

```
In [1]: %load_ext autoreload
        %autoreload 2

        %matplotlib inline
```

```
In [2]: import sys
        import os
        # add library module to PYTHONPATH
        sys.path.append(f"{os.getcwd()}/../")
```

```
In [3]: import sklearn
        from sklearn.tree import DecisionTreeClassifier, DecisionTreeRegressor
        from dtreeviz.models.sklearn_decision_trees import ShadowSKDTree

        import graphviz
        import pandas as pd
        from dtreeviz.models.shadow_decision_tree import ShadowDecTree
        from dtreeviz.models.sklearn_decision_trees import ShadowSKDTree
        from dtreeviz import trees
        from sklearn import tree
```

## Classifier

```
In [4]: random_state = 1234
        dataset = pd.read_csv("../data/titanic/titanic.csv")
        # Fill missing values for Age
        dataset.fillna({"Age":dataset.Age.mean()}, inplace=True)
        # Encode categorical variables
        dataset["Sex_label"] = dataset.Sex.astype("category").cat.codes
        dataset["Cabin_label"] = dataset.Cabin.astype("category").cat.codes
        dataset["Embarked_label"] = dataset.Embarked.astype("category").cat.codes
        features = ["Pclass", "Age", "Fare", "Sex_label", "Cabin_label", "Embarked_label"]
        target = "Survived"
        tree_classifier = DecisionTreeClassifier(max_depth=4, random_state=random_state)
        tree_classifier.fit(dataset[features], dataset[target])
```

```
Out[4]: DecisionTreeClassifier(max_depth=4, random_state=1234)
```

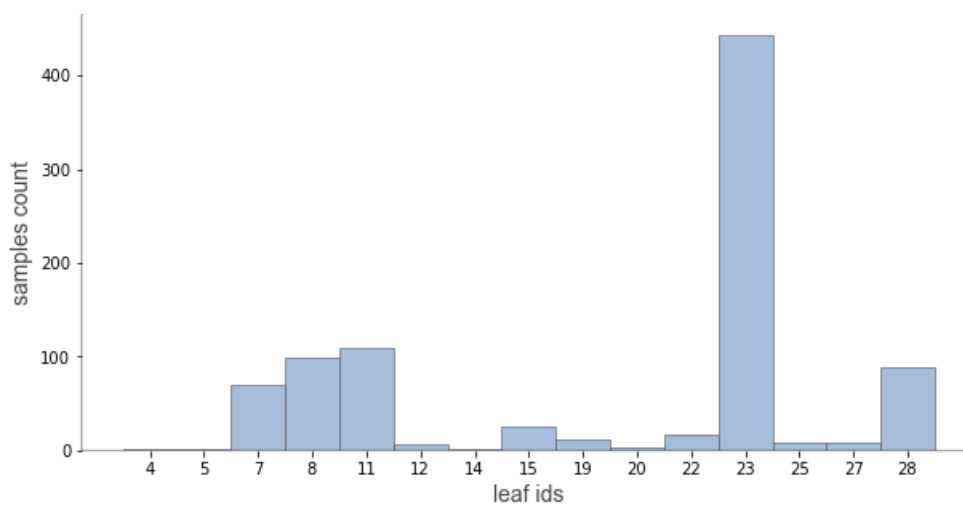
## Initialize shadow tree

```
In [5]: sk_dtree = ShadowSKDTree(tree_classifier, dataset[features], dataset[target], features, target, [0, 1])
```

