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# **CAPSTONE PROJECT**

## **IMPROVED SOURCE OF DRINKING WATER**

**Presented By:**  
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*Department of Information Science and Engineering*

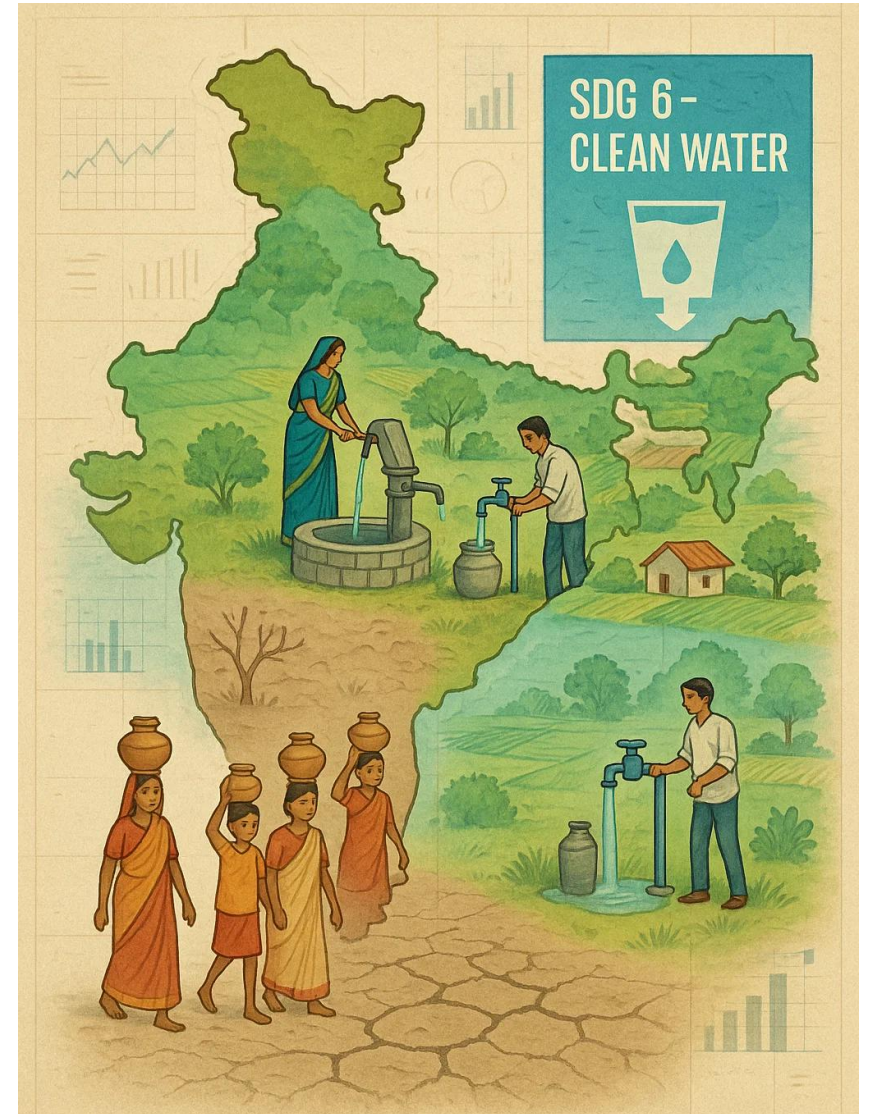
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# OUTLINE

- **Problem Statement**
- **Proposed System**
- **System Development Approach**
- **Algorithm & Deployment**
- **Result**
- **Conclusion**
- **Future Scope**
- **References**
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# PROBLEM STATEMENT

Access to safe and improved sources of drinking water remains a critical issue, especially in rural and underdeveloped regions of India. Despite government initiatives and efforts under Sustainable Development Goals (SDGs), significant disparities still exist across states and communities. There is a need for a data-driven approach to identify and predict areas lacking access to safe drinking water, enabling better planning and targeted interventions.



