	<pre>import nltk nltk.download('punkt') nltk.download('wordnet') nltk.download('averaged_perceptron_tagger') [nltk_data] Error loading punkt: <urlopen 11001]="" [errno="" [nltk_data]="" error="" failed="" getaddrinfo=""> [nltk_data] Error loading wordnet: <urlopen 11001]="" [errno="" [nltk_data]="" error="" failed="" getaddrinfo=""> [nltk_data] Error loading averaged_perceptron_tagger: <urlopen error<="" pre=""></urlopen></urlopen></urlopen></pre>
Out[2]: In [3]:	<pre>[nltk_data]</pre>
[n [4]: [n [5]:	<pre>df_new['lyrics'] = [word_tokenize(entry) for entry in df_new['lyrics']] df['lyrics'] = [entry.lower() for entry in df['lyrics']] df['lyrics'] = [word_tokenize(entry) for entry in df['lyrics']]</pre>
[n [6]:	<pre>tag_map['J'] = wn.ADJ tag_map['V'] = wn.VERB tag_map['R'] = wn.ADV for index,entry in enumerate(df_new['lyrics']): Final_words = [] word_Lemmatized = WordNetLemmatizer() for word, tag in pos_tag(entry):</pre>
[n [7]: Out[7]:	<pre>if word not in stopwords.words('english') and word.isalpha(): word_Final = word_Lemmatized.lemmatize(word,tag_map[tag[0]] Final_words.append(word_Final) df_new.loc[index,'text_final'] = str(Final_words)</pre> df.tail()
In [8]:	1478 [rip, them, down, hold, me, up, tell, them, th 3 1479 [it, 's, way, too, late, to, think, of, someon 4 1480 [got, no, place, to, go, but, there, 's, a, gi 2 1481 [every, light, in, the, night, flicker, in, an 1 tag_map = defaultdict(lambda : wn.NOUN) tag_map['J'] = wn.ADJ
	<pre>tag_map['V'] = wn.VERB tag_map['R'] = wn.ADV for index,entry in enumerate(df['lyrics']): Final_words = [] word_Lemmatized = WordNetLemmatizer() for word, tag in pos_tag(entry): if word not in stopwords.words('english') and word.isalpha(): word_Final = word_Lemmatized.lemmatize(word,tag_map[tag[0]] Final_words.append(word_Final) df.loc[index,'text_final'] = str(Final_words)</pre>
In [9]:	Creating Training data, training label, test data, test label
	Part 3 : Feature Engineering We cannot give input as a list of words in Machine Learning models. Feature Engineering needs to be done for the lyrics column. We will use 3 types of NLP models for this - CountVectorizer
. [40].	TfidfVectorizer Tfidf-NGram Model At first, Label Encoding -
n [10]: n [11]:	<pre>Encoder = LabelEncoder() train_y = Encoder.fit_transform(train_y.ravel()) valid_y = Encoder.fit_transform(valid_y.ravel()) all_texts = [] for items in train_x: all_texts.append(items) for items in valid_x: all_texts.append(items) print(all_texts[0])</pre>
	['god', 'need', 'friend', 'god', 'come', 'end', 'god', 'lose', 'mind', 'god', 'find', 'wan', 'na', 'wan', 'na', 'wan', 'na', 'lov e', 'felt', 'like', 'lake', 'hate', 'felt', 'like', 'love', 'say', 'fea r', 'say', 'scar', 'wan', 'na', 'wan', 'na', 'wan', 'na', 'wan', 'na', 'tell', 'believe', 'tell', 'see', 'cause', 'know', 'trust', 'heart', 'fil', 'disgust', 'ca', 'take', 'tell', 'believe', 'tell', 'believe', 'tell', 'believe', 'tell', 'believe', 'tell', 'disgust', 'tell', 'believe', 'tell', 'tell', 'believe', 'tell', 'tell', 'believe', 'tell',
1 [12]:	· · · · · · · · · · · · · · · · · · ·
n [13]:	CountVectorizer Model
n [14]:	<pre>import string import re porter_stemmer = nltk.stem.porter.PorterStemmer() def porter_tokenizer(text, stemmer=porter_stemmer):</pre>
