

EmeraldApp: An AI-Driven Tool for Detecting Greenwashing in Sustainability Claims

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Abstract

Greenwashing, i.e., presenting misleading environmental claims, undermines the trustworthiness of sustainability reporting. We present EmeraldApp, a tool designed to verify corporate sustainability claims. EmeraldApp classifies claims as greenwashing, not greenwashing, or abstains, and generates fact-based justifications. The tool supports different LLM pipelines: EM-RAG, which retrieves evidence from *EmeraldDB*, a vector database of ESG (Environmental, Social, Governance) report chunks; EM-KGRAG, which retrieves evidence from *EmeraldGraph*, a domain-specific knowledge graph; EM-HYBRID, which combines the other two retrieval pipelines; and EM-NR, without retrieval capabilities. EmeraldApp supports quick claim checks for end users, while advanced configuration enables pipeline comparison and deeper inspection of factual evidence and justification. Previously checked claims are stored in *ClaimsDB* and are accessible through the EmeraldApp to improve auditability and transparency.

CCS Concepts

• **Computing methodologies** → **Knowledge representation and reasoning**; • **Information systems** → *Information retrieval*.

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Keywords

Greenwashing Detection, Retrieval Augmented Generation (RAG), Knowledge Graph, Responsible AI, Evidence-based Explanations

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1 Introduction

Greenwashing, the deceptive practice of presenting misleading claims about an organization’s environmental performance, has become a pervasive obstacle to genuine sustainability progress [2], creating a verification gap: given a corporate sustainability claim (e.g., “We have achieved 100% renewable energy sourcing”), stakeholders often lack the tools to efficiently cross-check such assertions against the vast, dispersed volume of corporate ESG (Environmental, Social, and Governance) disclosures. Manual inspection and verification remain prohibitively labor-intensive and time-consuming [1, 9], while the ambiguity of corporate language makes it difficult to distinguish between verifiable facts and marketing noise.

While large language models (LLMs) excel in general text tasks, they falter in specialized verification due to: (1) hallucination of plausible but false details [3, 4]; (2) absence of domain-specific grounding [10]; and (3) opaque outputs lacking traceable evidence trails. Current solutions like ecoOops [5] and Greenwashing Identifier [7] primarily function as compliance auditors or risk assessment engines, scanning entire corporate reports to assign high-level compliance scores or flag general regulatory discrepancies, with limited



support for interactive, evidence-traceable assessment of individual claims.

In contrast, EmeraldApp is an AI-driven tool for automatic greenwashing verification and justification at the level of individual sustainability claims. EmeraldApp is powered by EmeraldMind [8], which introduces multiple RAG-LLM verification pipelines for greenwashing detection. Unlike prior efforts, which mainly provide document-level compliance scores or coarse risk flags, EmeraldApp provides an interactive, evidence-traceable workflow: users submit a claim, run it under multiple verification modes (retrieval-based pipelines EM-RAG, EM-KGRAG, EM-HYBRID, or a no-retrieval mode EM-NR), and receive a claim-specific verdict and a short justification. For the RAG pipelines, the justification is grounded in domain-specific evidence stores. These can be in a textual or subgraph format, depending on the retrieval mode. The system also logs verified claims and their justifications, enabling users to search, reopen, compare, and reuse past analyses for auditing and ongoing investigations.

2 EmeraldApp Overview

The EmeraldApp architecture (Figure 1) separates a user-centric frontend from a modular backend that orchestrates claim verification over shared evidence stores. The frontend provides components for claim input and verification configuration, search and inspection of past claims, and a results and visualization area. The backend provides access to several verification pipelines. *EmeraldDB* and *EmeraldGraph* hold ESG evidence in text and graph form, respectively. *ClaimsDB* indexes previously verified claims for search and reuse through the Claim Manager. The source code for EmeraldApp is publicly available at: <https://github.com/ai4greenwashing/EmeraldApp>.

