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**Artificial intelligence and inflammatory bowel disease: Where are we going?**

Da Rio L *et al.* Artificial intelligence and IBD

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**Abstract**

Inflammatory bowel diseases, namely ulcerative colitis and Crohn’s disease, are chronic and relapsing conditions that pose a growing burden on healthcare systems worldwide. Because of their complex and partly unknown etiology and pathogenesis, the management of ulcerative colitis and Crohn’s disease can prove challenging not only from a clinical point of view but also for resource optimization. Artificial intelligence, an umbrella term that encompasses any cognitive function developed by machines for learning or problem solving, and its subsets machine learning and deep learning are becoming ever more essential tools with a plethora of applications in most medical specialties. In this regard gastroenterology is no exception, and due to the importance of endoscopy and imaging numerous clinical studies have been gradually highlighting the relevant role that artificial intelligence has in inflammatory bowel diseases as well. The aim of this review was to summarize the most recent evidence on the use of artificial intelligence in inflammatory bowel diseases in various contexts such as diagnosis, follow-up, treatment, prognosis, cancer surveillance, data collection, and analysis. Moreover, insights into the potential further developments in this field and their effects on future clinical practice were discussed.

**Key Words:** Inflammatory bowel disease; Artificial intelligence; Machine learning; Crohn’s disease; Ulcerative colitis; Computer-aided diagnosis

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**Core Tip:** Management of patients with inflammatory bowel disease is complex and costly. Therefore, in this field being able to improve clinical efficiency and optimize healthcare resources is of paramount importance. In this regard, artificial intelligence appears to be an extremely promising tool with a significantly wide range of potential applications that encompass diagnosis, treatment, and follow-up. This review summarized the most recent significant scientific findings regarding the application of artificial intelligence in inflammatory bowel diseases, providing a picture of the current state of the field and future perspectives.

**INTRODUCTION**

Ulcerative colitis (UC) and Crohn’s disease (CD) are chronic and relapsing conditions that affect the gastrointestinal tract and are labelled as inflammatory bowel diseases (IBD)[1,2]. In the last few decades, IBD has emerged as a global disease with a conspicuous burden on public health and healthcare costs[3].

The diagnosis and management of IBD is complex and implies the interplay and synergy of various specialists, including clinical gastroenterologists, gastrointestinal endoscopists, radiologists, pathologists, surgeons, and clinical nutritionists[4].

Along with histological assessment, endoscopy has an important role in the diagnosis and the follow-up of IBD while also being the mainstay for colorectal cancer surveillance[5]. Due to the essential role of endoscopy and imaging in gastroenterology, and particularly in IBD, artificial intelligence (AI)-based image analysis can be utilized in numerous applications such as evaluation of endoscopic lesions, cancer detection, and assessment of disease activity (*e.g.*, prognosis and response to treatment)[6].

**Artificial intelligence**

AI is an umbrella term that encompasses any cognitive function developed by machines for learning or problem solving. A particular subset of AI is represented by machine learning (ML), a discipline that uses large datasets as an input in order to identify patterns of interaction among variables, allowing the possibility to apply these findings to new data[7].

A further subset and evolution of ML is deep learning (DL), which mimics the neuronal interaction in the human brain to develop artificial neural networks and subsequently convolution neural networks (CNN) that are then able to use the input data in an autonomous fashion with the aim of assessing predictive factors of a specific outcome through the development of multiple levels of abstractions[8].

Among the numerous implementations of AI in medicine, a promising field is that of automatic collection of complex and nuanced clinical data from electronic medical records through natural language processing, the subset of AI that studies the interpretation of the human language made by the computer[9,10].

Another fertile ground of application of AI is the interpretation of radiologic imaging. A computer model based on computed tomography enterography allows for accurate characterization of intestinal fibrosis in CD, in some instances outperforming human radiologists[11].

In the context of gastrointestinal endoscopy, AI has found application in two main fields, namely in the detection of mucosal lesions, with computer-aided detection (CADe) and in the characterization of mucosal lesions, with computer-aided diagnosis, along with the evaluation of the quality of the endoscopic procedure itself, with computer aided monitoring[12-20].

