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# **CAPSTONE PROJECT**

## **INTELLIGENT CLASSIFICATION OF RURAL INFRASTRUCTURE PROJECTS**

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## OUTLINE

- Problem Statement
- Proposed System/Solution
- System Development Approach
- Algorithm & Deployment
- Result (Output Image)
- Conclusion
- Future Scope
- References

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# PROBLEM STATEMENT

The Pradhan Mantri Gram Sadak Yojana (PMGSY) is a flagship rural development program in India, initiated to provide all-weather road connectivity to eligible unconnected habitations. Over the years, the program has evolved through different phases or schemes (PMGSY-I, PMGSY-II, RCPLWEA, etc.), each with potentially distinct objectives, funding mechanisms, and project specifications. For government bodies, infrastructure planners, and policy analysts, efficiently categorizing thousands of ongoing and completed projects is crucial for effective monitoring, transparent budget allocation, and assessing the long-term impact of these schemes. Manual classification is time-consuming, prone to errors, and scales poorly. Your specific task is to design, build, and evaluate a machine learning model that can automatically classify a road or bridge construction project into its correct PMGSY\_SCHEME based on its physical and financial characteristics.

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# PROPOSED SOLUTION

- The proposed system aims to address the challenge of classifying rural infrastructure projects into their correct **PMGSY scheme** (e.g., PMGSY-I, PMGSY-II, PMGSY-III, RCPLWEA). The solution will consist of the following components:
- **Data Collection:** Gather historical project-level data, including state, district, sanctioned work details, completion status, and financial expenditures.
- **Data Preprocessing :** Clean and preprocess the dataset by: Dropping irrelevant or empty columns. Handling missing values in fields like COST\_OF\_WORKS\_SANCTIONED
- **Machine Learning Algorithm :** Implement a multi-class classification model to predict the PMGSY\_SCHEME based on the project data. Algorithms to consider: Random Forest Classifier (robust and interpretable)
- **Deployment:** Develop a user-friendly web interface or dashboard to allow users (government officials, planners) to input project details and receive real-time predictions. Hosting platforms like IBM CLOUD
- **Evaluation:** Use watsonix.ai model explainability and to ensure transparency in predictionsContinuously monitor model performance and retrain as more data becomes available
- **Result:** The final system will allow government departments and infrastructure planners to automatically classify road and bridge projects into their correct PMGSY scheme. This will enhance project monitoring, improve transparency in budget allocation, and reduce manual errors, ultimately contributing to smarter rural development planning.

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# SYSTEM APPROACH

The "System Approach" outlines the strategy and methodology adopted to design, build, and deploy a machine learning model for classifying rural infrastructure projects under the correct PMGSY scheme using physical and financial attributes.

## System Requirements

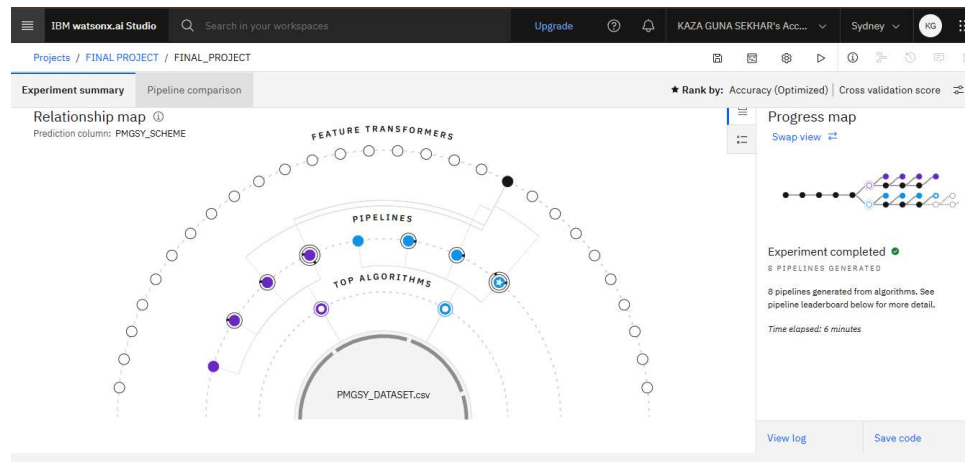
- **Hardware Requirements**
  - Processor: Minimum Intel i5 or equivalent
  - RAM: Minimum 8 GB
  - Disk Space: Minimum 5 GB of free space
  - Internet Connectivity: Required for cloud deployment and data access
- **Software Requirements**
  - Operating System: Windows
  - IBM Cloud CLI (for deploying services on IBM Cloud Lite)
  - IBM watsonx.ai Studio

