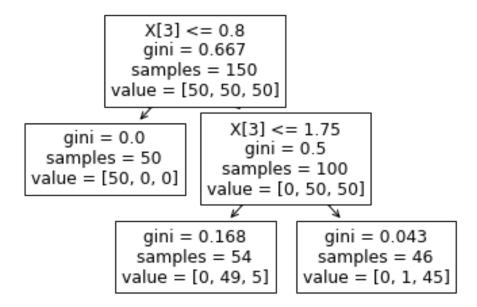
```
!pip install dtreeviz
import graphviz.backend as be
from sklearn.datasets import *
from dtreeviz.trees import *
from IPython.display import Image, display svg, SVG
clas = tree.DecisionTreeClassifier()
iris = load iris()
X train = iris.data
y train = iris.target
clas.fit(X_train, y_train)
DecisionTreeClassifier(ccp alpha=0.0, class weight=None,
criterion='gini',
                                                                                                                     max depth=None, max features=None,
max leaf nodes=None,
                                                                                                                     min impurity decrease=0.0,
min impurity split=None,
                                                                                                                     min samples leaf=1, min samples split=2,
                                                                                                                     min weight fraction leaf=0.0,
presort='deprecated',
                                                                                                                     random state=None, splitter='best')
from sklearn.tree import plot tree
plot tree(clas)
 [Text(133.9200000000000, 181.2, 'X[3] \le 0.8 \ngini = 0.667 \nsamples = 0.667 \nsam
150 \setminus \text{nvalue} = [50, 50, 50]'),
  Text(66.9600000000001, 108.72, 'gini = 0.0\nsamples = 50\nvalue =
 [50, 0, 0]'),
  Text(200.88000000000002, 108.72, 'X[3] \le 1.75 \cdot gini = 0.5 \cdot samples = 0.5 \cdot
100 \setminus \text{nvalue} = [0, 50, 50]'),
   Text(133.9200000000000, 36.239999999999, 'gini = 0.168\nsamples =
54\nvalue = [0, 49, 5]'),
   Text(267.8400000000003, 36.239999999999, 'gini = 0.043\nsamples =
46 \cdot nvalue = [0, 1, 45]')
```



1. Classification

2. Regression

```
)
Viz
```

3. Horizontal Decision Tree

4. Show prediction path

```
clas = tree.DecisionTreeClassifier()
iris = load iris()
X train = iris.data
y train = iris.target
clas.fit(X train, y train)
X = iris.data[np.random.randint(0, len(iris.data)),:]
viz = dtreeviz(clas,
               X_train,
               y_train,
               feature names=iris.feature names,
               class_names=["setosa", "versicolor", "virginica"],
               X=X)
viz
/usr/local/lib/python3.7/dist-packages/numpy/core/_asarray.py:83:
VisibleDeprecationWarning: Creating an ndarray from ragged nested
sequences (which is a list-or-tuple of lists-or-tuples-or ndarrays
with different lengths or shapes) is deprecated. If you meant to do
this, you must specify 'dtype=object' when creating the ndarray
  return array(a, dtype, copy=False, order=order)
```

```
X
array([5.5, 2.3, 4. , 1.3])
```