Therefore, implementation of AI in IBD is a promising tool for improving the assessment of disease activity and reducing the interobserver variability in grading such activity[21]. In addition, similar to what is already happening in the general population, CADe systems could eventually improve detection of IBD-associated dysplasia[22]. Finally, AI may also allow for application of precision medicine through the analysis of large databases, correlating differences in the biology of the patient with differences in the susceptibility to develop IBD, the activity of the disease, and the response to specific therapies[23] (Figure 1). Such considerations are further supported by the growing number of clinical studies on the application of AI for IBD in recent years (Figure 2).

Table 1 summarized the most impactful studies on AI in several fields of IBD that were identified after a literature review using PubMed (MEDLINE) from inception to November 6, 2022. The impact measure of the articles cited was assessed *via* the *Reference Citation Analysis* (RCA; Baishideng Publishing Group Inc.) tool.

**AI in etiology, pathogenesis, and diagnosis of IBD**

Although it is not yet thoroughly understood, the etiology of IBD is known to depend upon the complex interplay of genetic, microbial, and environmental factors and the immune system[24]. In this context, AI allows for more effective data analysis to evaluate the role of specific genes in the predisposition and development of IBD.

Genome-wide association studies have been employed in order to identify sequence variations related with specific conditions, with over 200 genes found to be potentially implicated in IBD etiology[25]. Such complex and large genomic data, however, are difficult to assess with standard analytical tools. In this context, studies have shown that ML and DL can effectively analyze genome-wide association study data and overcome part of the inherent limitation of such methodology through algorithms of minimum Redundancy-Maximum Relevance and incremental feature selection[26-28].

Gut microbiota is thought to play a relevant role in the complex etiology and pathogenesis of IBD, partly concurring in providing a substrate that may range from protective to proinflammatory[29]. Moreover, in the context of dysbiosis, the taxonomic composition appears to vary between patients affected by IBD and unaffected subjects and between IBD subtypes, UC, and CD. Promising studies in this field show that the application of ML algorithms on the analysis of gut microbiome data may assist the clinician in diagnosing IBD[30,31].

AI has also been employed in the attempt of supporting the conventional diagnosis of IBD. In this regard, isolated and combined endoscopic and histological parameters were used to develop ML models, which was accurate in the classification of pediatric patients affected by IBD[32].

Additional models based on confocal laser endomicroscopy were developed and by quantitative analysis through cryptometry were able to provide a diagnosis of IBD with high sensitivity and specificity. Moreover, these models managed to successfully differentiate between UC and CD[33].

**AI applications for endoscopic assessment in IBD**

In order to rigorously define endoscopic disease activity through specific parameters and limit interobserver variability, a plethora of endoscopic scores have been developed in the field of IBD. The most relevant scores are the CD endoscopic index of severity and the simple endoscopic score for CD, the Mayo endoscopic score, the UC endoscopic index of severity, and the UC colonoscopic index[34-38].

Implementation of AI in this field might be a further step towards reproducibility and homogeneity of endoscopic findings. The first successful attempts in identifying the presence of mucosal remission or activity *via* AI were made using a dedicated CADe system based on CNN trained on large datasets of endoscopic still images in patients affected by UC[39,40].

A subsequent development was represented by the implementation of neural networks in order to assess disease activity not only on still images but on the entirety of colonoscopy videos in real time. Also in this field, AI was efficient in recognizing active disease, calculating scores, predicting the risk of clinical relapse, and supporting clinicians in real time decision making for treatment[41-44].

Moreover, deep neural networks trained on endoscopic images and histological reports of UC managed to identify patients in endoscopic remission and histological remission with such an accuracy that it may potentially obviate the need for biopsy collection and analysis to identify patients in remission[45].

With regards to prediction of *in vivo* histological activity, the first dedicated CADe system was recently developed for application in endocytoscopy, yielding a high accuracy when compared with the gold standard of the pathologist’s assessment[46].