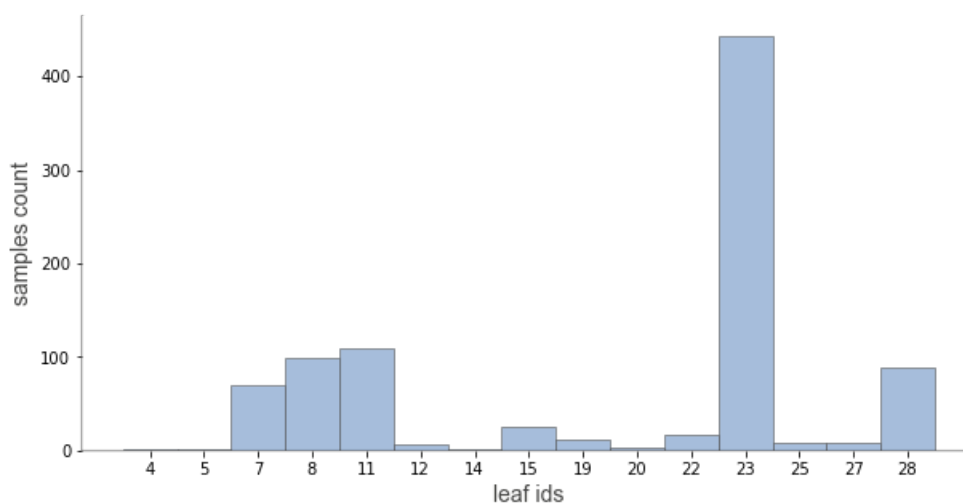
## Visualizations

### viz\_leaf\_samples

```
In [6]: trees.viz_leaf_samples(tree_classifier, dataset[features], features)
```

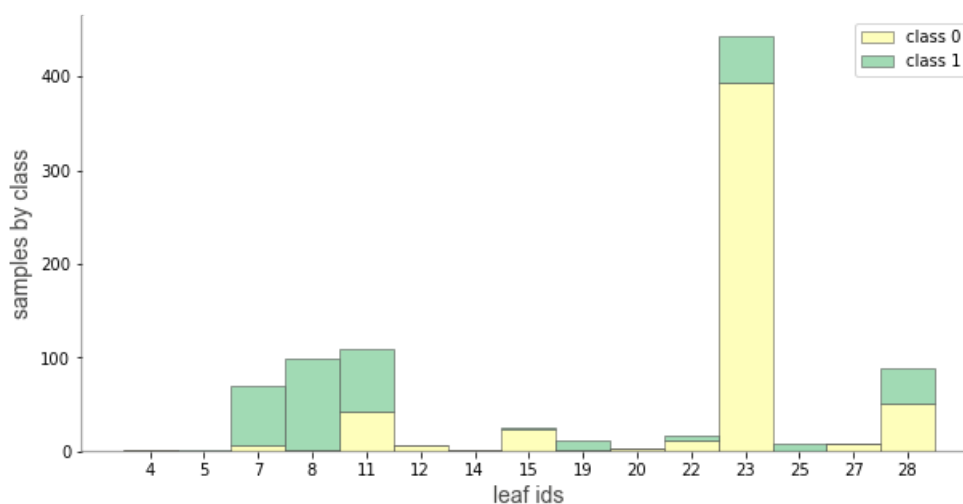


```
In [7]: trees.viz_leaf_samples(sk_dtree)
```

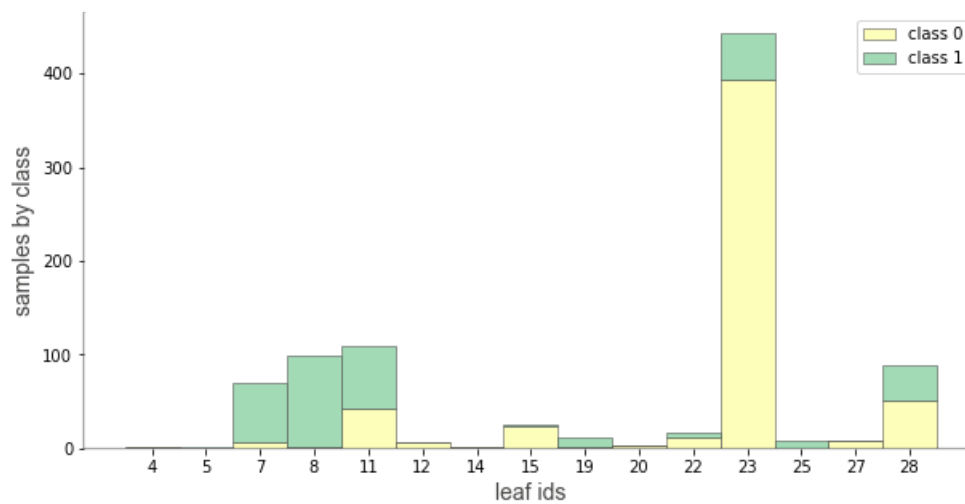


## ctreeviz\_leaf\_samples

```
In [8]: trees.ctreeviz_leaf_samples(tree_classifier, dataset[features], dataset[target], features)
```



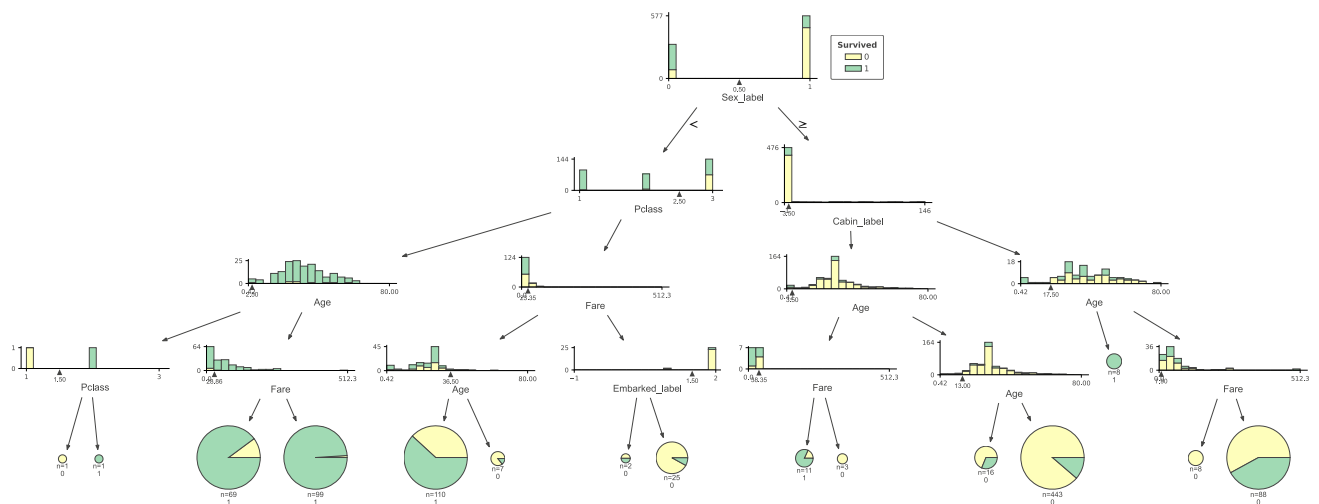
```
In [9]: trees.ctreeviz_leaf_samples(sk_dtree)
```



## dtreeviz

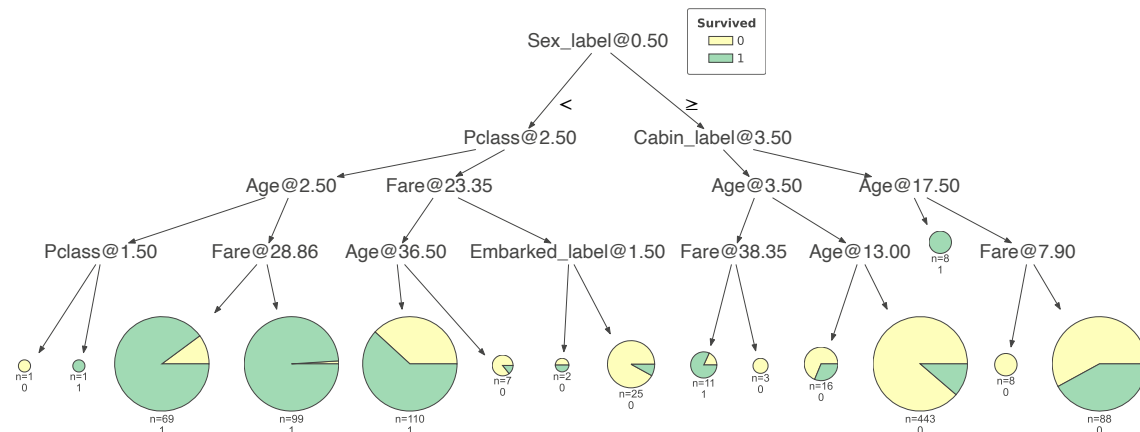
```
In [10]: trees.dtreeviz(tree_classifier, dataset[features], dataset[target], features, target, class_names=[0, 1])
```