	<pre>lower_txt = text.lower() tokens = nltk.wordpunct_tokenize(lower_txt) stems = [porter_stemmer.stem(t) for t in tokens] no_punct = [s for s in stems if re.match('^[a-zA-Z]+\$', s) is not Note ne] return no_punct from sklearn.feature_extraction.text import TfidfVectorizer tfidf_vect = TfidfVectorizer(</pre>
	analyzer='word', binary=False, stop_words="english", tokenizer=porter_tokenizer) Tfidf-NGram Model
n [15]:	<pre>#tfidf_vect = TfidfVectorizer(analyzer='word', max_features=7000) tfidf_vect.fit(all_texts) xtrain_tfidf = tfidf_vect.transform(train_x) xvalid_tfidf = tfidf_vect.transform(valid_x) tfidf_vect_ngram = TfidfVectorizer(analyzer='word', ngram_range=(2,2), nax_features=7000) tfidf_vect_ngram.fit(all_texts) xtrain_tfidf_ngram = tfidf_vect_ngram.transform(train_x) xvalid_tfidf_ngram = tfidf_vect_ngram.transform(valid_x)</pre>
	Now 3 Training Dataset, 3 Test Dataset are ready to fit into Machine Learning models, one dataset per model defined above. Part 4: Mood Prediction Model
	<pre>valid, is_neural_net=False): classifier.fit(feature_vector_train, label) predictions = classifier.predict(feature_vector_valid) return metrics.accuracy_score(predictions, valid_y)</pre> Multinomial Naive Bayes Model
n [17]:	<pre>accuracy_count_nb = train_model(naive_bayes.MultinomialNB(), xtrain_count t, train_y, xvalid_count) print ("NB, Count Vectors: ", accuracy_count_nb) accuracy_word_nb = train_model(naive_bayes.MultinomialNB(), xtrain_tfid t, train_y, xvalid_tfidf) print ("NB, WordLevel TF-IDF: ", accuracy_word_nb) accuracy_ngram_nb = train_model(naive_bayes.MultinomialNB(), xtrain_tfid f_ngram, train_y, xvalid_tfidf_ngram) print ("NB, N-Gram Vectors: ", accuracy_ngram_nb)</pre>
າ [18]:	<pre>print ("NB, N-Gram Vectors: ", accuracy_ngram_nb) NB, Count Vectors: 0.5836909871244635 NB, WordLevel TF-IDF: 0.4592274678111588 NB, N-Gram Vectors: 0.5793991416309013 Random Forest Classifier accuracy_count_rf = train_model(RandomForestClassifier(n_estimators=100))</pre>
21:	<pre>accuracy_count_rf = train_model(RandomForestClassifier(n_estimators=100),xtrain_count,train_y,xvalid_count) print("RF, Count Vectors: ",accuracy_count_rf) accuracy_word_rf = train_model(RandomForestClassifier(n_estimators=100) xtrain_tfidf,train_y,xvalid_tfidf) print("RF, WordLevel TF-IDF Vectors: ",accuracy_word_rf) accuracy_ngram_rf = train_model(RandomForestClassifier(n_estimators=100),xtrain_tfidf_ngram,train_y,xvalid_tfidf_ngram) print("RF, N-Gram Vectors: ",accuracy_ngram_rf)</pre>
n [19]:	RF, Count Vectors: 0.630901287553648 RF, WordLevel TF-IDF Vectors: 0.6437768240343348 RF, N-Gram Vectors: 0.6180257510729614 Logistic Regression Model accuracy_count_lc = train_model(linear_model.LogisticRegression(), xtrain_count, train_y, xvalid_count)
	<pre>n_count, train_y, xvalid_count) print ("LR, Count Vectors: ", accuracy_count_lc) accuracy_word_lc = train_model(linear_model.LogisticRegression(), xtrain_tfidf, train_y, xvalid_tfidf) print ("LR, WordLevel TF-IDF: ", accuracy_word_lc) accuracy_ngram_lc = train_model(linear_model.LogisticRegression(), xtrain_tfidf_ngram, train_y, xvalid_tfidf_ngram) print ("LR, N-Gram Vectors: ", accuracy_ngram_lc)</pre>
ո [20]։	<pre>LR, Count Vectors: 0.6351931330472103 LR, WordLevel TF-IDF: 0.6266094420600858 LR, N-Gram Vectors: 0.5836909871244635 XGBoost Classifier accuracy_count_bo = train_model(xgboost.XGBClassifier(), xtrain_count.tecsc(), train_y, xvalid_count.tocsc())</pre>
