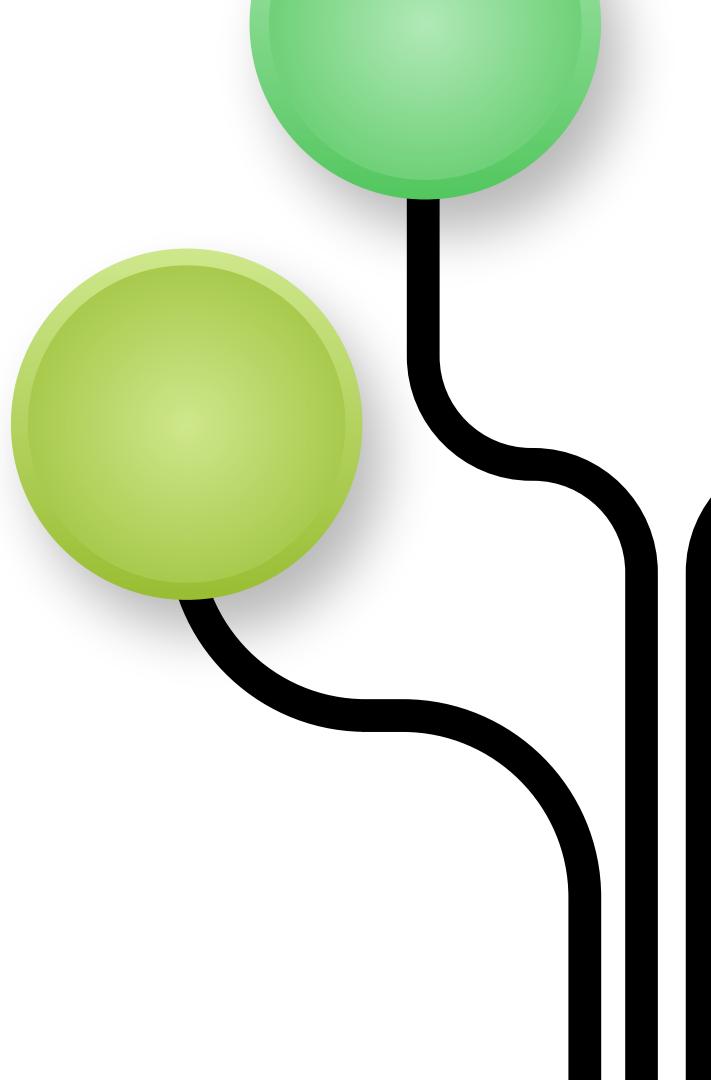


Adaptive Branching: A Constraint Satisfaction Approach

By Victoria and Leah Zhang



Overview

Introduction

Preprocessing

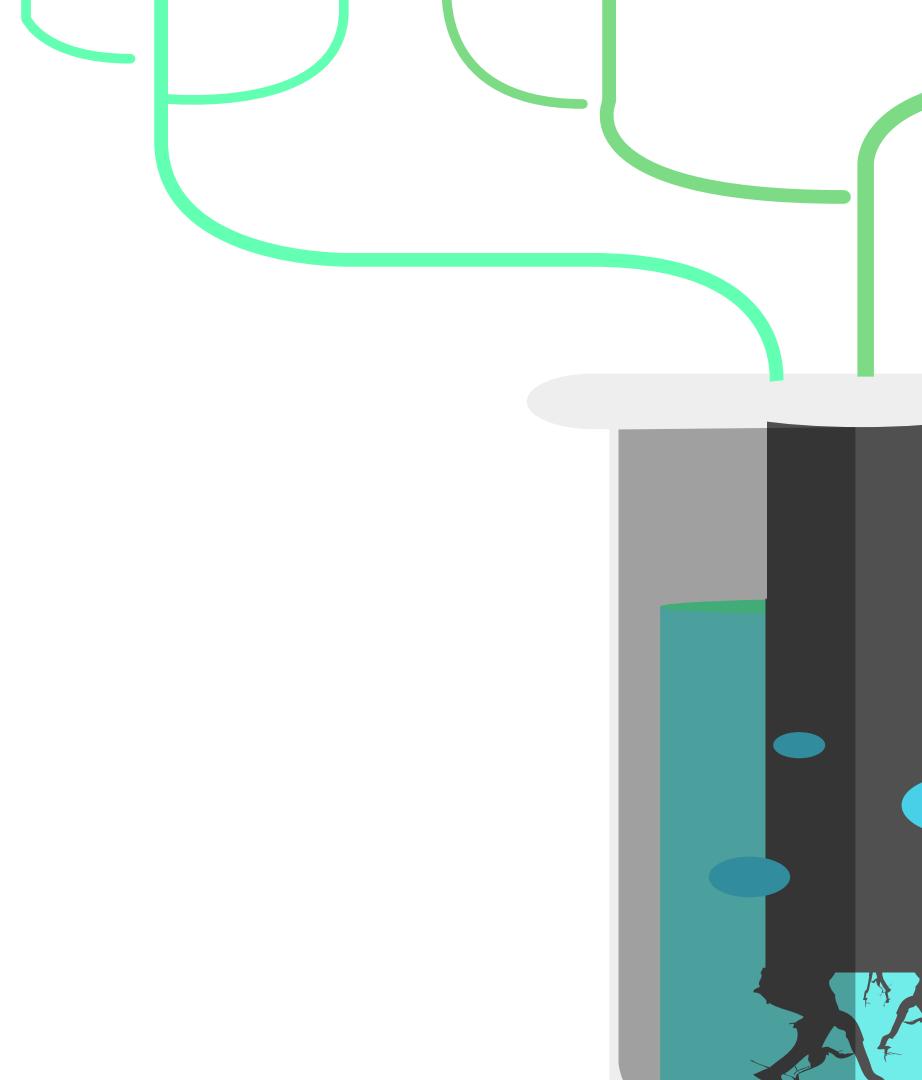
Methods

Results

Conclusion

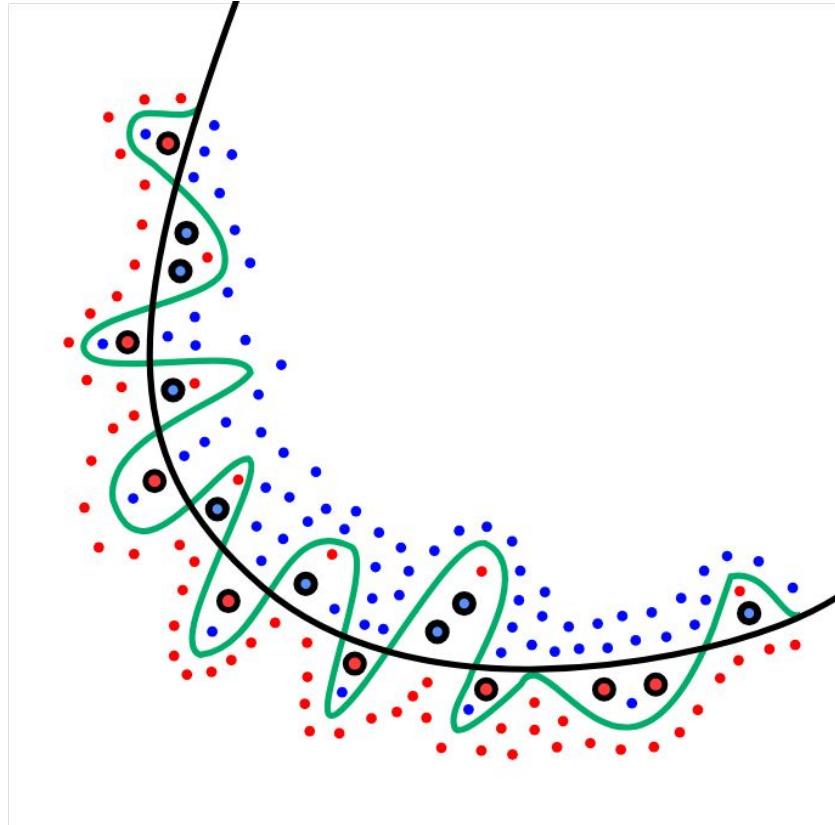


Introduction



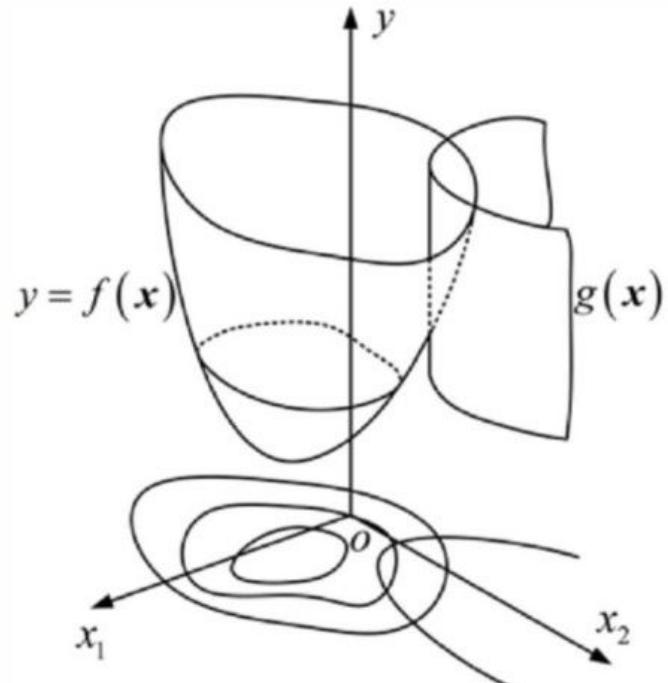
Our Problem

- Decision Trees struggle with *Overfitting*
- Pre Existing Solutions such as Ensemble Learning and Gradient Boosting lack *Efficiency and Interpretability*