Finally, in order to overcome the fact that conventional DL systems train CNN based on the subjective scoring of images done by clinicians, novel approaches have been developed based on algorithms that only analyze parameters such as pixel color data and vascular pattern recognition. Objective computer-based and operator-independent tools provided a dedicated score that significantly correlated with both endoscopic and histological scoring systems[47].

Video capsule endoscopy (VCE) represents one of the main modalities of investigating the small bowel in suspect or established CD[48]. The main scoring systems used for disease activity quantification are the Lewis score and the capsule endoscopy CD activity index, which consider parameters such as extension, grading of inflammation, and presence of strictures[49,50].

VCE video review, however, is time-consuming and requires a high level of attention during the observation of thousands of frames. In order to simplify this task, AI has been implemented with the objective of reducing the time needed for image assessment by selecting the most relevant frames or portions of the video. Several CNN models have been developed in order to recognize pathologic findings such as erosions, ulcers, and strictures with very high accuracy[51-54]. Moreover, a study that assessed the review time for VCE showed how the employment of AI systems allowed clinicians to complete the task in a fraction of the time with no differences in overall accuracy[55].

In addition to recognition of disease activity, AI may find a relevant role in the surveillance of colorectal neoplasia in patients with IBD since a history of long-standing disease is a significant risk factor in developing colorectal cancer[5]. While AI is proving increasingly useful in the detection of colonic neoplasia in the general population with dedicated CADe systems, first attempts at developing specific applications for surveillance in IBD are being made[13]. As of today, successful detection of dysplasia has been described in case reports regarding the application of EndoBRAIN, a CADe system, in endoscopy and endocytoscopy[56,57].

Along with CADe systems, AI may prove to be an effective and safe tool for real-time quality improvement of the examination as well (*i.e.* withdrawal time, checking for blind spots), thereby potentially increasing the adenoma detection rate also in IBD patients[58].

**AI in IBD histology assessment**

Histological remission is now considered an adjunct target of treatment in UC and arguably the most stringent way to assess remission[59]. Growing evidence shows that persistence of histologic disease activity, even in the absence of macroscopic endoscopic inflammation, is associated with worse clinical outcome and risk of relapse[60]. More than 30 histological scores have been proposed to grade UC histological activity, but their application in clinical practice remains minimal, mainly due to the impracticality of the scores[61,62]. Even when the scores are applied, for example in clinical trials, the interobserver variability is very high, limiting comparison and reproducibility. Indeed, clinical trials increasingly resort to expensive central reading systems so that all biopsies are evaluated by few highly-qualified pathologists to reduce variability. Therefore, AI-based systems to automatically read UC biopsies would be of great help standardizing the assessment and reducing interobserver variability[63].

Trials in this field are ongoing, and initial results are promising. The first attempt to develop a CADe model to assess UC biopsies focused on eosinophils. The system had a good agreement compared to the manual count performed by human pathologists (interclass correlation coefficient = 0.81-0.92) but did not demonstrate an association between eosinophils counted and overall inflammatory activity[64]. More recently, Gui *et al*[65] proposed to simply assess UC activity by taking into consideration the sole presence or absence of neutrophils, the hallmark of active inflammation. They proposed a simplified score, the PICaSSO histologic remission index that was then embedded into a CADe system that was able to distinguish histological activity from remission in biopsies of UC with good accuracy. Further improvements to the same CADe have been recently presented showing that the neutrophil-only assessment by the CADe is largely consistent with mainstay scores such as Robarts and Nancy histological indexes[66].

Other studies on CADe systems for assessment of UC are ongoing and preliminary results are promising[67,68] (Table 2). Further applications of CNN in computational pathology have been studied in order to empower the pathologist’s accuracy and efficiency, with models trained on whole slide images that were able to effectively support the pathologists by excluding non-diagnostic slides, while retaining optimal sensitivity[69].

**AI in prognosis and prediction of response to treatment**

Being able to predict the course of disease with regards to severity and progression is of the utmost importance in IBD patients in order to implement specific management strategies accordingly[70].