2.1 Backend

This section describes the core backend components that enable evidence-grounded claim verification in EmeraldApp: the evidence stores that hold ESG knowledge, the pipelines that retrieve and reason over it, and the mechanisms for revisiting past analyses.

2.1.1 Evidence Stores. The backbone of EmeraldApp is its evidence stores for claim verification. *EmeraldDB* is a vectorized document store. It contains ESG report passages as embeddings, along with contextual metadata (e.g., report, company, year, page) and provenance needed for traceable retrieval. *EmeraldGraph* is a domain-specific property graph. It contains ESG entities and typed relations that link organizations to KPI observations, goals, claims, and supporting documents. These stores surface ESG-specific evidence that is often missing from generic knowledge bases. They also enable auditable reasoning by combining retrievable text context with explicit graph paths.

The stores are populated automatically from corporate ESG reports using a parsing and normalization pipeline. KPI definitions [6] from standards such as GRI provide consistent grounding for indicator names, units, and numerical interpretation. A predefined schema (see details in [8]) constrains entity types, relations, and attributes and separates conceptually distinct disclosures (e.g., targets vs. realized performance). This structure supports fine-grained retrieval of evidence (or counter-evidence) for a claim and enables transparent, evidence-backed justifications.

Finally, the annotated sustainability claims form a dedicated *ClaimsDB*, which stores natural language assertions together with their labels, provenance, and, when available, human- or model-generated justifications. This repository serves as a source of linguistic templates for typical corporate statements and as a benchmark for evaluating new predictions, enabling EmeraldApp to support activities such as searching past decisions, reusing verified evidence, and comparing system outputs across pipelines.

2.1.2 Claim Verification Pipelines. EmeraldApp supports four modes for verifying sustainability claims: EM-RAG, EM-KGRAG, EM-HYBRID, and EM-NR, implemented according to the EmeraldMind framework [8]. After typing a claim, a user can select one of these pipelines, which determines how evidence is retrieved and how the final verdict is produced. EM-RAG is the default configuration.

The EM-RAG pipeline focuses on retrieving textual evidence from ESG reports stored in *EmeraldDB*. Given a user-claim and the company name, the *Document Retriever* queries the *EmeraldDB* for semantically relevant text chunks from the corresponding ESG reports. The claim and the retrieved textual evidence are then passed to an *LLM Verifier*, which reasons over this context to generate a verdict and justification returned to the user.

The EM-KGRAG pipeline leverages the domain-specific *EmeraldGraph* to provide structured factual context. For a given claim, the system locates the relevant entities in *EmeraldGraph*, and *Graph Retriever* uses a graph-based context retrieval algorithm to extract an evidence subgraph. This subgraph, together with the claim, is passed as structured context to the *LLM Verifier*, which produces a verdict and justification analogous to the EM-RAG case.

To maximize classification accuracy and coverage, the EM-HYBRID pipeline combines both text and graph retrieval, following the hybrid design introduced in EmeraldMind [8]. When a user selects EM-HYBRID, the system runs the EM-RAG and EM-KGRAG pipelines in parallel and forwards their verdicts and justifications to an *LLM Judge*. The *LLM Judge* compares the alternatives and returns a final, consolidated verdict and justification to the user, together with the corresponding evidence.

The EM-NR pipeline serves as a no-retrieval baseline. The user claim is passed directly to the *LLM Verifier*, which produces a verdict and justification based only on the prompt instructions and the model's internal knowledge. This mode is used to contrast retrieval-grounded verification with unguided LLM reasoning.

2.1.3 Past Claims Management. The *Claims Manager* provides a persistent layer for logging, indexing, and reusing previously verified sustainability claims within EmeraldApp. It stores each claim together with its final verdict, justification, associated evidence (text passages or subgraphs), and metadata such as company, year, and pipeline used. This enables users to search and filter historical decisions, quickly retrieve past analyses for similar or recurring statements, and compare how different pipelines behaved on the same or related claims. By turning past verifications into a queryable knowledge base, EmeraldApp supports audibility, monitoring of model behavior over time, and interactive exploration of emerging greenwashing patterns across companies and reporting periods.