5. Show node number

6. Without Any graphs

7. Show just prediction path, nothing else

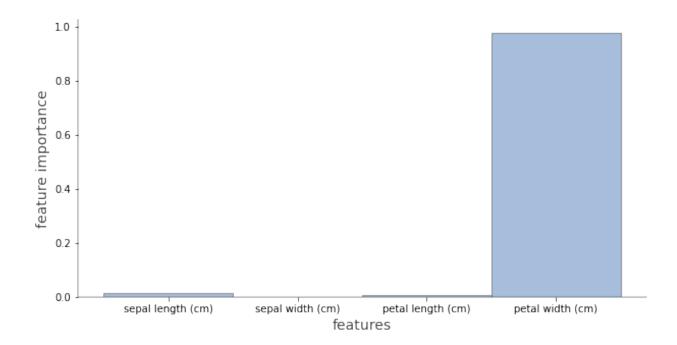
```
clas = tree.DecisionTreeClassifier()
iris = load iris()
X train = iris.data
y_train = iris.target
clas.fit(X train, y train)
X = iris.data[np.random.randint(0, len(iris.data)),:]
viz = dtreeviz(clas,
               X train,
               y_train,
               target name='price',
               feature names=iris.feature names,
               class names=["setosa", "versicolor", "virginica"],
               show_just_path=True)
viz
/usr/local/lib/python3.7/dist-packages/numpy/core/ asarray.py:83:
VisibleDeprecationWarning: Creating an ndarray from ragged nested
sequences (which is a list-or-tuple of lists-or-tuples-or ndarrays
with different lengths or shapes) is deprecated. If you meant to do
this, you must specify 'dtype=object' when creating the ndarray
  return array(a, dtype, copy=False, order=order)
```

8. Prediction Path in Plain english

```
print(explain_prediction_path(clas, X,
feature_names=iris.feature_names, explanation_type="plain_english"))
sepal length (cm) < 5.95
petal length (cm) < 4.85
1.75 <= petal width (cm)</pre>
```

9. Feature Importance

```
print(explain_prediction_path(clas, X,
feature_names=iris.feature_names, explanation_type="sklearn_default"))
AxesSubplot(0.125,0.125;0.775x0.755)
```



10. Univariate Regression

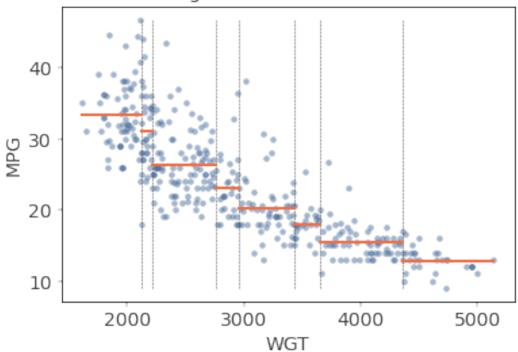
```
import pandas as pd
import matplotlib.pyplot as plt
from sklearn.tree import DecisionTreeRegressor
from dtreeviz.trees import *

df_cars = pd.read_csv("cars.csv")
X, y = df_cars[['WGT']], df_cars['MPG']

dt = DecisionTreeRegressor(max_depth=3, criterion="mae")
dt.fit(X, y)

fig = plt.figure()
ax = fig.gca()
rtreeviz_univar(dt, X, y, 'WGT', 'MPG', ax=ax)
plt.show()
```

Regression tree depth 3, samples per leaf 1, Training R^2 =0.7257759664142766



11. 3-D Regression

```
df_cars = pd.read_csv("cars.csv")
df_cars
     MPG CYL
                 ENG
                       WGT
            8 307.0 3504
0
     18.0
            8 350.0 3693
     15.0
1
2
     18.0
            8 318.0 3436
3
     16.0
            8 304.0 3433
4
     17.0
            8 302.0 3449
            4 140.0
387
    27.0
                     2790
            4 97.0 2130
388
    44.0
    32.0
            4 135.0 2295
389
    28.0
            4 120.0 2625
390
391
    31.0
               119.0 2720
[392 rows x 4 columns]
from mpl toolkits.mplot3d import Axes3D
from sklearn.tree import DecisionTreeRegressor
from dtreeviz.trees import *
```

```
X = df_cars[['WGT','ENG']]
y = df_cars['MPG']
dt = DecisionTreeRegressor(max depth=3, criterion="mae")
dt.fit(X, y)
figsize = (6,5)
fig = plt.figure(figsize=figsize)
ax = fig.add subplot(111, projection='3d')
t = rtreeviz bivar 3D(dt,
                      feature_names=['Vehicle Weight', 'Horse Power'],
                      target name='MPG',
                      fontsize=14,
                      elev=20,
                      azim=25,
                      dist=8.2,
                      show={'splits','title'},
                      ax=ax)
plt.show()
```