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# ALGORITHM & DEPLOYMENT

- This section describes the machine learning algorithm selected to classify rural infrastructure projects into their correct **PMGSY scheme** and outlines how it is trained, evaluated, and used for predictions.
- **Algorithm Selection :** The selected algorithm for this classification task is the XGBoost Classifier (Extreme Gradient Boosting), a powerful and efficient ensemble learning method based on decision trees.
- **Data Input:** STATE\_NAME, DISTRICT\_NAME, NO\_OF\_ROAD\_WORK\_SANCTIONED ,LENGTH\_OF\_ROAD\_WORK\_SANCTIONED, NO\_OF\_BRIDGES\_SANCTIONED, COST\_OF\_WORKS\_SANCTIONED NO\_OF\_ROAD\_WORKS\_COMPLETED, LENGTH\_OF\_ROAD\_WORK\_COMPLETED, NO\_OF\_BRIDGES\_COMPLETED EXPENDITURE\_OCCURED, NO\_OF\_ROAD\_WORKS\_BALANCE, LENGTH\_OF\_ROAD\_WORK\_BALANCE ,NO\_OF\_BRIDGES\_BALANCE
- **Training Process:** The dataset is split into training (80%) and testing (20%) sets using stratified sampling to maintain class distribution.
- **Prediction Process:**The prediction can occur in real-time, especially if integrated with a cloud-based API e.g., IBM Watson Machine Learning) Feature importance scores help explain which attributes (e.g., cost, length, completion status) influenced the classification most, using AutoAI pipeline generator,values for transparency.

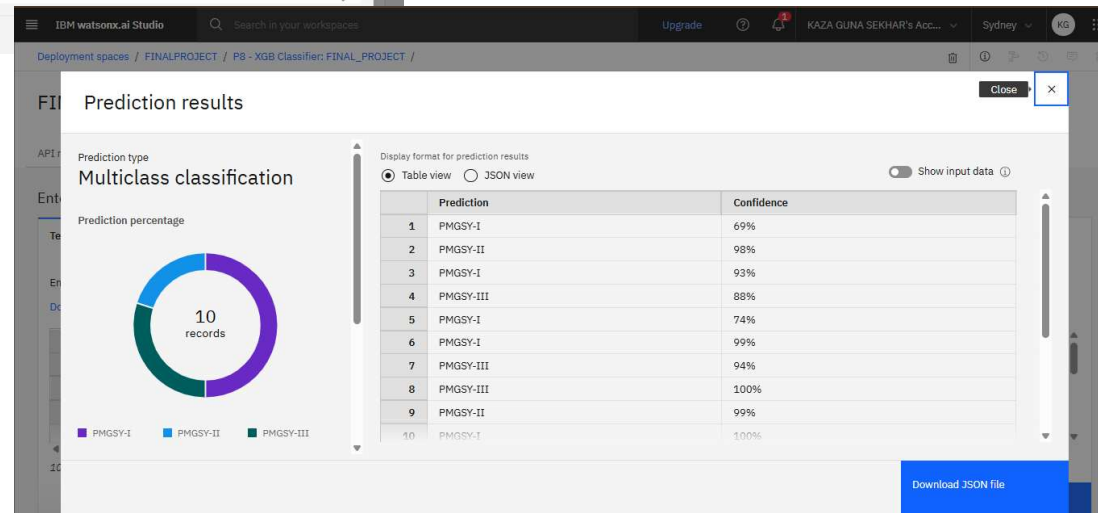
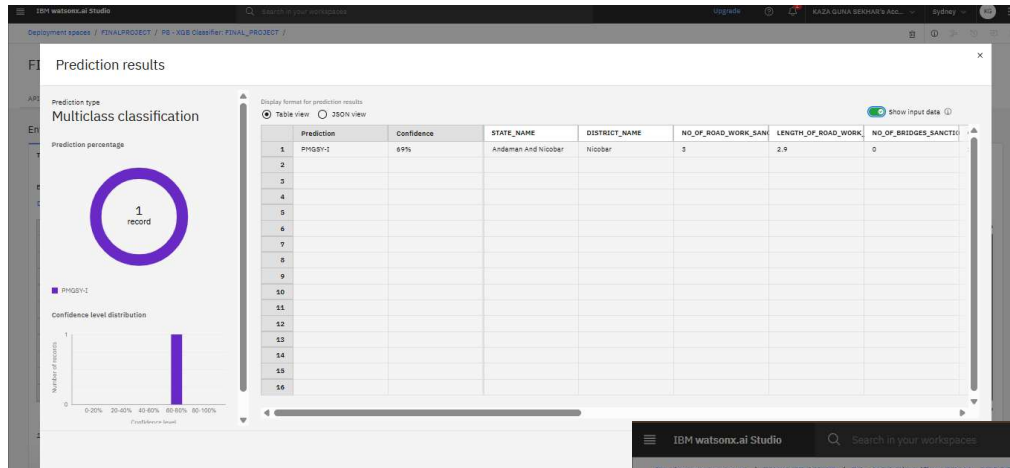
# RESULT