# PROPOSED SOLUTION

- The system aims to **predict access to improved sources of drinking water** using data analytics and machine learning. This enables policymakers to identify areas at risk and make informed decisions.
- **Data Collection:**
  - Data from the **78th Round of NSSO survey** on household drinking water sources.
  - Includes location-wise data (state/UT), socio-economic indicators, and access levels.
- **Data Preprocessing:**
  - Handle missing values, normalize data, and perform feature selection for impactful attributes.
  - Performed feature selection and encoding to prepare the dataset for modeling.
- **Machine Learning Algorithm:**
  - Use AutoAI in IBM watsonx.ai to automatically train and select the best regression model (Snap Boosting Machine Regressor).
  - Model trained to predict the *Indicator* column — a numeric representation of water source status.
- **Deployment:**
  - Model deployed using **IBM Watsonx.ai Studio**.
  - Accessible via REST API for real-time prediction across regions.
- **Evaluation:**
  - Performance assessed using **R<sup>2</sup> score** and **RMSE**
  - Model showed high accuracy in predicting improved source availability across states..

# SYSTEM APPROACH

## ➤ System requirements:

- Platform: IBM Watsonx.ai
- Dataset: NSSO 78th Round (Drinking Water Survey)
- Prediction Target: Indicator (Improved Water Access)

## ➤ Library required to build the model

- AutoAI: Automated model selection and training
- Runtime: 8 CPU, 32 GB RAM (Watsonx.ai)

## watsonx.ai Studio

Date of last update: 05/06/2025 • [Docs](#)



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About

## Summary

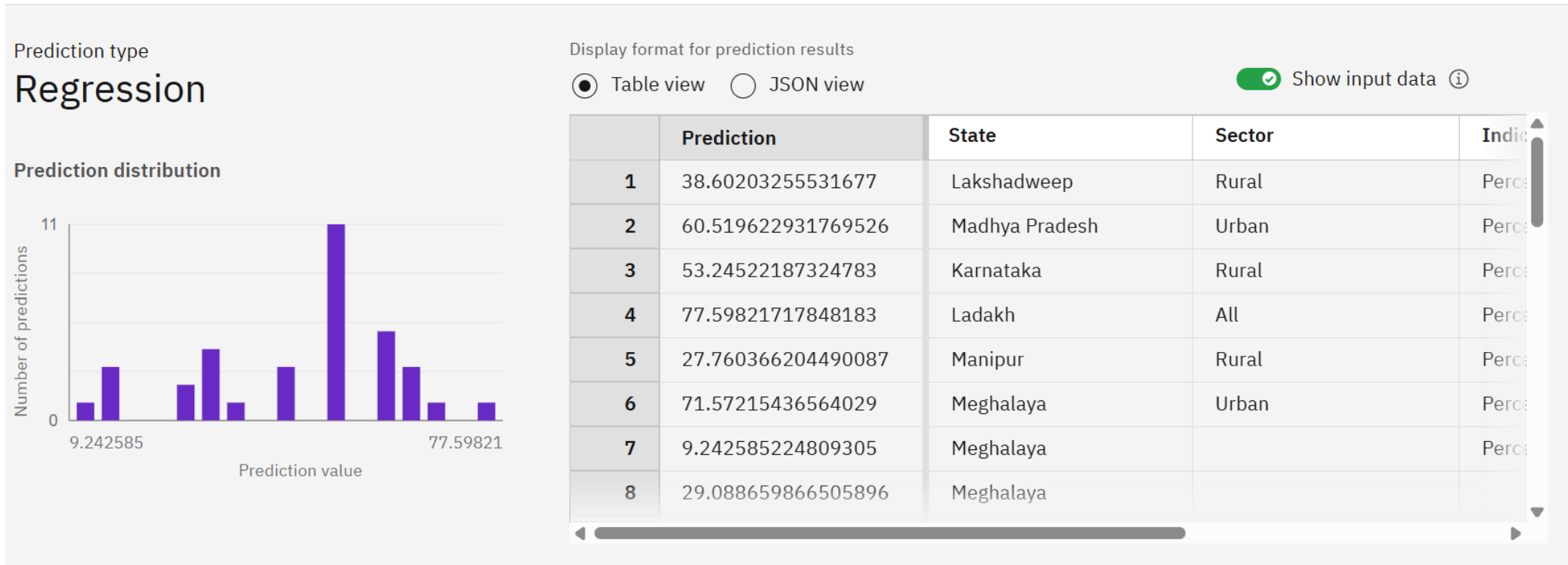
IBM watsonx.ai Studio provides the set of integrated tools for IBM watsonx.ai and Cloud Pak for Data as a Service. IBM watsonx.ai Studio is powered by IBM watsonx.ai Runtime. With a suite of tools for all skill levels, everyone can collaborate to develop machine learning solutions and on watsonx.ai, develop generative AI solutions. You can write code, visually code on a graphical canvas, or automatically build AI solutions.

