

# Tree Viewer

A visualization of classification and regression trees.

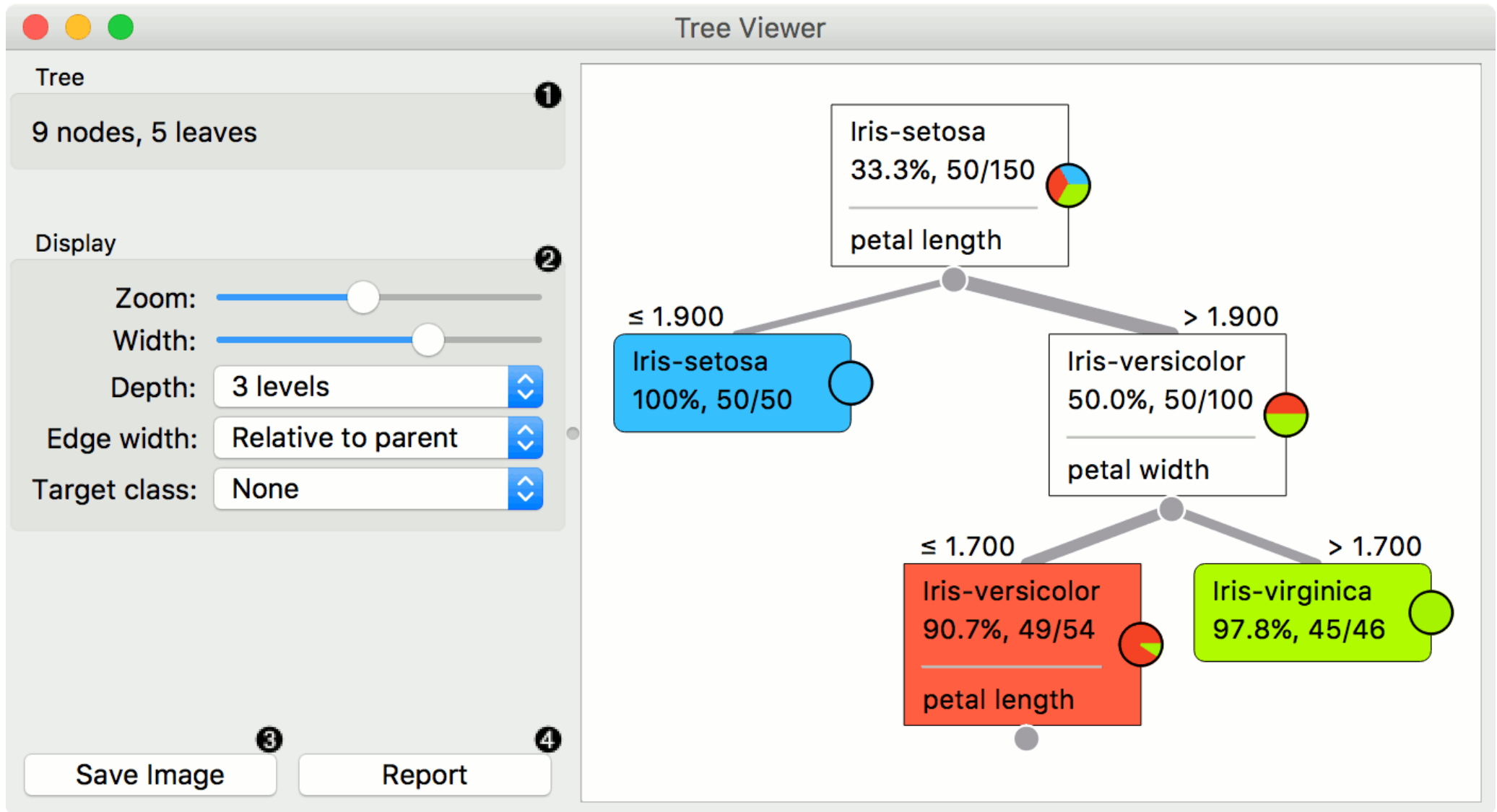
## Inputs

- Tree: decision tree

## Outputs

- Selected Data: instances selected from the tree node
- Data: data with an additional column showing whether a point is selected

This is a versatile widget with 2-D visualization of **classification and regression trees**. The user can select a node, instructing the widget to output the data associated with the node, thus enabling explorative data analysis.



1. Information on the input.

2. Display options:

- Zoom in and zoom out
- Select the tree width. The nodes display information bubbles when hovering over them.
- Select the depth of your tree.
- Select edge width. The edges between the nodes in the tree graph are drawn based on the selected edge width.
  - All the edges will be of equal width if *Fixed* is chosen.

