#### NLP-TA

Team Members: Manan Khandelwal Aakanshi Gera Sagar Sindhu Lakshya Bindal

## 1. Title of the Project

**SamvidhanAl** (Al-Based Legal Contract Generator)

## 2. Problem Statement & Objectives

#### **Problem Statement:**

Legal contract drafting is a critical but time-consuming for businesses, legal professionals, and individuals. Traditional contract generation relies heavily on manual drafting, which introduces several challenges:

- Time-Consuming: Lawyers spend hours drafting and reviewing standard contracts.
- **High Legal Costs:** Small businesses and startups often struggle with the cost of hiring legal experts for routine contracts.
- **Human Error & Inconsistencies:** Manual entry can lead to errors, missing clauses, and discrepancies.
- Scalability Issues: Large firms handling multiple contracts daily struggle to manage consistency
  and
  compliance.

#### **Real-World Example Scenarios:**

- 1. **Startups & Freelancers:** A startup needs to draft NDAs, employment agreements, and partnership contracts frequently. Hiring a lawyer for each document is costly.
- 2. **Real Estate Industry:** Property lease agreements must be generated with accurate tenant details, payment terms, and jurisdictional laws.
- 3. **E-commerce & SaaS Companies:** Subscription-based services require contracts with consistent and legally sound terms.

4. **HR Departments:** Employment contracts for new hires often follow a standard format but require customization for names, roles, and benefits.

#### **Novelty in Our Project:**

- Rule-Based NLP for Contract Generation: Unlike traditional template-based methods, our system intelligently extracts relevant entities and ensures they fit logically within predefined
- 2. **Automation Without LLMs:** Unlike expensive generative Al models, this system is lightweight, cost-effective, and interpretable.

## 3. Step-by-Step Methodology

#### **Step 1: Data Collection & Preprocessing**

• Dataset: CaseHOLD (Case Holdings On Legal Decisions)

**CaseHOLD** is a legal dataset with **53,000+ multiple-choice questions** designed to test Al's ability to identify the correct legal ruling (holding) of a cited case. Each question provides a **legal excerpt** with five possible rulings—one correct and four incorrect. It helps train Al models for legal reasoning and case law analysis.

- Data Cleaning and Preprocessing:
  - Remove stop words and legal jargon.
  - Normalize text (lowercasing, punctuation removal, etc.).
- Tokenization & Entity Tagging:
  - Apply Spacy for Named Entity Recognition (NER).
  - Use Regex for pattern matching (e.g., dates, names, contract duration).

#### **Step 2: Named Entity Recognition (NER) for Clause Extraction**

- Tools Used: Spacy, Regex, NLTK
- Entities to Extract:
  - → **PARTIES**: Names of individuals or companies

- → **DATES:** Contract start and end dates
- → **DURATION**: Contract validity
- → PAYMENT TERMS: Financial obligations
- → GOVERNING LAW: Applicable jurisdiction

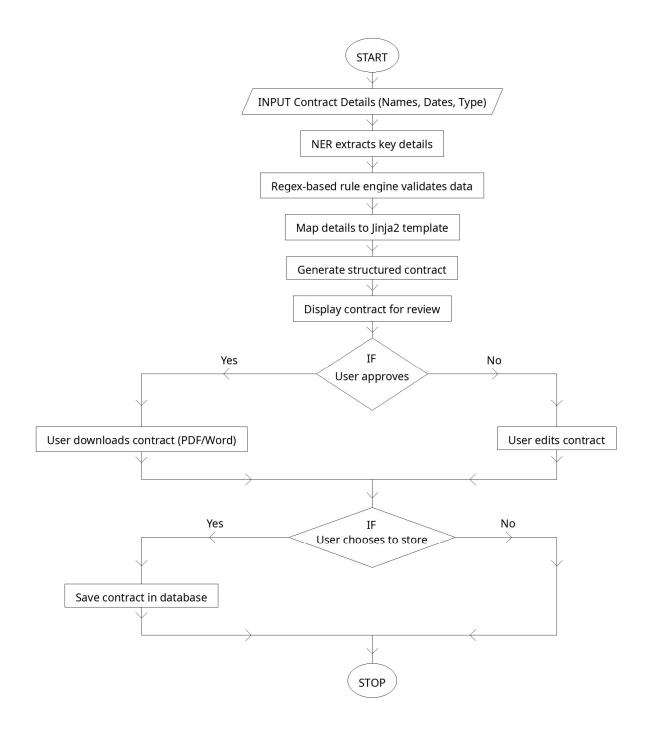
#### **Step 3: Contract Template Generation**

- **Technology Used:** Jinja2 Template ( A predefined format for legal documents )
- Process:
  - → Create predefined templates with placeholders.
  - → Dynamically fill placeholders with extracted data.

## **Step 4: Model Evaluation & Improvement**

- Performance Metrics:
  - → Precision & Recall (for NER accuracy)
  - → Execution Time (contract generation speed)
  - → User Feedback & Iteration
- Enhancements:
  - → Improve regex rules.
  - → Expand contract templates for diverse legal use cases.

# 4. System Diagrams



# 5. Architecture Design

