

How to visualize a single Decision Tree from the Random Forest in Scikit-Learn (Python)?

June 29, 2020 by Piotr Płoński **Random forest**

The Random Forest is an ensemble of Decision Trees. [A single Decision Tree can be easily visualized in several different ways](#). In this post I will show you, how to visualize a Decision Tree from the Random Forest.

First let's train Random Forest model on Boston data set (it is house price regression task available in `scikit-learn`).

```
# Load packages
import pandas as pd
from sklearn.datasets import load_boston
from sklearn.ensemble import RandomForestRegressor
from sklearn import tree
from dtreeviz.trees import dtreeviz # will be used for tree visual
from matplotlib import pyplot as plt
plt.rcParams.update({'figure.figsize': (12.0, 8.0)})
plt.rcParams.update({'font.size': 14})
```

Load the data and train the Random Forest.

```
boston = load_boston()
X = pd.DataFrame(boston.data, columns=boston.feature_names)
y = boston.target
```

This site uses cookies. If you continue browsing our website, you accept these cookies.

[More info](#)

Accept

```
rf = RandomForestRegressor(n_estimators=100)
rf.fit(X, y)
```

```
RandomForestRegressor(bootstrap=True, ccp_alpha=0.0, criterion='ms
max_depth=None, max_features='auto', max_lea
max_samples=None, min_impurity_decrease=0.0,
min_impurity_split=None, min_samples_leaf=1,
min_samples_split=2, min_weight_fraction_lea
n_estimators=100, n_jobs=None, oob_score=Fal
random_state=None, verbose=0, warm_start=Fal
```

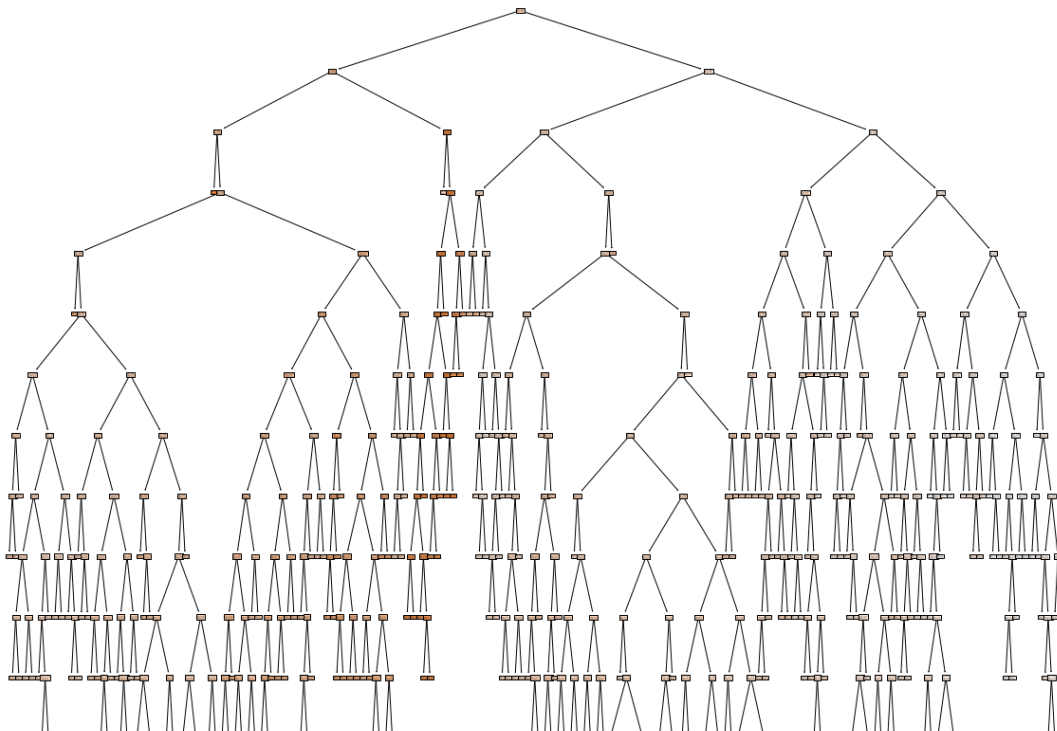
Decision Trees are stored in a `list` in the `estimators_` attribute in the `rf` model. We can check the length of the list, which should be equal to `n_estimators` value.

```
len(rf.estimators_)

>>> 100
```