Studies have showed that by employing natural language processing and ML algorithms trained on several clinical data from electronic medical records (*i.e.* demographics, laboratory tests, endoscopy reports) it was possible to predict disease severity and surrogate markers of disease flare such as outpatient steroid use or hospitalization[9,71,72].

Furthermore, initial studies on the use of ML to predict the prognosis of patients affected by UC have shown that the endoscopic mucosal healing predicted by a deep neural network model was associated with prognostic features (reduced risks of hospitalization and colectomy) with statistical significance[73].

Lack of response to biological therapies (around 1 out of 3 patients with anti-TNF-alpha therapy) or progressive decline in response over time is an issue of paramount importance in IBD, with a significant economic toll on healthcare costs[74]. For this reason, the possibility to anticipate the likelihood of a patient responding to a specific biological drug prior to its start represents a cost-effective approach to treatment individualization.

ML has been implemented to deal with the complexity of such topics with encouraging results. Through the use of ML on clinical data from trials, random forests were developed to predict the response to biologics such as infliximab, vedolizumab, and ustekinumab, both in CD and in UC with encouraging results[75-78]. The integration and analysis of patients’ molecular and clinical features *via* ML algorithms therefore appears to be a promising field with great potential in terms of patient management and optimization of healthcare costs[79].

**AI in preclinical settings and drug discovery**

Design and discovery of new drugs are complex processes hampered by high costs and time consumption. Furthermore, especially in recent years, the process of drug discovery has to deal with the increasing wealth and complexity of data from various “omics” (*i.e.* genomics, proteomics)[80].

AI, however, is proving to be a crucial tool in the development of novel drug candidates modernizing this field as well. Among the various implementations of ML and DL in the drug discovery process we can list peptide synthesis, virtual screening (structure-based and ligand-based), prediction of toxicity, drug monitoring and release, quantitative structure-activity relationship, and drug repositioning[81].

In the field of IBD AI has already been employed in drug discovery with encouraging initial results. Computational approaches have been used to identify metabolite-target interactions using the dataset of the IBD cohort from the Human Microbiome Project 2, followed by ML analyses aimed at ranking metabolites according to their importance in IBD and identifying possible human targets through virtual ligand-based screening. Overall, 983 high quality connections between metabolites from the gut microbiota and human proteins possibly relevant to IBD were identified, thus providing multiple novel drug targets for potential immune therapies[82].

Another relevant application of DL was the construction of a scaffold-based molecular design workflow aimed at developing drug candidates targeted towards the discoidin domain receptor 1. Through a deep generative model, molecular docking and virtual profiling, a high-quality scaffold-based molecular library was established, subsequently leading to the synthesis of a molecular compound that effectively inhibited the expression of proinflammatory cytokines, showing extraordinary kinase selectivity and significant therapeutic protection in *ex vivo* and *in vivo* animal models, respectively[83].

**CONCLUSION**

Based on the numerous advances in the field of IBD in recent years, it is foreseeable that AI will gain an ever-greater role in the standard patient care, ranging from evaluation of the risk of developing IBD to assistance in the assessment of mucosal activity or detection of dysplasia to support in histopathological reporting, prediction of disease course, and treatment efficacy (Table 1). At the same time, the integration of AI in daily practice will lead to changes in clinical practice itself, getting us closer to the concept of precision medicine and its subsequent improvement in the quality of care and optimization of healthcare costs (Figure 3).

As for research, AI is proving to be an irreplaceable aid in the collection and analysis of large data, while also limiting subjectiveness and interobserver variability, simplifying standardization and providing a feasible alternative to the need of independent central reading in clinical trials. The implementation of AI in everyday practice is expected to improve diagnostic accuracy and reproducibility by allowing for a better standardization of lesion features and classification and by increasing the detection rate of small and subtle mucosal abnormalities or lesions. Moreover, evidence shows how AI may have a role in advanced endoscopic techniques, such as confocal laser endomicroscopy or molecular imaging. Through these advanced application AI will support the clinician by detecting microscopic and even molecular details otherwise invisible, thereby paving the way to the potential incorporation of ultrastructural and molecular endpoints in IBD endoscopy.