Our Solution

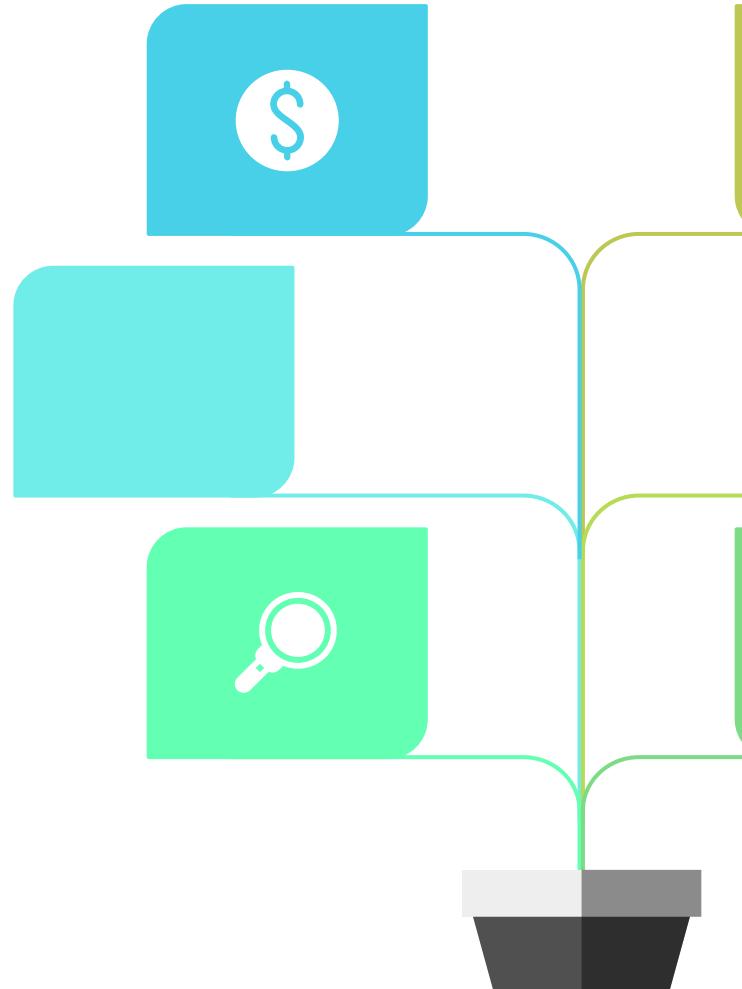
- Traditional Decision Trees algorithms utilize static thresholds for making splits and pruning
- To combat this we use *Dynamic Pruning* and *Constraint Optimization*



Our Goal

By building upon existing dynamic programming strategies while also emphasizing real-time adaptability, this project aims to construct decision trees without sacrificing either accuracy or efficiency, while also avoiding the lack of interpretability found in state-of-the-art ensemble methods.

Preprocessing



Dataset

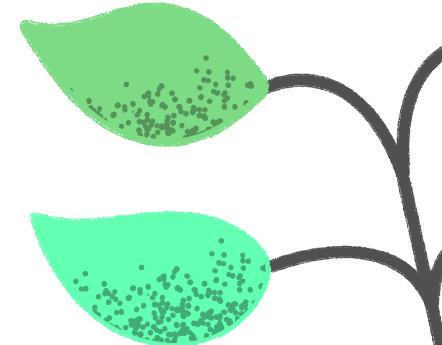
- Our dataset is on credit risk analysis.
- Class: likelihood of the loan defaulting
- ~32,800 instances
- 11 attributes



Attribute	Description
ID	Unique identifier for each loan applicant.
Age	Age of the loan applicant.
Income	Income of the loan applicant.
Home	Home ownership status (Own, Mortgage, Rent).
Emp_Length	Employment length in years.
Intent	Purpose of the loan (e.g., education, home improvement).
Amount	Loan amount applied for.
Rate	Interest rate on the loan.
Status	Loan approval status (Fully Paid, Charged Off, Current).
Percent_Income	Loan amount as a percentage of income.
Cred_Length	Length of the applicant's credit history.

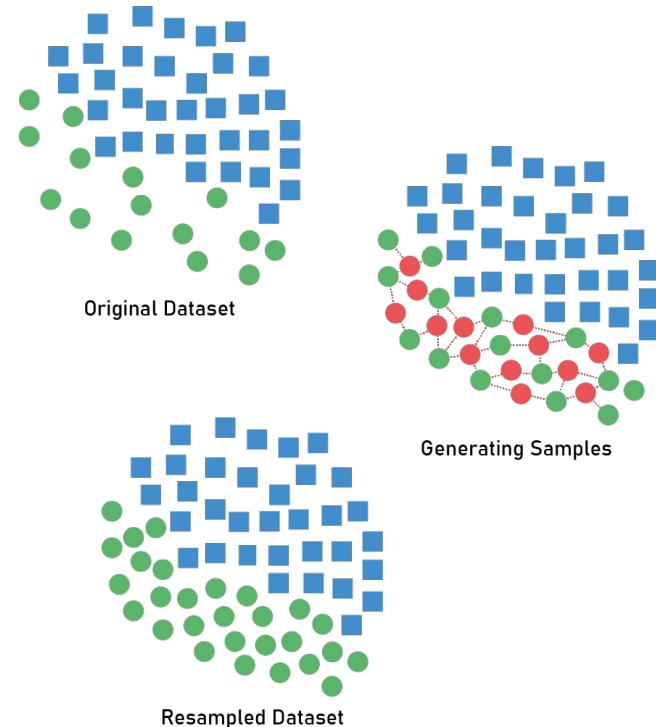
Preprocessing

- irrelevant variables, 'id', were removed.
- Missing values found in 'Emp_length' and 'Rate' were replaced
- Categorical variables, such as 'Home' and 'Intent' and including the class variable of 'Default' were numericized
- Numerical variables of 'Income,' 'Amount,' and 'Percent_income' were normalized through MinMaxScaler().



Synthetic Minority Oversampling Technique (SMOTE)

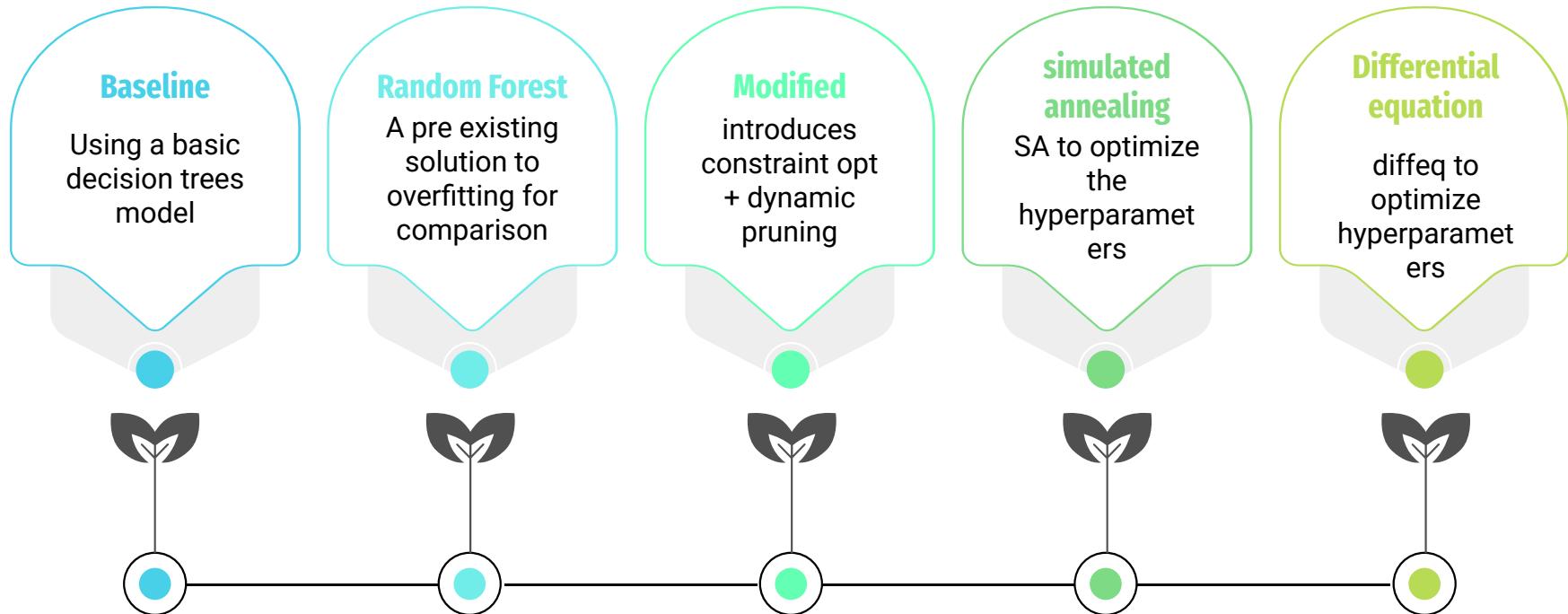
- 26.8k values for false, 82% of the total,
- 5745 values for true, 18% of the total.
- SMOTE creates synthetic data.
- Training set initially had 26063 data points with 21468 false and 4596 true
- Now 21468 for both false and true.



