



# Gaining Momentum

26–27 February 2025 — Virtual Event

GEOTHERMAL

2025

## aiION: A Machine Learning Approach to Geothermal Exploration

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## Geothermal Overview

1

### Renewable Source

Geothermal energy is derived from Earth's heat, providing stable, sustainable power.

2

### Global Growth

Geothermal energy could meet 15% of global electricity demand by 2050, with a potential global capacity of 800 gigawatts - equivalent to the current electricity demand of the US and India combined (IEA 2025).

3

### Key Countries

Significant in USA, Indonesia, Philippines, Türkiye, and New Zealand.



# Exploration Challenges

- 1** **Financial Risks**  
 High costs during predevelopment stages, including surface surveys and exploratory drilling.
- 2** **Hidden Resources**  
 Difficulty in identifying blind geothermal resources without surface manifestations.
- 3** **Expert Reliance**  
 Traditional methods heavily depend on expert knowledge, leading to uncertainties.



## Geochemical Analysis

### Cost-Effective

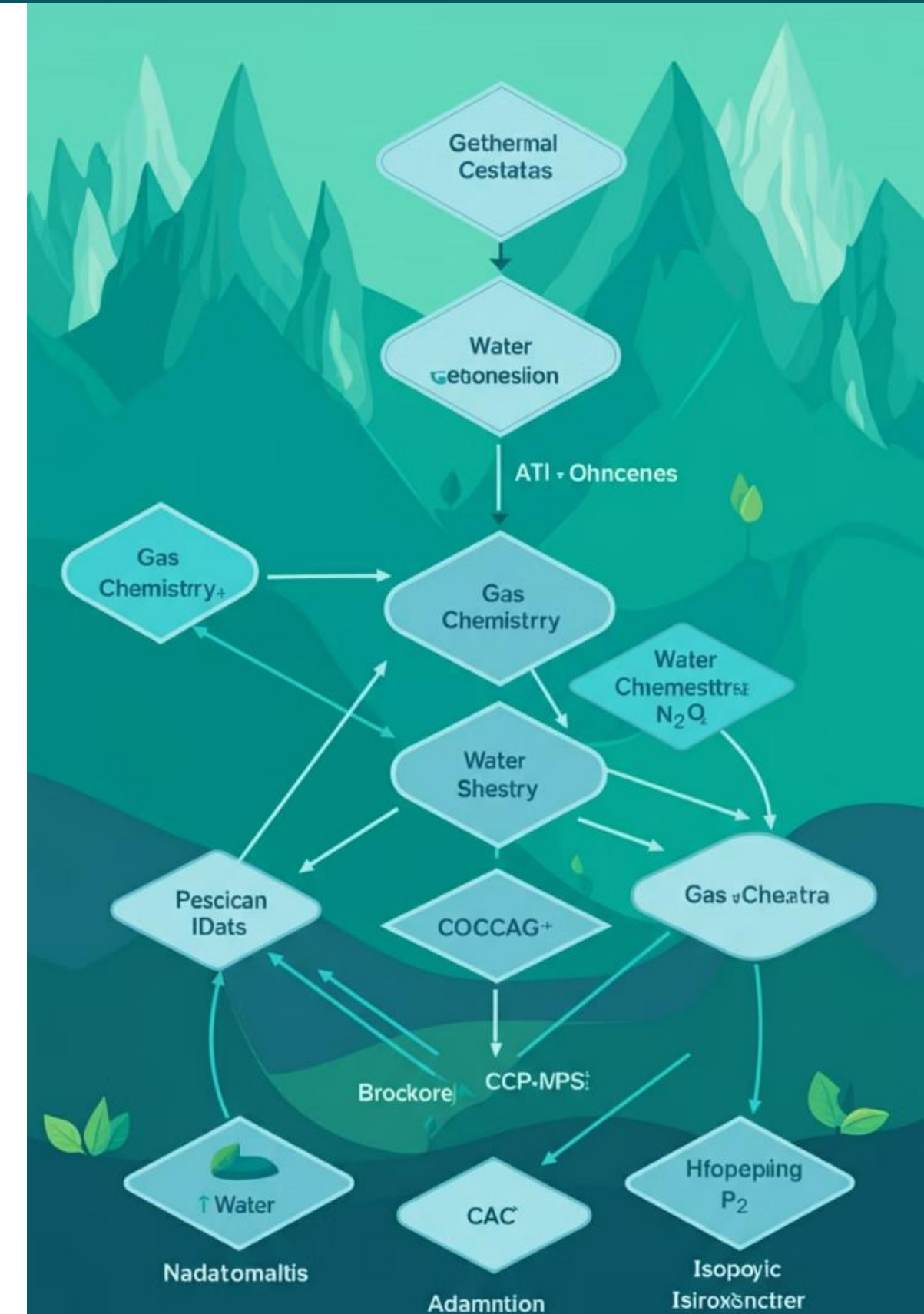
Geochemical data from groundwater samples are crucial in early exploration stages.

### Insightful

Provides valuable information on subsurface characteristics and reservoir properties.

### Analytical

Helps determine reservoir temperature, heat flow, and boundary conditions.



## Geochemical Applications



### Water Types

Aqueous species, major cations/anions, isotopes, and trace elements.

### Geothermometers

Estimate subsurface reservoir temperatures based on chemical composition.

### Reservoir Insights

Identify geothermal characteristics and potential for energy extraction.



# Geothermometry

1

## Classical Geothermometers

function based on temperature-dependent mineral-fluid equilibrium reactions, primarily utilizing silica concentrations and cation ratios (Na-K, Na-K-Ca, K-Mg) in geothermal waters.

2

## Multicomponent Geothermometry

analyzes the equilibrium between multiple minerals, focusing on the convergence of mineral saturation indices at the true reservoir temperature.