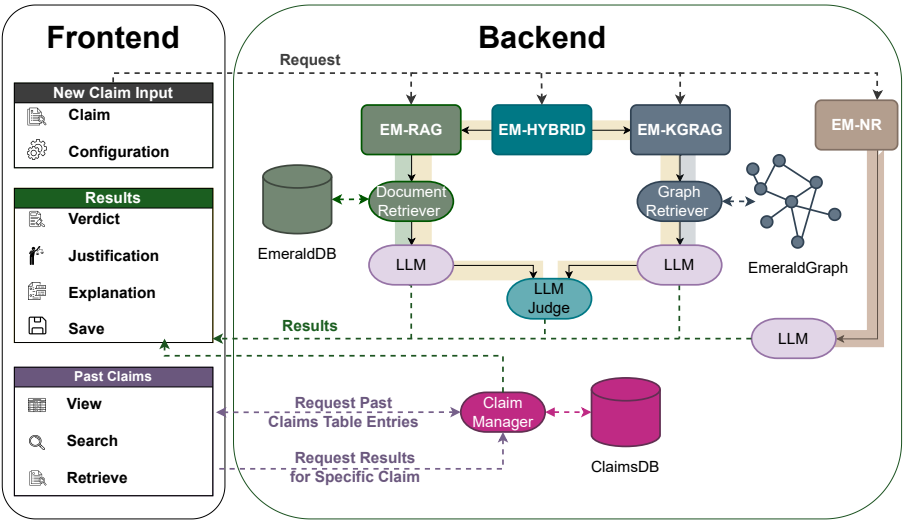


Figure 1: EmeraldApp Overview

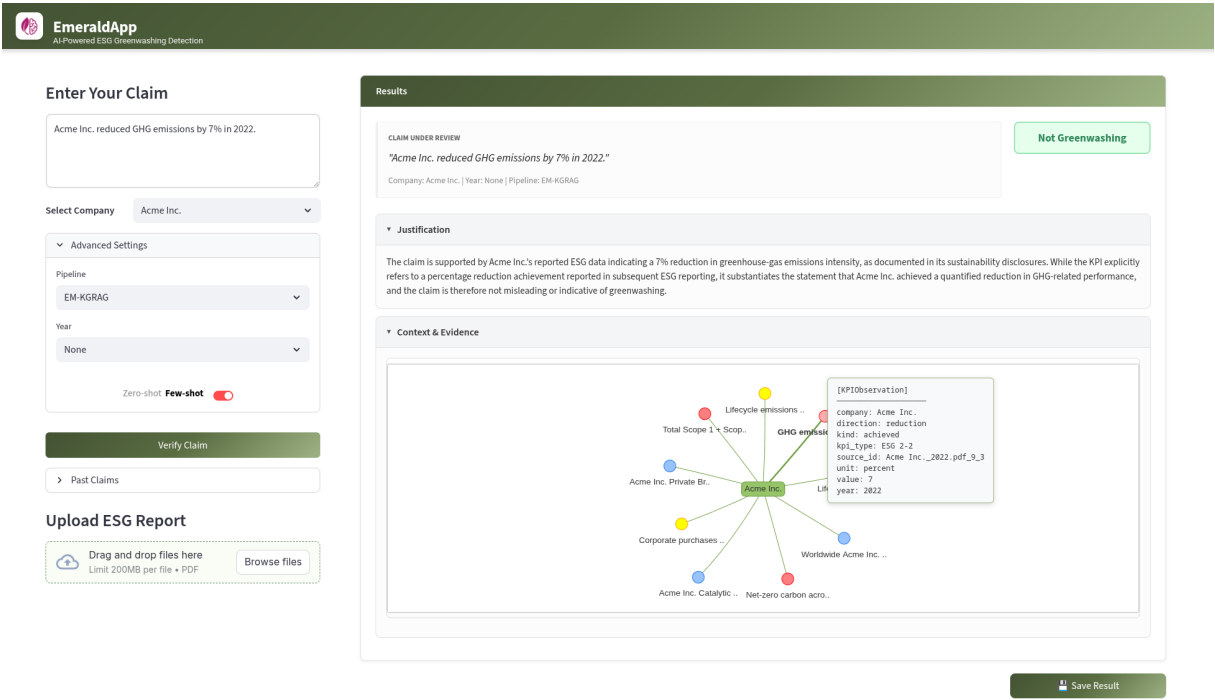


Figure 2: EmeraldApp interface with graph evidence for a non-greenwashing claim

2.2 Graphical User Interface

EmeraldApp’s interface follows a dual-panel layout, with the elements on the left dedicated to user input and past claims, and the right reserved for results. The *Enter Your Claim* area on the left lets users type a natural-language sustainability statement and select the company of interest. Optional advanced settings allow users to choose the reporting year, switch pipelines (EM-RAG, EM-KGRAG,

EM-HYBRID, or EM-NR), and adjust prompting (zero-shot vs. few-shot), as illustrated in Figure 2. A quick query can be performed using the default configuration strategy (EM-RAG with a zero-shot prompt) by pressing the *Verify Claim* button without any additional configuration (Figure 3).

The right-hand *Results* panel contains all output for the active claim. A header summarizes the claim under review, the selected company, and the active configuration (year, pipeline, and prompt),

EmeraldApp
AI-Powered ESG Greenwashing Detection

Enter Your Claim

Acme Corporation packaging is 92% recyclable or compostable, making it a sustainable choice.

Select Company: Acme Corporation

Advanced Settings

Verify Claim

Past Claims