Methods



Overview



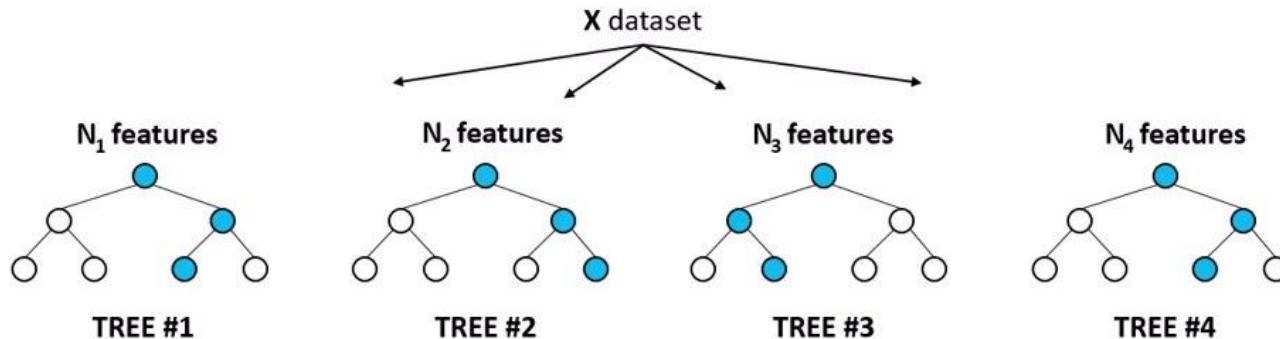
The Baseline Model

- Standard implementation of a decision tree using the scikit-learn library
- Gini impurity over information gain to reduce computation

$$Gini((D)) = 1 - \sum_{i=1}^n (p_i)^2$$

Random Forest

- Ensemble learning is a pre existing solution that addresses overfitting by using multiple decision trees
- Using the ‘RandomForestClassifier’ model from Scikit-learn. We decided to use 1000 trees and random state 42.



Modified Constraint Optimization Approach

Step	Classical Approach	Constraint-Based Approach
Feature Selection	Gini or entropy based splitting	Penalized gain with dynamic thresholds
Pruning	Cost-complexity pruning	Post-pruning using gain penalties
Computational Complexity	Greedy exploration of all branches	Prioritized search with reduced search space

Feature Selection with Dynamic Information Gain

- Dynamic feature selection process → custom calculation that includes a penalty for small gains.
- G is the standard gain or impurity calculation, P is a penalty factor, and T is a minimum gain threshold.
- Computation to choose features before splitting each node.

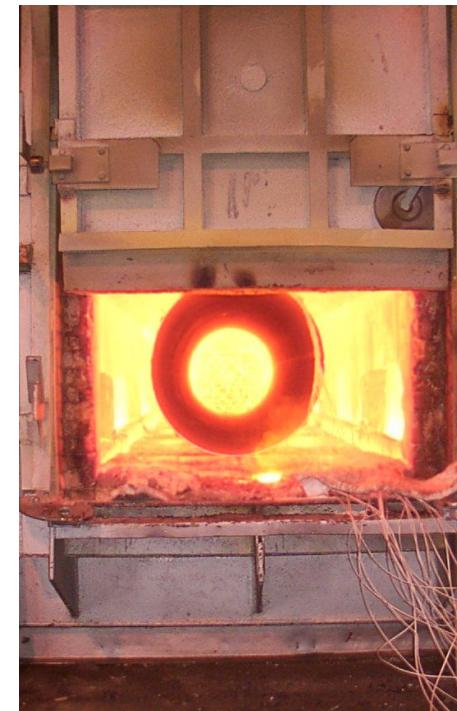
$$G_{adjusted} = G - \max(0, P * (T - G))$$

But how do we decide the best hyperparameters?

spoiler: we do some crazy shenanigans

Enhancing the Modified Tree with Simulated Annealing

- Inspired by metallurgical annealing \Rightarrow material is heated and slowly cooled to remove defects.
- “heat” = randomness in the search during the optimization process, which decreases over time to refine the solution.
- SA is a probabilistic algorithm whose strength lies in its ability to escape local minima, allowing it to find near-global optima



Enhancing the Modified Tree with Simulated Annealing

1. Initialize an initial solution
2. Calculate fitness
3. Generate new solutions by perturbing the current solution slightly
4. Evaluate new fitnesses
5. If the new solution is worse, accept it with probability $P(\text{accept}) = \exp\left(\frac{\Delta \text{fitness}}{\text{temp.}}\right)$
6. Update temperature using a set cooling rate: $T_{\text{new}} = T_{\text{old}} * \text{cooling rate.}$
7. Repeat steps 2-5 until a stopping criterion is met (max. number of iterations or a sufficiently low temperature is reached)
8. Return the best solution found after the final iteration.

Enhancing the Modified Tree with Simulated Annealing

- SA used to optimize the hyperparameters P (penalty factor) and T (minimum gain threshold)
- Fitness for this solution and subsequently-generated solutions was calculated with a function that evaluated the trade-off between accuracy and complexity via a Lagrangian objective
- It fulfills our initial objective

$$Fitness = Accuracy - \lambda * Tree\ Complexity$$

Alternative Using Differential Equation Optimization

What is a Differential Equation?

- A mathematical equation that relates a function to its derivatives.
- DiffEq a strong tool to model dynamic systems that change continuously.
- In optimization problems, DiffEq can be used to model the evolution of parameters

Alternative Using Differential Equation Optimization

Our Approach

- Uses similar function to fitness in simulated annealing
- Parameter updates based on gradient
- Stops when either the system reaches the final time point or when the gradients approach zero

$$J(P, T) = \text{Accuracy} - \lambda * \text{Number of Nodes}.$$

The actual equations

$$J(P, T) = \text{Accuracy} - \lambda * \text{Number of Nodes}.$$

$$\frac{dP}{dt} = -\frac{\partial J}{\partial P}, \quad \frac{dT}{dt} = -\frac{\partial J}{\partial T}$$

$$\frac{\partial J}{\partial P} \approx \frac{J(P+\epsilon, T) - J(P, T)}{\epsilon}, \quad \frac{\partial J}{\partial T} \approx \frac{J(P, T+\epsilon) - J(P, T)}{\epsilon}$$

Results



	Accuracy	F1 Score	Precision	Recall	ROC AUC
Baseline Tree	0.80	0.49	0.45	0.54	0.70
Random Forest	0.82	0.59	0.50	0.72	0.88
Constraint Optimization	0.80	0.50	0.45	0.55	0.68
Simulated Annealing	0.81	0.52	0.48	0.58	0.78
Differential Equation Optimization	0.8237	0	0	0	0



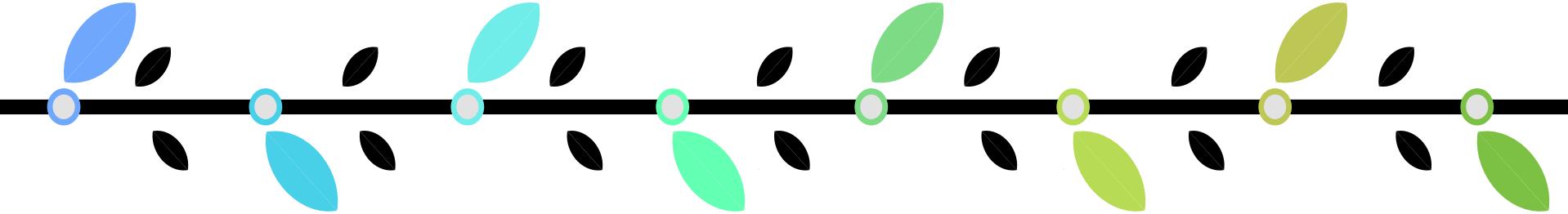
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Conclusion

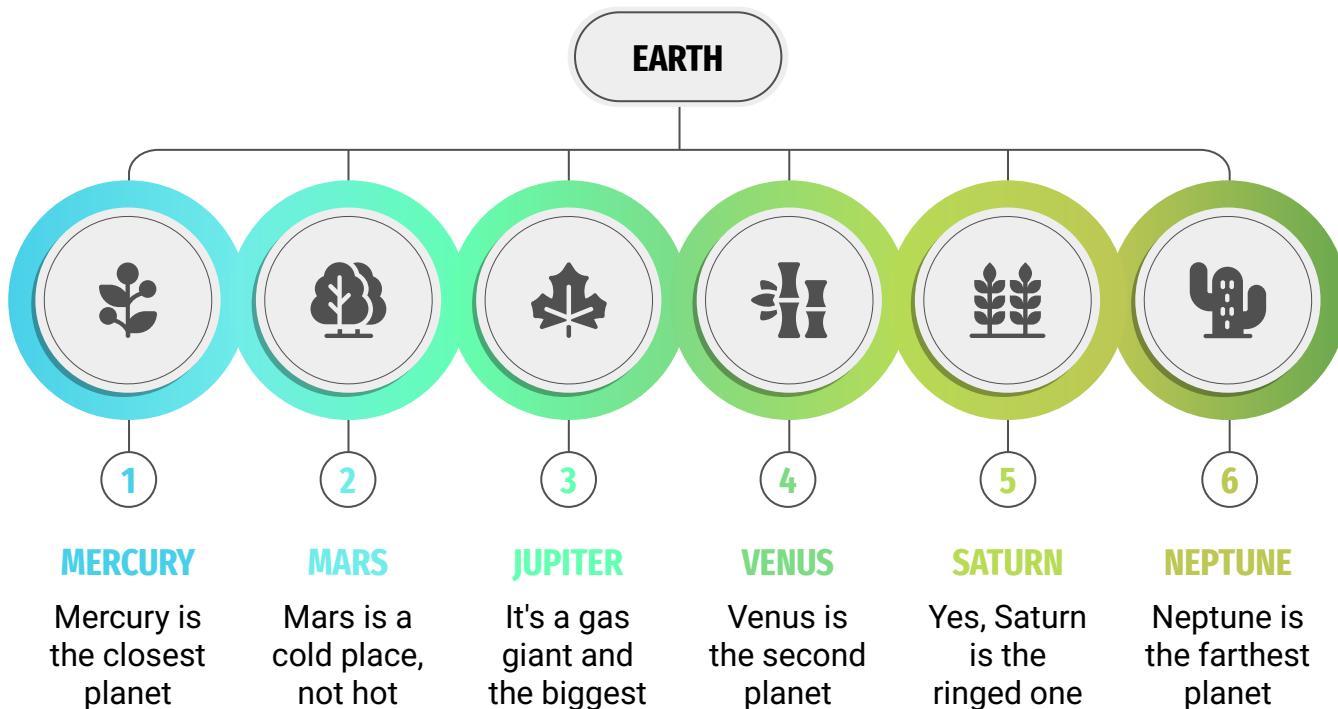
- Dynamic feature selection is possible
- Future Steps Include
 - Trying a smaller dataset
 - Using real world data prone to noise



