

Example of How a Single Decision Tree Will Predict a Tweet

Features of Tweet

keyword	apocalypse
location	empty string
link	False
num_hashtag	0
word_count	6-10
char_count	41-60
avg_char_per_word	6.1
keyword2	day
bigrams	none

Split 1 : location

Tweet => location => empty string

Entropy location : 0.50659025

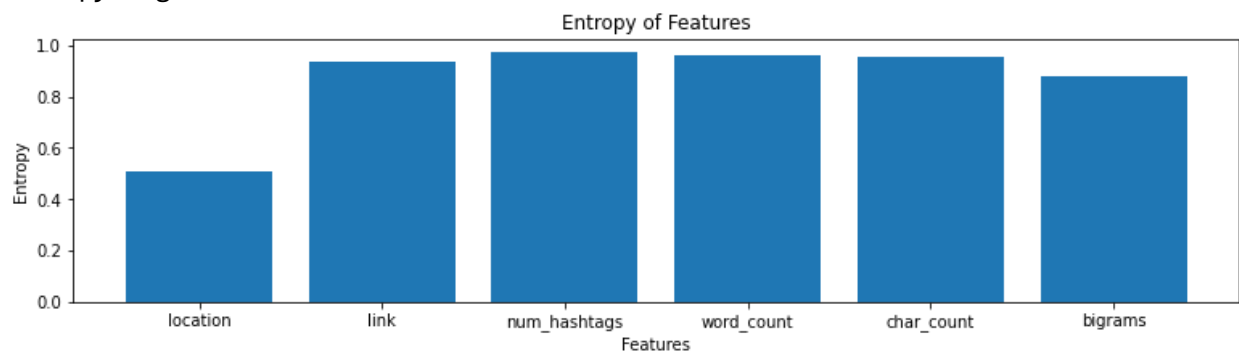
Entropy link : 0.9378582

Entropy num_hashtags : 0.97621673

Entropy word_count : 0.9635199

Entropy char_count : 0.9545876

Entropy bigrams : 0.882426



Split 2 : bigrams

Tweet => bigrams => none

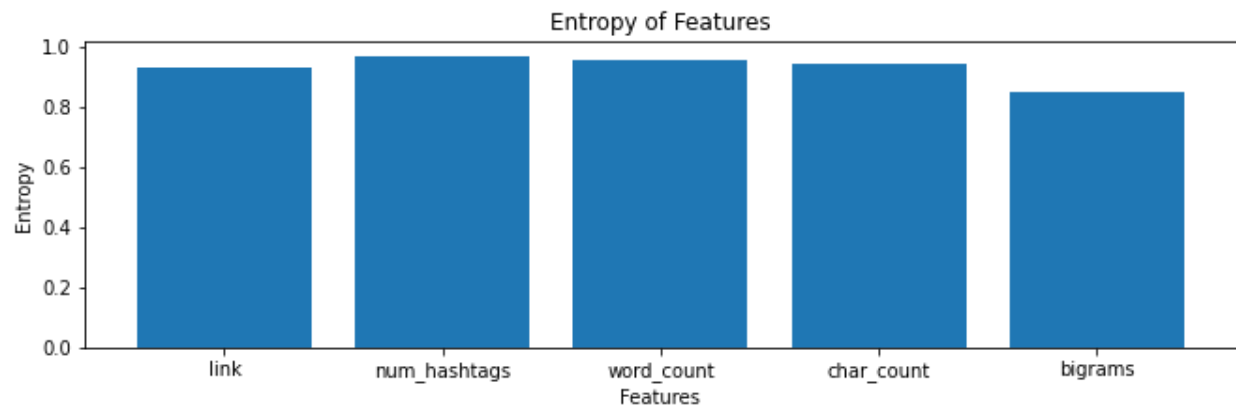
Entropy link : 0.92933375

Entropy num_hashtags : 0.96966213

Entropy word_count : 0.9549786

Entropy char_count : 0.93903965

Entropy bigrams : 0.85132957



