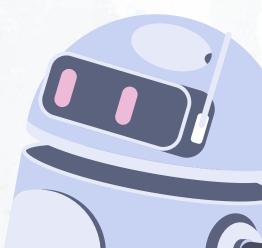
# AI Driven Entity Intelligence Risk→







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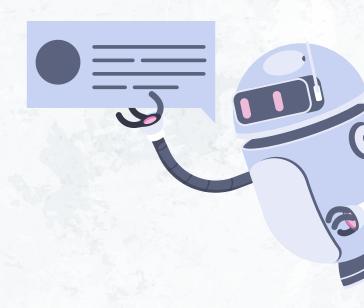
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# Goal of the challenge?

Develop an AI/ML system to automate entity identification, verification, and risk scoring from transaction data, reducing manual efforts and enhancing accuracy.

### Key Objectives:

- Extract and enrich entity data from multiple sources
- Detect fraudulent or high-risk entities
- Classify entities and assign risk scores.
- Provide evidence and confidence scores for analysis.



# Overview of the Technologies

## **Tech Stack**

#### Frontend:

HTML, CSS, JavaScript

#### **Backend:**

Python - DDGS, Langchain, getpass, pandas googlesearch, load\_dotenv



Workflow

## Flow Diagram

The diagram below illustrates the step-by-step process of our ML application.



# Intelligent Data Extraction

## Intelligent Data Extraction

Our system leverages AI-powered semantic search to enhance risk detection beyond simple keyword matching.

#### **How Semantic Search Improves Risk Analysis:**

√ Understands Context – Identifies risks even if exact terms like fraud or lawsuit aren't explicitly mentioned.

Deep Data Exploration – Analyzes Wikipedia & Google data for hidden regulatory, financial, and reputational risks.

Intelligent Pattern Recognition – Detects compliance violations, financial instability, and reputational threats using contextual insights.

By using semantic search, our system ensures accurate, context-aware risk evaluation for better decision-making.

## **Data Sources**

## **Our Data Sources**

- OFAC SDN List Using publicly available .csv files
- SEC Edgar data using publicly available bulk zip files
- Google searches
- Wikipedia



# Risk Scoring Criteria

## Risk Score Calculation



Our system evaluates entities based on multiple risk factors and assigns a final risk score to assess potential risks.

#### Key Risk Factors & Scoring:

- Sanctions Risk: 50 (Both sanctioned) | 30 (One sanctioned) | 0 (None)
- Regulatory Risk: 7 → HIGH | 3 → MEDIUM | 1 → LOW
- Financial Risk: 7 → HIGH | 3 → MEDIUM | 1 → LOW
- Reputation Risk: 7 → HIGH | 3 → MEDIUM | 1 → LOW

#### Output Includes:

- Sender & Receiver details
- ✓ Final Risk Score based on combined factors.
- ✓ Conclusion for informed decision-making

This structured scoring helps in accurate risk assessment while maintaining clarity and efficiency.

# **About Model**

## LLM

Our system leverages Llama 3.2 Vision, an open-source LLM developed by Meta, to enhance risk assessment through advanced AI techniques.

### Why Llama 3.2 Vision?

- Context-Aware Analysis Detects implicit risks beyond simple keyword searches.
- @ Accurate Entity Recognition Provides precise entity extraction and risk classification.
- Open-Source & Scalable Ensures flexibility for integration and future improvements.

By integrating Llama 3.2 Vision, our system achieves smarter, more efficient risk evaluation with Al-driven insights.

# Thanks! →

Any questions?