# ALGORITHM & DEPLOYMENT

- **Algorithm Selection:**
  - AutoAI on **IBM Watsonx.ai** was used to automatically select and train the best machine learning model for predicting water access indicators.
- **Data Input:**
  - Features: Household size, location, caste group, religion, water source type, distance to water source
  - Target: **Indicator** (Improved Drinking Water Access)
- **Training Process:**
  - Data split into training and testing sets
  - AutoAI performed preprocessing, feature engineering, and model tuning
  - Best model selected based on performance metrics like accuracy and ROC AUC
- **Prediction Process:**
  - Model deployed in Watson Studio
  - Accepts input features and returns predicted status of improved water access
  - Can be used for real-time policy planning and analysis

# RESULT

## Prediction results

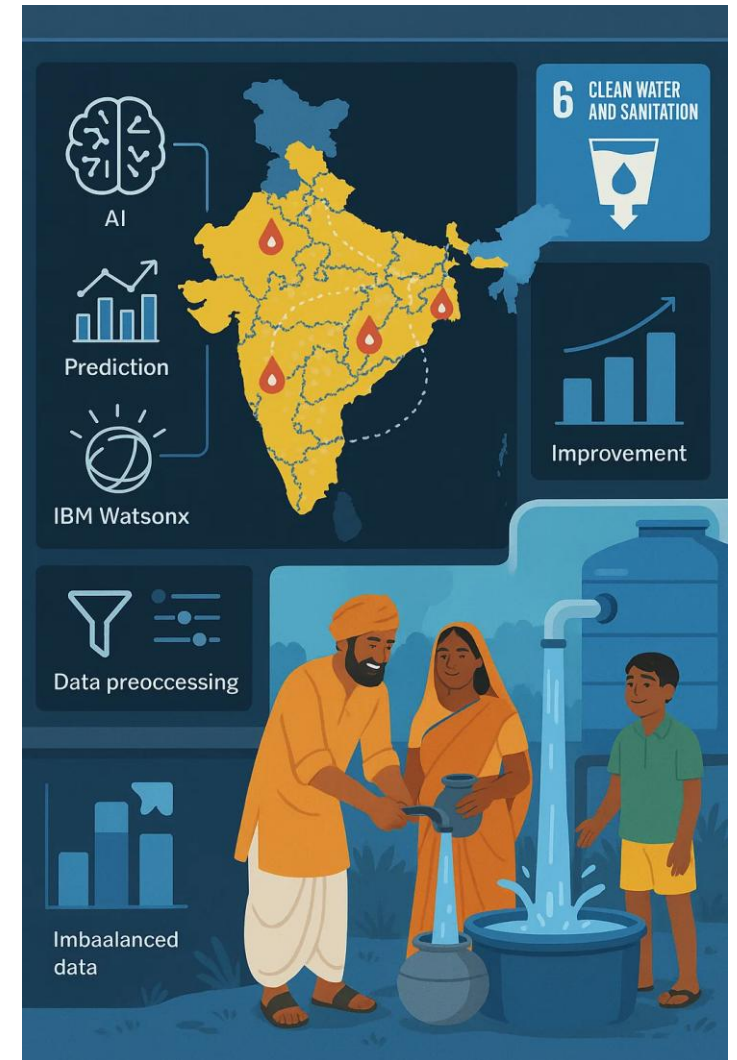


- The model effectively predicts whether a household has access to improved drinking water, aiding in data-driven decision-making.



# CONCLUSION

- The project successfully leveraged IBM Watsonx and cloud-based AI tools to predict access to improved drinking water with high accuracy.
- The solution helps identify regions and communities lacking proper access, enabling focused planning and resource allocation.
- Challenges included data imbalance and limited real-time features, which were mitigated through preprocessing and feature selection.
- This approach proves effective in supporting data-driven decisions for achieving SDG targets related to clean water access.