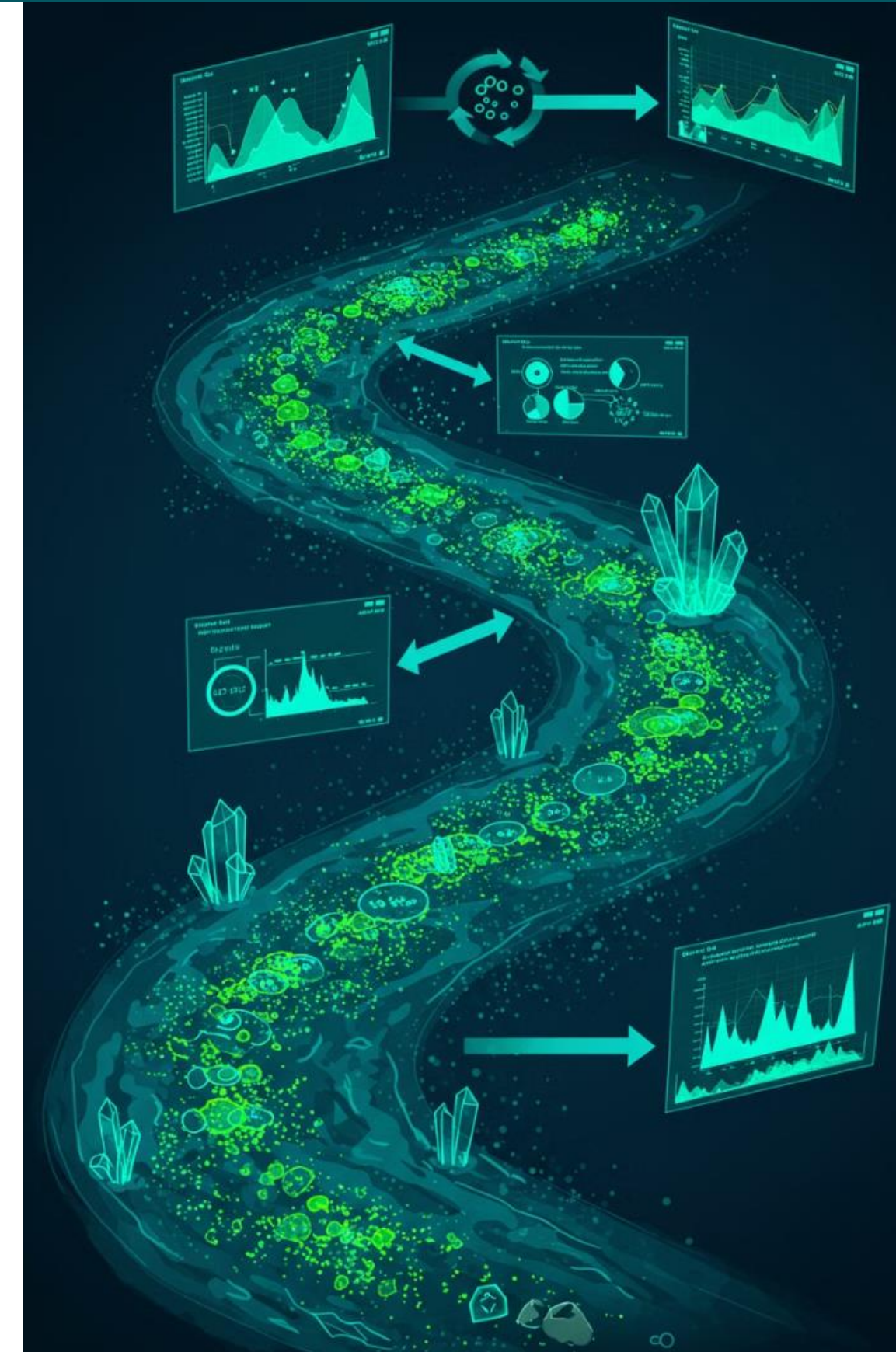
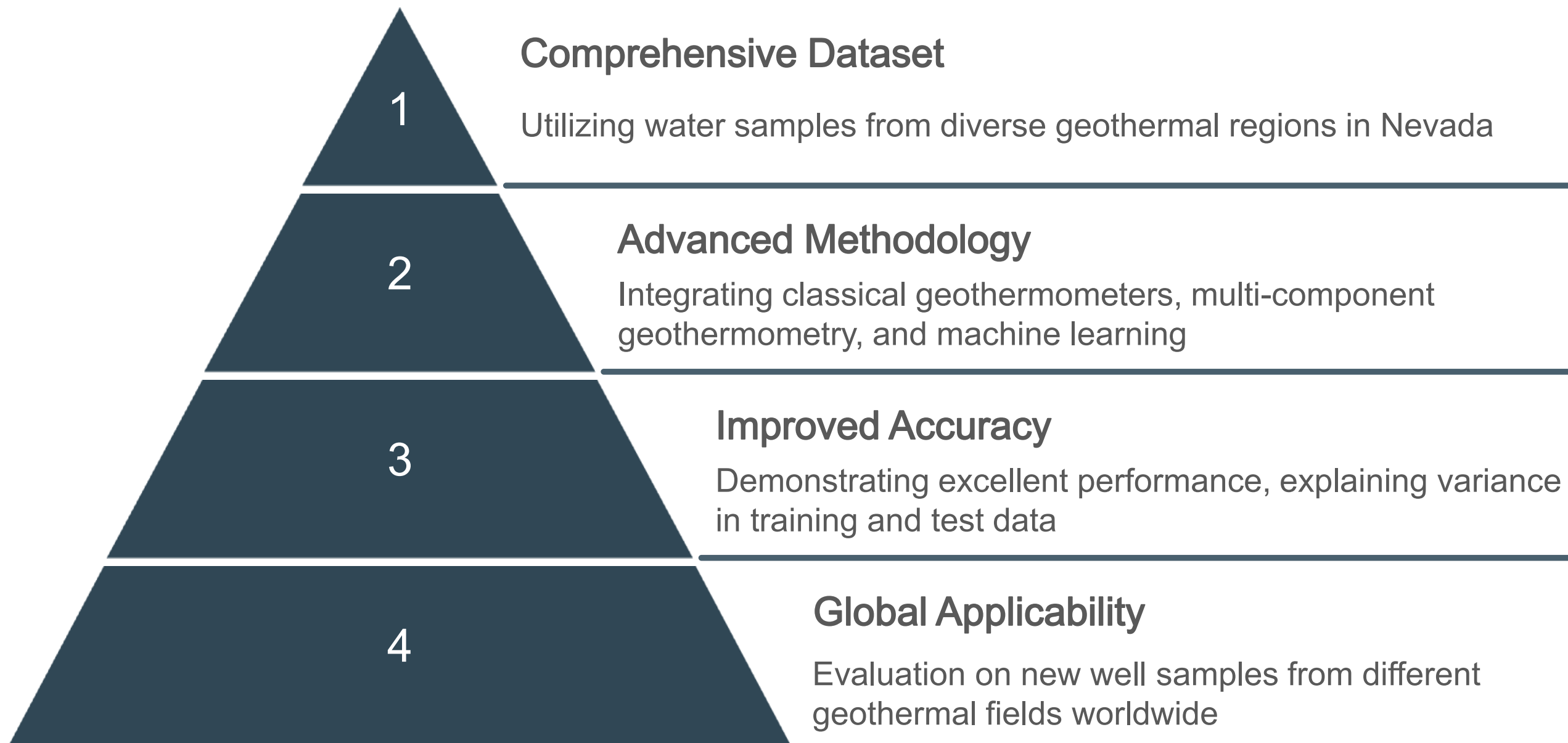
3

## Data-Driven Geothermometers

a modern approach that utilizes machine learning and statistical methods to establish correlations between fluid chemistry and reservoir temperatures.



