
CAPSTONE PROJECT

INTELLIGENT CLASSIFICATION OF RURAL INFRASTRUCTURE PROJECTS.

Presented By:

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OUTLINE

- **Problem Statement**
- **Proposed System/Solution**
- **System Development Approach** (IBM Cloud, Watsonx.ai Studio)
- **Algorithm & Deployment**
- **Result (Output Image)**
- **Conclusion**
- **Future Scope**
- **References**

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.

PROPOSED SOLUTION

- To automate the classification of rural infrastructure projects under various PMGSY schemes (PMGSY-I, PMGSY-II, PMGSY-III, etc.),
- We propose a **supervised machine learning model** trained on physical and financial characteristics of road and bridge projects.
- This solution will:
 - Increase classification speed and accuracy.
 - Eliminate manual errors and subjectivity.
 - Enable real-time monitoring and analysis.
- The solution is implemented using IBM Watsonx.ai Studio.

SYSTEM APPROACH

The project follows a **cloud-based machine learning development lifecycle** using IBM's **Watsonx.ai Studio**, structured as follows:

- **Data Collection & Upload**

The PMGSY dataset was sourced from AI Kosh and uploaded to IBM Cloud for processing.

- **Data Preprocessing**

Handled missing values, encoded categorical features (like State and District), and normalized numerical columns such as road length and cost.

- **Model Development in Watsonx.ai**

Used Watsonx.ai's **AutoAI** to automatically select and train the best classification model (e.g., Random Forest or Gradient Boosting) based on performance metrics.

- **Model Evaluation**

Evaluated model performance using metrics like **accuracy**, **precision**, and **F1-score** to ensure reliability.

- **Model Deployment**

The trained model was deployed as a **web service endpoint** in IBM Cloud, allowing real-time predictions through API calls.

ALGORITHM & DEPLOYMENT

We used IBM Watsonx.ai Studio's **AutoAI**, which automatically evaluated and selected the best-performing classification algorithm, such as **Random Forest**, **XGBoost**, or **Gradient Boosting**, suitable for structured tabular data.

■ Input Features:

- State Name
- District Name
- Number and Length of Road Works Sanctioned
- Number of Bridges Sanctioned
- Cost of Works Sanctioned
- Completion and Balance Details

■ Training Process:

- Data preprocessing and feature encoding
- Dataset split into training and test sets
- AutoAI handled model training, cross-validation, and hyperparameter tuning
- The model was evaluated using metrics like **accuracy**, **precision**, and **F1-score**

- **Prediction Process:**

The trained model predicts the appropriate PMGSY_SCHEME for new road or bridge projects based on financial and physical parameters.

- **Deployment:**

The final model was deployed as a REST API endpoint on IBM Cloud via Watsonx.ai Studio, enabling real-time project classification through cloud-based applications or user interfaces.

RESULT :-

Classification of Rural Infrastructure Projects ✔️ Deployed Online

API reference **Test**

Enter input data

Text

JSON

Enter data manually or use a CSV file to populate the spreadsheet. Max file size is 50 MB.

[Download CSV template](#) ⬇️

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| | STATE_NAME (other) | DISTRICT_NAME (other) | NO_OF_ROAD_WORK_SANCTIONED (double) | LENGTH_OF_ROAD_WORK_SANCTIONED (double) | NO_OF_BRIDGES_SANCTIONED (double) |
|---|--------------------|-----------------------|-------------------------------------|---|-----------------------------------|
| 1 | Andaman And Nicoba | Nicobar | 3 | 2.969 | 0 |
| 2 | Andaman And Nicoba | South Andaman | 24 | 41.992 | 0 |
| 3 | Andhra Pradesh | Krishna | 28 | 174.92 | 10 |
| 4 | | | | | |
| 5 | | | | | |

3 rows, 14 columns

Predict

Classification of Rural Infrastructure Projects ✓ Deployed Online

API reference

Test

Enter input data

Text

JSON

Enter data manually or use a CSV file to populate the spreadsheet. Max file size is 50 MB.