- When *Relative to root* is selected, the width of the edge will correspond to the proportion of instances in the corresponding node with respect to all the instances in the training data. Under this selection, the edge will get thinner and thinner when traversing toward the bottom of the tree.
  - *Relative to parent* makes the edge width correspond to the proportion of instances in the nodes with respect to the instances in their parent node.
  - Define the target class, which you can change based on classes in the data.
3. Press *Save image* to save the created tree graph to your computer as a *.svg* or *.png* file.
  4. Produce a report.

## Examples

Below, is a simple classification schema, where we have read the data, constructed the decision tree and viewed it in our **Tree Viewer**. If both the viewer and **Tree** are open, any re-run of the tree induction algorithm will immediately affect the visualization. You can thus use this combination to explore how the parameters of the induction algorithm influence the structure of the resulting tree.

The screenshot displays the Orange Data Mining software interface. At the top, a workflow canvas shows three widgets connected in a sequence: 'File', 'Tree', and 'Tree Viewer'. Below the canvas, the 'Tree' widget's settings are visible, including a name field set to 'Tree', parameters for inducing a binary tree, and classification settings. To the right, the 'Tree Viewer' widget displays a decision tree for the Iris dataset. The tree structure is as follows:

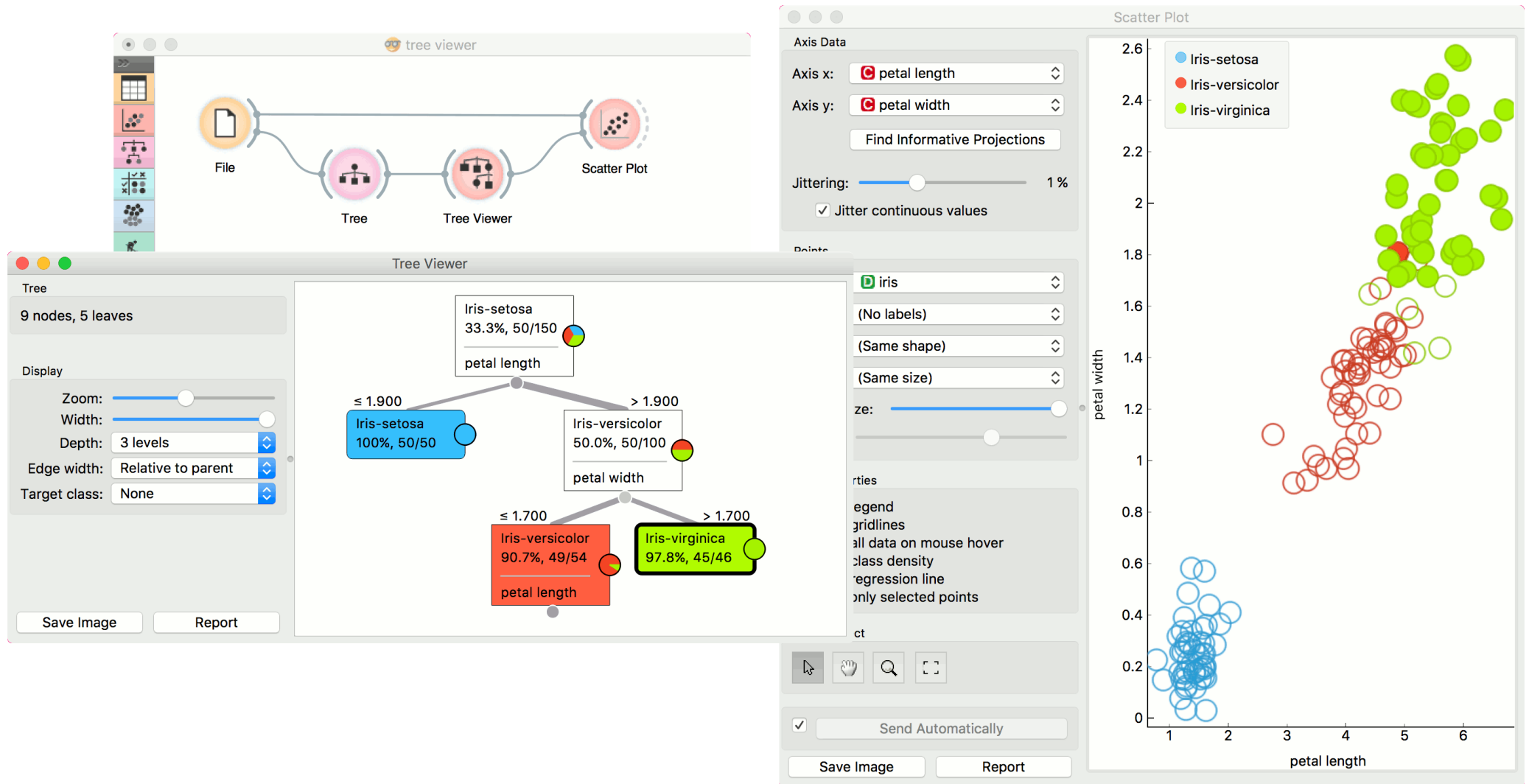
```

graph TD
    Root["Iris-setosa  
33.3%, 50/150  
petal length"]
    Root -- "≤ 1.900" --> L["Iris-setosa  
100%, 50/50"]
    Root -- "> 1.900" --> R["Iris-versicolor  
50.0%, 50/100  
petal width"]
    R -- "≤ 1.700" --> RL["Iris-versicolor  
90.7%, 49/54  
petal length"]
    R -- "> 1.700" --> RR["Iris-virginica  
97.8%, 45/46"]
  
