -1.6667 - 2.5826 \cdot DIS\_LOC + 0.0147 \cdot CDAI + + 0.0257 \cdot (DIS\_LOC \times CDAI)$$
(2)

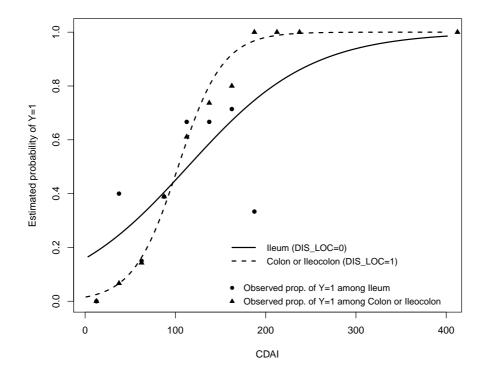


Figure 1 – Estimated probability curves and observed proportions stratified for  $DIS\_LOC$  versus CDAI score

#### 3.1. Model performance assessment

For the evaluation of model performance we consider overall performance, calibration and discrimination.

The model likelihood ratio, which assesses the overall significance of the model, is L.R. = 59.23 (p-value< 0.0001) and the goodness of fit test is z = 0.1384, p-value= 0.8899 (le Cessie and van Houwelingen, 1991). Therefore we can conclude that there is no evidence of lack of fit (Harrell, 2014; Agresti, 2002).

The model calibration, i.e. the agreement between predicted probabilities and observed frequencies of the event of interest, can be assessed by means of Figures 1 and 2.

Figure 1 shows the estimated probability curves for subjects with  $DIS\_LOC = 0$  (Ileum) and for those with  $DIS\_LOC = 1$  (Colon or Ileocolon) as a function of the CDAI score. Figure 1 also shows the observed proportions of subjects with  $FC \ge 150\mu g/g$  among those patients with  $DIS\_LOC = 0$  (triangles) and patients with  $DIS\_LOC = 1$  (circles) obtained by grouping the data into CDAI intervals ( $\le 25$ ], (25, 50], ..., (400, 425]).

Figure 2 shows the validation graph for the estimated model: "Non-parametric" is a smoothed function of observed events plotted versus predicted probabilities while "Ideal" is the ideal 45-degree line and the triangles indicate the outcomes

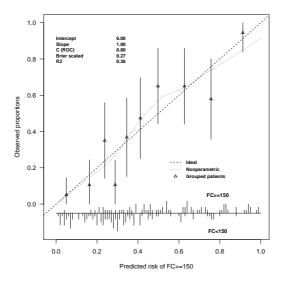


Figure 2 – Validation graph for the estimated model: "Non-parametric" is a smoothed function of observed events plotted versus predicted probabilities; "Ideal" is the ideal 45-degree line; triangles indicate the outcomes for deciles of prediction with their 95% confidence intervals

for deciles of prediction with their 95% confidence intervals. Several performance measures complete the legend (Steyerberg *et al.*, 2010; Royston and Altman, 2010). Examining Figures 1 and 2 we note a very appreciable agreement between predicted probabilities and observed frequencies of the event of interest.

The model's ability to discriminate between those subjects who experience the outcome of interest versus those who do not, can be assessed by means of several measures. The concordance (*c*-index) statistic is the most commonly used performance measure to indicate the discriminative ability of generalized linear regression models (Royston and Altman, 2010). It is the proportion, among all pairs of patients with different outcomes, in which the event probabilities predicted by the model are in the same order as the actual outcomes. The *c*index is thus the probability, across all patients, that a model will be correct in predicting that one patient has a higher probability of the outcome than another patient. For a binary outcome, the *c*-index is identical to the AUC: the Area Under the Receiver Operating Characteristic (ROC) Curve.