## Methodology



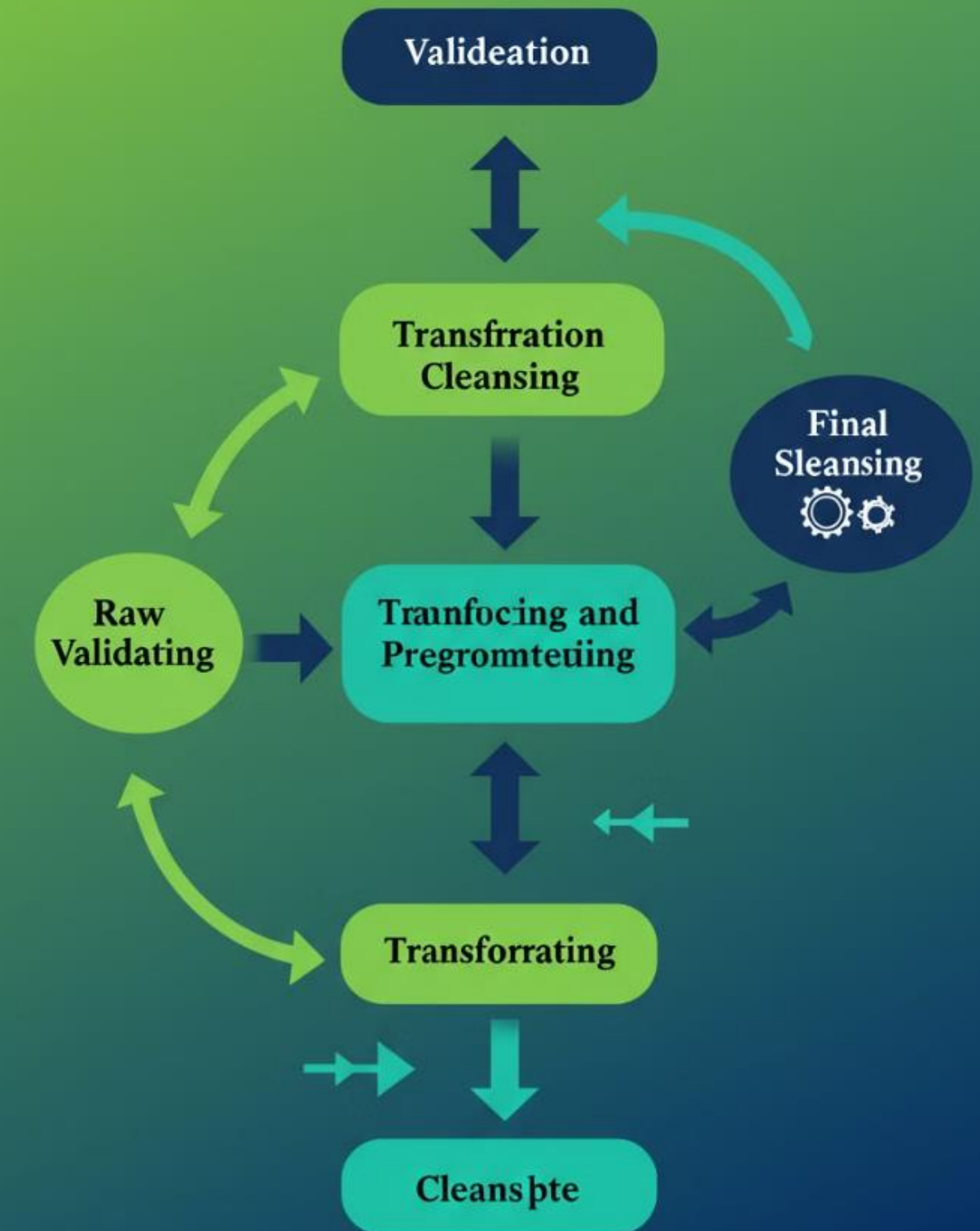
## Exploratory Data Analysis

- 1** **Extensive Dataset**

Analysis of over 14,400 geochemical samples from the Great Basin Groundwater Geochemical Database. This vast collection of information forms the backbone of the platform's predictive capabilities.
- 2** **Data Integrity**

To ensure data quality, the model performs a charge balance error calculation, rejecting samples outside the acceptable range of  $\pm 5\%$ . This rigorous approach guarantees the reliability of the input data.
- 3** **Feature Selection**

Through exploratory data analysis, eight key features were identified as strong influencers of temperature: **potassium, sodium, magnesium, calcium, chloride, fluorine, silica, and pH**. These elements form the core of aiION's predictive model.





## Data Processing

### Data Transformation

Employing various data transformation methods, including z-score, logarithmic transformation, and quantile normalization. These techniques optimize model performance and ensure accurate predictions.

### ML Model Development

Several machine learning algorithms were evaluated to determine each algorithm's predictive ability and to determine the best model.