Out[10]:



```
In [11]: trees.dtreeviz(sk_dtree, fancy=False)
```

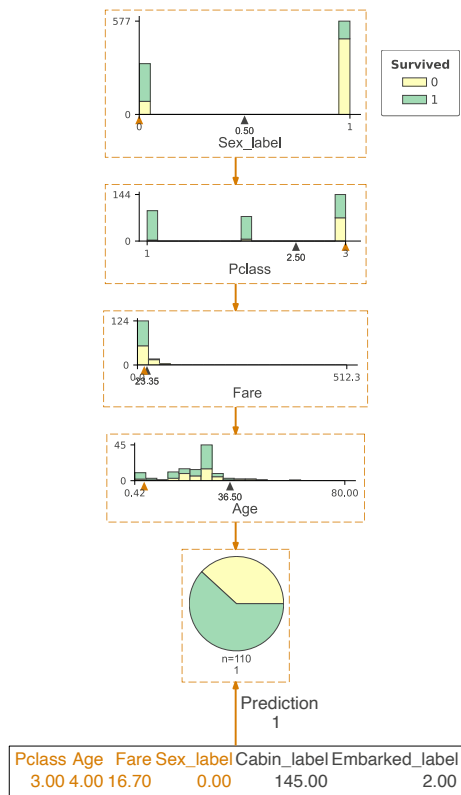
Out[11]:



## show just path

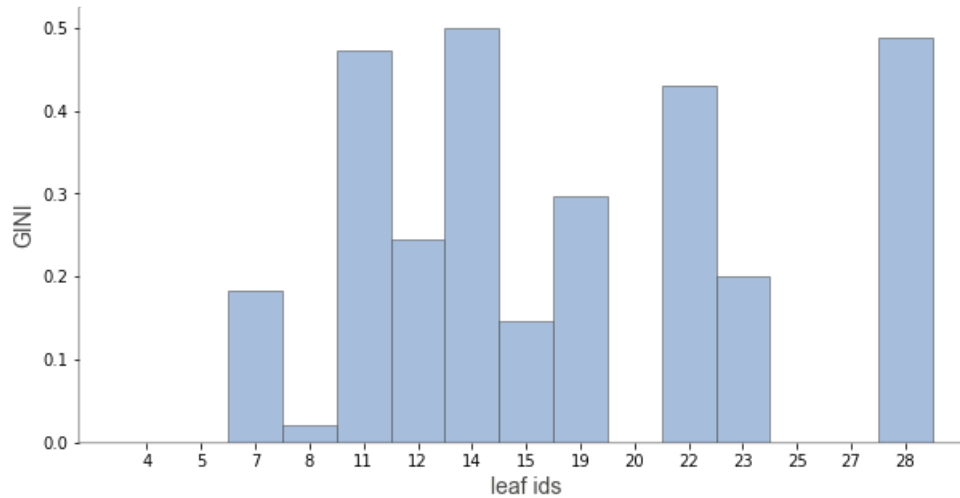
```
In [12]: trees.dtreeviz(sk_dtree, show_just_path=True, X = dataset[features].iloc[10])
```

Out[12]:

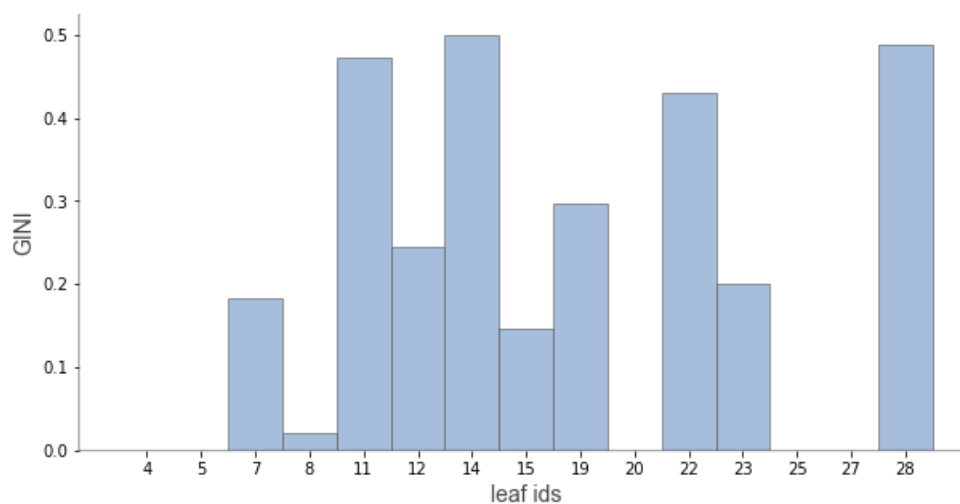


## viz\_leaf\_criterion

```
In [13]: trees.viz_leaf_criterion(tree_classifier)
```



```
In [14]: trees.viz_leaf_criterion(sk_dtree)
```



## describe\_node\_sample

```
In [15]: trees.describe_node_sample(tree_classifier, node_id=10, x_data=dataset[features], feature_names=features)
```

```
Out[15]:
```