Usually, the outcome prediction is  $\hat{Y} = 1$  when  $\hat{\pi}(\mathbf{x}) > \pi_0$  and  $\hat{Y} = 0$  when  $\hat{\pi}(\mathbf{x}) \leq \pi_0$ , where  $\pi_0$  is the cut-off probability. Let us denote with *TP* the number of true positives (i.e.  $\hat{Y} = 1$  when Y = 1), *TN* the true negatives (i.e.  $\hat{Y} = 0$  when Y = 0), *FP* the false positives ( $\hat{Y} = 1$  when Y = 0) and *FN* the false negatives ( $\hat{Y} = 0$  when Y = 1). Sensitivity and specificity are defined in terms of *TP*, *TN*, *FN* and *FP*:

$$sensitivity = \frac{TP}{TP + FN} \tag{3}$$

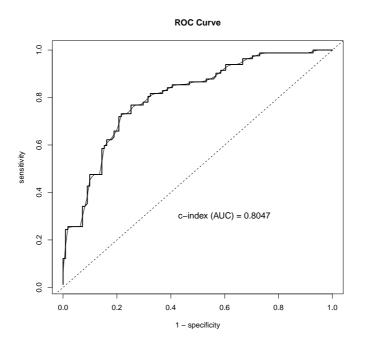


Figure 3 - ROC curve for the fitted logistic regression model

$$specificity = \frac{TN}{TN + FP}$$
 (4)

The ROC curve plots the *sensitivity* against one minus *specificity* for a range of possible cut-off probabilities.

The ROC curve for the fitted model is shown in Figure 3. In our case AUC=0.8047 (95%CI: 0.7429 - 0.8666) which indicates that the estimated model has an excellent discrimination ability.

#### 4. **Results and discussion**

The estimation results show that  $DIS\_LOC$  and CDAI are significantly associated with  $FC \ge 150 \mu g/g$  and that a significant interaction effect is present: for  $DIS\_LOC = 0$  (Ileum location) the estimated risk of  $FC \ge 150 \mu g/g$  (Y = 1) is relatively high for low values of CDAI and shows a moderate increasing-rate; for  $DIS\_LOC = 1$  (Colon or Ileocolon) the estimated risk of  $FC \ge 150 \mu g/g$  has a very different behaviour since it assumes low values for low CDAI scores but increases rapidly when CDAI increases (Figure 1).

The presence of a significant statistical interaction have important implications: the relationship between each of the interacting covariates (CDAI and Disease location) and the risk of having a mucosal inflammation depends on the value of the other interacting predictor (i.e. is not simply additive). This is evident in the odds ratio for  $DIS\_LOC$  which is not constant over the levels of CDAI.

TABLE 3 Estimated Odds Ratios (OR), with 95% confidence limits, for DIS\_LOC as a function of CDAI

CDAI	Estimated OR	95% CI
50	0.27346	0.09116 - 0.82027
100	0.98938	0.50121 - 1.95301
110	1.27955	0.62106 - 2.63623
120	1.65483	0.73796 - 3.71085
130	2.14018	0.85057 - 5.38502
140	2.76786	0.96042 - 7.97682
150	3.57965	$1.06975 { cdot} 11.97841$
200	12.95139	$1.66996 { cdot} 100.44462$

TABLE 4

Estimated odds ratio (with 95% confidence limits) for several increments in the predictor CDAI for disease locations Ileum and Colon or Ileocolon

	$DIS\_LOC=0$		DIS_LOC=1	
CDAI increments	OR	95% CI	OR	95% CI
5	1.07603	1.01784 - 1.13755	1.22369	1.13537 - 1.31889
10	1.15785	1.03600 - 1.29402	1.49743	1.28907 - 1 - 73947
15	1.24588	$1.05448  ext{-} 1.47202$	1.83240	$1.46357 \ 2.29416$
<b>20</b>	1.34061	$1.07329  ext{-} 1.67450$	2.24229	1.66170 - 3.02574
<b>25</b>	1.44254	$1.09244  extrm{-}1.90483$	2.74388	1.88665 - 3.99062
30	1.55222	1.11193 - 2.16684	3.35768	2.14205 - 5.26318

In Table 3 we show the estimated odds ratios and the corresponding 95% confidence intervals for  $DIS\_LOC = 1$ , versus  $DIS\_LOC = 0$  for several values of CDAI. It can be noted that the estimated odds ratio increases from a value of about 0.27 at CDAI = 50 and becomes significant approximately at CDAI = 150, where the lower confidence limit drops above 1.