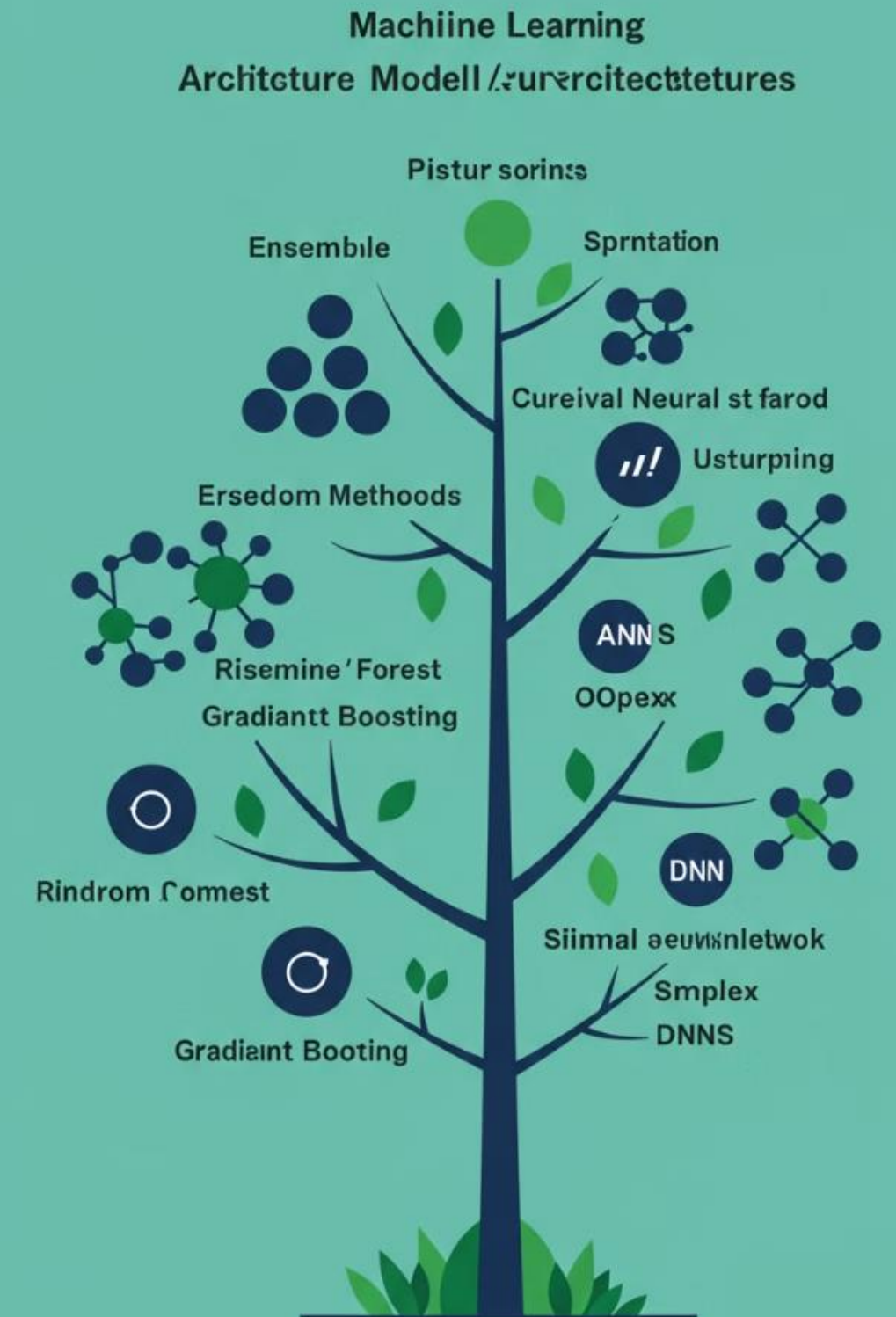
### Clustering Analysis

The software utilized clustering techniques such as K-means and Hierarchical Clustering to identify patterns within the dataset. This approach helps in understanding the underlying structure of the geochemical data.



## Evaluated ML Models

- 1** — **Random Forest (RF)**  
 Ensemble learning method using multiple decision trees
- 2** — **Gradient Boosting (XGB)**  
 Builds new trees sequentially to reduce bias from previous trees
- 3** — **Artificial Neural Network (ANN)**  
 Simple backpropagation neural network with four layers
- 4** — **Deep Neural Network (DNN)**  
 More complex architecture with three hidden layers and advanced techniques



## DNN Model Performance Metrics

# 0.9784

DNN R<sup>2</sup> (Train)

Coefficient of determination for training data

# 0.9783

DNN R<sup>2</sup> (Test)