We can plot a first Decision Tree from the Random Forest (with index `0` in the list):

```
plt.figure(figsize=(20,20))
_ = tree.plot_tree(rf.estimators_[0], feature_names=X.columns, fil
```



This site uses cookies. If you continue browsing our website, you accept these cookies.

[More info](#)



Do you understand anything? The tree is too large to visualize it in one figure and make it readable.

Let's check the depth of the first tree from the Random Forest:

```
rf.estimators_[0].tree_.max_depth

>>> 16
```

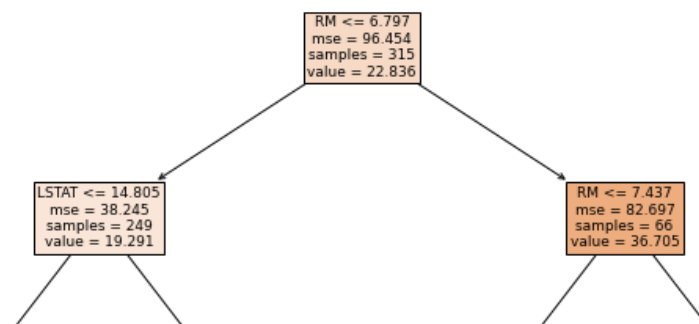
Our first tree has `max_depth=16`. Other trees have similar depth. To make visualization readable it will be good to limit the depth of the tree. In MLJAR's open-source AutoML package [mljar-supervised](#) the Decision Tree's depth is set to be in range from 1 to 4. Let's train the Random Forest again with `max_depth=3`.

```
rf = RandomForestRegressor(n_estimators=100, max_depth=3)
rf.fit(X, y)
```

```
RandomForestRegressor(bootstrap=True, ccp_alpha=0.0, criterion='mse',
                      max_depth=3, max_features='auto', max_leaf_nodes=None,
                      max_samples=None, min_impurity_decrease=0.0,
                      min_impurity_split=None, min_samples_leaf=1,
                      min_samples_split=2, min_weight_fraction_leaf=0.0,
                      n_estimators=100, n_jobs=None, oob_score=False,
                      random_state=None, verbose=0, warm_start=False)
```

The plot of first Decision Tree:

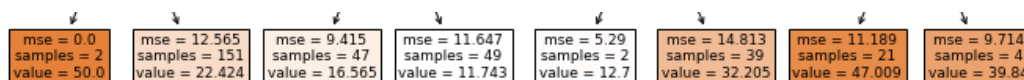
```
_ = tree.plot_tree(rf.estimators_[0], feature_names=X.columns, fig=fig)
```



This site uses cookies. If you continue browsing our website, you accept these cookies.