## FUTURE SCOPE

- Integrate real-time IoT sensor data (e.g., water quality, flow rate) for more dynamic and localized predictions.
- Expand the system to cover more states and rural regions with higher water scarcity levels.
- Apply advanced ML models like Gradient Boosting or Deep Learning for increased prediction accuracy.
- Utilize edge computing for faster decision-making in remote areas with low connectivity.
- Partner with local authorities and NGOs to drive real-world impact through targeted interventions.

# REFERENCES

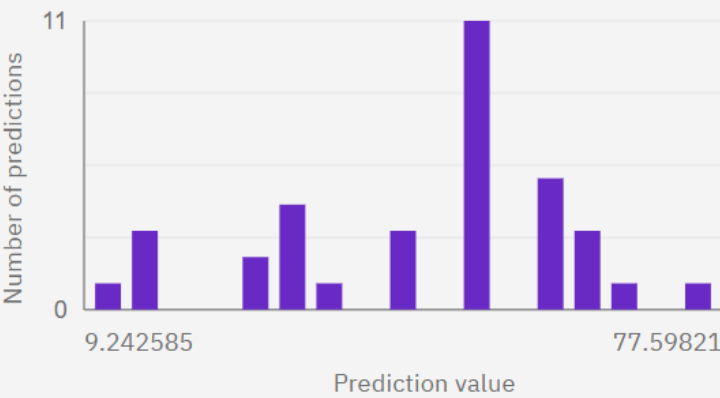
- **AI Kosh Dataset** – Improved Source of Drinking Water (78th Round Survey)  
[https://aikosh.indiaai.gov.in/web/datasets/details/improved\\_source\\_of\\_drinking\\_water\\_multiple\\_indicator\\_survey\\_78th\\_round.html](https://aikosh.indiaai.gov.in/web/datasets/details/improved_source_of_drinking_water_multiple_indicator_survey_78th_round.html)
- **IBM Cloud Lite Documentation** – Use of IBM Watson Studio, IBM Cloud Object Storage, and IBM Machine Learning Services – <https://cloud.ibm.com>
- **Python Libraries** – *Pandas, NumPy*

# SCREENSHOTS OF PROJECT

## Prediction results

Prediction type  
**Regression**

Prediction distribution



Display format for prediction results

☒ Table view ☐ JSON view

☒ Show input data ⓘ

	Prediction	State	Sector	Indic
1	38.60203255531677	Lakshadweep	Rural	Perce
2	60.519622931769526	Madhya Pradesh	Urban	Perce
3	53.24522187324783	Karnataka	Rural	Perce
4	77.59821717848183	Ladakh	All	Perce
5	27.760366204490087	Manipur	Rural	Perce
6	71.57215436564029	Meghalaya	Urban	Perce
7	9.242585224809305	Meghalaya		Perce
8	29.088659866505896	Meghalaya		