Coefficient of determination for test data

# 4.0097

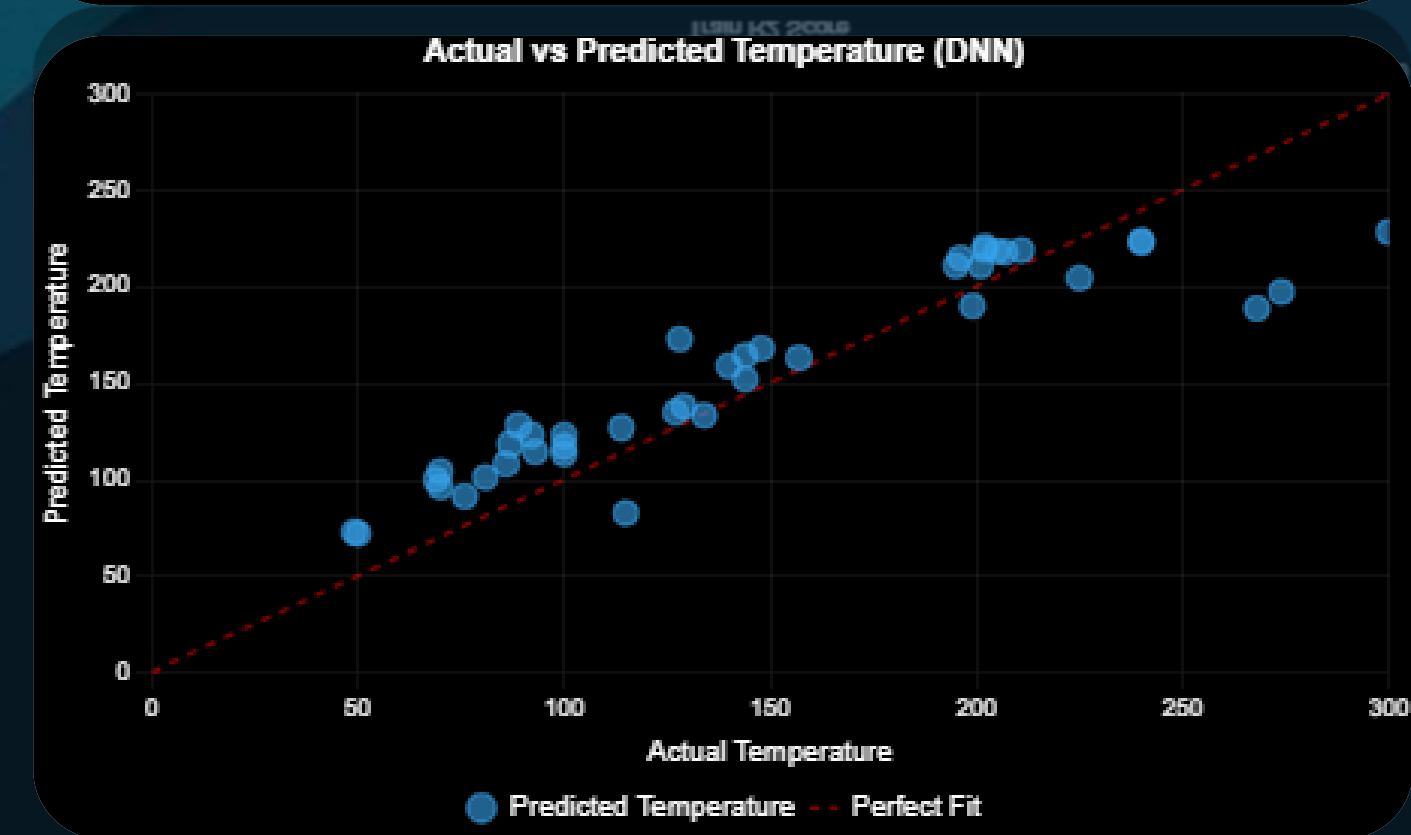
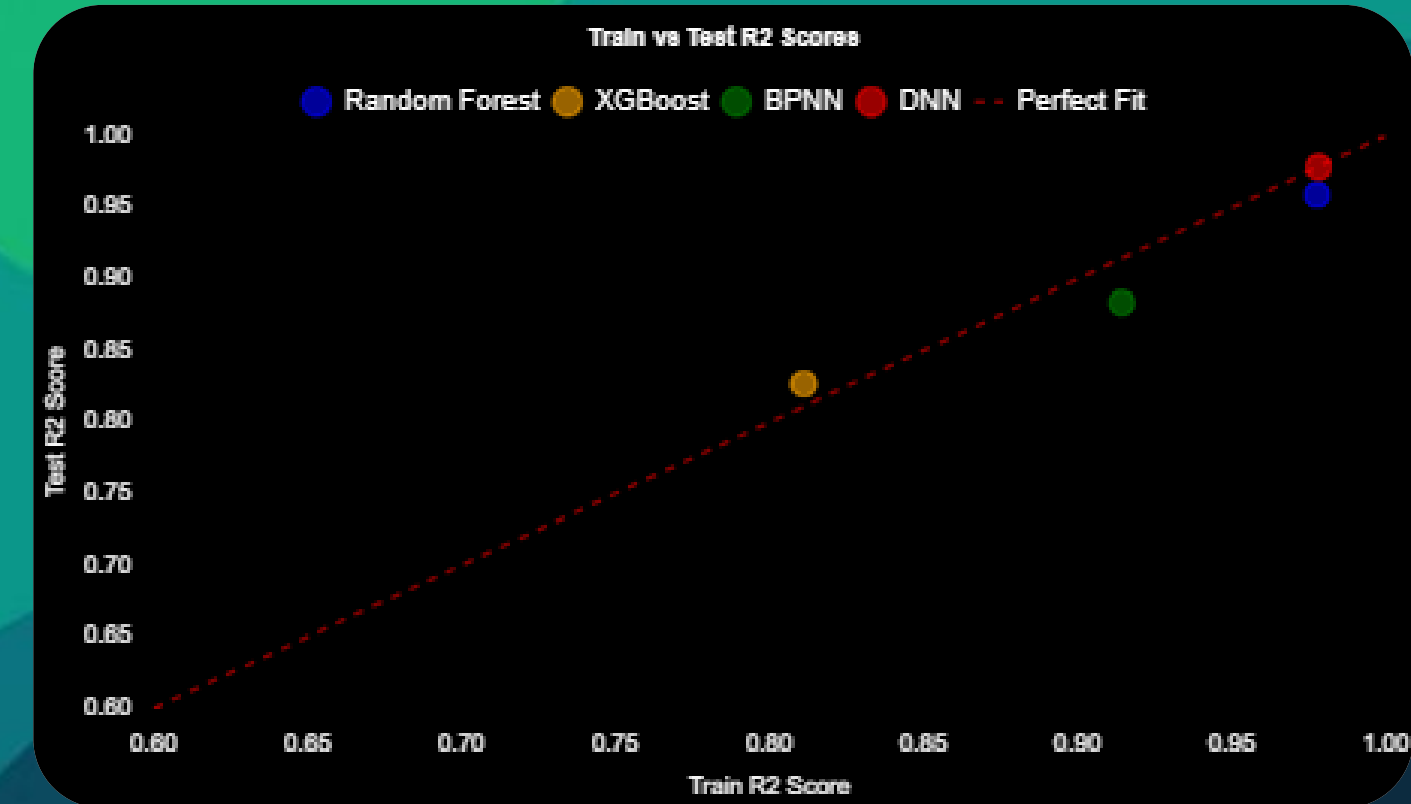
DNN RMSE

