

# gb\_algorithm

January 31, 2022

```
[2]: import pandas as pd
import math
from sklearn.tree import DecisionTreeRegressor
import graphviz
from sklearn import tree
```

Dataset:

```
[3]: d = {'category': ['hiphop', 'hiphop', 'hiphop', 'hiphop', 'hiphop', 'jazz', 'jazz', 'jazz'],
        'track':      ['h1', 'h2', 'h3', 'h4', 'h5', 'j1', 'j2', 'j3'],
        'feature_danceability' : [0.949, 0.743, 0.913, 0.810, 0.434, 0.654, 0.593, 0.234],
        'feature_acousticness' : [0.132, 0.234, 0.394, 0.504, 0.198, 0.534, 0.312, 0.341],
        'label'           : [1, 1, 1, 1, 1, 0, 0, 0]}
df = pd.DataFrame(data=d)

df
```

```
[3]:
```

	category	track	feature_danceability	feature_acousticness	label
0	hiphop	h1	0.949	0.132	1
1	hiphop	h2	0.743	0.234	1
2	hiphop	h3	0.913	0.394	1
3	hiphop	h4	0.810	0.504	1
4	hiphop	h5	0.434	0.198	1
5	jazz	j1	0.654	0.534	0
6	jazz	j2	0.593	0.312	0
7	jazz	j3	0.234	0.341	0

Initial Prediction:

```
[4]: hiphop = 0
jazz = 0

for i, row in enumerate(df.values):
    # hiphop
    if row[4] == 1:
```

```

        hiphop = hiphop + 1
    # jazz
    else:
        jazz = jazz + 1

# calculate initial predictions

f0x = math.log(hiphop/jazz)

f0x

```

[4]: 0.5108256237659907

```

[5]: # append initial prediction to dataset (equal for every sample)
f0x_array = []

for i, row in enumerate(df.values):
    f0x_array.append(f0x)

df['f0x'] = f0x_array

df

```

[5]:	category	track	feature_danceability	feature_acousticness	label	f0x
0	hiphop	h1	0.949	0.132	1	0.510826
1	hiphop	h2	0.743	0.234	1	0.510826
2	hiphop	h3	0.913	0.394	1	0.510826
3	hiphop	h4	0.810	0.504	1	0.510826
4	hiphop	h5	0.434	0.198	1	0.510826
5	jazz	j1	0.654	0.534	0	0.510826
6	jazz	j2	0.593	0.312	0	0.510826
7	jazz	j3	0.234	0.341	0	0.510826

Iterative Process:

First Iteration:

```

[6]: # calculate probabilities
prob_array = []
e = 2.718

for i, row in enumerate(df.values):
    prob = math.pow(e, row[5]) / (1 + math.pow(e, row[5]))
    prob_array.append(prob)

df['probability_f0x'] = prob_array

```

```
[7]: # calculate pseudo residuals
residual_array = []

for i, row in enumerate(df.values):
    residual = row[4] - row[6]
    residual_array.append(residual)

df['pseudo_residuals_1'] = residual_array
```

```
[8]: # build regression tree

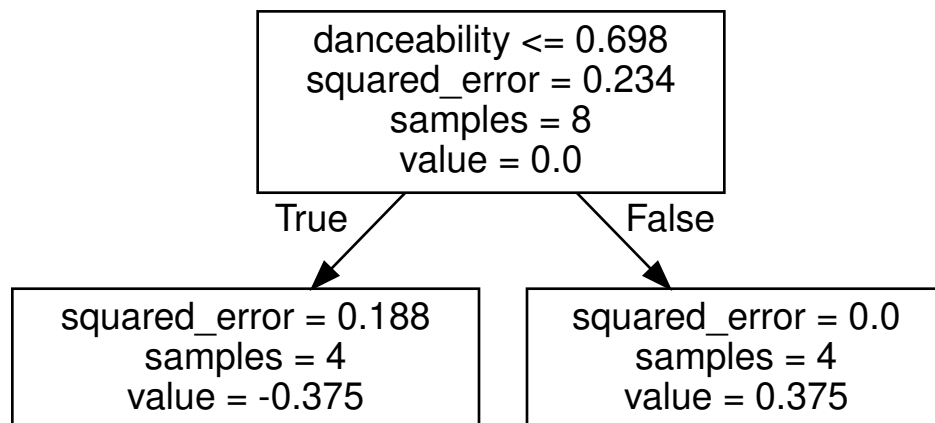
# input values
features = ['feature_danceability']
Y = df['pseudo_residuals_1']
X = df[features]

# build tree
dt = DecisionTreeRegressor(max_depth = 1)
dt.fit(X, Y)

# display
dot_data = tree.export_graphviz(
    dt,
    feature_names = ['danceability']
)
graph = graphviz.Source(dot_data)

graph
```