THANK YOU



Decision Tree Diagrams



Decision Tree Diagrams



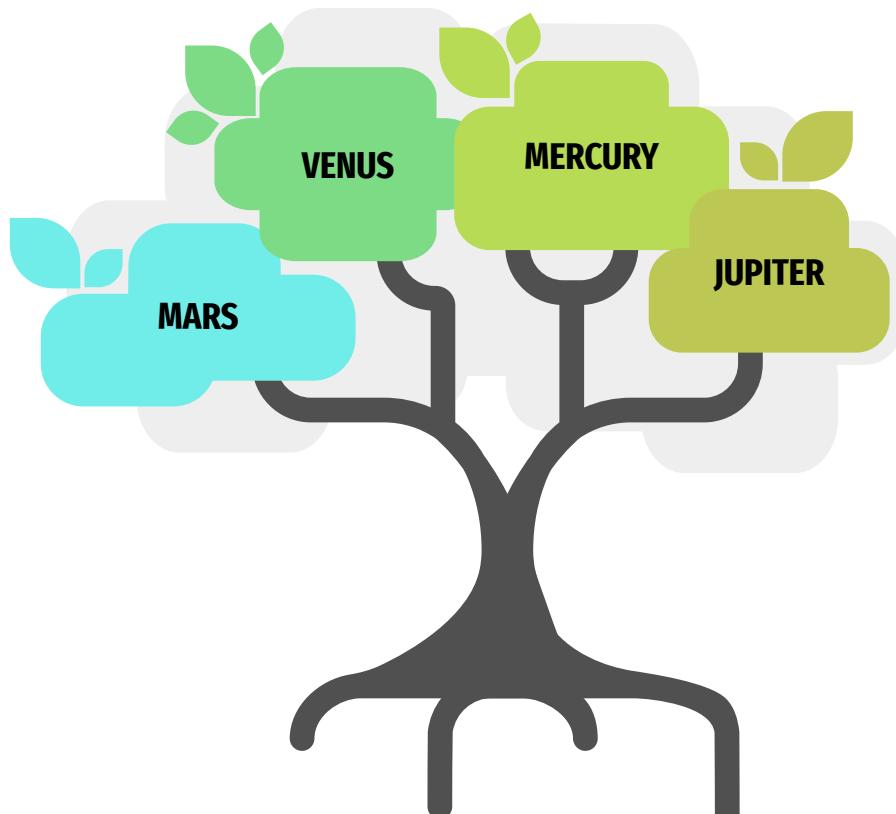
MARS

Despite being red, Mars is a cold place



VENUS

Venus has a very poisonous atmosphere



MERCURY

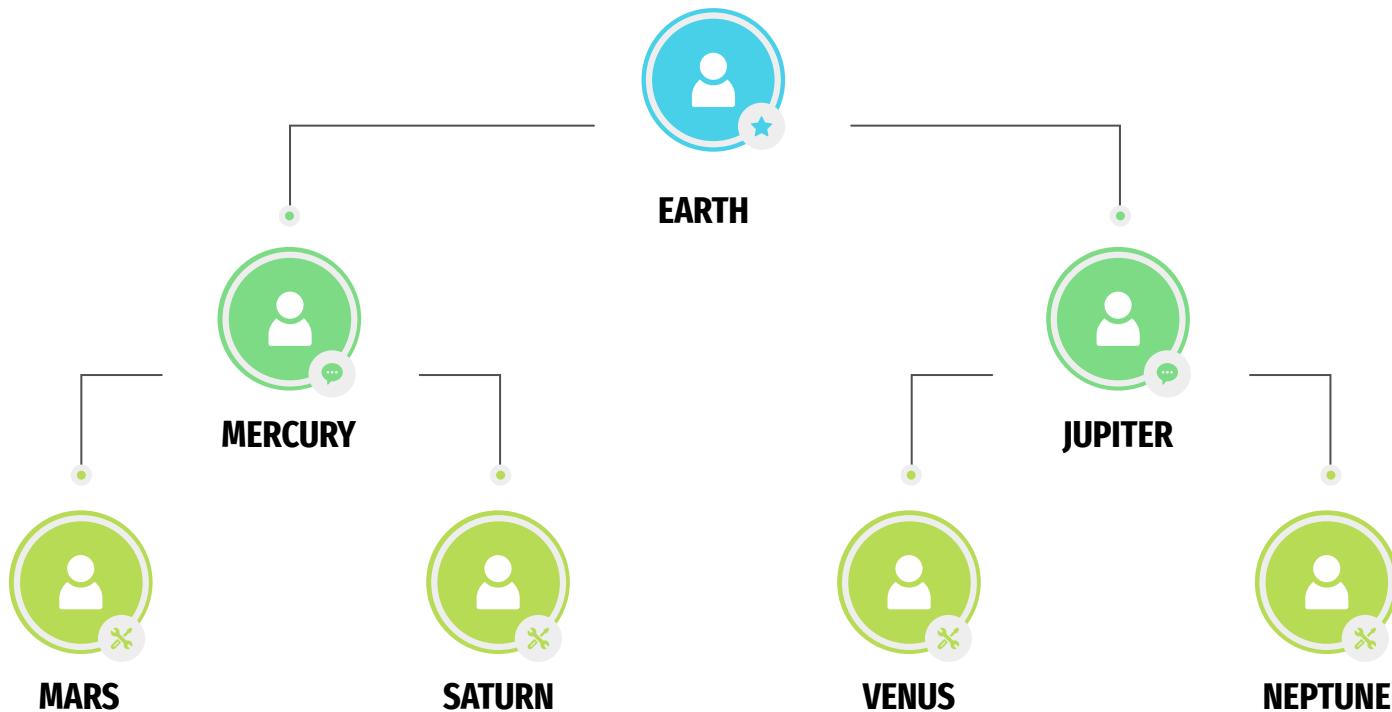
Mercury is the closest object to the Sun



JUPITER

Jupiter is a gas giant and the biggest planet

Decision Tree Diagrams



Decision Tree Diagrams

MERCURY

It's the smallest planet in the Solar System

VENUS

Venus has a very poisonous atmosphere

JUPITER

Jupiter is the biggest planet of them all

MARS

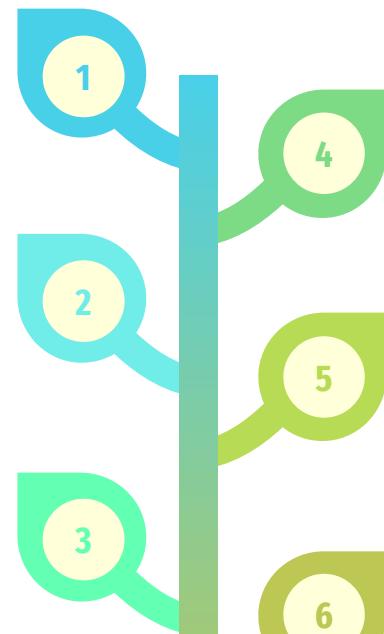
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NEPTUNE

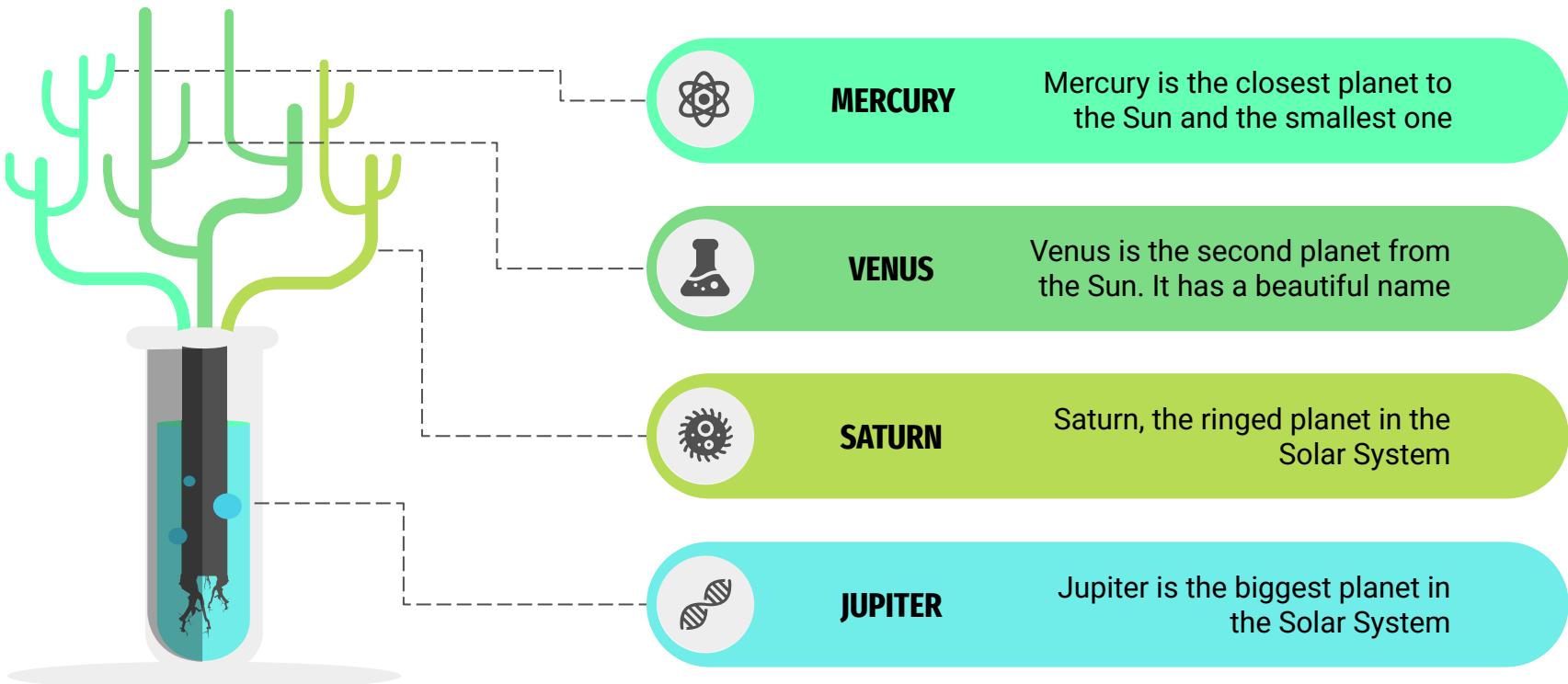
Neptune is the farthest planet from the Sun

