CAPSTONE PROJECT

DIGIDHARMA: AUTOAI-POWERED PREDICTION OF TELE-LAW ENGAGEMENT

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PROBLEM STATEMENT

- The problem focuses on the inequality in access to Tele-Law services across India. While the initiative has expanded to many states and districts, there's a noticeable lack of understanding about how different demographic groups and regions are using these services.
- Certain marginalized communities—such as Scheduled Castes (SC), Scheduled Tribes (ST), and Other Backward Classes (OBC)—appear to be underrepresented in the data. Similarly, some districts show very low levels of outreach or engagement, which raises questions about how fairly the legal aid services are reaching the population.
- Another complication is that the number of CSCs (Common Service Centres) varies from region to region, making it difficult to make direct comparisons between areas. This variation creates complexity in assessing whether legal aid is being distributed equitably across all districts and communities.



PROPOSED SOLUTION

In this project, a machine learning model is developed to **predict the total number of Tele-Law case registrations** (Total) for each district. The prediction is based on input features such as:

- Number of CSCs in the district
- Gender-wise data (Female, Male)
- Caste-wise data (SC, ST, OBC, General)
- Geographic identifiers (State/UT, District)

The target variable is the **Total case registrations**, which represents the volume of legal aid usage in each region. The model learns how demographic and infrastructural factors influence service utilization, allowing for forecasting of future case registration volumes and identifying patterns of underutilization.



SYSTEM APPROACH

This section outlines the overall strategy, tools, and methodology used to develop and deploy a machine learning model that predicts the **total number of Tele-Law case registrations** across districts using IBM Watsonx.ai's AutoAl.

System Requirements:

Component	Specification
Device	Laptop/Desktop with minimum 4GB RAM (8GB+ recommended for smoother operation)
Internet Speed	Stable internet connection (minimum 10 Mbps recommended)
Browser	Google Chrome / Mozilla Firefox (latest version recommended)
Cloud Environment	IBM Cloud Academic Portal
Account Access	IBM Cloud account with access to Watsonx.ai and Cloud Object Storage
Platform	IBM Watsonx.ai (AutoAI)
Runtime	Watsonx.ai Runtime service (provisioned for model building and deployment)
Deployment	AutoAl-generated model deployed in a Watsonx.ai Deployment Space
Dataset Format	CSV (Tele-Law Case Registrations.csv)



SYSTEM APPROACH

Although IBM AutoAl handles model creation without manual coding, the backend utilizes several essential libraries and frameworks for data processing, model training, and evaluation:

Libraries Required:

Library/Tool	Purpose
pandas	Data manipulation and analysis
numpy	Numerical computations
scikit-learn	Machine learning models, preprocessing, pipeline creation
xgboost / lightgbm	Advanced tree-based boosting algorithms (AutoAl uses automatically)
watson-machine-learning-client	Interacting with deployed models on IBM Cloud

Note: All of these are handled automatically within the AutoAl pipeline; no manual installation or coding is required unless additional customization is needed outside the AutoAl interface.



ALGORITHM & DEPLOYMENT

ALGORITHM OVERVIEW

- This project uses **IBM Watsonx.ai AutoAl**, which automates model training and selection. It performs:
- Automatic preprocessing: Handles missing values, data types, and feature engineering.
- Model generation: Builds multiple pipelines using algorithms like:
 - Random Forest Regressor
 - Gradient Boosting (XGBoost/LightGBM)
 - Decision Tree, Linear, and Ridge/Lasso Regression
- Evaluation: Ranks models using metrics like R², MAE, and RMSE.
- Selection: Chooses the best-performing pipeline for deployment.



ALGORITHM & DEPLOYMENT

Model Deployment

After identifying the best-performing pipeline, the model is deployed to the cloud using IBM Watsonx.ai's integrated tools:

Model Promotion:

The selected model is promoted to a deployment space within Watsonx.ai.

