The aim of this notebook is to make use of the word2vec model to find similar songs

## WORD2VEC - Exploracion con un Corpus de canciones en Español usando tokenizer y stopwords para crear el corpus

Esto va a ser menos exacto que los anteriores porque hay menos palabras. Pero es interesante verlo.

```
In [34]: 
import pandas as pd
import numpy as np
import gensim.models.word2vec as w2v
import multiprocessing
import os
import re
import pprint
import sklearn.manifold
import matplotlib.pyplot as plt
```

Cargamos las el quijote

Limpiamos los caractes raros y quitamos acentos

To train the word2vec model, we first need to build its vocabulary. To do that, I iterated over each song and added it to an array that can later be fed to the model.

## Usamos tokenizer y stopwords para tener un mejor vocabulario

```
import nltk
from nltk.corpus import stopwords

text_corpus = []
for line in clean_file:
    tokenizer = nltk.tokenize.RegexpTokenizer(r'\w+') #para dividir por words y quitar puntuacion
    lower_case = line.lower()
    tokens_sin_puntuacion = tokenizer.tokenize(lower_case)
    tokens = [i for i in tokens_sin_puntuacion if (len(i)>1) and i not in stopwords.words('spanish')] #quitamos stop

text_corpus.append(tokens)
```

```
In [100]: ▶ # Dimensionality of the resulting word vectors.
              #more dimensions, more computationally expensive to train
              #but also more accurate
              #more dimensions = more generalized
              num features = 50
              # Minimum word count threshold.
              min_word_count = 1
              # Number of threads to run in parallel.
              #more workers, faster we train
              num workers = multiprocessing.cpu count()
              # Context window Length.
              context_size = 7
              downsampling = 1e-1
              # Seed for the RNG, to make the results reproducible.
              #random number generator
              #deterministic, good for debugging
              seed = 1
              songs2vec = w2v.Word2Vec(
                  sg=1,
                  seed=seed,
                  workers=num_workers,
                  size=num_features,
                  min_count=min_word_count,
                  window=context size,
                  sample=downsampling
              songs2vec.build vocab(text corpus)
              print (len(text_corpus))
```

32339

## Entrenamiento:

```
In [101]: | import time
    start_time = time.time()

songs2vec.train(text_corpus, total_examples=songs2vec.corpus_count, epochs=2)
    if not os.path.exists("trained"):
        os.makedirs("trained")

songs2vec.save(os.path.join("trained", "songs2vectors.w2v"))

print("--- %s seconds ---" % (time.time() - start_time))

--- 0.6159968376159668 seconds ---
In [102]: | songs2vec = w2v.Word2Vec.load(os.path.join("trained", "songs2vectors.w2v"))
```

## Let's explore our model

Find similar words

```
■ songs2vec.wv.most similar("sancho")
In [105]:
   Out[105]: [('asi', 0.9668811559677124),
               ('quijote', 0.9574342966079712),
               ('cura', 0.9552509784698486),
               ('pues', 0.945313036441803),
               ('decir', 0.9427224397659302),
               ('senora', 0.9385666847229004),
               ('si', 0.9359698295593262),
               ('verdad', 0.9331722259521484),
               ('habia', 0.9328069686889648),
               ('bien', 0.9322810769081116)]
In [106]:
              songs2vec.wv.most similar("caballo")
   Out[106]: [('pie', 0.9992693066596985),
               ('palabra', 0.999201774597168),
               ('mesmo', 0.9991865754127502),
               ('primo', 0.9990447759628296),
               ('cuenta', 0.998988687992096),
               ('nunca', 0.9989795088768005),
               ('lugar', 0.9989739656448364),
               ('iba', 0.9988583326339722),
               ('camino', 0.9988481998443604),
               ('ama', 0.9988390207290649)]
              songs2vec.wv.most similar("camino")
In [107]:
   Out[107]: [('mismo', 0.9994622468948364),
               ('mucha', 0.9994367957115173),
               ('canonigo', 0.9994223713874817),
               ('primero', 0.9993078708648682),
               ('fin', 0.9992329478263855),
               ('entender', 0.9992038607597351),
               ('nunca', 0.9991317391395569),
               ('libro', 0.9990227222442627),
               ('tenia', 0.9989996552467346),
               ('contento', 0.9989832043647766)]
```

```
In [110]:

■ songs2vec.wv.most similar("libro")
   Out[110]: [('dijese', 0.9994610548019409),
              ('aventura', 0.9994574189186096),
              ('mucha', 0.9993853569030762),
              ('entender', 0.9993103742599487),
              ('mismo', 0.9992714524269104),
              ('duena', 0.9992218613624573),
              ('desa', 0.9992120265960693),
              ('nombre', 0.9991418123245239),
              ('adonde', 0.9991415739059448),
              ('oyo', 0.9991327524185181)]
         Words out of context
          songs2vec.wv.doesnt match("sancho quijote aventura".split())
In [111]:
             C:\Users\jhern\Anaconda3\lib\site-packages\gensim\models\keyedvectors.py:877: FutureWarning: arrays to stack must b
             e passed as a "sequence" type such as list or tuple. Support for non-sequence iterables such as generators is depre
             cated as of NumPy 1.16 and will raise an error in the future.
               vectors = vstack(self.word vec(word, use norm=True) for word in used words).astype(REAL)
   Out[111]: 'aventura'
          In [117]:
   Out[117]: 'camino'
          songs2vec.wv.doesnt_match("comida bebida árbol".split()) #se equivoca
In [120]:
   Out[120]: 'bebida'
```

```
In [121]:
          songs2vec.most similar(positive=['sancho', 'panza'], negative=['quijote'])
              #queen
              C:\Users\jhern\Anaconda3\lib\site-packages\ipykernel launcher.py:1: DeprecationWarning: Call to deprecated `most si
              milar` (Method will be removed in 4.0.0, use self.wv.most similar() instead).
                """Entry point for launching an IPython kernel.
   Out[121]: [('vuesa', 0.9589300751686096),
               ('mio', 0.95602947473526),
               ('digo', 0.956020176410675),
               ('senor', 0.9536997079849243),
               ('decir', 0.9526917934417725),
               ('respondio', 0.9521059989929199),
               ('dijo', 0.9520894885063171),
               ('bien', 0.9513685703277588),
               ('dice', 0.9496554136276245),
               ('dios', 0.9490358829498291)]
          Semantic distance between words
           def nearest_similarity_cosmul(start1, end1, end2):
In [122]:
                  similarities = songs2vec.wv.most similar cosmul(
                      positive=[end2, start1],
                      negative=[end1]
                  start2 = similarities[0][0]
                  print("{0} es a {1}, lo que {2} es a {3}".format(start1, end1, start2, end2))
In [124]:
           ▶ nearest similarity cosmul("sancho", "panza", "don") #bien
              sancho es a panza, lo que quijote es a don
           ▶ nearest similarity cosmul("libro", "historia", "camino") #mas o menos
In [125]:
              libro es a historia, lo que llegando es a camino
```