SATURN

Saturn is composed of hydrogen and helium



Decision Tree Diagrams



Decision Tree Diagrams

MERCURY

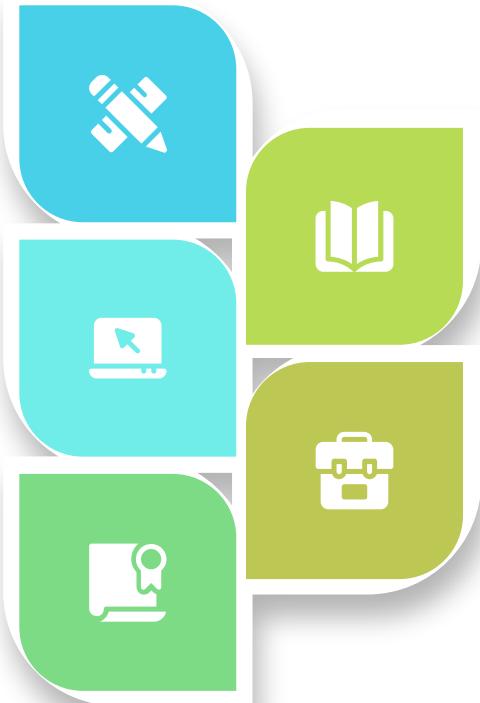
Mercury is the closest planet to the Sun and the smallest one

VENUS

Venus has a beautiful name and is the second planet from the Sun

JUPITER

Jupiter is a gas giant and the biggest planet in the Solar System



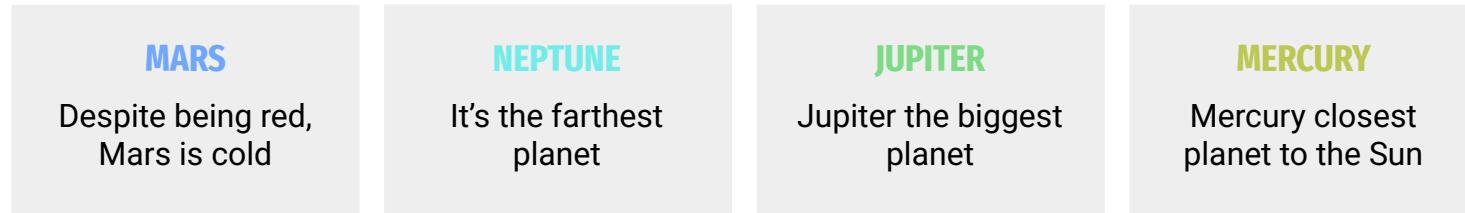
MARS

Despite being red, Mars is a cold place full of iron oxide dust

NEPTUNE

Neptune is the farthest planet from the Sun and the fourth-largest one

Decision Tree Diagrams



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SATURN

Yes, Saturn is the
ringed one



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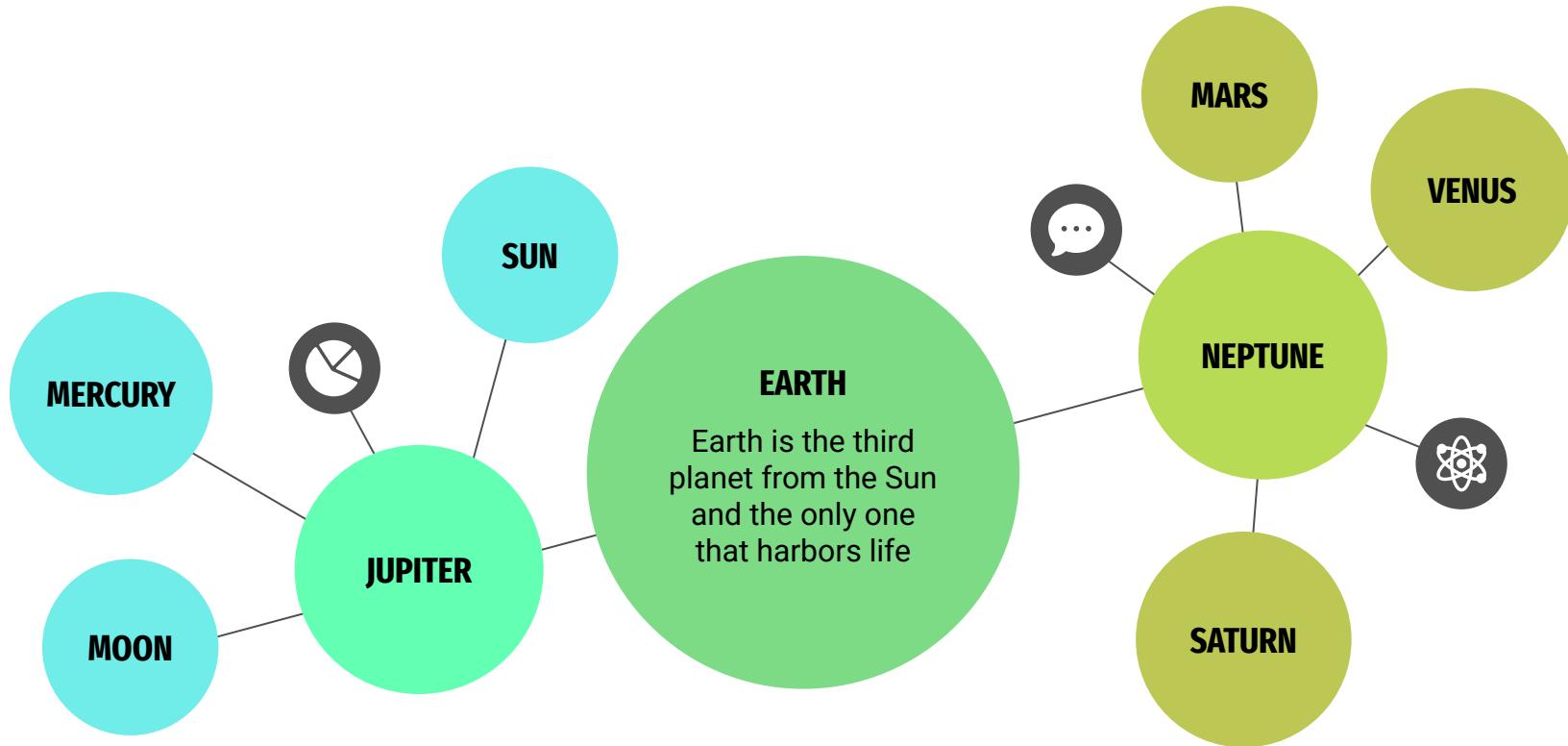
NEPTUNE

Neptune is the farthest planet from the Sun and the fourth-largest one

SATURN

Planet Saturn is a gas giant composed mostly of hydrogen and helium

Decision Tree Diagrams



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Despite being red, Mars is a cold place



