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ImageCLEF 2025: Multimedia Retrieval in Medical, Social Media and Content Recommendation Applications

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Abstract. ImageCLEF has been a part of CLEF (Conference and Labs of the Evaluation Forum) for more than 20 years. Started in 2003, ImageCLEF is an evaluation initiative which promotes the evaluation of technologies for annotation, indexing, retrieval, or generation of multi-modal data. It provides access to large amounts of challenging data in very diverse use cases like medicine, argumentation, reasoning, generation, or content recommendation. In its 23rd edition, ImageCLEF will

* apart from the general organisers, the authors are listed in alphabetical order.

have four main tasks: (i) a *Medical* task involving concept detection and caption prediction in radiology images, synthetic medical images created with Generative Adversarial Networks (GANs), Visual Question Answering for improving the diagnosis and classification of real medical gastrointestinal images, and multimodal dermatology response generation, (ii) a joint ImageCLEF-Touché task *Image Retrieval/Generation for Arguments* to convey the premise of an argument, (iii) the *ToPicto* task which involves converting either text or speech into a meaningful sequence of pictograms and (iv) a new Multimodal Reasoning task addressing question answering and reasoning generation. In its last edition in 2024, 90 users and 31 unique teams submitted runs, totaling 257 runs, revealing a good impact in the community, similar to previous years.

Keywords: Medical AI, image captioning, GANs, Visual Question Answering, response generation, argumentation

1 Introduction

Since 2003, the ImageCLEF evaluation campaign has provided the scientific community a space for research and benchmark activities in fields like annotation, indexing, classification and retrieval of multimodal data. The 23rd edition of ImageCLEF takes place in Madrid, Spain, 9-12 September 2025¹⁸, as part of the Conference and Labs of the Evaluation Forum (CLEF) [16, 17]. A series of benchmarking activities have been developed to evaluate various aspects of mono-language and cross-language information retrieval systems [13, 16, 17]. Some of the fields of interest for ImageCLEF include information retrieval in multimedia, machine learning, deep learning, data mining, computer vision, natural language processing, data generation, and multimodal data fusion. Over the years, both ImageCLEF and the CLEF campaign have demonstrated significant scholarly impact [26, 27] and impact in the community overall, with the term ImageCLEF mentioned in 7,330 papers since 2003, according to Google Scholar. This paper presents the four tasks planned for 2025. Some examples can be seen in Fig. 1

2 ImageCLEFmedical

The ImageCLEFmedical task is currently at its 21st edition [17]. The 2025 edition introduces 4 medical sub-tasks, namely: (i) the *Caption* task with medical concept detection and caption prediction, (ii) the *GAN* task on synthetic medical images, (iii) *MEDVQA* regarding Visual Question Answering for gastrointestinal data, and (iv) *Mediqa-magic* with a new use-case on multimodal dermatology response generation.

*ImageCLEFmedical-Caption*¹⁹. The *Caption* task consists of the interpretation of the insights gained from radiology images, and in this 9th edition [7, 8,

¹⁸ <https://clef2025.clef-initiative.eu/>

¹⁹ <https://www.imageclef.org/2025/medical/caption>

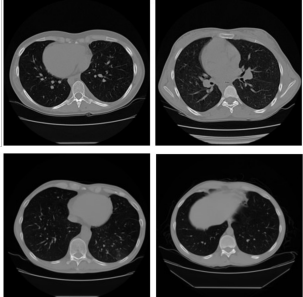
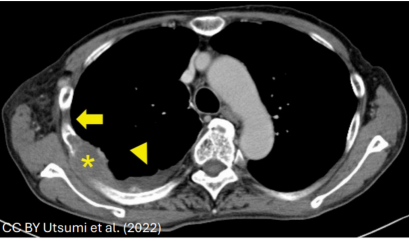