Nonetheless, with regard to future applications of AI, there are several issues that will need solving, such as how to provide more transparency of the AI algorithms, how to assess and choose among different models with similar purposes, and how to effectively allocate resources in the plethora of models that will be available, to only list a few[84]. To the best of our knowledge, this is the first review on the roles and the future perspectives of AI in IBD that considers the first clinical applications in a real world setting that are now available from the most recent clinical studies. Nevertheless, the chief limitation is represented by the limited amount of evidence that support our review, which is inevitably due to the scarcity of data currently available for the topic itself from the literature. Undoubtedly further studies are direly needed to build a more robust and comprehensive foundation for future analyses. In brief, while the human component is unlikely to be substituted altogether in future clinical practice and while a few questions still await an answer, AI is an extremely promising means of improvement in patient care and resource optimization.

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**Footnotes**

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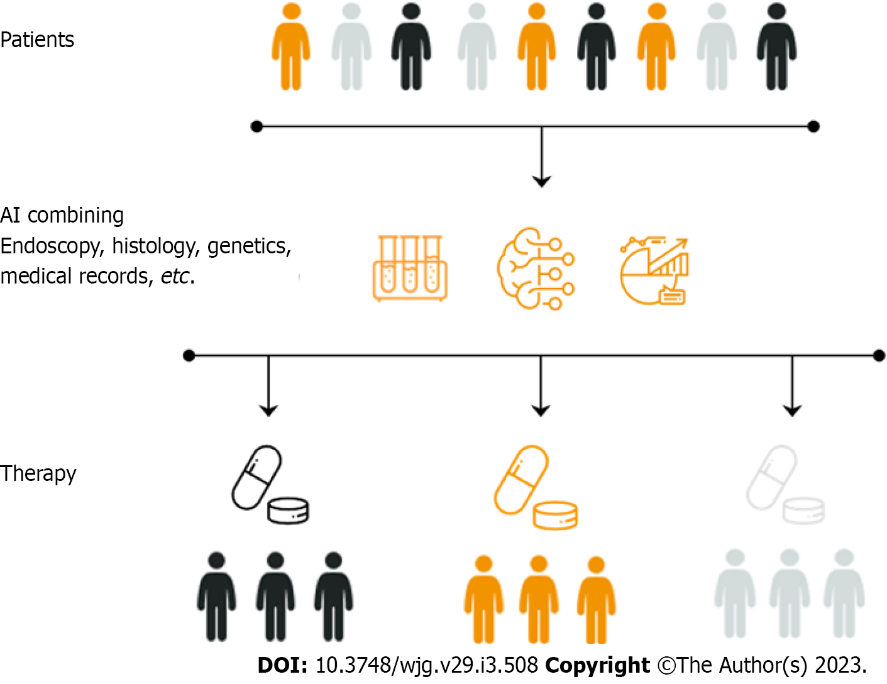
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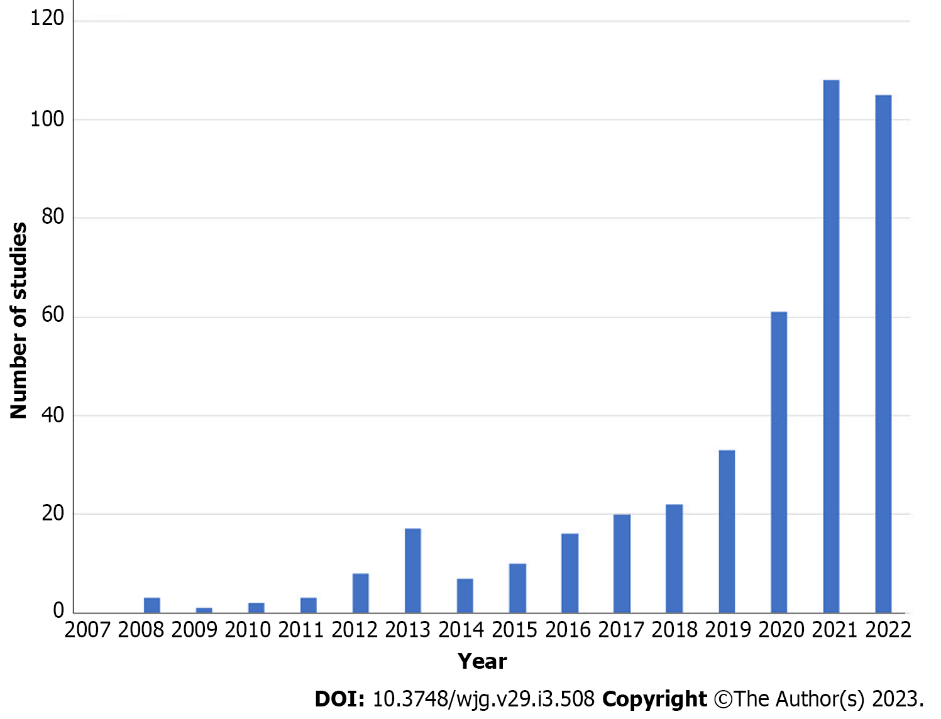
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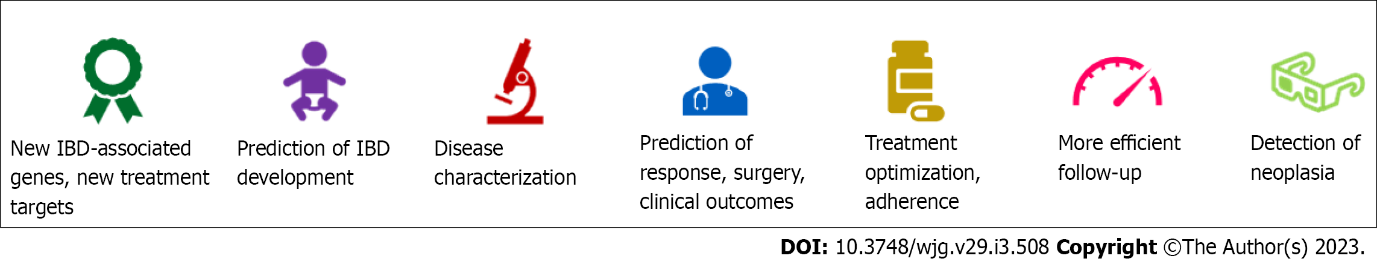
**Figure Legends**