As far as the covariate CDAI is concerned we have that also the odds ratio for CDAI increments are influenced by the disease location. In Table 4 are reported the estimated odds ratios for several increments in CDAI for the different disease locations considered. Examining these results, it can be seen that for an increment of 20 points in the CDAI score, the odds of having a mucosal inflammation are estimated to increase: of 1.34 times (95%CI: 1.07 - 1.67) for Ileal disease location; of 2.24 times (95%CI: 1.66 - 3.03) for colonic or ileocolonic disease location (Table 4).

Furthermore it is worth noting that the estimated model is able to quantify the risk of symptomless but ongoing inflammation since CDAI scores ranging from 100 to 150, which conventionally indicate a "clinically non-active" disease, are actually associated with a relevant probability of persistent mucosal inflammation. Specifically our model shows that when CDAI = 120 the estimated probability that  $FC \geq 150 \mu g/g$  is:

- 0.52 (95% CI: 0.38 0.66) if  $DIS\_LOC = 0$ ;
- 0.64 (95% CI: 0.51 0.76) if  $DIS\_LOC = 1$ .

When CDAI = 140 these probabilities are:

- 0.60 (95% CI: 0.45 0.72) if  $DIS\_LOC = 0$ ;
- 0.80 (95% CI: 0.70 0.88) if  $DIS\_LOC = 1$ .

These results conveniently contextualized in a clinical field are of significant value: as an example a symptomless CD patient with a CDAI = 140 with an ileal disease location has a 60% chance of having an endoscopically and/or histologically-evident "active disease"; while for a patient with the same CDAI score, but with a colonic or ileocolonic disease, the estimated risk having a mucosal inflammation rises up to 80%. A similar reasoning can be applied with regard CDAI = 120.

#### 4.1. Cut-off analysis

The discrimination ability of the estimated model is very appreciable (AUC = 0.8047). Therefore, even with several cautions, we shall explore the potentialities of a diagnostic decision rule based on the estimated model.

The optimal probability cut-point  $\pi_0$  for the purposes of classification is usually chosen according several criteria among which we mention:

• a) the criterion based on Youden's *J*-statistic (Youden, 1950) where the optimal cut-off point is the one that maximizes the sum of *sensitivity* and *specificity* minus one

$$J = max(sensitivity + specificity - 1)$$
(5)

• b) the cut-off which minimizes the distance d between the curve and the upper left corner of the graph, where

$$d = \sqrt{(1 - sensitivity)^2 + (1 - specificity)^2} \tag{6}$$

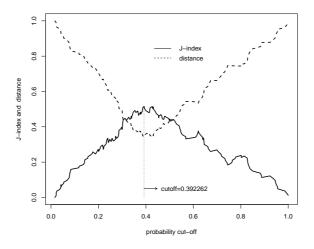
• c) the cut-off which maximizes both *sensitivity* and *specificity*.

For the fitted model: the probability cut-off that satisfies criteria a) and b) is  $\pi_0 = 0.392262$  for which J = 0.51604 and d = 0.342519 (sensitivity = 0.768293, specificity = 0.747748); the cut-off which maximizes both sensitivity and specificity is  $\pi_0 = 0.399818$ , this "optimal" choice for the probability cut-point is approximately where the sensitivity and specificity curves cross (sensitivity = 0.743902 and specificity = 0.747748). These results are summarized in Figure 4 and 5 respectively.

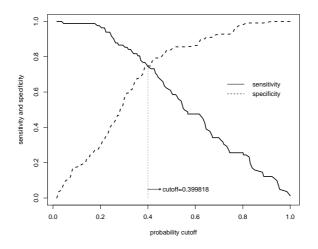
Thus, if the objective is to choose an optimal probability cut-off for the purposes of classification, one might select as cut-off point  $\pi_0 = 0.392262$  (criteria based on *J*-index and upper left corner distance), which corresponds to a 77% of correct classification for true positives and a 75% of true negatives.

Choosing as cut-off  $\pi_0 = 0.392262$ , the corresponding value of the risk score is  $g_{\pi_0}(\mathbf{x}) = -0.43781$ . The estimated risk-score (2) equals the value  $g_{\pi_0}(\mathbf{x}) = -0.43781$  when CDAI = 94.40 for subjects with  $DIS\_LOC = 0$  and when CDAI = 83.84 for subjects with  $DIS\_LOC = 1$ .

Thus, contextualizing these findings in a diagnostic framework, a patient with a CDAI score less than 94.40 with an ileal disease location is classified by



Figure~4 – Plot of Youden's J statistic and d (top-left corner distance) versus all possible probability cut-off points.