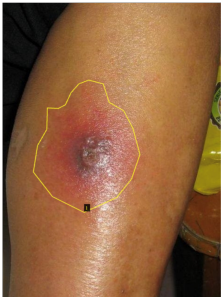
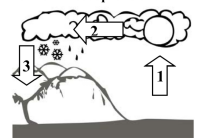
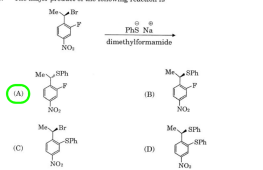
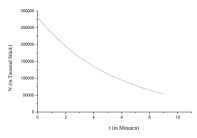
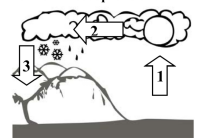
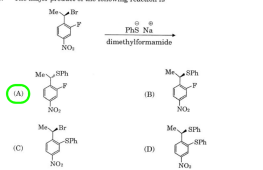
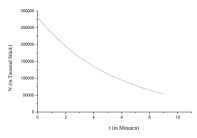
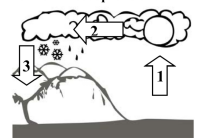
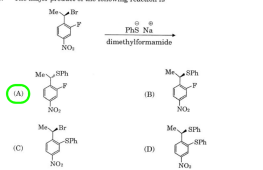
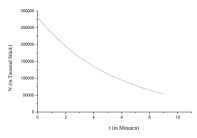
<p>1. Real or generated?</p> 	<p>2. Concepts:</p> <p>X-Ray Computed Tomography (CUI C0040405);</p> <p>Computed tomography imaging - action (CUI C0729619);</p> <p>Chest (CUI C0817096);</p> <p>Neoplasms (CUI C0027651);</p> <p>Pleural effusion (disorder) (CUI C0032227);</p> <p>Mediastinum (CUI C0025066);</p>  <p>Caption: Computed tomography images after treatment. Thoracic SMARCA4-deficient undifferentiated tumor showing osteolytic changes in the ribs (asterisk) is noted. However, pleural thickening (yellow arrow) disappears and pleural effusion (yellow arrowhead) decreases in the mediastinal window setting.</p>						
<p>3. Topic: AI in Medicine Argument: AI helps doctors with complicated operations</p>  <p>Relevance</p> <p>2 1 0</p>	<p>4.</p>  <p>Clinical History: Femal patient, 60-years old. Hospitalized in the Department of Internal Medicine due to hepatitis B. She has a medical history of 15 days and had the same disease two months ago. Painful.</p> <p>Question/Answers:</p> <p>How much of the body is affected? [single spot], limited area, widespread</p> <p>How large are the affected areas? size of thumb nail, [size of palm], larger area</p> <p>Pain? [yes], no</p> <p>Discharge? yes, [no]</p>						
<p>5.</p> <table border="1"> <thead> <tr> <th data-bbox="365 966 682 997">Bulgarian</th> <th data-bbox="682 966 982 997">English</th> <th data-bbox="982 966 1258 997">German</th> </tr> </thead> <tbody> <tr> <td data-bbox="365 997 682 1207"> <p>10. Разгледай схемата. С коя стрелка е означен процесът на преминаване на водата от течно в газообразно състояние?</p>  <p>(a) 1 (b) 2 (c) 3</p> <p>Subject: Biology Subfield: Man and Nature Grade: 4</p> </td> <td data-bbox="682 997 982 1207"> <p>48. The major product of the following reaction is</p>  <p>(A) (B) (C) (D)</p> <p>Subject: Chemistry Subfield: Chemistry Grade: 12</p> </td> <td data-bbox="982 997 1258 1207"> <p>20. In der Abbildung ist die Zerfallskurve eines radioaktiven Stoffes zu sehen. Lesen Sie vom Diagramm die Halbwertszeit ab!</p>  <p>(a) 3 Minuten. (b) 4 Minuten. (c) 5 Minuten.</p> <p>Subject: Physics Subfield: Physics Grade: 12</p> </td> </tr> </tbody> </table>		Bulgarian	English	German	<p>10. Разгледай схемата. С коя стрелка е означен процесът на преминаване на водата от течно в газообразно състояние?</p>  <p>(a) 1 (b) 2 (c) 3</p> <p>Subject: Biology Subfield: Man and Nature Grade: 4</p>	<p>48. The major product of the following reaction is</p>  <p>(A) (B) (C) (D)</p> <p>Subject: Chemistry Subfield: Chemistry Grade: 12</p>	<p>20. In der Abbildung ist die Zerfallskurve eines radioaktiven Stoffes zu sehen. Lesen Sie vom Diagramm die Halbwertszeit ab!</p>  <p>(a) 3 Minuten. (b) 4 Minuten. (c) 5 Minuten.</p> <p>Subject: Physics Subfield: Physics Grade: 12</p>
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Fig. 1. Sample images from (1) ImageCLEFmedical-GAN, with real and generated images, (2) ImageCLEFmedical-Caption, presenting an image with associated concepts and captions, (3) Image Retrieval/Generation for Arguments, showing an argument with possible image submissions, (4) MEDIQA-MAGIC, examples of segmentation and question answers, (5) MultimodalReason, with examples of questions and answers in different languages and at different levels.