**Figure 1 Application of artificial intelligence in inflammatory bowel diseases for precision medicine.** AI: Artificial intelligence.



**Figure 2 Clinical studies on the application of artificial intelligence in inflammatory bowel diseases in the last 25 years (as of November 6, 2022).**



**Figure 3 Fields of application of artificial intelligence in inflammatory bowel diseases.** IBD: Inflammatory bowel disease.

**Table 1 Main studies of artificial intelligence application in the various fields of inflammatory bowel diseases**

|  |  |  |  |
| --- | --- | --- | --- |
| **Ref.** | **Field** | **Study features** | **Main finding** |
| Isakov *et al*[26], 2017 | IBD genetics | ML model to assess 16390 genes in IBD and healthy patients | Identified 347 IBD-risk genes (67 newly identified) |
| Cheng *et al*[27], 2019 | IBD genetics | Software analysis to assess the genetics of 32713 IBD patients | Identified several genes potentially involved in UC; identification of 11 common Gene Ontology terms for UC |
| Yuan *et al*[28], 2017 | IBD genetics | Software analysis to assess 12754 genes in IBD and healthy patients | Identified 41 genes closely associated with IBD |
| Mihajlović *et al*[30], 2021 | IBD and microbiota | ML classification algorithm to identify IBD from 1638 fecal samples | Confirmed strong connection between IBD and specific fecal microbial species |
| Manandhar *et al*[31], 2021 | IBD and microbiota | ML model analysis of fecal microbiota from 729 IBD patients and 700 healthy controls | Identified of 117 bacterial taxa with a potential role in diagnostic screening of IBD |
| Mossotto *et al*[32], 2017 | IBD diagnosis | ML model to assess 287 pediatric patients with IBD | Accuracy of 83.3% of the combined endoscopy-histology ML model in the classification of pediatric IBD patients |
| Quénéhervé *et al*[33], 2019 | IBD diagnosis | AI analysis of CLE images from 50 IBD patients and 9 healthy controls | AI analysis had 100% sensitivity and specificity for IBD diagnosis, 92% sensitivity and 91% specificity of IBD differential diagnosis |
| Ananthakrishnan *et al*[9], 2013 | IBD diagnosis and data collection | NLP model trained and validated on 700 UC patients and 700 CD patients to improve case definition and identification from EMRs | NLP model provided better accuracy (AUC 0.94-0.95) than models using only the International Classification of Diseases 9th revision for IBD case definition and identification |
| Stidham *et al*[39], 2019 | IBD endoscopy | DL model for UC severity trained on 16514 endoscopic images | Similar performance of the DL model and experienced human reviewers in grading UC endoscopic severity |
| Ozawa *et al*[40], 2019 | IBD endoscopy | CNN-based CADe system for UC severity trained on 26304 endoscopic images | CADe system had AUCs of 0.86 and 0.98 in the identification of Mayo score 0 and 0-1, respectively |
| Maeda *et al*[41], 2021 | IBD endoscopy | Endoscopic AI model used in real time on 135 UC patients in clinical remission | Endoscopic applications of real time AI predicted clinical relapse of UC with statistical significance |
| Gottlieb *et al*[42], 2021 | IBD endoscopy | DL algorithm to assess UC severity on 795 full-length endoscopy videos | DL algorithm showed significant inter-rater agreement to human central readers for prediction of UC severity |
| Yao *et al*[43], 2021 | IBD endoscopy | Endoscopic AI model (CNN) to assess UC grading used on 169 endoscopy videos and compared to dual central reader review | AI model approximated the scoring of experienced reviewers for grading of UC endoscopic activity |
| Byrne *et al*[44], 2021 | IBD endoscopy | DL model (CNN) to detect and assess UC activity leveraged on > 375000 frames | DL model resulted in well aligned scoring guidelines and experts’ performances |
| Takenaka *et al*[45], 2020 | IBD endoscopy | DL algorithm trained on endoscopic images and biopsy results and tested on 875 UC patients | DL model identified with an accuracy > 90% patients in endoscopic and histologic remission |