	<pre>csc(), train_y, xvalid_count.tocsc()) print ("Xgb, Count Vectors: ", accuracy_count_bo) accuracy_word_bo = train_model(xgboost.XGBClassifier(), xtrain_tfidf.tocsc(), train_y, xvalid_tfidf.tocsc()) print ("Xgb, WordLevel TF-IDF: ", accuracy_word_bo) accuracy_ngram_bo = train_model(xgboost.XGBClassifier(), xtrain_tfidf_ngram.tocsc(), train_y, xvalid_tfidf_ngram.tocsc()) print ("Xgb, Ngram Level Vectors: ", accuracy_ngram_bo) [14:36:10] WARNING: C:/Users/Administrator/workspace/xgboost-win64_relean</pre>
	[14:36:10] WARNING: C:/Users/Administrator/workspace/xgboost-win64_release_1.3.0/src/learner.cc:1061: Starting in XGBoost 1.3.0, the default evaluation metric used with the objective 'multi:softprob' was changed from 'merror' to 'mlogloss'. Explicitly set eval_metric if you'd like to restore the old behavior. Xgb, Count Vectors: 0.6866952789699571 [14:36:12] WARNING: C:/Users/Administrator/workspace/xgboost-win64_release_1.3.0/src/learner.cc:1061: Starting in XGBoost 1.3.0, the default evaluation metric used with the objective 'multi:softprob' was changed from 'merror' to 'mlogloss'. Explicitly set eval_metric if you'd like to restore the old behavior.
	ore the old behavior. Xgb, WordLevel TF-IDF: 0.630901287553648 [14:36:16] WARNING: C:/Users/Administrator/workspace/xgboost-win64_release_1.3.0/src/learner.cc:1061: Starting in XGBoost 1.3.0, the default evaluation metric used with the objective 'multi:softprob' was changed from 'merror' to 'mlogloss'. Explicitly set eval_metric if you'd like to restore the old behavior. Xgb, Ngram Level Vectors: 0.5665236051502146
n [21]:	<pre>Bagging - Random Forest accuracy_count_bg = train_model(ensemble.RandomForestClassifier(), xtrain_count, train_y, xvalid_count) print ("RF, Count Vectors: ", accuracy_count_bg) accuracy_word_bg = train_model(ensemble.RandomForestClassifier(), xtrain_tfidf, train_y, xvalid_tfidf) print ("RF, WordLevel TF-IDF: ", accuracy_word_bg)</pre>
n [22]:	
	<pre>accuracy_count_svm = train_model(LinearSVC(multi_class="ovr"), xtrain_count, train_y, xvalid_count) print ("SVM, Count Vectors: ", accuracy_count_svm) accuracy_word_svm = train_model(LinearSVC(multi_class="ovr"), xtrain_tfdf, train_y, xvalid_tfidf) print ("SVM, WordLevel TF-IDF: ", accuracy_word_svm) accuracy_ngram_svm = train_model(LinearSVC(multi_class="ovr"), xtrain_traidf_ngram, train_y, xvalid_tfidf_ngram)</pre>
n [23]:	print ("SVM, Ngram Vectors: ", accuracy_ngram_svm) SVM, Count Vectors: 0.630901287553648 SVM, WordLevel TF-IDF: 0.6523605150214592 SVM, Ngram Vectors: 0.6223175965665236 import matplotlib as mpl import numpy as np
	<pre>def plot_cm(X, y, clf, title): cm = metrics.confusion_matrix(y, clf.predict(X)) print(metrics.classification_report(y, clf.predict(X))) np.set_printoptions(suppress=True) mpl.rc("figure", figsize=(10,10)) hm = sns.heatmap(cm,</pre>
	<pre>square=True, fmt='d', yticklabels=['happy','sad','angry','relaxed'], xticklabels=['happy','sad','angry','relaxed'], cmap='Blues') plt.title(title) plt.ylabel('actual class') plt.xlabel('predicted class') plt.tight_layout()</pre>
n [24]:	<pre>#plt.savefig("rf_tfidf.jpg") plt.show()</pre>
	<pre>import matplotlib.pyplot as plt claf=ensemble.RandomForestClassifier() claf.fit(xtrain_tfidf, train_y) plot_cm(xvalid_tfidf, valid_y, claf, 'Training Word Level - TfidfVectors</pre>