18–23], there are two subtasks: concept detection and caption prediction. The *concept detection* subtask aims to develop competent systems that are able to predict the Unified Medical Language System (UMLS[®]) Concept Unique Identifiers (CUIs) based on the visual image content. The F1-Score [10] will be used for evaluation. This year, the updated ROCov2 [24] dataset will be used.

In the *caption prediction* subtask, domain-specific models such as BioBERT [14] are utilised to enhance the evaluation using BERTScore. Furthermore, experiments are conducted with other models, including ClinicalBLEURT and Med-

BERTScore [3], to explore alternative scoring methods. Readability scores such as LENS [15] and the Coleman-Liau Index [4] (CLI) are also incorporated to assess the clarity and readability of the generated captions. An explainability extension is offered for both tasks, where participants are asked to provide explanations, such as heatmaps and Shapley values, for a small subset of images that will be manually evaluated.

*ImageCLEFmedical-GAN*²⁰. The 3rd edition of the GANs task will continue to explore the quality of synthetic medical images generated by various models, including Generative Adversarial Networks (GANs). Building on the previous two editions [1,2], the sub-tasks will address both the challenges of ensuring privacy and confidentiality in medical imaging and the quality of the generated data itself. The focus remains on investigating potential risks related to data’s integrity and the potential for sensitive information to be embedded within the generated images. The sub-task involves analyzing test image datasets to evaluate the likelihood that specific real patient images were used to train the generative models. The provided data will include both real and generated axial slices of CT scans from tuberculosis patients, captured at various stages of their treatment. Metrics like Accuracy, F1 score will be used to determine if specific real images were used in the training of generative models or to identify the particular dataset of real images that contributed to the training of the generative model responsible for creating each synthetic image.

*ImageCLEFmedical-MEDVQA*²¹. The 3rd MEDVQA-GI challenge focuses on the integration of Visual Question Answering (VQA) with synthetic gastrointestinal (GI) data to enhance diagnostic accuracy and AI learning. Building on the previous two editions [11,12], this year’s challenge asks participants to complete two subtasks. The first subtask involves developing algorithms to accurately interpret and respond to questions related to GI images. The aim is for these algorithms to understand the context and details in the images and provide precise answers that assist in medical diagnostics. The second subtask focuses on generating high-fidelity synthetic GI images that closely resemble real medical images. The dataset for this year is an extended version of the collections used in the last two years, including GI images with VQA annotations [9]. The evaluation will differ for each task, which will be a combination of subjective and objective evaluations.

*ImageCLEFmedical-MEDIQA-MAGIC*²². In the 2nd MEDIQA-MAGIC task, we will extend on the previous year’s dataset and challenge based on multimodal dermatology response generation. Participants will be given a clinical narrative context along with accompanying images. The task is divided into two relevant sub-parts: (i) segmentation of dermatological problem regions, and (ii) providing answers to closed-ended questions.