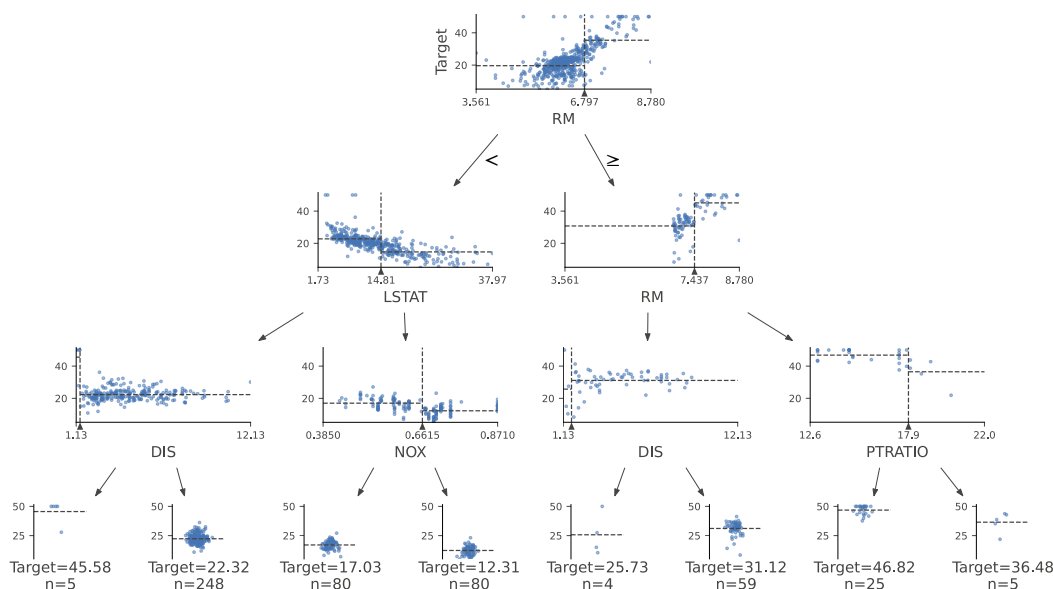
[More info](#)

Accept



We can use `dtreeviz` package to visualize the first Decision Tree:

```
viz = dtreeviz(rf.estimators_[0], X, y, feature_names=X.columns, t
viz
```



Summary

I show you how to visualize the single Decision Tree from the Random Forest. Trees can be accessed by integer index from `estimators_` list. Sometimes when the tree is too deep, it is worth to limit the depth of the tree with `max_depth` hyper-parameter. What is interesting, limiting the depth of the trees in the Random Forest will make the final model much smaller in terms of used RAM memory and disk space needed to save the model. It will also change the performance of the default Random Forest (with full trees), it will help or not, depending on the data set.

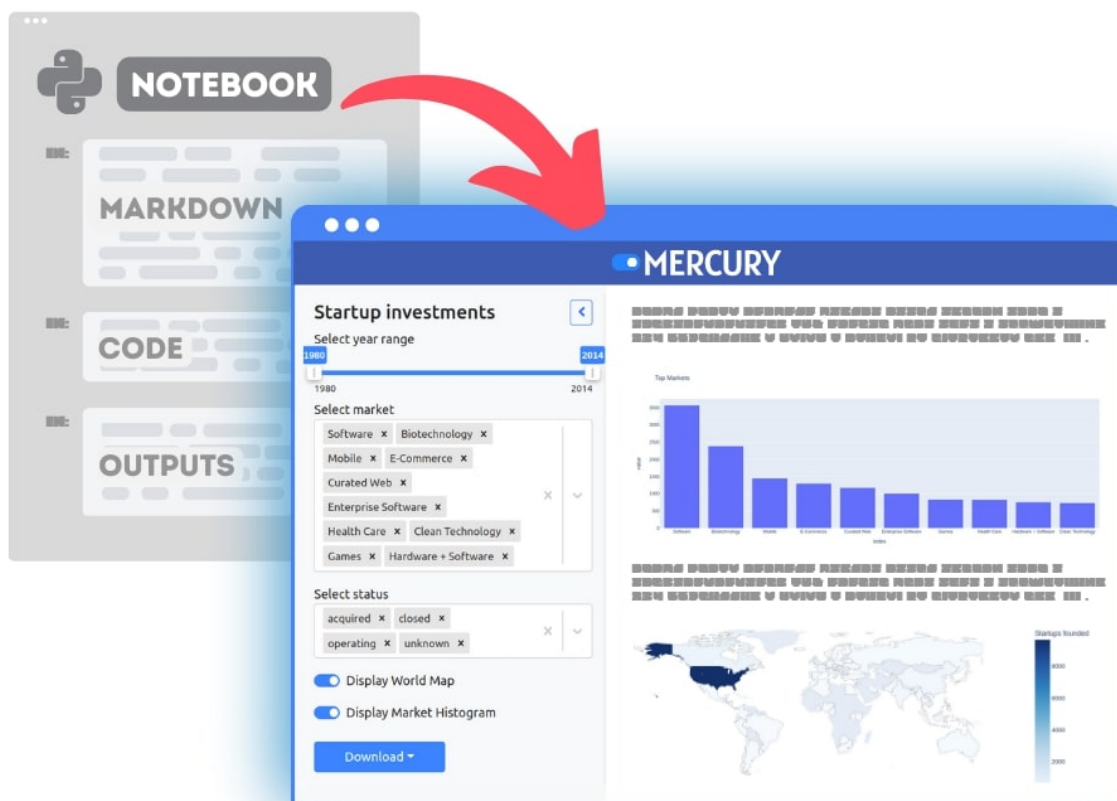
« Random Forest Feature Importance Computed in 3 Ways with Python