	Pclass	Age	Fare	Sex_label	Cabin_label	Embarked_label
<b>count</b>	117.0	117.000000	117.000000	117.0	117.000000	117.000000
<b>mean</b>	3.0	23.976667	11.722829	0.0	6.196581	1.341880
<b>std</b>	0.0	10.534377	4.695136	0.0	31.167855	0.789614
<b>min</b>	3.0	0.750000	6.750000	0.0	-1.000000	0.000000
<b>25%</b>	3.0	18.000000	7.775000	0.0	-1.000000	1.000000
<b>50%</b>	3.0	27.000000	9.587500	0.0	-1.000000	2.000000
<b>75%</b>	3.0	29.699118	15.500000	0.0	-1.000000	2.000000
<b>max</b>	3.0	63.000000	23.250000	0.0	145.000000	2.000000

```
In [16]: trees.describe_node_sample(sk_dtree, node_id=10)
```

```
Out[16]:
```

	Pclass	Age	Fare	Sex_label	Cabin_label	Embarked_label
<b>count</b>	117.0	117.000000	117.000000	117.0	117.000000	117.000000
<b>mean</b>	3.0	23.976667	11.722829	0.0	6.196581	1.341880
<b>std</b>	0.0	10.534377	4.695136	0.0	31.167855	0.789614
<b>min</b>	3.0	0.750000	6.750000	0.0	-1.000000	0.000000
<b>25%</b>	3.0	18.000000	7.775000	0.0	-1.000000	1.000000
<b>50%</b>	3.0	27.000000	9.587500	0.0	-1.000000	2.000000
<b>75%</b>	3.0	29.699118	15.500000	0.0	-1.000000	2.000000
<b>max</b>	3.0	63.000000	23.250000	0.0	145.000000	2.000000

## explain\_prediction\_path

```
In [17]: X = dataset[features].iloc[10]
X
```

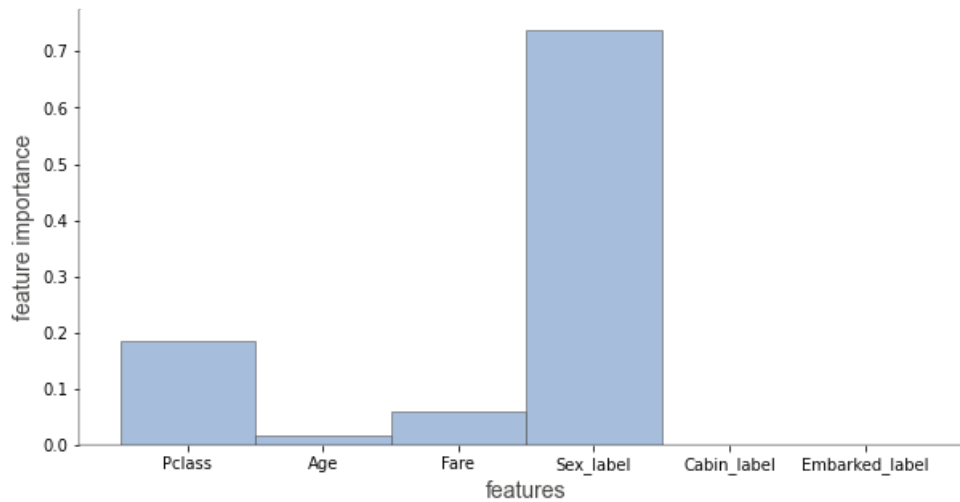
```
Out[17]: Pclass      3.0
Age          4.0
Fare        16.7
Sex_label    0.0
Cabin_label  145.0
Embarked_label 2.0
Name: 10, dtype: float64
```

```
In [18]: print(trees.explain_prediction_path(tree_classifier, X, feature_names=features, explanation_type="plain_

2.5 <= Pclass
Age < 36.5
Fare < 23.35
Sex_label < 0.5
```

```
In [19]: trees.explain_prediction_path(tree_classifier, X, feature_names=features, explanation_type="sklearn_defa
```

```
Out[19]: <AxesSubplot:xlabel='features', ylabel='feature importance'>
```



## Regressor

```
In [20]: features_reg = ["Pclass", "Fare", "Sex_label", "Cabin_label", "Embarked_label", "Survived"]
target_reg = "Age"
tree_regressor = DecisionTreeRegressor(max_depth=3, random_state=random_state, criterion="mae")
tree_regressor.fit(dataset[features_reg], dataset[target_reg])
```

```
Out[20]: DecisionTreeRegressor(criterion='mae', max_depth=3, random_state=1234)
```

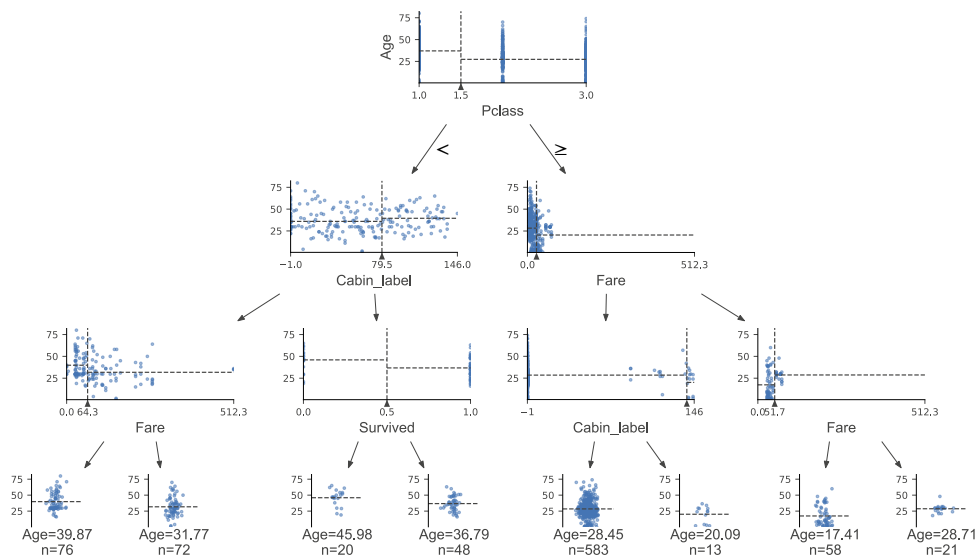
```
In [21]: sk_dtreg = ShadowSKDTree(tree_regressor, dataset[features_reg], dataset[target_reg], features_reg, t
```