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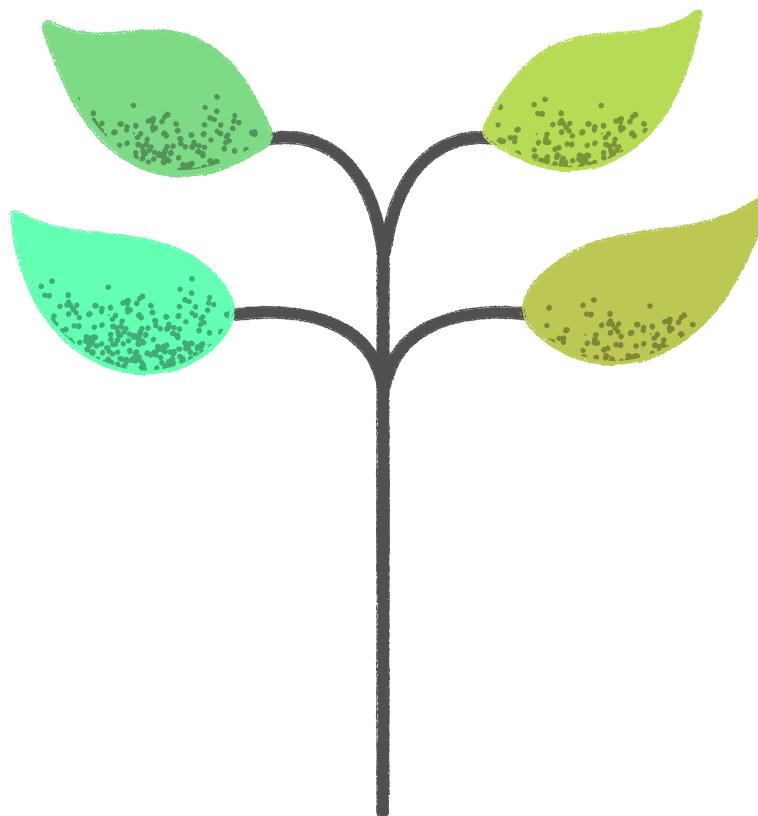
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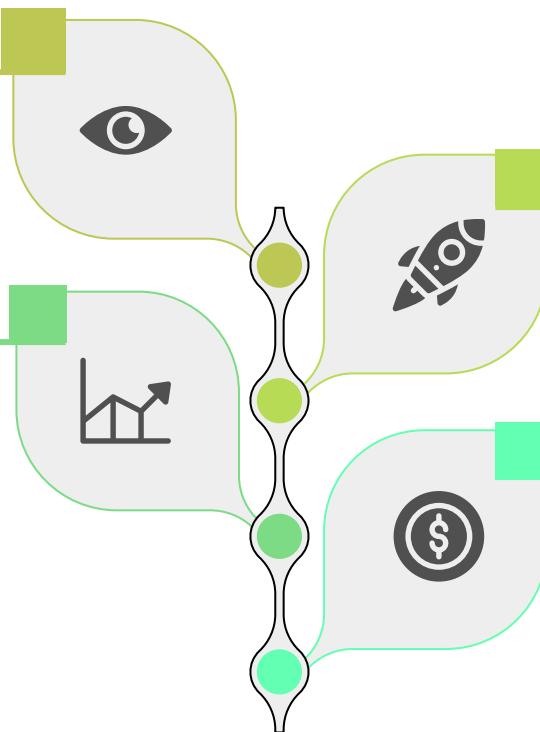
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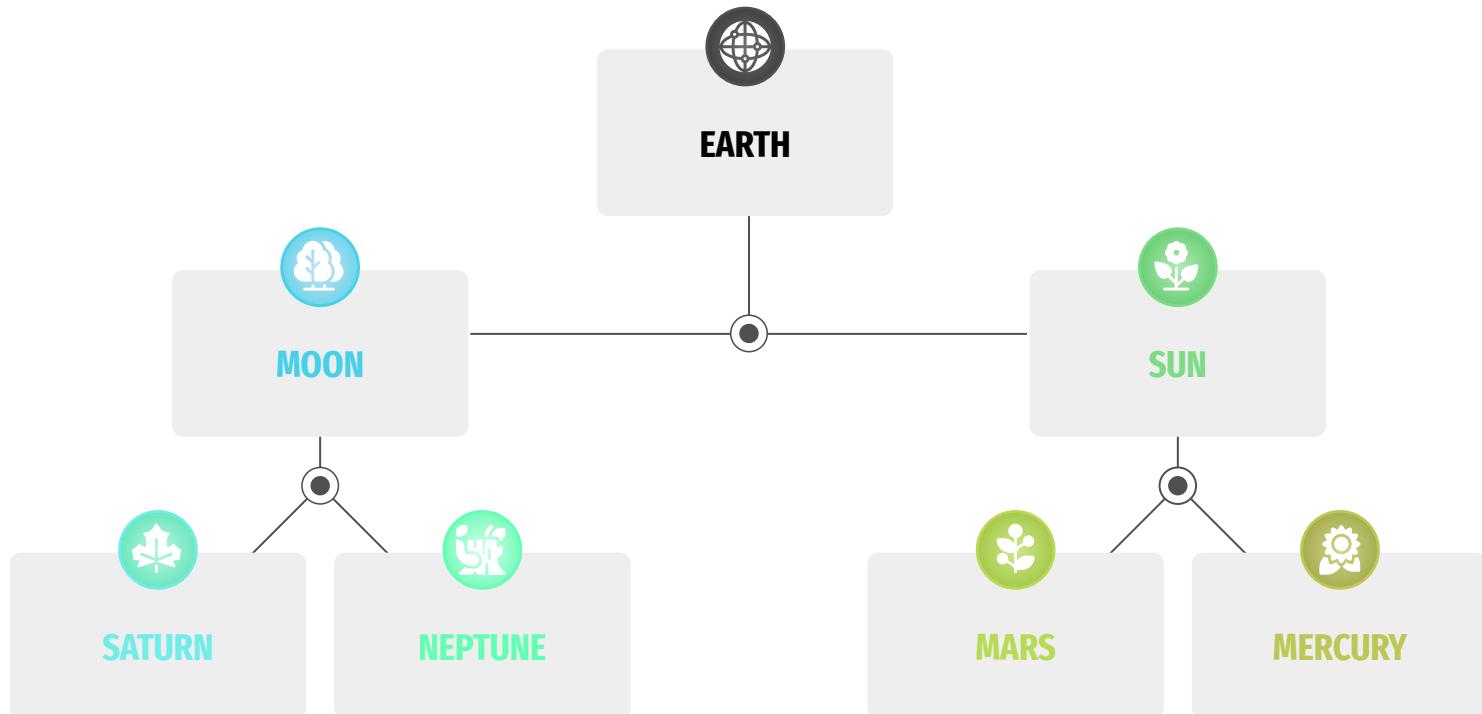
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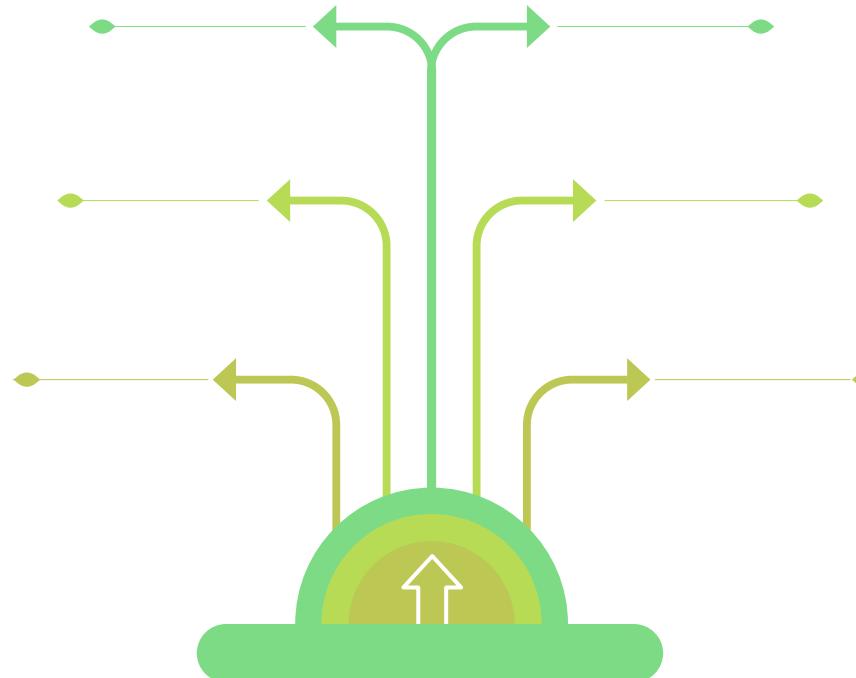
