# CODRELATE-2025

**PROBLEM STATEMENT:** Ai-powered content analysis and recommendation.

#### **DATASET:**

Link: <a href="https://drive.google.com/drive/folders/1uF6Bzle6YF6cth\_LaYVKcEb-bPSeX4lt?usp=drive\_link">https://drive.google.com/drive/folders/1uF6Bzle6YF6cth\_LaYVKcEb-bPSeX4lt?usp=drive\_link</a>

> Data description

Each row in the data is a different article published on Medium. For each article, you have the following features:

- **title** [string]. The title of the article.
- **text** /string). The text content of the article.
- url /string). The URL associated to the article.
- authors /list of strings]: The article authors.
- **timestamp** [string]: The publication datetime of the article.
- tags /list of strings/: List of tags associated to the article.

#### TASK:

- > Tag Modeling Identify the tags based on the title and text.
- ➤ Engagement Prediction Predict article popularity based on features like title, tags, and reading time.
- **Keyword Extraction** Identify the most relevant keywords to summarize articles effectively.

#### **INNOVATION:**

- ➤ Personalized Article Recommendations Suggest articles based on user reading history and preferences.
- ➤ Author Influence Analysis Evaluate the impact of authors by analyzing engagement metrics across their articles.
- ➤ Content Optimization Assistant Provide recommendations to authors for improving article reach and engagement.

### Round 1: Data Analysis & Exploration

#### **Problem Statement**

Participants will be provided with a raw dataset containing real-world data. The objective is to **clean**, **explore**, **and extract meaningful insights** from the data to support decision-making. Teams must apply **data preprocessing techniques**, **exploratory data analysis (EDA)**, **and feature selection** to uncover patterns, trends, and relationships within the dataset. The goal is to prepare a well-structured dataset that can be used effectively in the next round for model building.

# Workflow / Methodology

## 1) Understanding the Dataset

- Identify the type of data (structured, unstructured, categorical, numerical).
- Analyze column names, data types, and the significance of each feature.
- Check for inconsistencies, missing values, and data imbalances.

### 2) Data Cleaning & Preprocessing

- Handle missing values using imputation techniques (mean, median, mode, interpolation).
- Detect and manage outliers using statistical methods (IQR, Z-score).
- Standardize or normalize numerical data if necessary.
- Encode categorical variables using techniques like one-hot encoding or label encoding.

### 3) Exploratory Data Analysis (EDA)

- Generate summary statistics (mean, median, standard deviation, skewness, kurtosis).
- Identify relationships between features using correlation matrices.
- Analyze class distribution and detect any data imbalances.
- Apply dimensionality reduction techniques (if needed) to improve interpretability.

#### 4) Data Visualization & Insight Extraction

- Use histograms, box plots, scatter plots, and bar charts to visualize trends.
- Utilize heatmaps and pair plots to analyze relationships between variables.
- Identify key factors affecting the problem statement using interactive dashboards.

#### 5) Feature Engineering

- Identify the most relevant features that contribute to predictions.
- Perform feature scaling (Min-Max Scaling, Standardization) to improve consistency.
- Create new meaningful features through domain knowledge or data transformations.

#### 6) Efficient Use of Data

- Ensure that no unnecessary features or redundant data points are included.
- Optimize memory usage to handle large datasets efficiently.
- Justify each preprocessing step with logical reasoning.

# 7) Final Report Submission

- Document all findings, insights, and decisions in a structured format.
- Present key visualizations to support the analysis.
- Provide a well-commented Jupyter Notebook or script explaining each step.

#### **Evaluation Criteria**

# Problem Understanding & Approach (20%)

Clarity in defining the problem and relevance of chosen techniques.

# → Data Cleaning & Preprocessing (20%)

Efficient handling of missing values, outliers, and inconsistencies.

# ★ Exploratory Data Analysis (EDA) (20%)

Quality and depth of statistical analysis and insights.

# ★ Visualization & Interpretation (15%)

• Use of effective charts and visual storytelling to communicate findings.

# ★ Feature Engineering & Selection (15%)

Identification of key features contributing to model performance.

# Report & Code Quality (10%)

• Well-documented code and structured report with justifications.