## Visualizations

### dtreeviz

```
In [22]: trees.dtreeviz(tree_regressor, dataset[features_reg], dataset[target_reg], features_reg, target_reg)
```

```
Out[22]:
```

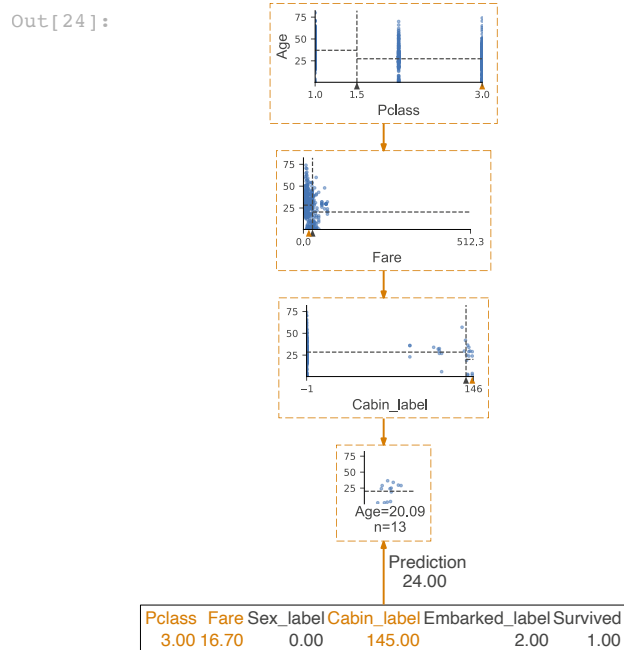


show just path

```
In [23]: x = dataset[features_reg].iloc[10]
x
```

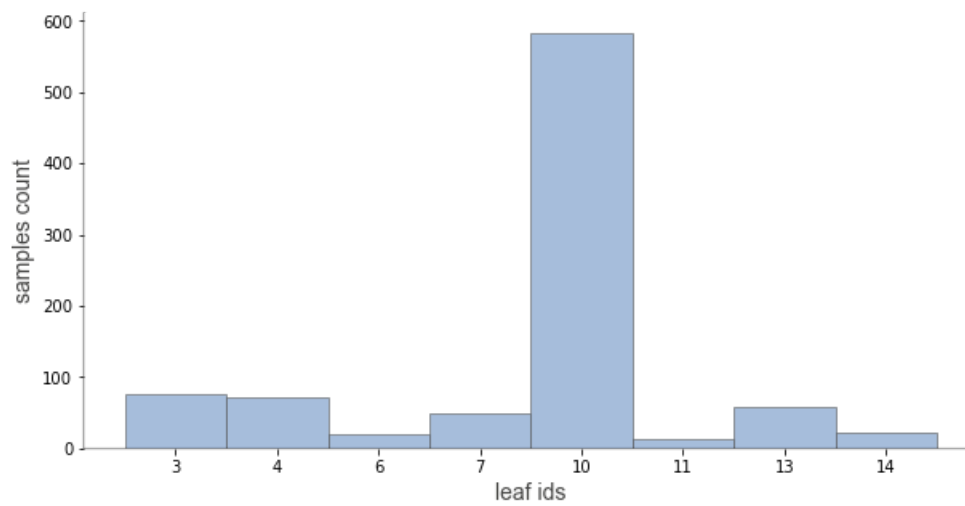
```
Out[23]: Pclass      3.0
Fare       16.7
Sex_label   0.0
Cabin_label 145.0
Embarked_label 2.0
Survived    1.0
Name: 10, dtype: float64
```

```
In [24]: trees.dtreeviz(sk_dtree_reg, show_just_path=True, X = x)
```

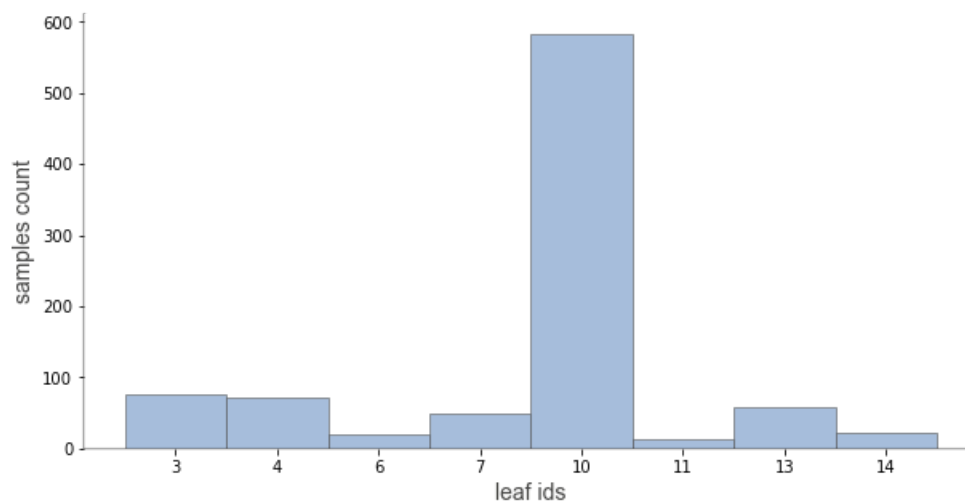


viz\_leaf\_samples

```
In [25]: trees.viz_leaf_samples(tree_regressor, dataset[features_reg], features_reg)
```