```

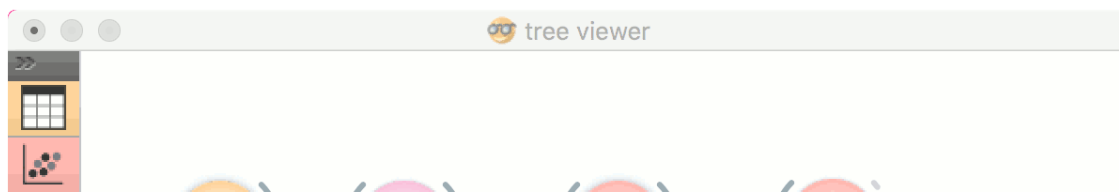
The 'Tree Viewer' widget also includes a 'Display' section with sliders for Zoom, Width, and Depth, and a 'Target class' dropdown set to 'None'. Buttons for 'Save Image' and 'Report' are located at the bottom of the widget.

Clicking on any node will output the related data instances. This is explored in the schema below that shows the subset in the data table and in the **Scatter Plot**. Make sure that the tree data is passed as a data subset; this can be done by connecting the **Scatter Plot** to the **File** widget first, and connecting it to the **Tree Viewer** widget next. Selected data will be displayed as bold dots.

**Tree Viewer** can also export labeled data. Connect **Data Table** to **Tree Viewer** and set the link between widgets to *Data* instead of *Selected Data*. This will send the entire data to **Data Table** with an additional meta column labeling selected data instances (Yes for selected and No for the remaining).



Finally, **Tree Viewer** can be used also for visualizing regression trees. Connect **Random Forest** to **File** widget using *housing.tab* dataset. Then connect **Pythagorean Forest** to **Random Forest**. In **Pythagorean Forest** select a regression tree you wish to further analyze and pass it to the **Tree Viewer**. The widget will display the constructed tree. For visualizing larger trees, especially for regression, **Pythagorean Tree** could be a better option.



File Random Forest Pythagorean Forest Tree Viewer

**Pythagorean Forest**

Forest

Trees: 10

Display

Depth: 21

Target class: Class mean

Size: Normal

Zoom: [Slider]

**Tree Viewer**

Tree

249 nodes, 125 leaves

Display

Zoom: [Slider]

Width: [Slider]

Depth: 3 levels

Edge width: Relative to parent

Color by: Default

Save Image Report

Tree structure:

```

graph TD
    Root["22.8 ± 84.4  
328 instances  
RM"]
    Root -- "≤ 6.941" --> LSTAT["20.1 ± 40.3  
280 instances  
LSTAT"]
    Root -- "> 6.941" --> RM_R["38.1 ± 79.7  
48 instances  
RM"]
    LSTAT -- "≤ 6.941" --> DIS["23.4 ± 26.1  
165 instances  
DIS"]
    LSTAT -- "≤ 6.941" --> NOX["15.4 ± 19.6  
115 instances  
NOX"]
    RM_R -- "≤ 6.941, 7.445]" --> CRIM["31.1 ± 41.3  
26 instances  
CRIM"]
    RM_R -- "> 7.445" --> TAX["44.9 ± 36.6  
22 instances  
TAX"]
  
```