Applying word2vec

```
Import data and packages:
In [ ]: #import data frames with cleaned data from previous notebook
        %store -r textdata
        %store -r titledata
In [ ]: #import packages
        import pandas as pd
        import numpy as np
        from sklearn.model_selection import train_test_split
        import gensim
        from gensim.models import Word2Vec
In [ ]: #import pre-trained word2vec model trained on Google News data
        import gensim.downloader as api
        wv = api.load('word2vec-google-news-300')
        Apply word2vec to text column:
In [ ]: #split into train and test sets
        text_train, text_test = train_test_split(
           textdata,
            test_size=0.3,
           random_state=42
```

In []:	#add column to dataframes containing average vector for each row's text
	<pre>text_train['vectors'] = text_train['text'].apply(lambda x: wv.get_mean_vector(x, ignore_missing=True))</pre>
In []:	<pre>text_test['vectors'] = text_test['text'].apply(lambda x: wv.get_mean_vector(x, ignore_missing=True))</pre>
In []:	#view results
	text_train

	text	label	vectors
7768	[swirl, revelation, allegation, russian, invol	0	[-0.0003381749, 0.004170802, 8.591608e-05, 0.0
20151	[despite, problem, plague, donald, trump, fail	1	[0.0077793393, 0.019553736, 0.0042430665, 0.03
1003	[tune, alternate, current, radio, network, acr	1	[0.0028786482, 0.008605114, 0.0070612496, 0.03
60898	[kiev, reuters, ukrainian, president, petro, p	0	[-0.02019227, -0.0072652902, 0.0150532145, 0.0
18689	[madman, merkel, demand, internet, publicly, r	1	[0.0044224965, 0.002993722, 0.005907909, 0.022
•••			
66235	[aboard, aquarius, rescue, ship, reuters, migr	0	[0.011924975, 0.010988319, 0.0011468495, 0.025
44512	[moscow, reuters, senior, russian, lawmaker, s	0	[-0.006364392, 0.0013059269, 0.018351823, 0.01
921	[trump, want, win, rust, belt, need, practice,	1	[0.0034884035,0.010201223,0.00054162065,0.0
17458	[dubai, reuters, former, egyptian, prime, mini	0	[-0.0059219073, 0.0037472998, 0.011917742, 0.0
69180	[tim, kaine, cheer, end, white, majority, span	1	[-0.00827276, 0.0054527237, 0.01909309, 0.0335

Out[]:

```
In [ ]: text_test
Out[ ]:
                                                      text label
                                                                                                     vectors
          33113
                  [new, york, reuters, michael, moore, left-wing...
                                                              0 [-0.00028460205, 0.0022248349, -0.0014604458, ...
                                                              1 [0.021414263, 0.006217401, 0.010361887, 0.0372...
          56520
                    [catholic, church, decade, long, likely, even,...
          24790
                                                              1 [0.008622794, 0.010029217, -0.0054355743, 0.02...
                    [megyn, kelly, fox, news, anchor, definitely, ...
                   [peter, certo, wordseven, election, year, shoo...
           5431
                                                              1 [0.007797616, 0.005972583, 0.013798478, 0.0367...
          16964
                    [something, interest, unz, review, recipient, ...
                                                              1 [0.009118676, 0.008844791, 0.008179908, 0.0458...
          38902
                    [australia, stop, donation, corrupt, clinton, ...
                                                              1 [-0.006731601, -0.0014126656, 0.0031150207, 0....
          48226
                                                              0 [0.005291016, 0.004906544, 0.01436485, 0.02873...
                   [ten, progressive, judge, virginia, decide, mu...
          19398
                     [fort, lauderdale, fla, reuters, florida, gove...
                                                              0 [-0.005333254, 0.006447841, 0.0089601865, 0.02...
                                                              1 [0.001132972, 0.009400117, 0.009536006, 0.0369...
           4343
                   [donald, trump, stun, neglect, disavow, nazi, ...
                                                              1 [-0.0077695693, 0.011566349, 0.008971804, 0.02...
          36825 [president, trump, nominate, christopher, wray...
         17567 rows × 3 columns
In [ ]: #split train and test data into x and y
          x_text_train_wv = text_train.vectors
          x_text_test_wv = text_test.vectors
          y_text_train_wv = text_train.label
          y_text_test_wv = text_test.label
In [ ]: # make vector arrays usable for modelling
          x_text_train_wv_2d = np.stack(x_text_train_wv)
          x_text_test_wv_2d = np.stack(x_text_test_wv)
In [ ]:
          #store variables for later use
          %store x_text_train_wv_2d
          %store x_text_test_wv_2d
          %store y_text_train_wv
          %store y_text_test_wv
          Stored 'x_text_train_wv_2d' (ndarray)
          Stored 'x_text_test_wv_2d' (ndarray)
          Stored 'y_text_train_wv' (Series)
          Stored 'y_text_test_wv' (Series)
          Apply word2vec to title column:
In [ ]: #split title data into training and test sets
          title_train, title_test = train_test_split(
              titledata,
              test_size=0.3,
              random_state=42
          title_test['vectors'] = title_test['title'].apply(lambda x: wv.get_mean_vector(x, ignore_missing=True))
         title_train['vectors'] = title_train['title'].apply(lambda x: wv.get_mean_vector(x, ignore_missing=True))
In [ ]:
In [ ]:
          #split train and test data into x and y
          x_title_train_wv = title_train.vectors
          x_title_test_wv = title_test.vectors
          y title train wv = title train.label
          y_title_test_wv = title_test.label
In [ ]: # make vector arrays usable for modelling
```

x_title_train_wv_2d = np.stack(x_title_train_wv)

Stored 'y_title_test_wv' (Series)