Root Mean Square Error for test data

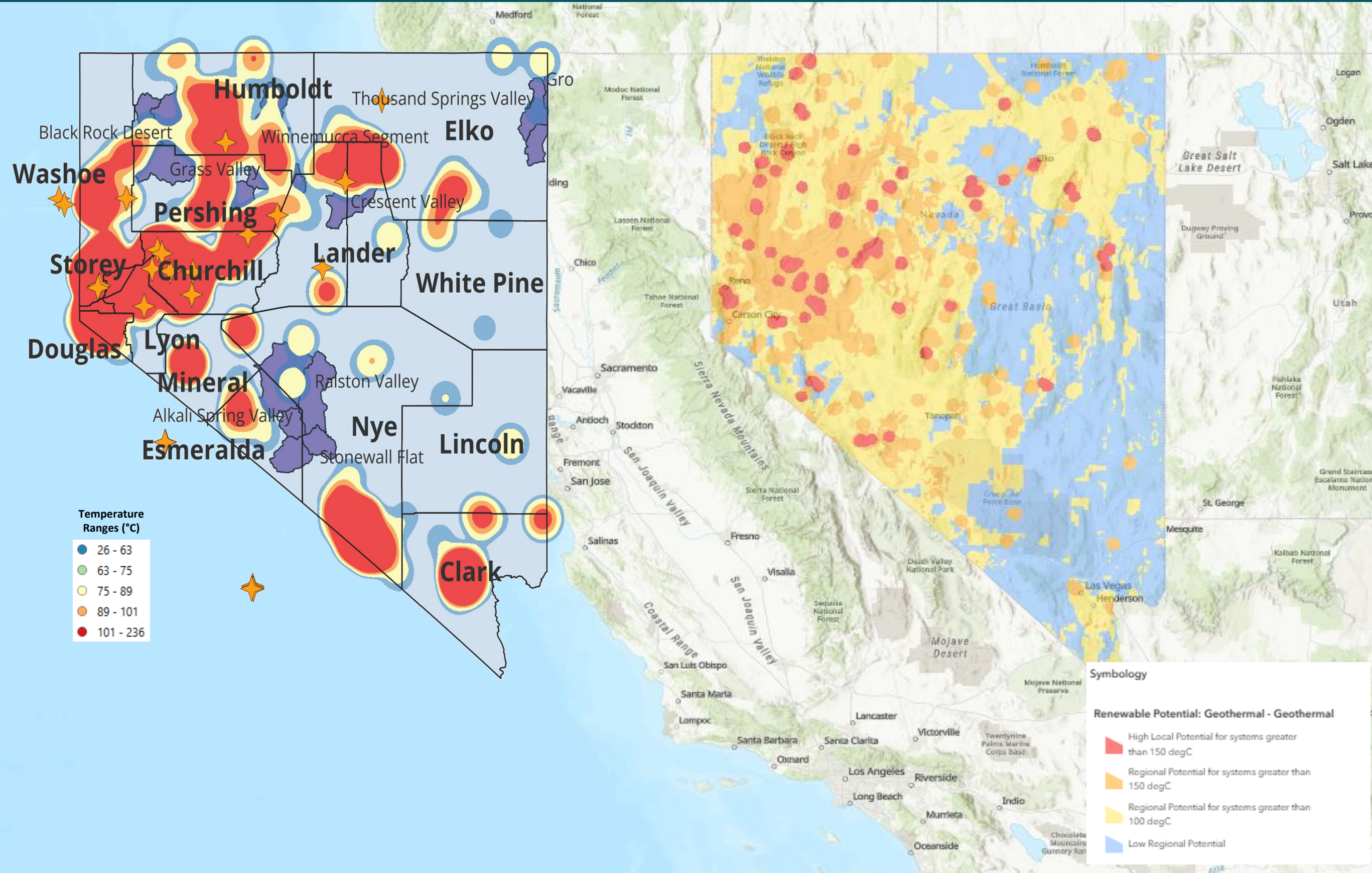
# 2.6363

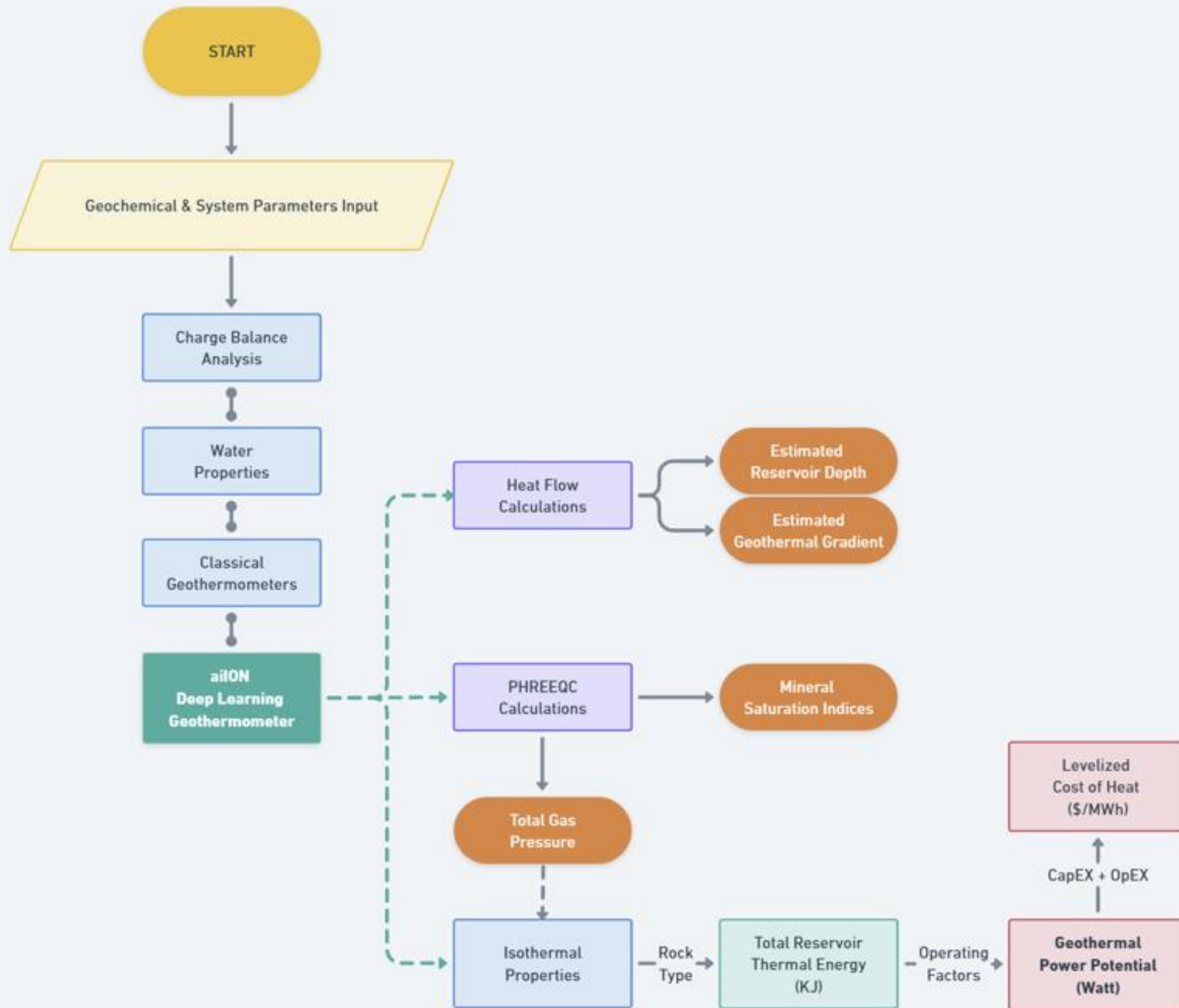
DNN MAE

Mean Absolute Error for test data



## Nevada Geothermal Potential Map





## Water Properties Module

### ■ Classification

Categorize water based on its properties, including class, type, and description. This classification helps in understanding the nature and origin of the geothermal fluid.