Deployment Creation:

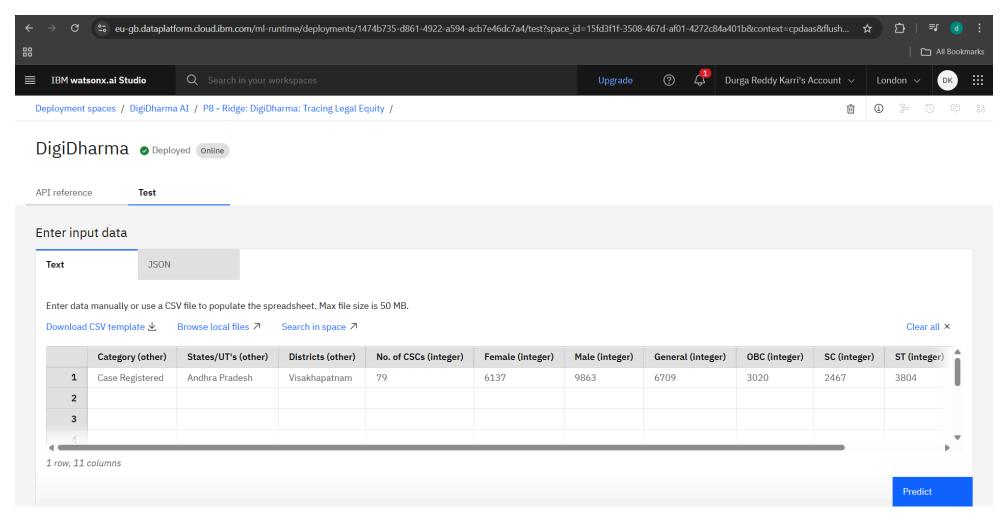
- A new deployment is created (real-time or batch) to make the model accessible via API.
- The deployment is named and version-controlled.

Model Testing:

- The deployed model is tested using new input values directly from the cloud UI.
- The predicted output (i.e., total case registrations) is generated in real time.