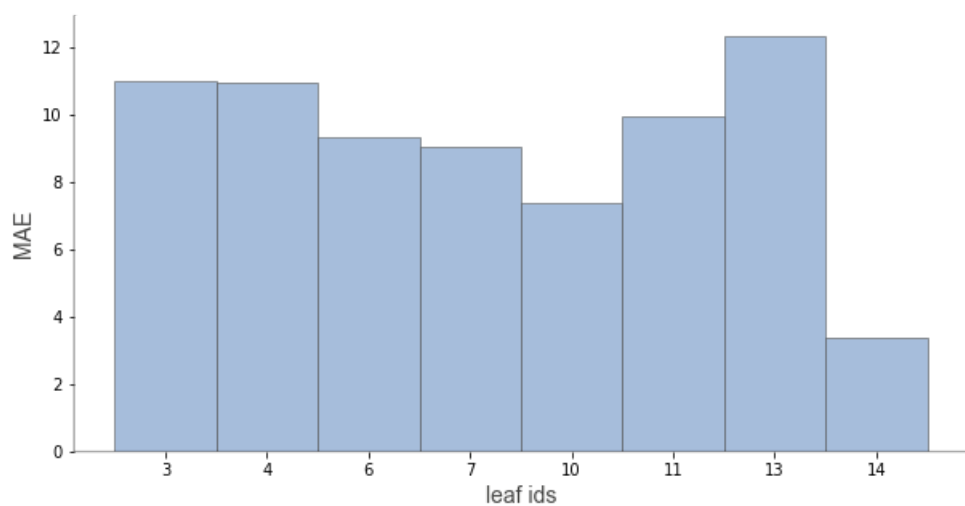


```
In [26]: trees.viz_leaf_samples(sk_dtree_reg)
```



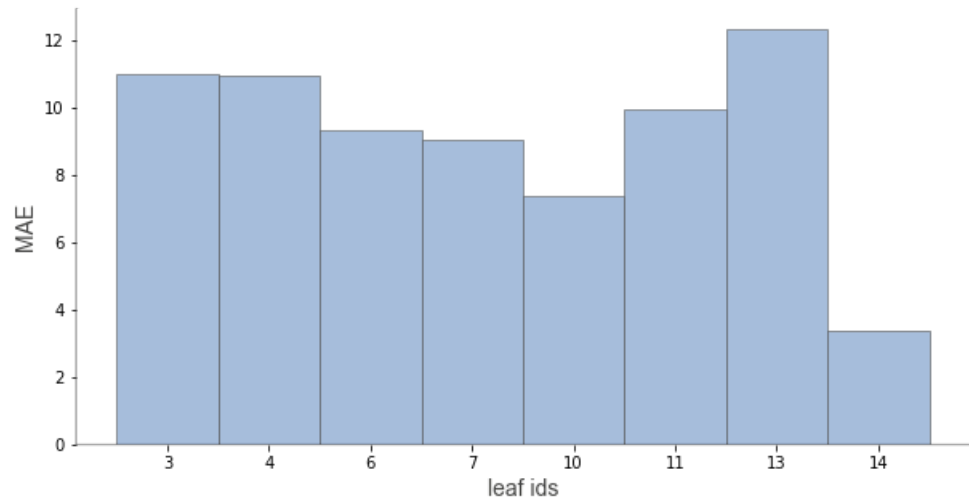
## viz\_leaf\_criterion

```
In [27]: trees.viz_leaf_criterion(tree_regressor)
```



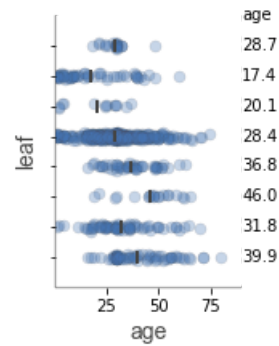
```
In [28]: trees.viz_leaf_criterion(sk_dtree_reg)
```



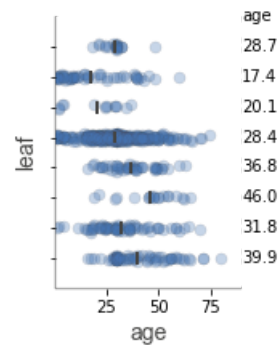


viz\_leaf\_target

```
In [29]: trees.viz_leaf_target(tree_regressor, dataset[features_reg], dataset[target_reg], features_reg, target_r
```



```
In [30]: trees.viz_leaf_target(sk_dtree_reg)
```



describe\_node\_sample

```
In [31]: trees.describe_node_sample(tree_regressor, node_id=1, x_data=dataset[features], feature_names=features_r
```

Out[31]:

	Pclass	Fare	Sex_label	Cabin_label	Embarked_label	Survived
count	216.0	216.000000	216.000000	216.000000	216.000000	216.000000
mean	1.0	37.048118	84.154687	0.564815	54.773148	1.175926
std	0.0	14.046369	78.380373	0.496933	43.781906	0.996073
min	1.0	0.920000	0.000000	0.000000	-1.000000	-1.000000
25%	1.0	29.000000	30.923950	0.000000	13.750000	0.000000

	Pclass	Fare	Sex_label	Cabin_label	Embarked_label	Survived
<b>50%</b>	1.0	35.000000	60.287500	1.000000	53.500000	2.000000
<b>75%</b>	1.0	47.250000	93.500000	1.000000	91.250000	2.000000
<b>max</b>	1.0	80.000000	512.329200	1.000000	146.000000	2.000000

```
In [32]: trees.describe_node_sample(sk_dtree_reg, node_id=1)
```

```
Out[32]:
```