### ■ Hydrogeochemical Processes

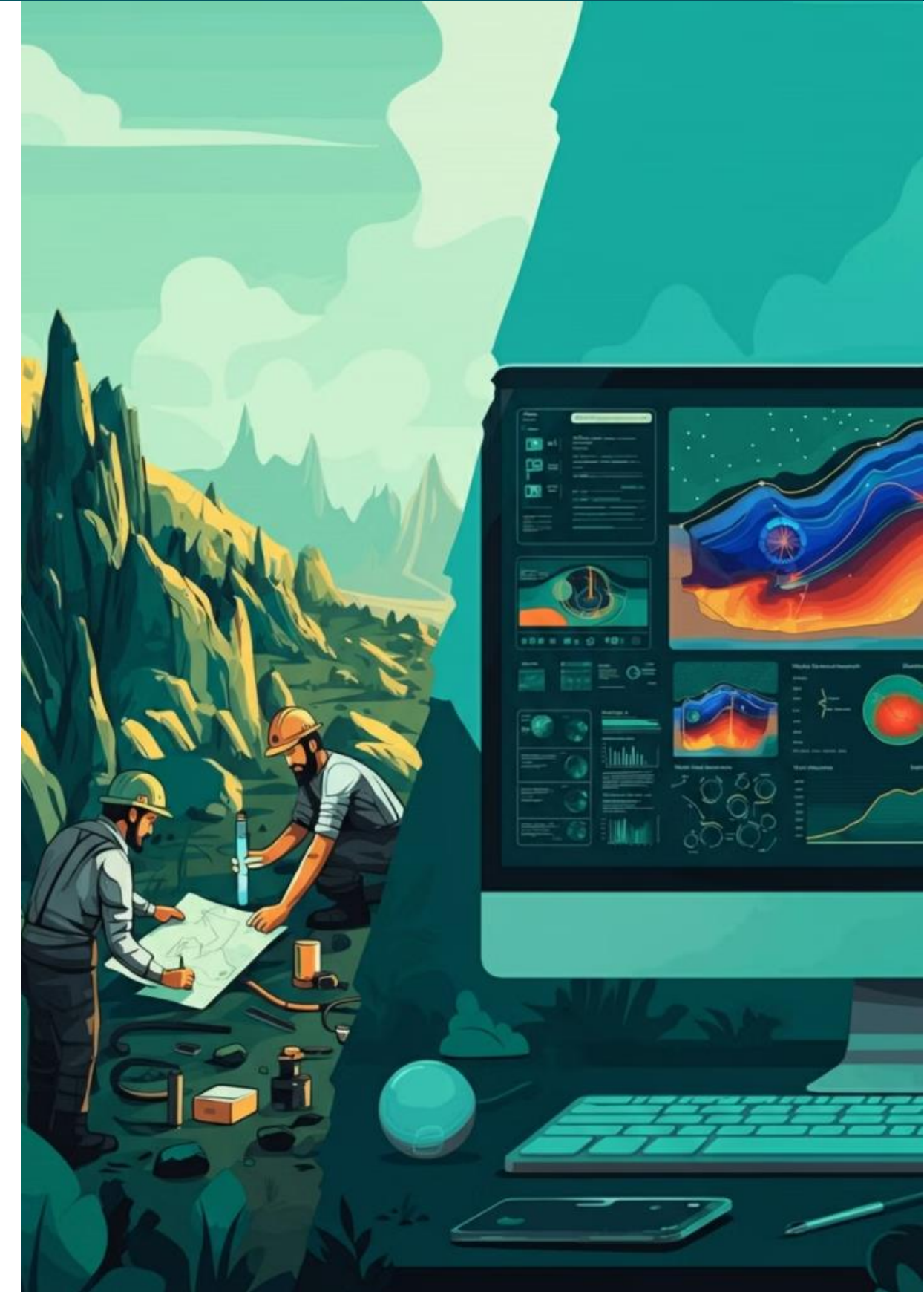
Understanding the mechanisms controlling water chemistry. It helps identify the dominant processes affecting the geothermal fluid composition and controlling water chemistry and rock weathering, categorizing samples into precipitation-dominated, rock weathering-dominated, and evaporation/crystallization-controlled types.

### ■ Chemical Equilibrium State

Determine the chemical equilibrium state of the water, providing insights into its maturity and potential for geothermal energy production.

### ■ Trace Elements Analysis

Ternary Diagrams to visualize the relationships between  $\text{Cl}^-$ , B, and  $\text{F}^-$  concentrations and  $\text{Cl}^-$ , B, and  $\text{Li}^+$  concentrations in the geothermal fluid. This analysis helps in understanding the fluid's origin and evolution.



## Solution Analysis Module

### Molality and Moles

Lists the molality and moles of various elements, including C, Ca, Cl, F, K, Mg, Na, S, and Si.

### Solution Description

Provides detailed information including Sample ID, pH, pe, Activity of water, Ionic strength, Mass of water, Total alkalinity, Total CO<sub>2</sub>, Electrical balance and Percent error.

### Species Distribution

Lists the molality, activity, and log activity of various species, such as OH<sup>-</sup>, H<sup>+</sup>, H<sub>2</sub>O, HCO<sub>3</sub><sup>-</sup>, Ca<sup>2+</sup>, Cl<sup>-</sup>, F<sup>-</sup>, K<sup>+</sup>, Mg<sup>2+</sup>, Na<sup>+</sup>, SO<sub>4</sub><sup>2-</sup>, and SiO<sub>2</sub>.

### Mineral Saturation Analysis

This module provides a comprehensive summary of mineral saturation indices, indicating the extent of their saturation in the geothermal fluid. This information is crucial for understanding the chemical equilibrium of the system.

### Gas Fugacity and Pressure

ailON calculates and reports the fugacity and partial pressure of gases in the solution, such as CO<sub>2</sub> and H<sub>2</sub>O. This data is essential for understanding the behavior of gases in the geothermal reservoir.



# Thermophysical Module



## Comprehensive Solution Properties

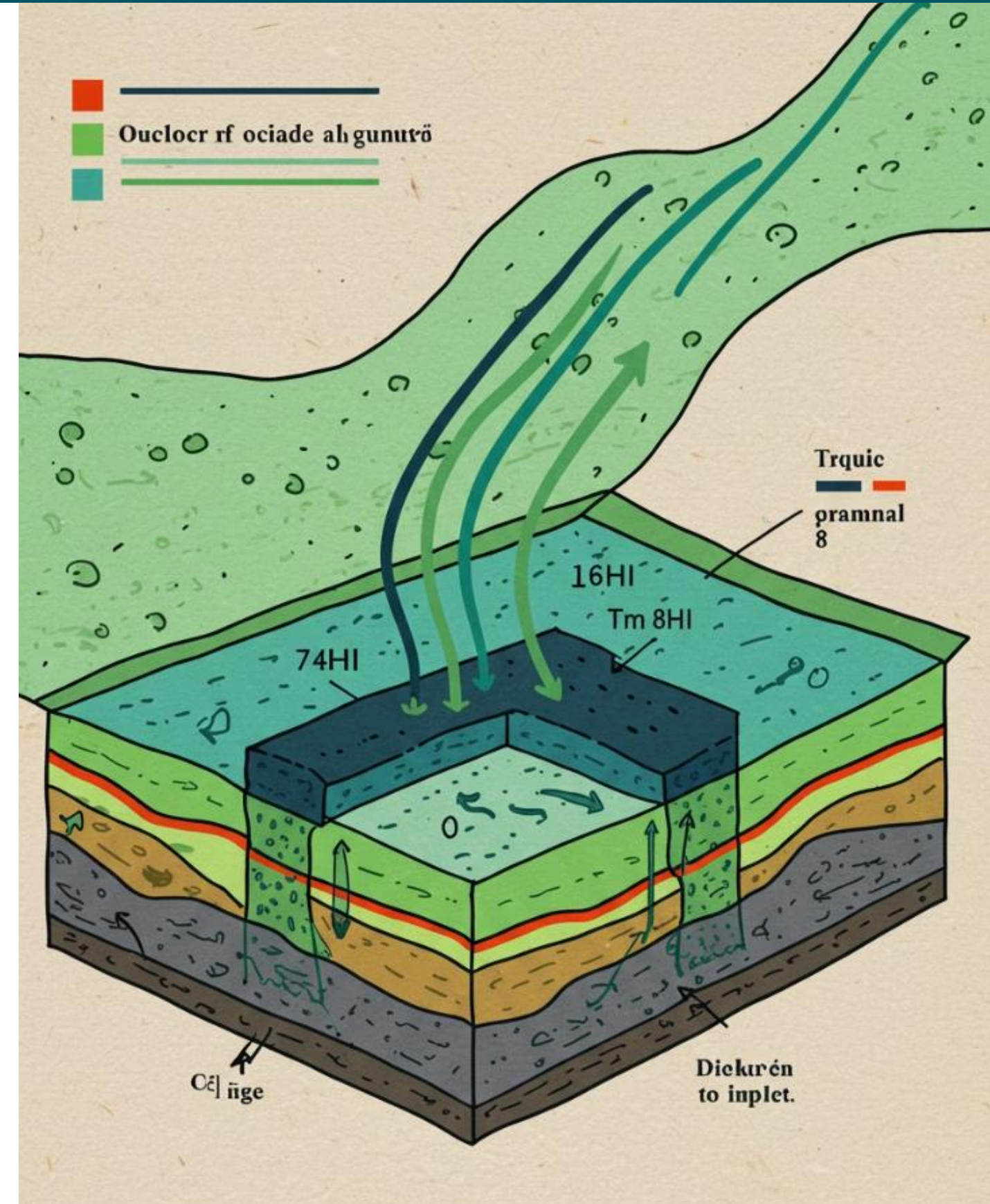
aiON calculates crucial thermophysical properties including solution quality, density, specific volume, dynamic viscosity, thermal conductivity, internal energy, entropy, enthalpy, and heat capacity.