| Maeda *et al*[46], 2019 | IBD endoscopy (endocytoscopy) | CADe system to predict persistent histologic phlogosis from endocytoscopy validated on 100 UC patient | CADe system provided a diagnostic accuracy of 91% with perfect reproducibility for identification of persistent histologic inflammation |
| Bossuyt *et al*[47], 2020 | IBD endoscopy | AI algorithm based on pixel color data and pattern recognition from endoscopic images tested on 55 patients | AI algorithm (“red density”) provided an objective computer-based assessment of UC disease activity with good correlation with endoscopic and histological scoring systems |
| Aoki *et al*[51], 2019 | IBD endoscopy (VCE) | AI system (CNN) tested on 10440 small bowel images for detection of erosions and ulcers in CD | AI system showed an accuracy of 90.8% for detection of erosions and ulcers |
| Klang *et al*[52], 2020 | IBD endoscopy (VCE) | DL algorithm applied on 17640 VCE images for ulcer detection in CD | DL algorithm provided an accuracy ranging from 95.4% to 96.7% with an AUC of 0.99 for ulcer detection |
| Klang *et al*[53], 2021 | IBD endoscopy (VCE) | DL model applied on 27892 VCE images for identification of intestinal strictures in CD | DL model showed an accuracy of 93.5% in stricture identification and excellent differentiation between strictures and other lesions |
| Ferreira *et al*[54], 2022 | IBD endoscopy (VCE) | DL model trained and validated on 8085 VCE images for detection of erosions and ulcers in CD | DL model provided an accuracy of 92.4% and a precision of 97.1% for lesion detection |
| Aoki *et al*[55], 2020 | VCE | Comparison between standard endoscopist reading and reading after AI model screening of 20 full-length VCE videos | The mean VCE video reading time was significantly shorter after AI model (CNN) screening compared to standard reading |
| Maeda *et al*[56], 2021 | IBD endoscopy (surveillance) | Case report of dysplasia detection by AI system in a patient with long standing UC | AI system (EndoBRAIN) identified 2 colonic lesions that harbored low-grade dysplasia upon histological examination |
| Fukunaga *et al*[57], 2021 | IBD endoscopy (surveillance) | Case report of dysplasia detection by AI system in a patient with long standing UC | AI system (EndoBRAIN) identified rectal lesions that harbored high-grade dysplasia upon histological examination |
| Reddy *et al*[72], 2019 | IBD prognosis | ML model employed on 82 CD patients’ EMRs to predict disease course | ML model predicted inflammation severity with high accuracy (AUC 92.8%) from EMR data |
| Takenaka *et al*[73], 2022 | IBD prognosis | ML model validated on endoscopic images and biopsy results from 875 UC patients to predict disease course | Histologic remission detected by the ML model correlated with a significant reduction in clinical relapse, steroid use, hospitalization, and colectomy |
| Li *et al*[75], 2021 | Response to treatment | ML model employed on 174 CD patients to predict response to infliximab | ML model based on clinical and serological parameters showed an accuracy of 0.85 for prediction of response to infliximab |
| Waljee *et al*[76], 2018 | Response to treatment | AI model employed on 472 CD patients to predict response to vedolizumab | AI model based on clinical and serological parameters was able to identify patients that achieved a corticosteroid-free biologic remission at week 52 of vedolizumab |
| Waljee *et al*[77], 2018 | Response to treatment | ML algorithm employed on 491 UC patients to predict response to vedolizumab | ML algorithm based on clinical and serological parameters was able to identify patients that achieved a corticosteroid-free biologic remission at week 52 of vedolizumab |
| Doherty *et al*[78], 2018 | Response to treatment | AI model to assess response to treatment with ustekinumab in 306 patients with CD | AI model detected patients in remission based on clinical data and fecal microbiota at week 6 and 22 of ustekinumab |