	Pclass	Fare	Sex_label	Cabin_label	Embarked_label	Survived
<b>count</b>	216.0	216.000000	216.000000	216.000000	216.000000	216.000000
<b>mean</b>	1.0	84.154687	0.564815	54.773148	1.175926	0.629630
<b>std</b>	0.0	78.380373	0.496933	43.781906	0.996073	0.484026
<b>min</b>	1.0	0.000000	0.000000	-1.000000	-1.000000	0.000000
<b>25%</b>	1.0	30.923950	0.000000	13.750000	0.000000	0.000000
<b>50%</b>	1.0	60.287500	1.000000	53.500000	2.000000	1.000000
<b>75%</b>	1.0	93.500000	1.000000	91.250000	2.000000	1.000000
<b>max</b>	1.0	512.329200	1.000000	146.000000	2.000000	1.000000

## explain\_prediction\_path

```
In [33]: X_reg = dataset[features_reg].iloc[10]
X_reg
```

```
Out[33]: Pclass      3.0
Fare      16.7
Sex_label  0.0
Cabin_label 145.0
Embarked_label 2.0
Survived  1.0
Name: 10, dtype: float64
```

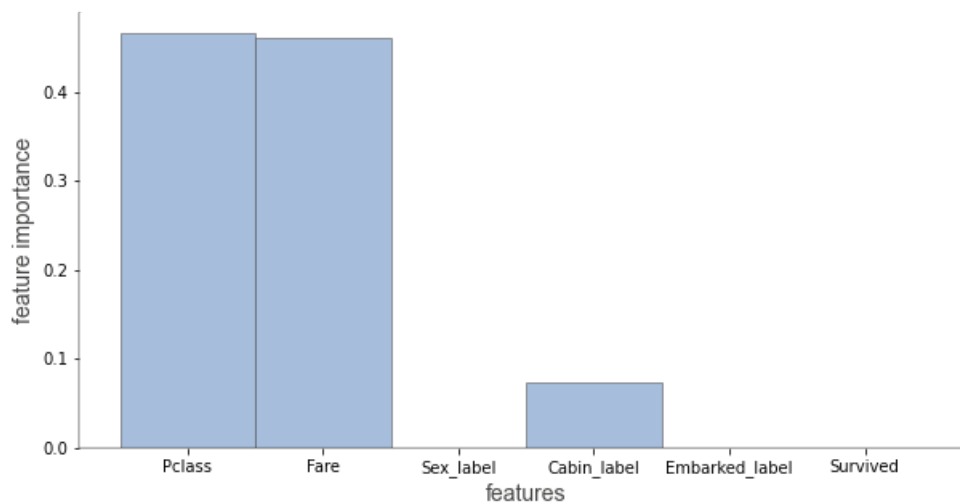
```
In [34]: print(trees.explain_prediction_path(tree_regressor, X_reg, feature_names=features_reg, explanation_type=
1.5 <= Pclass
Fare < 27.82
139.5 <= Cabin_label
```

```
In [35]: print(trees.explain_prediction_path(sk_dtree_reg, X_reg, "plain_english"))

1.5 <= Pclass
Fare < 27.82
139.5 <= Cabin_label
```

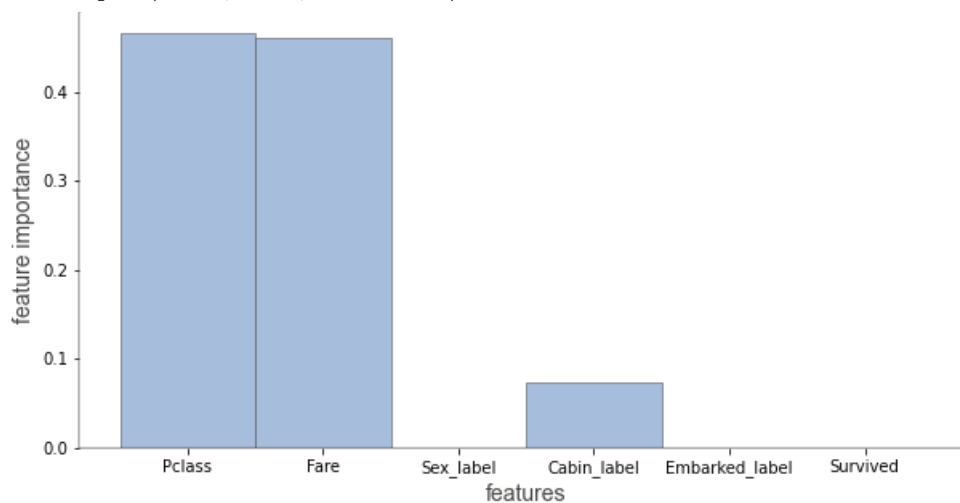
```
In [36]: trees.explain_prediction_path(tree_regressor, X_reg, feature_names=features_reg, explanation_type="sklea
```

```
Out[36]: <AxesSubplot:xlabel='features', ylabel='feature importance'>
```



```
In [37]: print(trees.explain_prediction_path(sk_dtree_reg, X_reg, explanation_type="sklearn_default"))
```

AxesSubplot(0.125,0.125;0.775x0.755)



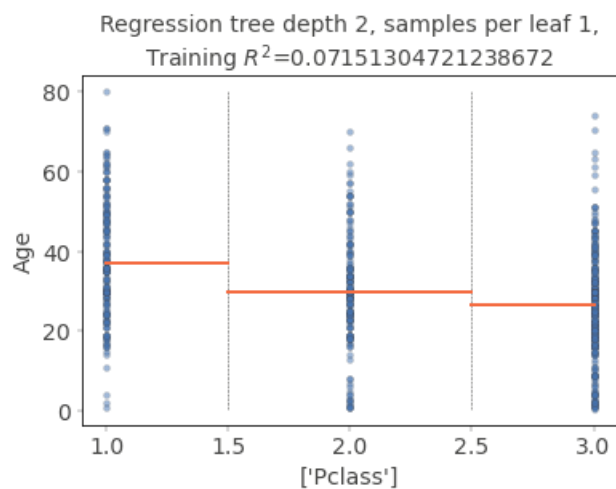
## rtreeviz\_univar

```
In [38]: features_reg_univar = ["Pclass"]
target_reg = "Age"
dtr_univar = DecisionTreeRegressor(max_depth=2, random_state=random_state, criterion="mae")
dtr_univar.fit(dataset[features_reg_univar], dataset[target_reg])
```