 $\label{eq:Figure 5-Plot of sensitivity and specificity versus all possible probability cut-off points.$ 

the model as  $\hat{Y} = 0$ : a subject that likely does not have an ongoing mucosal inflammation. A patient with ileal disease location with  $CDAI \ge 94.40$  is classified by the model as  $\hat{Y} = 1$ : a subject that likely does have an endoscopically or histologically active disease.

Similarly, a person who has a colonic or ileocolonic disease with a CDAI score less than 83.84 is classified as  $\hat{Y} = 0$ : a patient that likely does not have an inflammation. Finally, a patient who has the same disease location, but with  $CDAI \geq 83.84$  likely does have an active mucosal inflammation.

It should be noted that in this work we do not propose changing the CDAI threshold currently used in clinical practice. Rather, the results emphasize that values of CDAI usually considered as not clinically relevant (< 150) actually correspond to situations in which there is a high risk that Crohn's Disease is endoscopically and/or histologically active.

# 5. Concluding Remarks

To our knowledge, this is the first study that investigated the relationship among Crohn's Disease Activity Index, Disease location and Faecal calprotectin in a large cohort of CD patients.

Naturally we are aware that the results obtained are valid only for the patients of our population study, and that therefore some caution is required before extending our results in general. Keeping this caution in mind, we can conclude that our proposal could provide a useful clinical decision aid.

The estimated model with CDAI and disease location as predictors allows to estimate the probability of  $FC \geq 150 \mu g/g$ : a situation that is strongly indicative of endoscopically and /or histologically active disease. Furthermore, it is worth noting that the proposed model allows to take into account that the effect of one predictor variable on the response variable is different at different values of the other predictor. To facilitate the use and interpretation of our results we provide a nomogram (Appendix) which allows to compute the probability of having faecal calprotectin  $\geq 150 \mu g/g$ .

Finally, the estimated model, besides having an excellent discrimination ability, is able to intercept the presence of a symptomless inflammation since the results indicate that CDAI scores ranging from 100 to 150, which conventionally indicate a "clinically non-active" disease, are actually associated with a high probability of the presence of mucosal inflammation.

This ability is potentially very useful for improving patient management, scheduling invasive procedures at the right time and tailoring the best treatment strategy in order to control the clinical evolution of CD in such a way to reduce hospitalization, CD-related surgery and future disability.

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# Appendix

A nomogram (Figure 6) is a convenient graphical tool to compute the probability of an outcome of interest (Harrell, 2014; R Core Team, 2013).

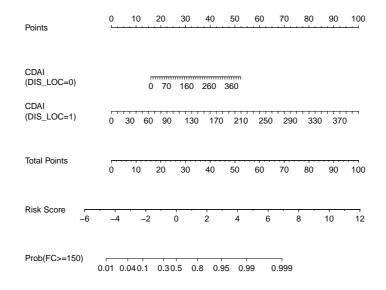


Figure 6 – Nomogram for computing the estimated risk of an active disese

The location disease establishes which CDAI scale must be used. Therefore based on the CDAI score the corresponding total point value can be read on the "Total Points" axis. Therefore, the probability of  $FC \geq 150 \mu g/g$  for each subject can be calculated from the scale presented on the bottom line.

For instance, let us consider a patient who has the disease localized in the ileum with a CDAI score of 130: this subject has about 28 points and the corresponding estimated probability of  $FC \geq 150 \mu g/g$  is about 0.56. Similarly, a patient who has colonic or ileocolonic disease and with a CDAI score of 130 has about 33 points and the corresponding estimated probability of having an active disease about 0.73.

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#### SUMMARY

In Crohn's Disease it is extremely important to detect the presence of symptomless mucosal inflammation in such a way as to prevent the evolution of the disease. The aim of this study is to identify predictor variables for estimating the risk of the presence of mucosal inflammation. The results show that the estimated model provides a clear picture of the relationship among the selected predictors and the outcome of interest and has a very appreciable ability for identifying patients at high risk of symptomless but persistent inflammation.

 $K\!eywords:$ Logistic Regression; Odds Ratio; Risk Assessment; Diagnostic Rule; Crohn's Disease.