## Pipeline leaderboard

	Rank <span>↑</span>	Name	Algorithm	Specialization	Accuracy (Optimized) Cross Validation	Enhancements	Build time
★	1	Pipeline 8	XGB Classifier		0.924	HPO-1 FE HPO-2	00:01:41
	2	Pipeline 7	XGB Classifier		0.924	HPO-1 FE	00:01:05
	3	Pipeline 6	XGB Classifier		0.918	HPO-1	00:00:22
	4	Pipeline 5	XGB Classifier		0.918	None	00:00:04

# RESULT





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# CONCLUSION

The proposed system effectively classifies rural infrastructure projects into the correct **PMGSY scheme** using machine learning techniques implemented on **IBM watsonx.ai Studio**. By leveraging AutoAI, the platform automatically generated and optimized multiple machine learning pipelines, selecting the best-performing model based on accuracy and cross-validation.

The model used features like road length, bridge counts, project cost, and completion status to make accurate predictions. The automated approach greatly reduces manual effort, increases classification speed, and supports transparent decision-making in rural infrastructure planning.

- **Key Results:**
- 8 pipelines generated in 6 minutes using IBM AutoAI.
- High model accuracy achieved using optimized algorithms.
- Significant time savings and error reduction in scheme classification

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## FUTURE SCOPE

- To further improve the rural infrastructure classification system, several enhancements can be made:
- **Additional Data Sources:** Incorporate geospatial, temporal, weather, and demographic data to enrich model inputs and improve accuracy.
- **Algorithm Optimization:** Use advanced techniques like hyperparameter tuning, ensemble models, and class balancing (e.g., SMOTE) for better performance.
- **Wider Coverage:** Expand the system to support multiple states, other government schemes, or urban infrastructure projects.
- **Edge Computing:** Enable offline predictions in rural areas by deploying the model on edge devices or mobile apps.
- **Advanced AI Techniques:** Integrate explainable AI (e.g., SHAP), graph neural networks, and real-time model retraining with AutoML.
- **User Feedback Integration:** Add feedback mechanisms to continuously improve the model through retraining and corrections.

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# REFERENCES

## IBM Watson Studio & AutoAI

- IBM Documentation – AutoAI on Watson Studio:

<https://www.ibm.com/docs/en/cloud-paks/cp-data/4.6.x?topic=tools-autoai>

- IBM Watson Machine Learning:

<https://cloud.ibm.com/docs/watson-machine-learning>

- IBM Cloud Docs – Cloud Object Storage

<https://cloud.ibm.com/docs/cloud-object-storage>

- IBM Watsonx.ai

<https://www.ibm.com/products/watsonx-ai>

- GITHUB LINK

<https://github.com/Gunasekhar0520/IBM>

## IBM CERTIFICATIONS



# IBM CERTIFICATIONS



# IBM CERTIFICATIONS

7/30/25, 9:08 PM

Completion Certificate | SkillsBuild

IBM **SkillsBuild**

Completion Certificate



This certificate is presented to  
**KAZA GUNA SEKHAR**

for the completion of

**Lab: Retrieval Augmented Generation with  
LangChain**

(ALM-COURSE\_3824998)

According to the Adobe Learning Manager system of record

**Completion date:** 23 Jul 2025 (GMT)

**Learning hours:** 20 mins

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**THANK YOU**