[8]:



```
[9]: # calculate output values
```

```
sum_pr = 0
```

```

sum_prob = 0

# leafs
left = df.loc[df['feature_danceability'] < 0.698]
right = df.loc[df['feature_danceability'] > 0.698]

# left leaf
for i, row in enumerate(left.values):
    sum_pr = sum_pr + row[7]
    sum_prob = sum_prob + (row[6] * (1 - row[6]))

ov_left = sum_pr / sum_prob

sum_pr = 0
sum_prob = 0

for i, row in enumerate(right.values):
    sum_pr = sum_pr + row[7]
    sum_prob = sum_prob + (row[6] * (1 - row[6]))

ov_right = sum_pr / sum_prob

# fill table
ov_array = []

for i, row in enumerate(df.values):
    if row[2] < 0.698:
        ov = ov_left
    else:
        ov = ov_right
    ov_array.append(ov)

print(ov_left)
print(ov_right)

df['output_value_1'] = ov_array

```

```

-1.599925851127446
1.6000317795910484

```

```

[10]: # new predictions
learning_rate = 0.6

fix_array = []

for i, row in enumerate(df.values):
    fix = row[5] + (learning_rate * row[8])

```

```
f1x_array.append(f1x)

df['f1x'] = f1x_array
```

Second Iteration:

```
[11]: # calculate probabilities
prob_array = []

for i, row in enumerate(df.values):
    prob = math.pow(e, row[9]) / (1 + math.pow(e, row[9]))
    prob_array.append(prob)

df['probability_f1x'] = prob_array
```

```
[12]: # calculate pseudo residuals
residual_array = []

for i, row in enumerate(df.values):
    residual = row[4] - row[10]
    residual_array.append(residual)

df['pseudo_residuals_2'] = residual_array
```

```
[13]: # build regression tree

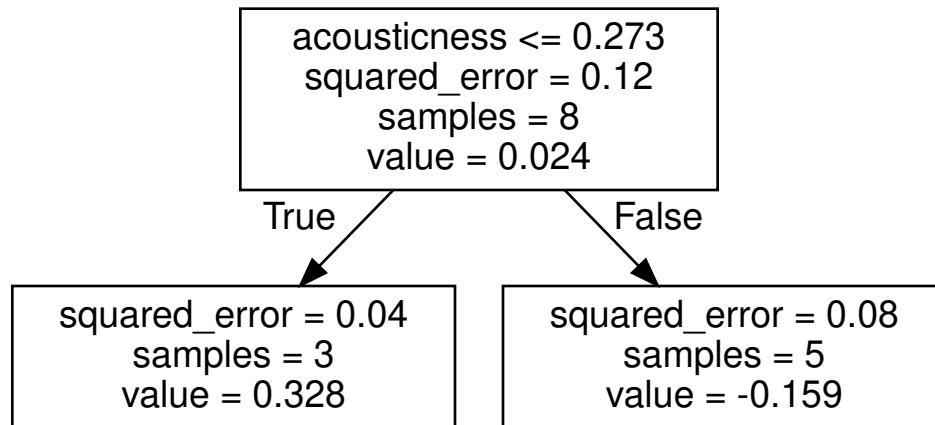
# input values
features = ['feature_acousticness']
Y = df['pseudo_residuals_2']
X = df[features]

# build tree
dt2 = DecisionTreeRegressor(max_depth = 1)
dt2.fit(X, Y)

# display
dot_data = tree.export_graphviz(
    dt2,
    feature_names = ['acousticness']
)
graph = graphviz.Source(dot_data)

graph
```