Search company, claim, or verdict...

Company	Claim	Verdict	Pipeline	Type
Acme Corp.	Acme Corp.	Greenw.	EM-RAG	Zero-shot
Acme Inc.	Acme Inc. h	Not Gre	EM-RAG	Zero-shot
Dunder Mifflin	Dunder Mifflin	Greenw.	EM-KGRAG	Zero-shot
Acme Inc.	Acme Inc. n	Greenw.	EM-RAG	Few-shot

Populate Form

Upload ESG Report

Drag and drop files here
Limit 200MB per file • PDF

Browse files

Results

CLAIM UNDER REVIEW

"Acme Corporation packaging is 92% recyclable or compostable, making it a sustainable choice."

Company: Acme Corporation | Year: None | Pipeline: EM-RAG

Greenwashing Detected

Justification

Snippet 7 states Acme Corporation goal is to convert "100% of packaging to be recyclable or compostable by 2025". However, Snippet 4 states that in 2022, only 58% of KDP plastic packaging fit the definition of "recyclable" as defined by the Ellen MacArthur Global Commitment (a 30% recycling rate across geographies of 400m or more inhabitants). Additionally, Snippet 3 states "check locally — not recycled in many communities", indicating that even packaging "designed" to be recyclable may not actually "be" recyclable depending on local infrastructure. Therefore, claiming 92% recyclability/compostability is misleading as it doesn't account for the limitations of recycling infrastructure and the fact that a significant portion of their packaging isn't actually recyclable by global standards.