AI: Artificial intelligence; AUC: Area under the curve; CADe: Computer-aided detection; CD: Crohn’s disease; CLE: Confocal laser endomicroscopy; CNN: Convolution neural networks; DL: Deep learning; UC: Ulcerative colitis; EMR: Electronic medical record; IBD: Inflammatory bowel disease; ML: Machine learning; NLP: Natural language processing; VCE: Video capsule endoscopy.

**Table 2 Artificial intelligence application for histological assessment of ulcerative colitis**

|  |  |  |  |  |
| --- | --- | --- | --- | --- |
| **Ref.** | **Study design** | **Population** | **Outcome** | **Results** |
| Vande Casteele *et al*[64], 2022 | Cohort study | Colonic biopsies from 88 UC patients with histologically active disease | To assess a DL machine in quantifying eosinophils in colonic biopsies and validate against a pathologist’s count | The AI system highly agreed with manual eosinophil count by pathologists (ICC 0.81-0.92) |
| Peyrin-Biroulet *et al*[67], 2022 | Cohort study | 200 histological images of UC biopsies | To evaluate an AI algorithm in assessing histological disease activity according to the Nancy index | The CNN model had an excellent agreement with pathologists in the assessment of the Nancy index (ICC 0.84) |
| Villanacci *et al*[66], 2022 | Cohort study | 614 biopsies from 307 UC patients | To test a CNN-based CADe system for evaluating HR based on PHRI, Robarts, and Nancy indexes | The CADe system accurately assessed HR (sensitivity 89%, specificity 85% for PHRI) and similar performance for Nancy and Robarts |

AI: Artificial intelligence; CADe: Computer-aided detection; CNN: Convolution neural networks; DL: Deep learning; HR: Histological remission; ICC: Interclass correlation coefficient; PHRI: PICaSSO histologic remission index; UC: Ulcerative colitis.



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