[13]:



[14]: *# calculate output values*

```

sum_pr    = 0
sum_prob  = 0

# leafs
left = df.loc[df['feature_acousticness'] < 0.273]
right = df.loc[df['feature_acousticness'] > 0.273]

# left leaf
for i, row in enumerate(left.values):
    sum_pr    = sum_pr + row[11]
    sum_prob  = sum_prob + (row[10] * (1 - row[10]))

ov_left = sum_pr / sum_prob

print(sum_pr)
print(sum_prob)

sum_pr    = 0
sum_prob  = 0

for i, row in enumerate(right.values):
    sum_pr    = sum_pr + row[11]
    sum_prob  = sum_prob + (row[10] * (1 - row[10]))

ov_right = sum_pr / sum_prob

# fill table
ov_array = []

for i, row in enumerate(df.values):

```

```

    if row[3] < 0.273:
        ov = ov_left
    else:
        ov = ov_right
    ov_array.append(ov)

print(ov_left)
print(ov_right)

df['output_value_2'] = ov_array

```

```

0.9840961179667522
0.5416655512088253
1.8167965745108998
-0.7815561900329465

```

```

[15]: # new predictions
f2x_array = []

for i, row in enumerate(df.values):
    f2x = row[9] + (learning_rate * row[12])
    f2x_array.append(f2x)

df['f2x'] = f2x_array

```

Final Calculation Predictions for samples

```

[16]: # calculate probabilites
prob_array = []

for i, row in enumerate(df.values):
    prob = math.pow(e, row[13]) / (1 + math.pow(e, row[13]))
    prob_array.append(prob)

prob_array

df['Final Classification Probability'] = prob_array

df

# probability that song belongs to category hiphop

```

```

[16]:
category track  feature_danceability  feature_acousticness  label  f0x \
0  hiphop  h1  0.949  0.132  1  0.510826
1  hiphop  h2  0.743  0.234  1  0.510826
2  hiphop  h3  0.913  0.394  1  0.510826
3  hiphop  h4  0.810  0.504  1  0.510826

```

4	hiphop	h5	0.434	0.198	1	0.510826
5	jazz	j1	0.654	0.534	0	0.510826
6	jazz	j2	0.593	0.312	0	0.510826
7	jazz	j3	0.234	0.341	0	0.510826

	probability_f0x	pseudo_residuals_1	output_value_1	f1x	\
0	0.624988	0.375012	1.600032	1.470845	
1	0.624988	0.375012	1.600032	1.470845	
2	0.624988	0.375012	1.600032	1.470845	
3	0.624988	0.375012	1.600032	1.470845	
4	0.624988	0.375012	-1.599926	-0.449130	
5	0.624988	-0.624988	-1.599926	-0.449130	
6	0.624988	-0.624988	-1.599926	-0.449130	
7	0.624988	-0.624988	-1.599926	-0.449130	

	probability_f1x	pseudo_residuals_2	output_value_2	f2x	\
0	0.813163	0.186837	1.816797	2.560923	
1	0.813163	0.186837	1.816797	2.560923	
2	0.813163	0.186837	-0.781556	1.001911	
3	0.813163	0.186837	-0.781556	1.001911	
4	0.389579	0.610421	1.816797	0.640948	
5	0.389579	-0.389579	-0.781556	-0.918064	
6	0.389579	-0.389579	-0.781556	-0.918064	
7	0.389579	-0.389579	-0.781556	-0.918064	

	Final Classification Probability
0	0.928286
1	0.928286
2	0.731414
3	0.731414
4	0.654953
5	0.285372
6	0.285372
7	0.285372