## Geothermal Gradient Analysis

Computing effective thermal conductivity and geothermal gradients, providing insights into the heat distribution within the reservoir.

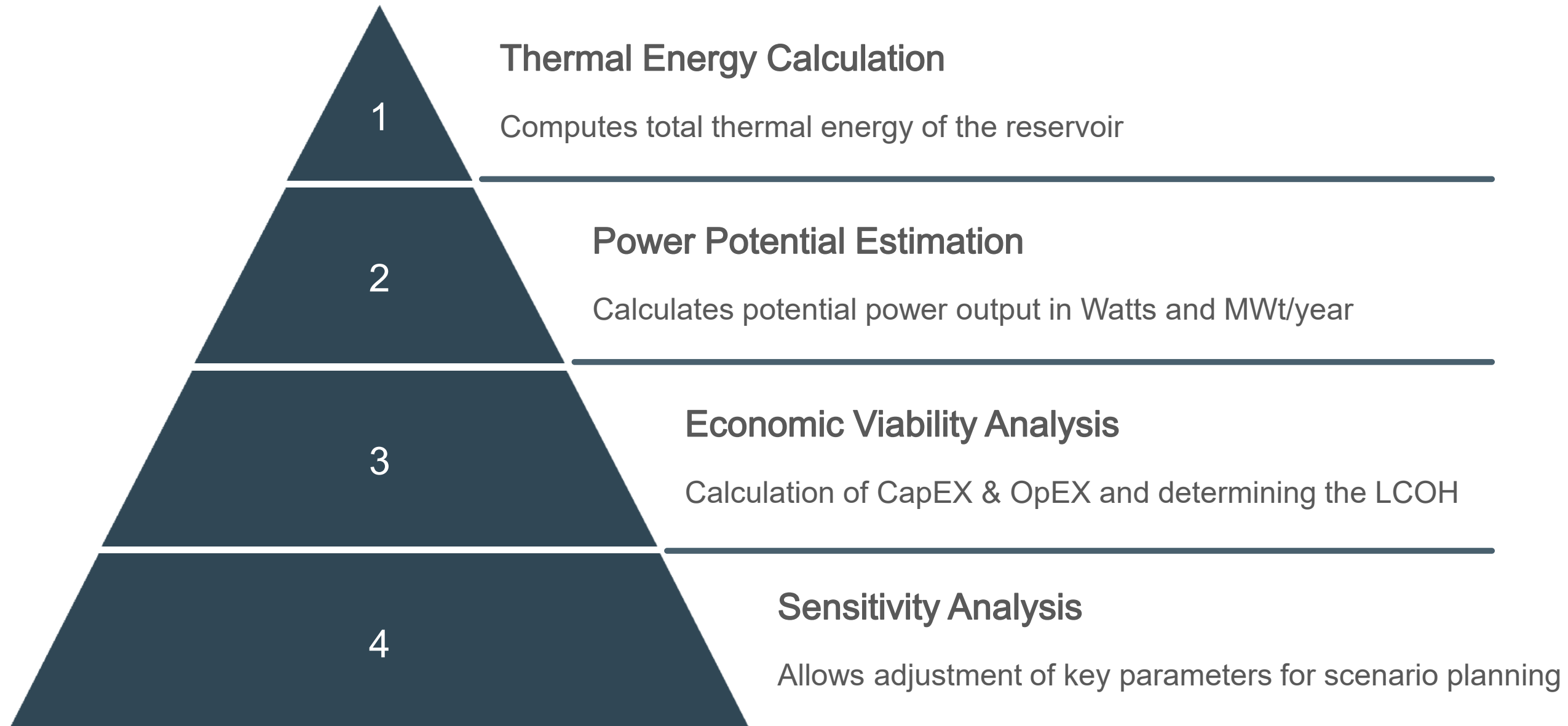
## Heat Flow Assessment

aiON calculates heat flow, a critical parameter for understanding the energy potential of a geothermal system.






# Geothermal Potential Module




# Comprehensive Technical Report



## THERMOCHEMICAL ANALYSIS REPORT

Organization	ABC Company
Country	Country
State/ Province	State
Site ID	ABC
Permit	XYZ
Generated by	User ID
Date	Monday, September 23, 2024



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## Key Challenges



### Data Availability

Scarcity of large, high-quality subsurface datasets limits AI model accuracy.



### Data Quality

Ensuring data quality is crucial for effective AI application.



### Data Accessibility

Improving access to geothermal data is necessary for wider AI adoption.



# Implications for Geothermal Exploration

1

## Improved Accuracy

ailON provides more reliable temperature predictions for geothermal reservoirs

2

## Cost-Effective Exploration

Reduces the need for expensive drilling and testing in early exploration stages

3

## Blind System Identification

Enhances ability to locate and assess "blind" geothermal systems without surface manifestations

4

## Resource Assessment

Facilitates more accurate estimation of geothermal potential in unexplored areas





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