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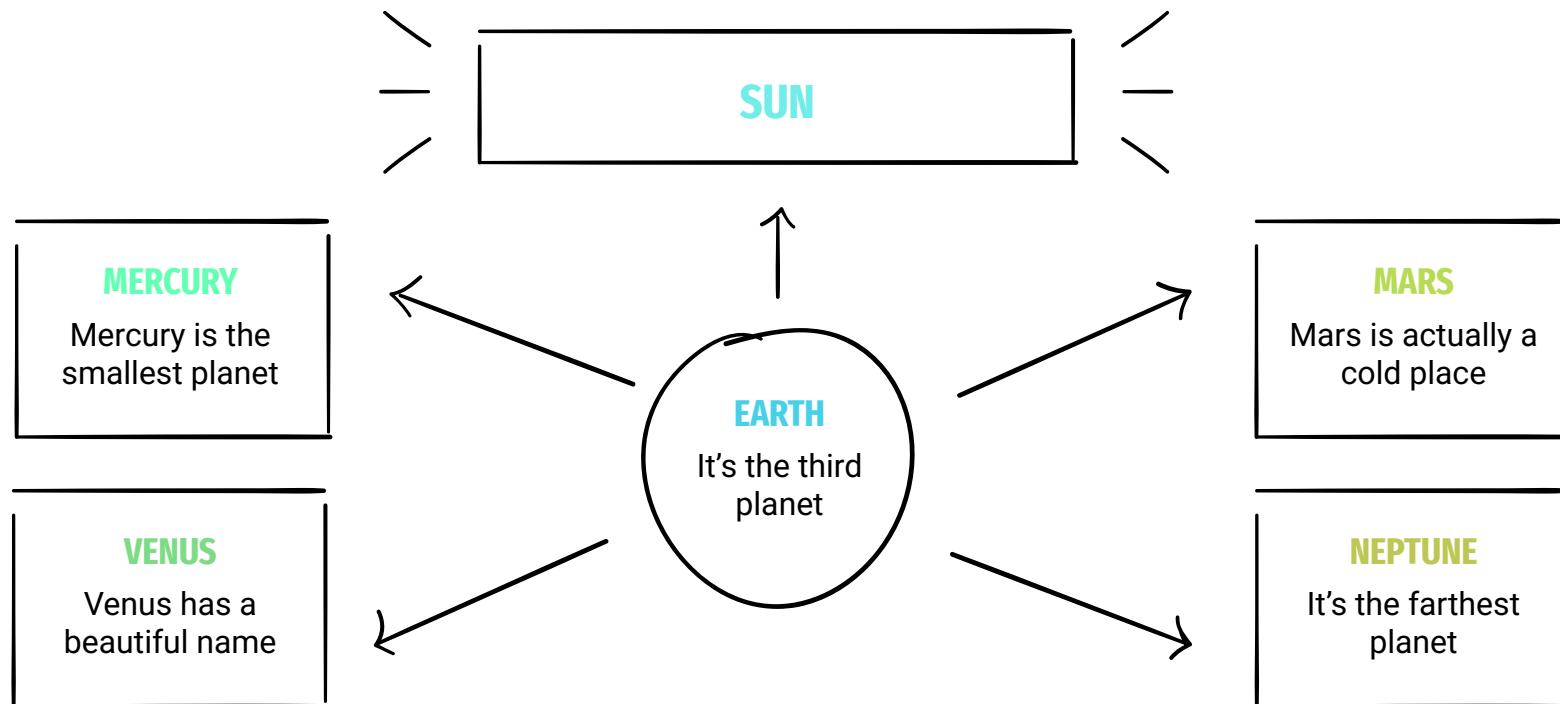
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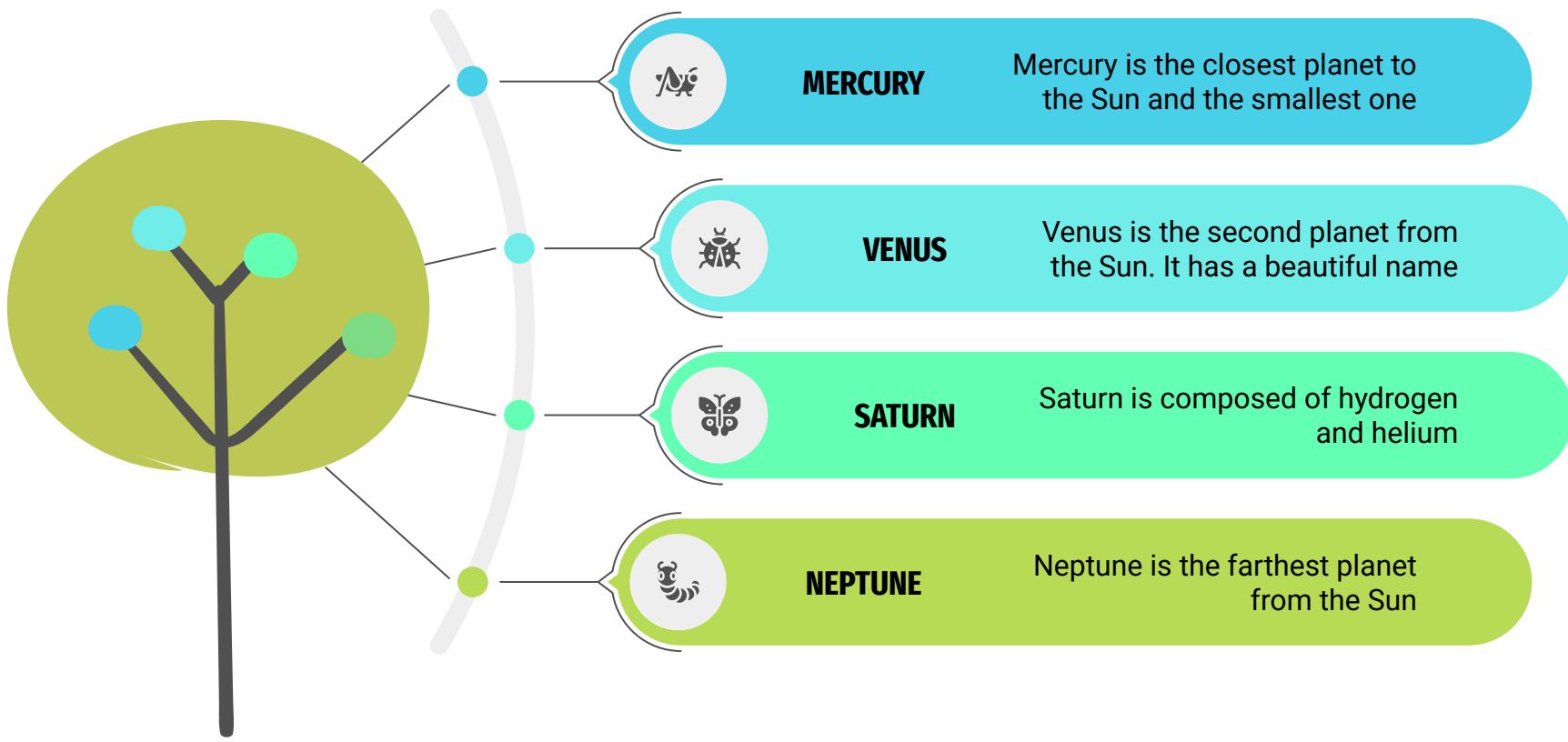
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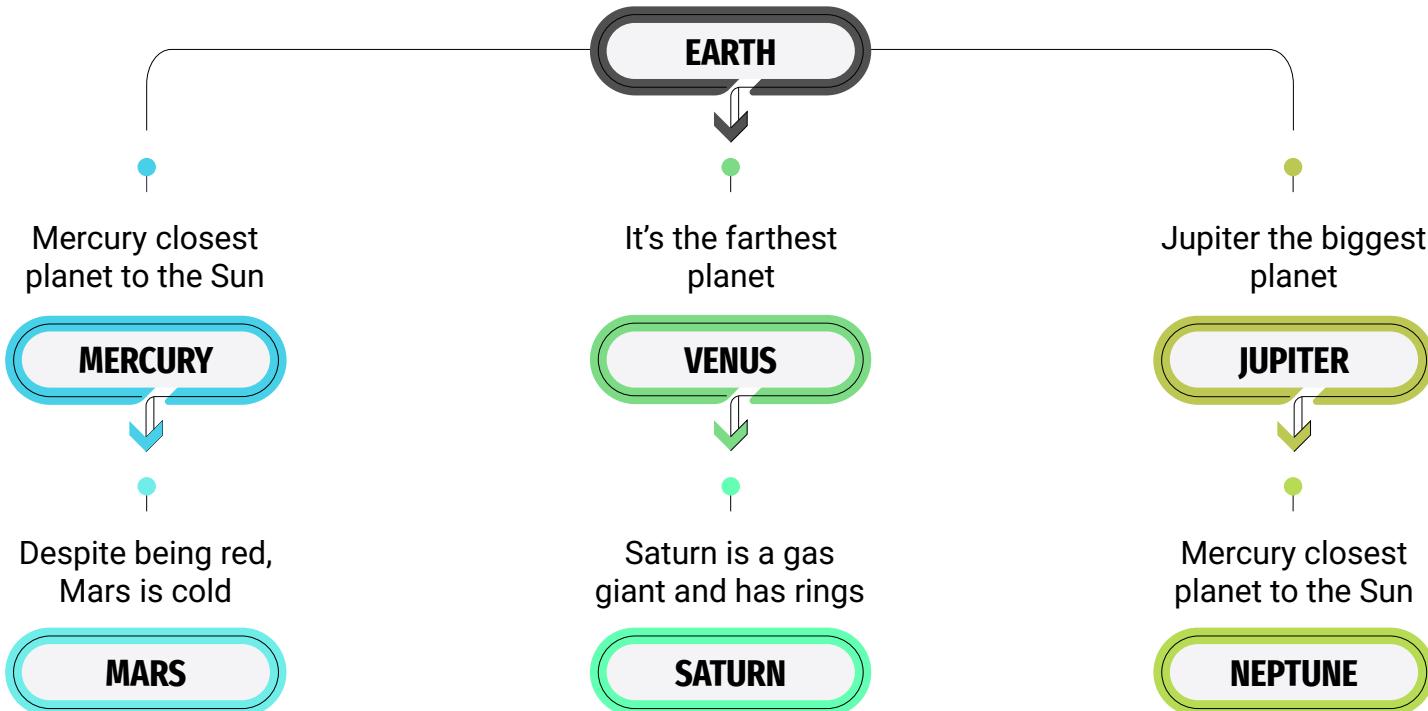
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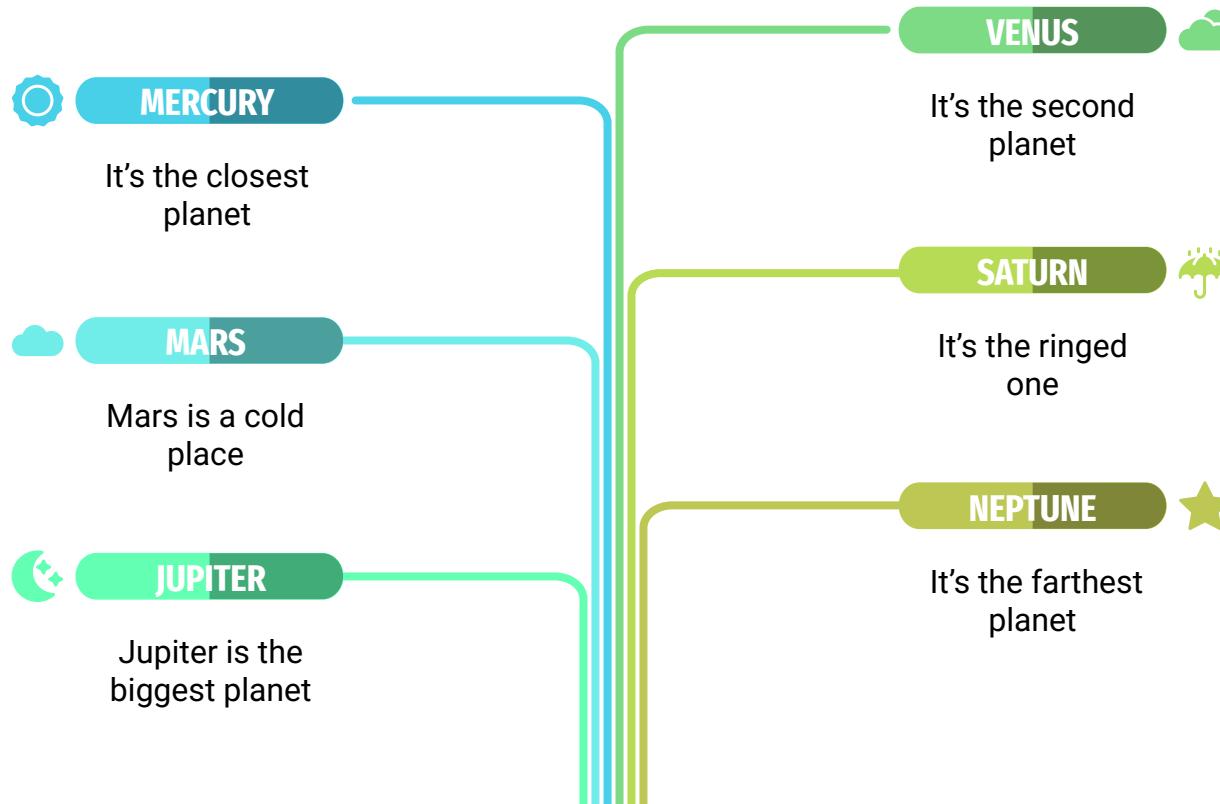
Decision Tree Diagrams