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| | COST_OF_WORKS_SANCTIONED (double) | NO_OF_ROAD_WORKS_COMPLETED (double) | LENGTH_OF_ROAD_WORK_COMPLETED (double) | NO_OF_BRIDGES_COMPLETED (double) | EXPENDITURE (double) |
|---|-----------------------------------|-------------------------------------|--|----------------------------------|----------------------|
| 1 | 1.3857 | 3 | 2.969 | 0 | 1.1082 |
| 2 | 23.42439 | 7 | 17.747 | 0 | 8.6343 |
| 3 | 09 | 11 | 80.703 | 0 | 49.6143 |
| 4 | | | | | |
| 5 | | | | | |

3 rows, 14 columns

Predict

Expected output from the dataset:-

| | | | | | | | | | | | | |
|----------------------|-----------|-----|----------|----|----------|-----|----------|----|----------|----|---------|----|
| Andaman / Nicobar | PMGSY-I | 3 | 2.969 | 0 | 1.3857 | 3 | 2.969 | 0 | 1.1082 | 0 | 0 | 0 |
| Andaman / North and | PMGSY-I | 32 | 60.169 | 0 | 24.6908 | 32 | 59.19 | 0 | 18.5754 | 0 | 0 | 0 |
| Andaman / North and | PMGSY-II | 24 | 54.67 | 0 | 27.7646 | 0 | 0 | 0 | 0 | 24 | 54.67 | 0 |
| Andaman / South And | PMGSY-I | 32 | 40.146 | 0 | 18.5375 | 31 | 39.287 | 0 | 11.6894 | 1 | 0.541 | 0 |
| Andaman / South And | PMGSY-II | 24 | 41.992 | 0 | 23.42439 | 7 | 17.747 | 0 | 8.6343 | 17 | 24.033 | 0 |
| Andhra Pra Anantapur | PMGSY-I | 619 | 2169.505 | 35 | 526.1277 | 619 | 2126.386 | 34 | 517.912 | 0 | 0 | 1 |
| Andhra Pra Anantapur | PMGSY-II | 14 | 125.7 | 0 | 56.0711 | 14 | 124.619 | 0 | 49.0126 | 0 | 0 | 0 |
| Andhra Pra Anantapur | PMGSY-III | 27 | 267.158 | 8 | 180.5204 | 24 | 234.902 | 0 | 111.5182 | 3 | 23.2 | 8 |
| Andhra Pra Chittoor | PMGSY-I | 283 | 889.681 | 6 | 188.3558 | 283 | 878.711 | 6 | 185.1493 | 0 | 0 | 0 |
| Andhra Pra Chittoor | PMGSY-II | 18 | 126.045 | 0 | 57.9668 | 18 | 121.403 | 0 | 46.136 | 0 | 0 | 0 |
| Andhra Pra Chittoor | PMGSY-III | 84 | 629.581 | 8 | 421.6158 | 68 | 476.394 | 0 | 193.0005 | 16 | 139.86 | 8 |
| Andhra Pra East Goda | PMGSY-I | 502 | 1402.046 | 20 | 341.4412 | 500 | 1308.782 | 17 | 322.73 | 2 | 13.7 | 3 |
| Andhra Pra East Goda | PMGSY-II | 18 | 123.049 | 0 | 61.7925 | 18 | 116.296 | 0 | 49.7726 | 0 | 0 | 0 |
| Andhra Pra East Goda | RCPLWEA | 26 | 340.896 | 10 | 249.8516 | 17 | 213.208 | 7 | 157.9498 | 9 | 123.276 | 3 |
| Andhra Pra East Goda | PMGSY-III | 27 | 183.925 | 2 | 105.3391 | 15 | 93.242 | 1 | 49.8617 | 12 | 87.78 | 1 |
| Andhra Pra East Goda | PM-JANM | 11 | 29.48 | 0 | | 0 | 0 | 0 | 0 | 11 | 29.48 | 0 |
| Andhra Pra Guntur | PMGSY-I | 329 | 996.013 | 13 | 236.5601 | 329 | 1000.144 | 13 | 242.894 | 0 | 0 | 0 |
| Andhra Pra Guntur | PMGSY-II | 13 | 141.44 | 1 | 78.0672 | 13 | 141.428 | 1 | 69.7068 | 0 | 0 | 0 |
| Andhra Pra Guntur | PMGSY-III | 34 | 223.804 | 10 | 246.7372 | 17 | 104.817 | 0 | 67.7782 | 17 | 115.164 | 10 |
| Andhra Pra Krishna | PMGSY-I | 286 | 689.37 | 9 | 157.6581 | 286 | 671.76 | 9 | 155.7358 | 0 | 0 | 0 |
| Andhra Pra Krishna | PMGSY-II | 14 | 112.79 | 0 | 74.989 | 14 | 110.499 | 0 | 64.8999 | 0 | 0 | 0 |
| Andhra Pra Krishna | PMGSY-III | 28 | 174.92 | 10 | 170.509 | 11 | 80.703 | 0 | 49.6143 | 17 | 90.375 | 10 |