How many trees in the Random Forest? »

This site uses cookies. If you continue browsing our website, you accept these cookies.

[More info](#)

Accept



Convert Python Notebooks to Web Apps

We are working on open-source framework [Mercury](#) for converting Jupyter Notebooks to interactive Web Applications.

[Read more](#)



This site uses cookies. If you continue browsing our website, you accept these cookies.

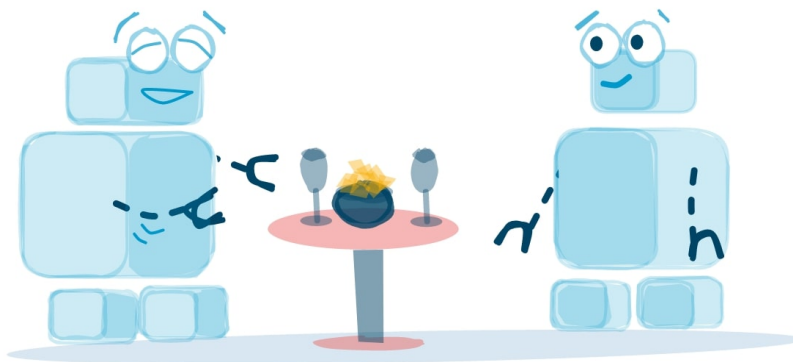
[More info](#)

Accept



Articles you might find interesting

1. [8 surprising ways how to use Jupyter Notebook](#)
 2. [Create a dashboard in Python with Jupyter Notebook](#)
 3. [Build Computer Vision Web App with Python](#)
 4. [Develop NLP Web App from Python Notebook](#)
 5. [Build dashboard in Python with updates and email notifications](#)
 6. [Share Jupyter Notebook with non-technical users](#)
-



Join our newsletter

Subscribe to our newsletter to receive product updates

Subscribe

This site uses cookies. If you continue browsing our website, you accept these cookies.

[More info](#)

Accept



**Outstanding Data
Science Tools**

[Blog](#)[About](#)[Brand Assets](#)[GitHub](#)[Twitter](#)[Mercury](#)[AutoML](#)[Pricing](#)[Compare Algorithms](#)[AutoML Comparison](#)[Decision Tree vs Random Forest](#)[What is AutoML?](#)[Random Forest vs Xgboost](#)[Golden Features](#)[Xgboost vs LightGBM](#)[K-Means Features](#)[CatBoost vs Xgboost](#)[Feature Selection](#)

© 2023 MLJAR, Sp. z o.o. • [Terms of service](#) • [Privacy policy](#) • [EULA](#) • [Contact](#) •

This site uses cookies. If you continue browsing our website, you accept these cookies.

[More info](#)[Accept](#)