In the first sub-task, given each image and the clinical history, participants will need to generate segmentations of the regions of interest for the described

²⁰ <https://www.imageclef.org/2025/medical/gans>

²¹ <https://www.imageclef.org/2025/medical/vqa>

²² <https://www.imageclef.org/2025/medical/mediqa>

dermatological problem. In the second sub-task, participants will be given a dermatological query, its accompanying images, as well as a closed-question with accompanying choices – the task is to select the correct answer to each question.

The dataset is created by using real consumer health users’ queries and images; the question schema was created by two certified dermatologists. Segmentation will be evaluated against common metrics such as Jaccard or IOU. Closed question-answering will be evaluated using metrics such accuracy and F1 score.

3 Image Retrieval/Generation for Arguments

*Touché-Argument-Images*²³ Images can drastically increase the impact of arguments and make them much more compelling. The 2025 edition marks the fourth edition of the task, having run at the Touché workshop since 2022. For this year’s edition, we have streamlined the task and focus on complete arguments. Given an argument, the participants shall find suitable images that convey the argument. Optionally, participants can add captions to explain the meaning of the submitted images. An example of an argument with possible image submissions can be seen in Fig. 1(3).

Participants can choose between two submission styles. (1) Retrieval: We deliver matching images from a focused crawl of around 10,000 images, together with website data and other data such as automatically generated image descriptions. (2) Prompted Generation: Based on the idea of the infinite index [6], participants can submit prompts for an image generator. A dedicated API is provided for this purpose. The submitted images are judged by human experts according to defined relevance criteria.

4 ImageCLEFtoPicto

In its 2nd edition, *ImageCLEFtoPicto*²⁴ involves converting either text or speech into a meaningful sequence of pictograms to be used as a communication aid. This year, the task will be split into 2 sub-tasks: (1) Text-to-Picto Translation, which involves the generation of a sequence of pictogram terms from a French text, using oral transcriptions from the ESLO corpus to create pictogram sequences, (2) Speech-to-Picto translation, a task that focuses on translating speech directly into pictogram sequences without going through the transcription. Using the ESLO corpus, participants will handle speech utterances and generate corresponding pictogram terms, challenging them to integrate speech recognition and translation technologies. This year, the ESLO Corpus [25] will be integrated to enhance the evaluation campaign, focusing specifically on the ESLO 2 sub-corpus, which contains contemporary interactions and offers modern language usage and a wide range of conversational contexts. For evaluation,

²³ <https://touche.webis.de/clef25/touche25-web/image-retrieval-generation-for-arguments.html>

²⁴ <https://www.imageclef.org/2025/topicto>

both automatic metrics like Picto-term Error Rate (PictoER) and the BLEU score will be used, as well as human expert evaluation, based on availability.

5 MultimodalReason

MultimodalReason is a new task focusing on Multilingual Visual Question Answering (VQA). Given an image of a question with four possible answers, participants must identify the single correct answer, as shown in Fig. 1(4). Furthermore, we provide OCR-extracted text and captions for each questions (but participants can extract their own). The questions available are in 11 different languages, across 20 school disciplines and different difficulty levels. The dataset for the task, called *Exams-V* [5], is publicly available²⁵. The goal of the task is to assess the reasoning capabilities of modern LLMs on complex inputs, presented in different languages, across various subjects. The official evaluation measure for the task will be *accuracy*. In addition to the existing dataset, we will provide the participants new unseen test data that follow the variety of languages, subjects, and difficulties of *Exams-V*.

6 Conclusions

This paper showcases the tasks proposed for the 23rd edition of the ImageCLEF evaluation campaign, in fields like medicine, data generation, multimodal data processing, data retrieval, natural language processing, just to name a few. ImageCLEF continues to provide very diverse and interesting challenges for researchers, giving them the possibility to assess the performance of their systems with a mix of both established and new tasks. All the tasks provide collections of test data end evaluation benchmarks, enabling researchers to assess the performance of their systems using realistic evaluation frameworks.

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²⁵ <https://huggingface.co/datasets/Rocktim/EXAMS-V/tree/main/data>

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