Predicted output:-

Prediction type

Multiclass classification

Prediction percentage

3 records

PMGSY-I

PMGSY-II

PMGSY-III

Confidence level distribution

Display format for prediction results

Table view

JSON view

Show input data

| | Prediction | Confidence |
|----|------------|------------|
| 1 | PMGSY-I | 69% |
| 2 | PMGSY-II | 88% |
| 3 | PMGSY-III | 96% |
| 4 | | |
| 5 | | |
| 6 | | |
| 7 | | |
| 8 | | |
| 9 | | |
| 10 | | |
| 11 | | |

Download JSON file

Projects / Intelligent Classification of Rural Infrastructure Projects / Rural Infrastructure Projects Classification

Experiment summary

Pipeline comparison

★ Rank by: Accuracy (Optimized) | Cross validation score

Relationship map ⓘ

Prediction column: PMGSY_SCHEME

Progress map

Swap view ↔

Experiment completed ●














10 PIPELINES GENERATED

10 pipelines generated from algorithms. See pipeline leaderboard below for more detail.

Time elapsed: 4 minutes

View log

Save code

| | Rank ↑ | Name | Algorithm | Specialization | Accuracy (Optimized) Cross Validation | Enhancements | Build time |
|---|--------|-------------|---|---|--|---|------------|
| ★ | 1 | Pipeline 10 |  Batched Tree Ensemble Classifier (XGB Classifier) |  | 0.924 |     | 00:01:53 |
| | 2 | Pipeline 9 |  XGB Classifier | | 0.924 |    | 00:01:49 |
| | 3 | Pipeline 8 |  XGB Classifier | | 0.924 |   | 00:01:11 |

CONCLUSION

- This project demonstrated an effective machine learning approach to classify rural infrastructure projects into their respective PMGSY schemes using project data. The model, built and deployed on **Watsonx.ai Studio**, achieved high accuracy and improved classification efficiency over manual methods.
- **Challenges** included handling missing values and high-dimensional categorical data. Potential improvements include adding geospatial features and building a user-friendly dashboard.
- Accurate classification is essential for **transparent budgeting, effective monitoring, and better policy decision-making** in rural development.

FUTURE SCOPE

- Use advanced models like XGBoost or CatBoost for better accuracy. Add geospatial and temporal data (e.g., terrain type, project year) to enhance predictions. Build a dashboard or mobile app for easy use by government officials.
- Enable real-time feedback to improve the model continuously. Extend the model to classify projects under other rural schemes.
- Add Explainable AI features to increase transparency and trust.

REFERENCES

- <https://www.ibm.com/cloud/watsonx-ai>
- AI Kosh dataset link:-
https://aikosh.indiaai.gov.in/web/datasets/details/pradhan_mantri_gram_sadak_yojna_pmgsy.html

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