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1.

Unigram without lemmatization and stemming

Training accuracy: 0.9999 Validation accuracy: 0.8216

original\_clean\_reviews=review\_cleaner(train['review'],lemmatize=False,stem=False) train\_predict\_sentiment(cleaned\_reviews=original\_clean\_reviews, y=train["sentiment"],ngram=1,max features=1000)

Unigram with lemmatization but without stemming

Training accuracy: 1.0 Validation accuracy: 0.8184

original\_clean\_reviews=review\_cleaner(train['review'],lemmatize=True,stem=False) train\_predict\_sentiment(cleaned\_reviews=original\_clean\_reviews, y=train["sentiment"],ngram=1,max features=1000)

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Unigram without lemmatization but with stemming

Training accuracy: 1.0 Validation accuracy: 0.82

original\_clean\_reviews=review\_cleaner(train['review'],lemmatize=False,stem=True) train\_predict\_sentiment(cleaned\_reviews=original\_clean\_reviews, y=train["sentiment"],ngram=1,max\_features=1000)

2.

Bigram without lemmatization and stemming

Training accuracy: 0.99995 Validation accuracy: 0.8208

original\_clean\_reviews=review\_cleaner(train['review'],lemmatize=False,stem=False) train\_predict\_sentiment(cleaned\_reviews=original\_clean\_reviews, y=train["sentiment"],ngram=2,max\_features=1000)

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Bigram with lemmatization but without stemming

Training accuracy: 0.99995 Validation accuracy: 0.8184

original\_clean\_reviews=review\_cleaner(train['review'],lemmatize=True,stem=False)

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train predict sentiment(cleaned reviews=original clean reviews,
y=train["sentiment"],ngram=2,max features=1000)
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Bigram without lemmatization but with stemming
Training accuracy: 1.0
Validation accuracy: 0.8264
original clean reviews=review cleaner(train['review'],lemmatize=False,stem=True)
train_predict_sentiment(cleaned_reviews=original_clean_reviews,
y=train["sentiment"],ngram=2,max features=1000)
3.
max features=10
Training accuracy: 0.87125
Validation accuracy: 0.5648
original clean reviews=review cleaner(train['review'],lemmatize=True,stem=False)
train predict sentiment(cleaned reviews=original clean reviews,
y=train["sentiment"],ngram=1,max features=10)
max features=100
Training accuracy: 0.99985
Validation accuracy: 0.7206
original_clean_reviews=review_cleaner(train['review'],lemmatize=True,stem=False)
train predict sentiment(cleaned reviews=original clean reviews,
y=train["sentiment"],ngram=1,max features=100)
max features=1000
Training accuracy: 1.0
Validation accuracy: 0.8184
original clean reviews=review cleaner(train['review'],lemmatize=True,stem=False)
train predict sentiment(cleaned reviews=original clean reviews,
y=train["sentiment"],ngram=1,max features=1000)
max features=5000
Training accuracy: 1.0
Validation accuracy: 0.8352
original clean reviews=review cleaner(train['review'],lemmatize=True,stem=False)
train predict sentiment(cleaned reviews=original clean reviews,
y=train["sentiment"],ngram=1,max features=5000)
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For original cleaned reviews using both unigram and bigram, the top ten most important features included similar words, ex: "worst" and "worse"; for lemmatized reviews for unigram and bigram, the most important features included more words, since "worst" and "worse" were lumped together; for stemmed reviews using unigram, the top ten most important features included words that are missing the ending letters, such as "wast" and "terribl", as well as "aw" which should actually be "awful," while using bigram resulted in "wast[e]" and "wast[e] time," which is redundant. The validation accuracy for all scenarios under unigram and all scenarios under bigram is similar.

As the number of <code>max\_features</code> increased, so did the validation accuracy. However, the difference between 1000 and 5000 <code>max\_features</code> is insignificant, and the only difference between the top ten important features for each is the first has "wonderful" while the second has "worse" instead, which is actually less helpful as "worst" and "worse" practically mean the same thing.