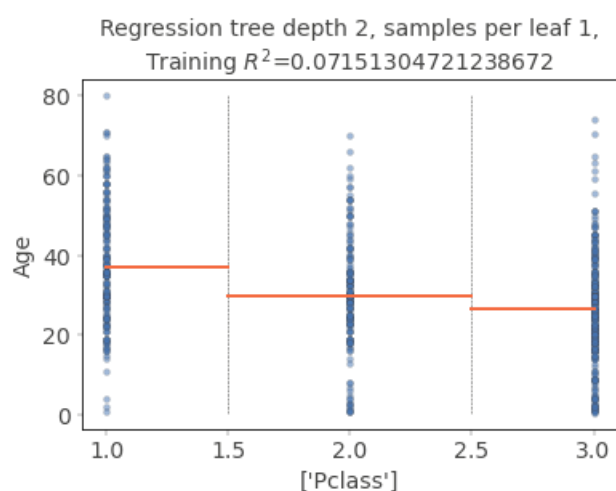
```
Out[38]: DecisionTreeRegressor(criterion='mae', max_depth=2, random_state=1234)
```

```
In [39]: skdtree_univar = ShadowSKDTree(dtr_univar, dataset[features_reg_univar], dataset[target_reg], features_r
```

```
In [40]: trees.rtreeviz_univar(dtr_univar, dataset[features_reg_univar], dataset[target_reg], features_reg_univar
```



```
In [41]: trees.rtreeviz_univar(skdtree_univar)
```



## rtreeviz\_bivar\_3D

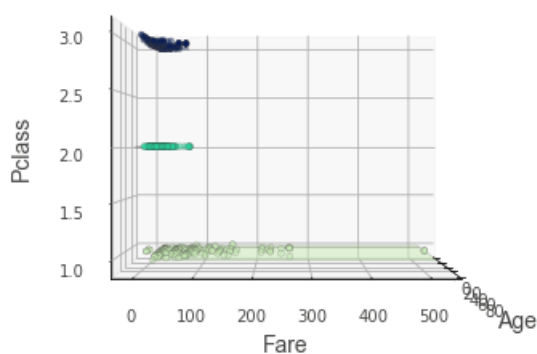
```
In [42]: features_reg_bivar = ["Age", "Fare"]
target_reg = "Pclass"
dtr_bivar = DecisionTreeRegressor(max_depth=3, random_state=random_state, criterion="mae")
dtr_bivar.fit(dataset[features_reg_bivar], dataset[target_reg])
```

```
Out[42]: DecisionTreeRegressor(criterion='mae', max_depth=3, random_state=1234)
```

```
In [43]: skdtree_bivar = ShadowSKDTree(dtr_bivar, dataset[features_reg_bivar], dataset[target_reg], features_reg_
```

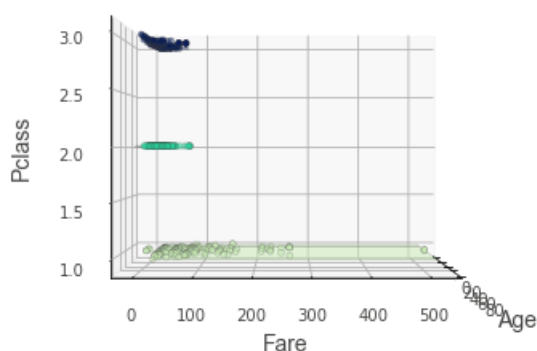
```
In [44]: trees.rtreeviz_bivar_3D(dtr_bivar, dataset[features_reg_bivar], dataset[target_reg], features_reg_bivar,
```

Regression tree depth 3, training  $R^2=0.568$



```
In [45]: trees.rtreeviz_bivar_3D(skdtree_bivar)
```

Regression tree depth 3, training  $R^2=0.568$



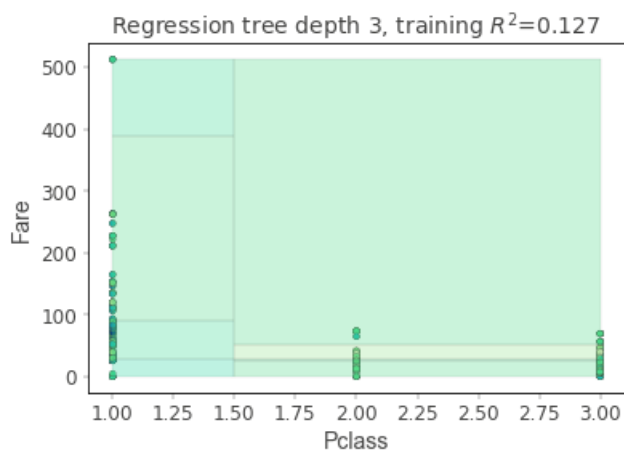
## rtreeviz\_bivar\_heatmap

```
In [46]: features_reg_bivar = ["Pclass", "Fare"]
target_reg = "Age"
dtr_bivar = DecisionTreeRegressor(max_depth=3, random_state=random_state, criterion="mae")
dtr_bivar.fit(dataset[features_reg_bivar], dataset[target_reg])
```

```
Out[46]: DecisionTreeRegressor(criterion='mae', max_depth=3, random_state=1234)
```

```
In [47]: skdtree_bivar = ShadowSKDTree(dtr_bivar, dataset[features_reg_bivar], dataset[target_reg], features_reg_b)
```

```
In [48]: trees.rtreeviz_bivar_heatmap(dtr_bivar, dataset[features_reg_bivar], dataset[target_reg], features_reg_b)
```



```
In [49]:
```

```
trees.rtreeviz_bivar_heatmap(skdtree_bivar)
```