Split 3 : link

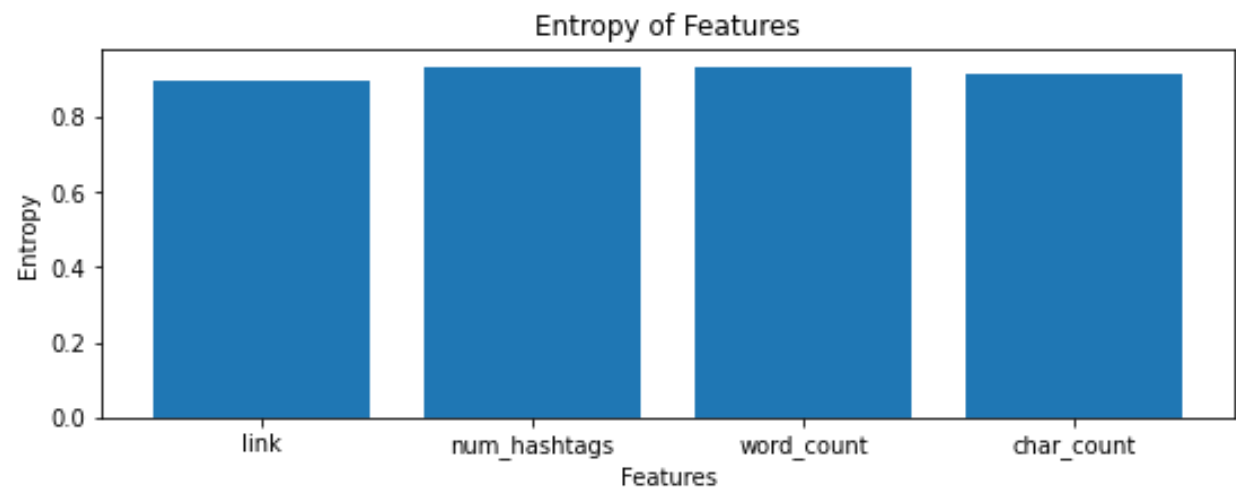
Tweet => link => False

Entropy link : 0.89738905

Entropy num_hashtags : 0.93562067

Entropy word_count : 0.9324234

Entropy char_count : 0.91463715



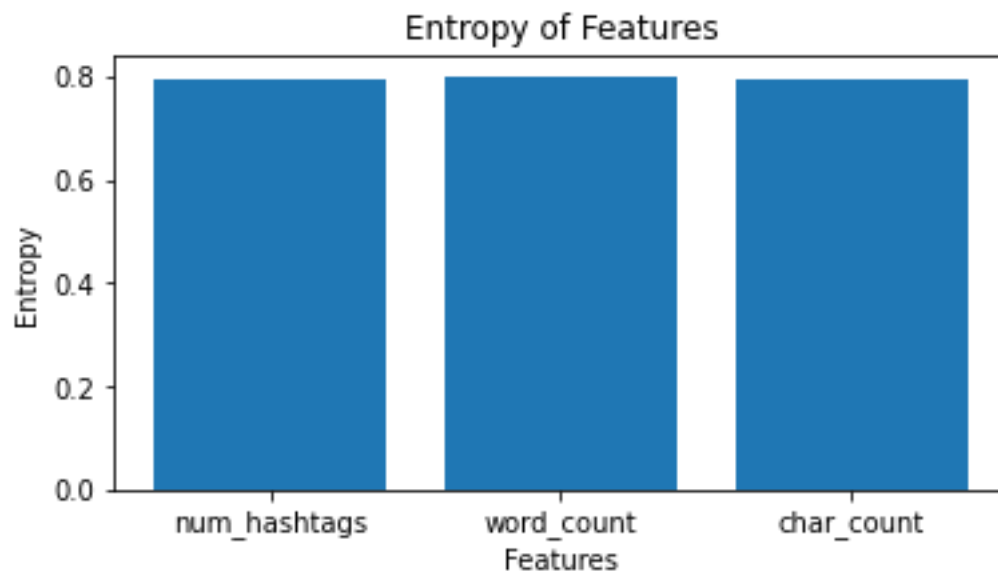
Split 4 : num_hashtags

Tweet => num_hashtags => 0

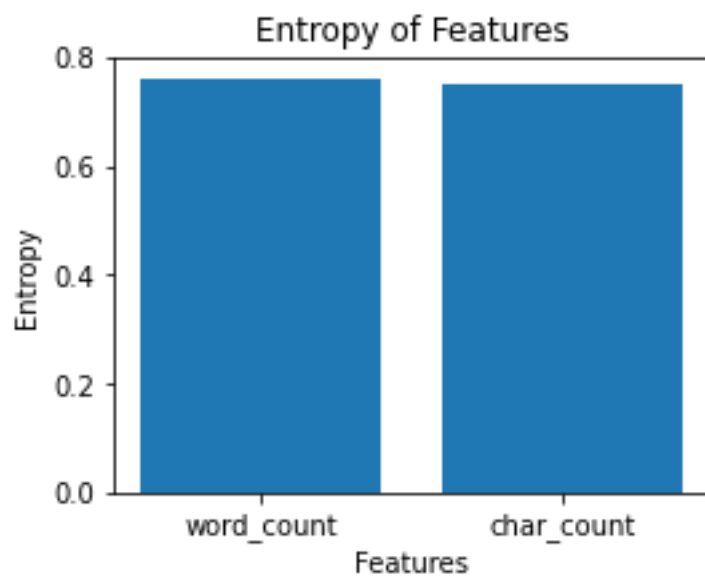
Entropy num_hashtags : 0.79276335

Entropy word_count : 0.8015714

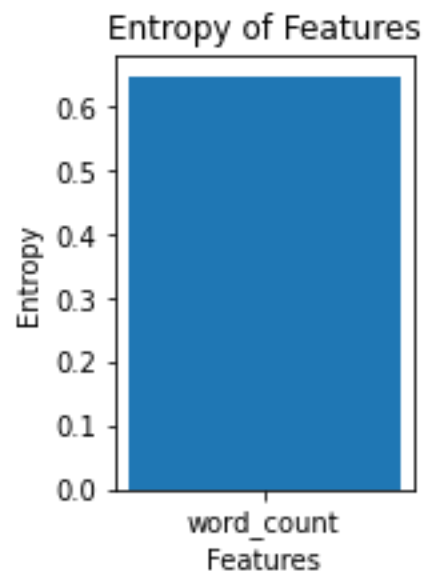
Entropy char_count : 0.7949071



Split 5 : char_count
Tweet => char_count => 41-60
Entropy word_count : 0.76196915
Entropy char_count : 0.7514149



Split 6 : word_count
Tweet => word_count => 6-10
Entropy word_count : 0.6495526



Tree: Leaf(value='0')
Prediction: 0
Actual Value: 0