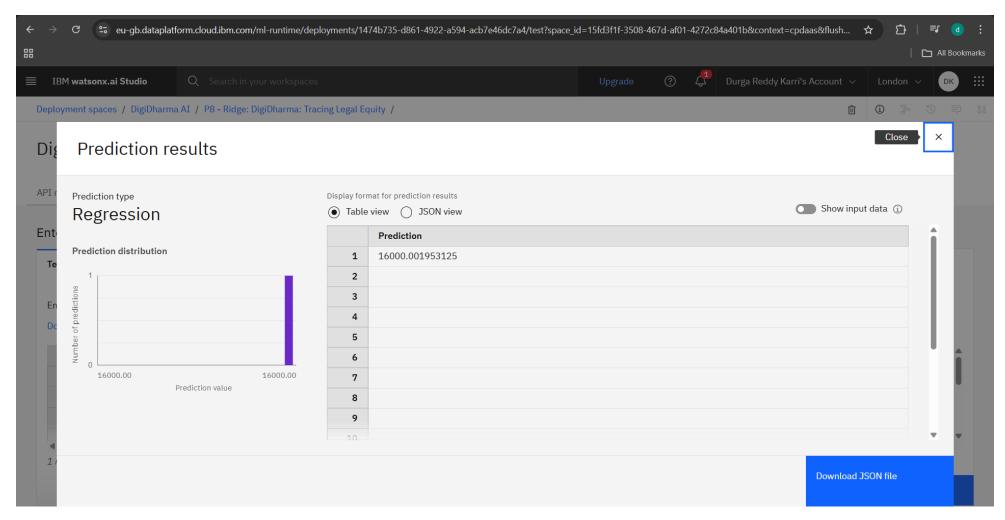
RESULT



INPUT DATA



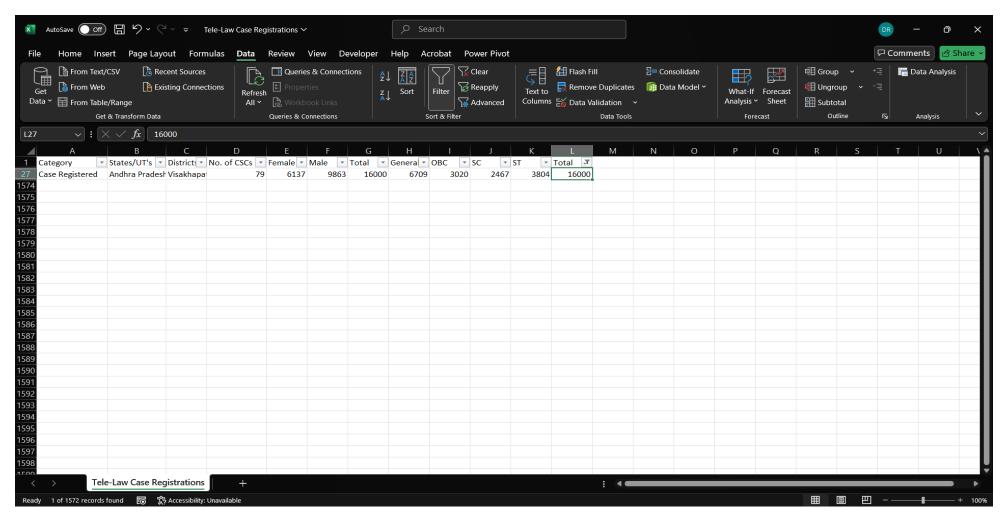
RESULT



PREDICTION RESULT



RESULT



MODEL PREDICTION ACCURACY: COMPARISON WITH ACTUAL DATA



CONCLUSION

- This project effectively used IBM Watsonx.ai's AutoAI to develop a predictive model for estimating Tele-Law case registrations across Indian districts. By leveraging demographic inputs like gender, caste, and CSC availability, the model uncovers key utilization patterns and highlights regions with possible disparities in legal aid access.
- AutoAl streamlined the entire process, from data preprocessing to model selection and evaluation, enabling the creation of a high-performing regression model with minimal manual effort. The insights gained from this model can help optimize CSC distribution, identify underrepresented communities, and support inclusive planning for legal service delivery—ultimately contributing to a more equitable justice system.
- Moving forward, this predictive system can be enhanced by integrating real-time data updates, incorporating socio-economic indicators, and expanding the feature set to include factors like literacy rate or digital accessibility.



FUTURE SCOPE

- Incorporate More Features: Enhance the model by adding socio-economic indicators such as literacy rate, income level, and digital connectivity to improve prediction accuracy.
- Real-Time Data Integration: Enable the system to work with live Tele-Law data feeds for continuous model updates and timely insights.
- District-Level Targeting: Use the model outputs to identify and prioritize low-performing districts for focused outreach and CSC deployment.
- Scalable Deployment: Extend the model across other government schemes to assess service accessibility and inclusion at a national scale.
- Dashboard Integration: Link predictions with an interactive dashboard to help policymakers visualize disparities and act accordingly.
- Predictive Alert System: Develop a system that flags regions at risk of low utilization in advance, enabling preventive action.

REFERENCES

- Official Dataset Source:
 - <u>District-wise Tele-Law Case Registration and Advice Enabled Data (FY 2021–22 to 2024–25) data.gov.in</u>
- IBM Cloud Platform: https://cloud.ibm.com/
- IBM Cloud: Watsonx.ai Studio (AutoAl Tool Documentation)



GITHUB LINK

https://github.com/durgareddy225/IBM-EdunetFoundation-Project



IBM CERTIFICATIONS

In recognition of the commitment to achieve Artificial Intelligence professional excellence Durga Reddy Karri Has successfully satisfied the requirements for: Getting Started with Artificial Intelligence Issued on: Jul 16, 2025 Issued by: IBM SkillsBuild Verify: https://www.credly.com/badges/55abc15c-665d-48d1-85f0-87cae0e34a55



IBM CERTIFICATIONS

In recognition of the commitment to achieve professional excellence Durga Reddy Karri Has successfully satisfied the requirements for: Journey to Cloud: Envisioning Your Solution Issued on: Jul 19, 2025 Issued by: IBM SkillsBuild Verify: https://www.credly.com/badges/9d94f5d0-6a59-4675-88ed-ad693765f42b



IBM CERTIFICATIONS

IBM SkillsBuild

Completion Certificate



This certificate is presented to

Durga Reddy Karri

for the completion of

Lab: Retrieval Augmented Generation with LangChain

(ALM-COURSE_3824998)

According to the Adobe Learning Manager system of record

Completion date: 01 Aug 2025 (GMT)

Learning hours: 20 mins



THANK YOU