## SCREENSHOTS OF PROJECT

IBM watsonx.ai Studio

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Experiment summary

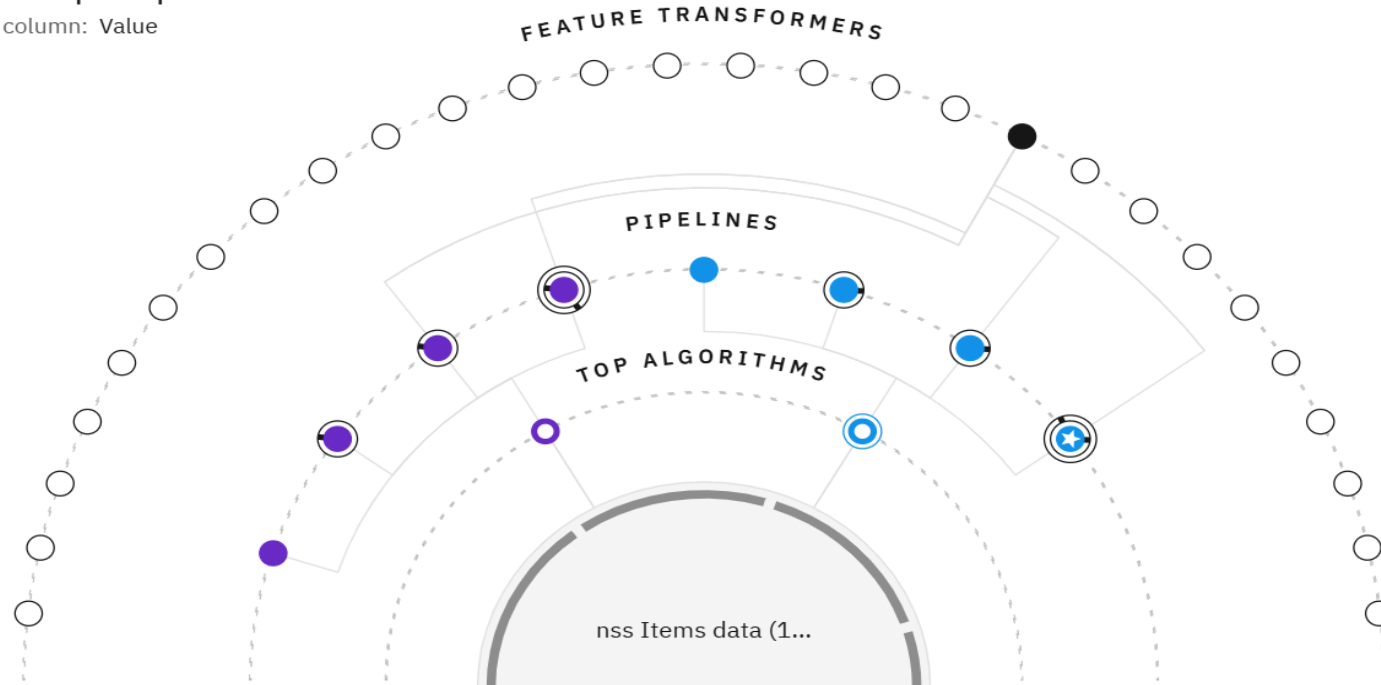
Pipeline comparison

★ Rank by:

Root mean squared error (RMSE) (... | Cross validation score

## Relationship map ⓘ

Prediction column: Value



## Progress map

Swap view  $\rightleftharpoons$



Experiment completed 

8 PIPELINES GENERATED

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

Time elapsed: 4 minutes

[View log](#)[Save code](#)