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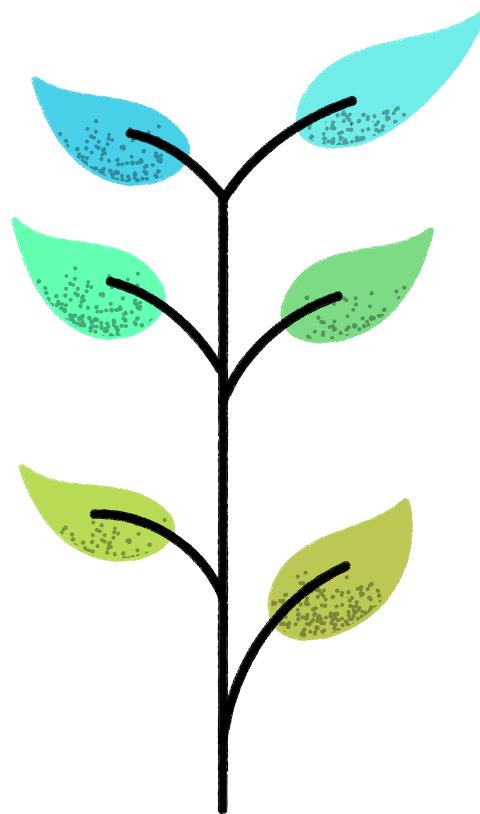
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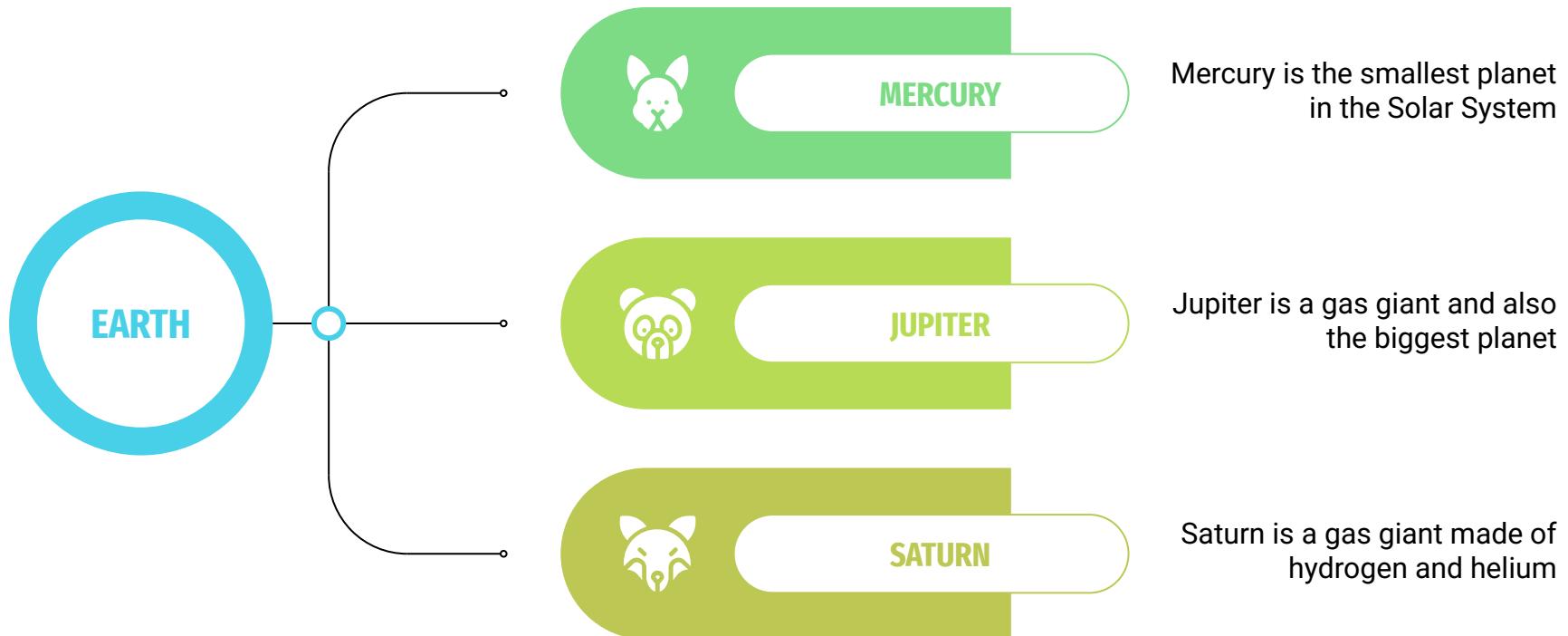
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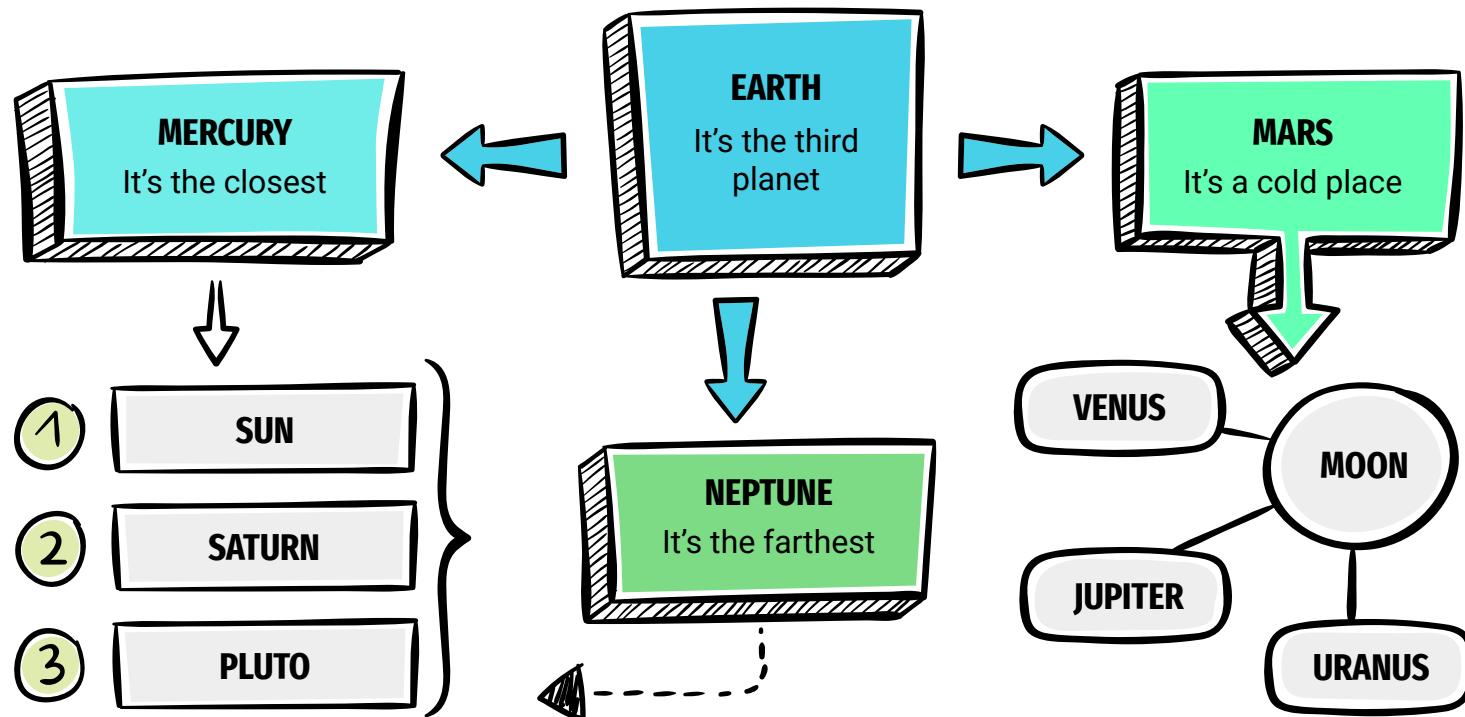
SATURN

Planet Saturn is a gas giant composed mostly of hydrogen and helium

Decision Tree Diagrams



Decision Tree Diagrams



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