Context & Evidence

trade use and based on purchases for both kdp and kdp - owned and partner brands. 12. kdp uses two data sources to determine a supplier's risk: elevate's s"

Passage 3

"made of paper and organic materials, we also partnered with academia, recyclers and composting facilities to run technical assessments to understand how the bottle would perform in existing systems, reducing virgin plastic use easy - peel k - cup pod lid we expanded production of k - cup pod varieties with a pull tab that allows consumers to more easily peel the lid and empty the grounds in order to prepare the pod for the recycling stream; check locally — not recycled in many communities. on - pack recycling messaging we made "recycle right" instructions more visible through the addition of the how2recycle label on k - cup pod, core hydration, snapple and select mont' s packages, encouraging consumer recycling behaviors 16 2022 corporate responsibility report overview environment supply chain people & communities governance appendix health & well - being"

Passage 4

"steel 10 % 11 % 12 % 14 % 14 % non - refillable glass bottles 24 % 23 % 22 % 10 % 10 % polypropylene 3 % 6 % 8 % 8 % 9 % other 1 % 4 % 4 % 5 % flexible packaging and plastic film 2 % 2 % 3 % 3 % 3 % refillable glass bottles 2 % 2 % 1 % 2 % 0 % polystyrene and epe 6 % 4 % 0. 20 % 0 % 0 % " data reported for 2019 and going forward includes most tertiary packaging and brewer packaging. " " reported data reflects plastics that are designed for recycling systems in the geographies where kdp distributes product. the ellen macarthur global commitment definition of "recyclable " specifies that a package type be recycled at a 30 % recycling rate across geographies of 400m or more inhabitants, which moves beyond the traditional definition of "recyclable ". in 2019, 46 % of kdp plastic packaging fit this definition. in 2020, 50 % of kdp plastic packaging fit this definition. in 2021, 58 % of kdp plastic packaging fit this definition. in 2022, 58 % of kdp"

Passage 5

Save Result

Figure 3: EmeraldApp interface with textual evidence for a detected greenwashing claim

and shows a color-coded verdict badge: red for “Greenwashing Detected” (Figure 3), green for “Not Greenwashing” (Figure 2), and a neutral color for “Abstain”. Below, the *Justification* section presents a concise justification grounded in retrieved ESG evidence; for example, in Figure 3, the justification references specific passages from the context. This is essential, as it helps users understand why a verdict was produced rather than treating it as a black-box label.

An expandable *Context & Evidence* view then reveals the underlying support: for text-based pipelines, it displays numbered passages from *EmeraldDB* (Figure 3); for graph-based pipelines, it shows an interactive subgraph from *EmeraldGraph* where hovering over nodes reveals detailed node metadata (Figure 2). These visualizations make the reasoning process tangible for users, allowing them to quickly assess whether the evidence is appropriate and how different configurations change the system’s behavior. Finally, a user can save the results of the query made through the save button (Figure 2).

The *Past Claims* collapsible section (Figure 3) includes a table that lists previously verified claims with their companies and verdicts. Search and column-view controls let users filter entries and reveal configuration details. By selecting a row and pressing *Populate Form*, the corresponding claim and settings are reloaded into the *Enter your claim* area, and the associated results are restored in the *Results* section, supporting fast re-runs and deeper evidence inspection. At the bottom, an *Upload ESG Report* widget (Figure 3) supports drag-and-drop or file browsing for PDF reports, which are ingested so that new disclosures become immediately available for verification.

3 Demo Scenarios

During the demo session, the user enters a claim, selects the organization, and runs verification. The system returns a verdict, a short justification, and the most relevant supporting evidence chunks. The user can then open the advanced settings to rerun the same claim with different configurations (e.g., pipeline and prompt) and view the corresponding results and evidence (text, graph). The user visits the past claims menu, filters the claims based on the company of interest, and reopens selected runs. Finally, the user downloads the results with the Save button.

4 Conclusions and Future Work

EmeraldApp is an AI-driven tool for greenwashing detection. It provides a verdict and a fact-based justification for each submitted claim by combining text- and graph-based retrieval from ESG knowledge. As future work, EmeraldApp can be extended with user authentication and personalized workspaces, allowing users to log in, view, and manage their own history of verified claims, selectively save important analyses, and upload and analyze proprietary data. Finally, a dedicated comparison view for running the same claim through multiple pipelines side by side would further help users understand how different configurations affect verdicts, evidence, and justifications.

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