# SCREENSHOTS OF PROJECT

	A	B	C	D	E
1	State	Sector	Indicator	Sub Indicator	Value
2	All India	Rural	Percentage of Persons Reported Access to Drinking Water,	Piped Water into Dwelling or Yard/plot Which was Sufficiently Available	22.5
3	All India	Urban	Percentage of Persons Reported Access to Drinking Water,	Piped Water into Dwelling or Yard/plot Which was Sufficiently Available	58.2
4	All India	All	Percentage of Persons Reported Access to Drinking Water,	Piped Water into Dwelling or Yard/plot Which was Sufficiently Available	32.9
5	All India	Rural	Percentage of Persons Reported Access to Drinking Water,	Improved Source of Drinking Water Located in the Household Premises	56.3
6	All India	Urban	Percentage of Persons Reported Access to Drinking Water,	Improved Source of Drinking Water Located in the Household Premises	76.3
7	All India	All	Percentage of Persons Reported Access to Drinking Water,	Improved Source of Drinking Water Located in the Household Premises	62.1
8	All India	Rural	Percentage of Persons Reported Access to Drinking Water,	Improved Source of Drinking Water Located in the Household Premises	59.6
9	All India	Urban	Percentage of Persons Reported Access to Drinking Water,	Improved Source of Drinking Water Located in the Household Premises	80.2
10	All India	All	Percentage of Persons Reported Access to Drinking Water,	Improved Source of Drinking Water Located in the Household Premises	65.6
11	All India	Rural	Percentage of Persons Reported Access to Drinking Water,	Exclusive Access to Improved Source of Drinking Water Located in the H	47.7
12	All India	Urban	Percentage of Persons Reported Access to Drinking Water,	Exclusive Access to Improved Source of Drinking Water Located in the H	63.1
13	All India	All	Percentage of Persons Reported Access to Drinking Water,	Exclusive Access to Improved Source of Drinking Water Located in the H	52.2
14	All India	Rural	Percentage of Persons Reported Access to Drinking Water,	Exclusive Access to Improved Source of Drinking Water Located in the H	50.6
15	All India	Urban	Percentage of Persons Reported Access to Drinking Water,	Exclusive Access to Improved Source of Drinking Water Located in the H	66.5
16	All India	All	Percentage of Persons Reported Access to Drinking Water,	Exclusive Access to Improved Source of Drinking Water Located in the H	55.3
17	Andaman & Nicobar Islands	Rural	Percentage of Persons Reported Access to Drinking Water,	Piped Water into Dwelling or Yard/plot Which was Sufficiently Available	56.7
18	Andaman & Nicobar Islands	Urban	Percentage of Persons Reported Access to Drinking Water,	Piped Water into Dwelling or Yard/plot Which was Sufficiently Available	70.2
19	Andaman & Nicobar Islands	All	Percentage of Persons Reported Access to Drinking Water,	Piped Water into Dwelling or Yard/plot Which was Sufficiently Available	62
20	Andaman & Nicobar Islands	Rural	Percentage of Persons Reported Access to Drinking Water,	Improved Source of Drinking Water Located in the Household Premises	58
21	Andaman & Nicobar Islands	Urban	Percentage of Persons Reported Access to Drinking Water,	Improved Source of Drinking Water Located in the Household Premises	70.3
22	Andaman & Nicobar Islands	All	Percentage of Persons Reported Access to Drinking Water,	Improved Source of Drinking Water Located in the Household Premises	62.8
23	Andaman & Nicobar Islands	Rural	Percentage of Persons Reported Access to Drinking Water,	Improved Source of Drinking Water Located in the Household Premises	70.7
24	Andaman & Nicobar Islands	Urban	Percentage of Persons Reported Access to Drinking Water,	Improved Source of Drinking Water Located in the Household Premises	98.3
25	Andaman & Nicobar Islands	All	Percentage of Persons Reported Access to Drinking Water,	Improved Source of Drinking Water Located in the Household Premises	81.5
26	Andaman & Nicobar Islands	Rural	Percentage of Persons Reported Access to Drinking Water,	Exclusive Access to Improved Source of Drinking Water Located in the H	53.8
27	Andaman & Nicobar Islands	Urban	Percentage of Persons Reported Access to Drinking Water,	Exclusive Access to Improved Source of Drinking Water Located in the H	69.2
28	Andaman & Nicobar Islands	All	Percentage of Persons Reported Access to Drinking Water,	Exclusive Access to Improved Source of Drinking Water Located in the H	59.8
29	Andaman & Nicobar Islands	Rural	Percentage of Persons Reported Access to Drinking Water,	Exclusive Access to Improved Source of Drinking Water Located in the H	66.6
30	Andaman & Nicobar Islands	Urban	Percentage of Persons Reported Access to Drinking Water,	Exclusive Access to Improved Source of Drinking Water Located in the H	96.4
31	Andaman & Nicobar Islands	All	Percentage of Persons Reported Access to Drinking Water,	Exclusive Access to Improved Source of Drinking Water Located in the H	78.2
32	Andhra Pradesh	Rural	Percentage of Persons Reported Access to Drinking Water,	Piped Water into Dwelling or Yard/plot Which was Sufficiently Available	21.9
33	Andhra Pradesh	Urban	Percentage of Persons Reported Access to Drinking Water,	Piped Water into Dwelling or Yard/plot Which was Sufficiently Available	48

# SCREENSHOTS OF PROJECT

dwater\_ai

Overview

Assets

Deployments

Jobs

Manage

Find assets

Import assets

New asset

1 asset

All assets

Asset types

Models

All assets

Name	Last modified	
P8 - Snap Boosting Machine Regressor: dwp_ai Machine learning model from AutoAI	2 hours ago Service	



# SCREENSHOTS OF PROJECT

[Deployment spaces](#) / [dwater\\_ai](#) / P8 - Snap Boosting Machine Regressor: dwp\_ai

Deployments

**Model details**

Input (1)

Column	↑	Type
Indicator		other
Sector		other
State		other
Sub Indicator		other



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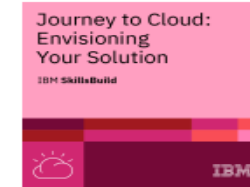
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According to the Adobe Learning Manager system of record

**Completion date:** 21 Jul 2025 (GMT)

**Learning hours:** 20 mins



**THANK YOU**