	<pre>import matplotlib.pyplot as plt claf=ensemble.RandomForestClassifier() claf.fit(xtrain_tfidf,train_y) plot_cm(xvalid_tfidf, valid_y, claf, 'Training Word Level - TfidfVector:</pre>
	<pre>import matplotlib.pyplot as plt claf=ensemble.RandomForestClassifier() claf.fit(xtrain_tfidf,train_y) plot_cm(xvalid_tfidf, valid_y, claf, 'Training Word Level - TfidfVector:</pre>
	import matplotlib.pyplot as plt claf=ensemble.RandomForestClassifier() claf.fit(xtrain_tfidf,train_y) plot_cm(xvalid_tfidf, valid_y, claf, 'Training Word Level - TfidfVector: zer on LR') precision recall f1-score support 0 0.73 0.63 0.67 97 1 0.50 0.85 0.63 68 2 1.00 0.33 0.50 12 3 0.79 0.41 0.54 56 accuracy 0.63 233 macro avg 0.75 0.56 0.59 233 weighted avg 0.69 0.63 0.62 233 Training Word Level - TfidfVectorizer on LR Training Word Level - TfidfVectorizer on LR
	import matplotlib.pyplot as plt claf=ensemble.RandomForestClassifier() claf.fit(xtrain_tfidf,train_y) plot_cm(xvalid_tfidf, valid_y, claf, 'Training Word Level - TfidfVector: zer on LR') precision recall f1-score support 0 0.73 0.63 0.67 97 1 0.50 0.85 0.63 68 2 1.00 0.33 0.50 12 3 0.79 0.41 0.54 56 accuracy macro avg 0.75 0.56 0.59 233 weighted avg 0.69 0.63 0.62 233 Training Word Level - TfidfVectorizer on LR
	import matplotlib.pyplot as plt claf=ensemble.RandomForestClassifier() claf.fit(xtrain_tfidf,train_y) plot_cm(xvalid_tfidf, valid_y, claf, 'Training Word Level - TfidfVector.zer on LR') precision recall f1-score support 0 0.73 0.63 0.67 97 1 0.50 0.85 0.63 68 2 1.00 0.33 0.50 12 3 0.79 0.41 0.54 56 accuracy 0.63 233 macro avg 0.75 0.56 0.59 233 weighted avg 0.69 0.63 0.62 233 Training Word Level - TfidfVectorizer on LR Training Word Level - TfidfVectorizer on LR
n [25]:	import matplotlib.pyplot as plt claf=ensemble.RandomForestClassifier() claf.fit(xtrain_tfaidf,train_y) plot_cm(xvalid_tfidf, valid_y, claf, 'Training Word Level - TfidfVector.zer on LR') precision recall f1-score support 0 0.73 0.63 0.67 97 1 0.50 0.85 0.63 68 2 1.00 0.33 0.59 12 3 0.79 0.41 0.54 56 accuracy macro avg 0.75 0.56 0.59 233 weighted avg 0.69 0.63 0.62 233 Training Word Level - TfidfVectorizer on LR Training Word Level - TfidfVectorizer on LR # import seaborn as sns import matplotlib.pyplot as plt claf=linear_model.LogisticRegression() claf.fit(xtrain_count, train_y) plot_cm(xvalid_count, valid_y, claf, 'Training CountVectorizer on LR') precision recall f1-score support
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Maine Roya song by Tanveer Evan

Tag: Sad

Tag: Sad

Tag: Sad

In [36]: pred("I cried: I cried you in the mist of mist, I didn't get you to slee
p in these nights. I've never thought of rotha in your memories. I never
thought you just cried.")

In [37]: pred1("Maine roya tujhe dhund dhunde Royatha main Tu na milaa Gehrisi in raaton mein soyaa naa tha main Rotatha teri yaadon mein woh hi raat")

In [38]: pred1("Jo dil mein bhara tune Dekhegi uss zeher ko \nBhugtegi mere gham
 ko \nMeri aah ke kahar ko\nApni khudgarzi ka ab anzaam dekhegi")

Major Project : Music Mood Prediction

In [1]: %matplotlib inline import numpy as np import pandas as pd import matplotlib.pyplot as plt

import seaborn as sns

warnings.filterwarnings("ignore")

Part 2 : Data Pre-processing

import warnings import requests import json

Krishi Agrahari, Rajsi Kesharwani, Nikhil Kamale, Kirti Mohitkar, Shazia Khan