

Univerzitet u Sarajevu

Elektrotehnički fakultet

Dubinska analiza podataka

Laboratorijska vježba 7. - Metode Predstavljanja Teksta

Izrada laboratorijske vježbe vrši se u ovom *Jupyter Notebook*-u. Isti je potrebno konvertovati u PDF dokument i predati na email adresu ekrupalija1@etf.unsa.ba.

Ime i prezime studenta, broj indeksa:

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Datum izrade izvještaja:

10.5.2021.

Za pomoć pri izradi izvještaja koristiti materijale za rad na predmetu *Predavanje 8*.

```
In [79]: !pip install ast
!pip install gensim==3.6.0
!pip install torch
!pip install transformers
```

Collecting ast

Using cached <https://files.pythonhosted.org/packages/4b/fb/2b954d2a38c9a0ef1da6a46737a75b4dbf6f60e5dad0f267a4ec5ece20de/AST-0.0.2.tar.gz>

ERROR: Command errored out with exit status 1: `python setup.py egg_info` Check the logs f or full command output.

Requirement already satisfied: gensim==3.6.0 in /usr/local/lib/python3.7/dist-packages (3.6.0)

Requirement already satisfied: numpy>=1.11.3 in /usr/local/lib/python3.7/dist-packages (from gensim==3.6.0) (1.19.5)

Requirement already satisfied: scipy>=0.18.1 in /usr/local/lib/python3.7/dist-packages (from gensim==3.6.0) (1.4.1)

Requirement already satisfied: six>=1.5.0 in /usr/local/lib/python3.7/dist-packages (from gensim==3.6.0) (1.15.0)

Requirement already satisfied: smart-open>=1.2.1 in /usr/local/lib/python3.7/dist-packages (from gensim==3.6.0) (5.0.0)

Requirement already satisfied: torch in /usr/local/lib/python3.7/dist-packages (1.8.1+cu101)

Requirement already satisfied: typing-extensions in /usr/local/lib/python3.7/dist-packages (from torch) (3.7.4.3)

Requirement already satisfied: numpy in /usr/local/lib/python3.7/dist-packages (from torch) (1.19.5)

Requirement already satisfied: transformers in /usr/local/lib/python3.7/dist-packages (4.5.1)

Requirement already satisfied: regex!=2019.12.17 in /usr/local/lib/python3.7/dist-packages (from transformers) (2019.12.20)

Requirement already satisfied: packaging in /usr/local/lib/python3.7/dist-packages (from transformers) (20.9)

Requirement already satisfied: filelock in /usr/local/lib/python3.7/dist-packages (from transformers) (3.0.12)

Requirement already satisfied: sacremoses in /usr/local/lib/python3.7/dist-packages (from transformers) (0.0.45)

Requirement already satisfied: importlib-metadata; python_version < "3.8" in /usr/local/lib/python3.7/dist-packages (from transformers) (3.10.1)

Requirement already satisfied: requests in /usr/local/lib/python3.7/dist-packages (from transformers) (2.23.0)

Requirement already satisfied: numpy>=1.17 in /usr/local/lib/python3.7/dist-packages (from transformers) (1.19.5)

Requirement already satisfied: tokenizers<0.11,>=0.10.1 in /usr/local/lib/python3.7/dist-packages (from transformers) (0.10.2)

Requirement already satisfied: tqdm>=4.27 in /usr/local/lib/python3.7/dist-packages (from transformers) (4.41.1)

Requirement already satisfied: pyparsing>=2.0.2 in /usr/local/lib/python3.7/dist-packages (from transformers) (2.4.7)

Requirement already satisfied: click in /usr/local/lib/python3.7/dist-packages (from sacremoses->transformers) (7.1.2)

Requirement already satisfied: joblib in /usr/local/lib/python3.7/dist-packages (from sacremoses->transformers) (1.0.1)

Requirement already satisfied: six in /usr/local/lib/python3.7/dist-packages (from sacremoses->transformers) (1.15.0)

Requirement already satisfied: zipp>=0.5 in /usr/local/lib/python3.7/dist-packages (from importlib-metadata; python_version < "3.8"->transformers) (3.4.1)

Requirement already satisfied: typing-extensions>=3.6.4; python_version < "3.8" in /usr/local/lib/python3.7/dist-packages (from importlib-metadata; python_version < "3.8"->transformers) (3.7.4.3)

Requirement already satisfied: idna<3,>=2.5 in /usr/local/lib/python3.7/dist-packages (from requests->transformers) (2.10)

Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.7/dist-packages (from requests->transformers) (2020.12.5)

Requirement already satisfied: urllib3!=1.25.0,!1.25.1,<1.26,>=1.21.1 in /usr/local/lib/python3.7/dist-packages (from requests->transformers) (1.24.3)

Requirement already satisfied: chardet<4,>=3.0.2 in /usr/local/lib/python3.7/dist-packages (from requests->transformers) (3.0.4)

Zadatak 1. - Učitavanje *dataseta*

U okviru ovog zadatka potrebno je izvršiti učitavanje *dataseta* u varijablu nad kojom će se izvršiti njegovo dalje pripremanje za korištenje metoda mašinskog učenja koristeći različite metode predstavljanja teksta.

Zadatak:

Preuzeti *dataset* koji se nalazi na [sljedećem linku](#).

Dataset ima sljedeći oblik:

	_unit_id	tokens	pos_tags	no_of_occurrences
0	842613455	[interest, new, two, bill, latest, citibank, p...	(NN, JJ, CD, NN, JJS, NN, NN, NN, VBN, NN, VBN...	[1, 2, 1, 1, 1, 1, 2, 1, 2, 1, 1, 1, 1, 1, 2, ...
1	842613456	[new, wait, email, aim, privat, recess, low, s...	(JJ, NN, NN, NN, NN, NN, JJ, VBD, JJ, JJ, NN, ...	[6, 1, 2, 1, 3, 1, 1, 1, 1, 1, 1, 1, 2, 1, 1, ...
2	842613457	[give, latest, secur, regul, tighten, began, p...	(VB, JJS, NN, NN, NN, VBD, JJ, NN, CD, JJ, NN, ...	[1, 1, 4, 2, 1, 1, 1, 1, 1, 1, 1, 1, 3, 1, 1, ...
3	842613458	[prefer, turnov, howev, result, requir, defam,...	(NN, NN, NN, NN, NN, NN, NN, FW, NN, NN, NN, V...	[1, 1, 1, 3, 1, 1, 2, 4, 1, 2, 2, 1, 1, 1, 1, ...
4	842613459	[soft, new, make, unchang, breath, day, indeci...	(JJ, JJ, NN, JJ, NN, NN, VB, NN, NN, NNS, VBP, ...	[1, 4, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, ...
...

Potrebno je učitati samo `tokens` i `no_of_occurences` kolone, jer u ovom koraku preprocesiranja podataka više nije neophodno koristiti ID instanci, kao ni POS tagove. Voditi računa o tome da se podaci ne učitaju kao jedan veliki string, nego kao lista stringova, na isti način na koji su bili reprezentirani prije spašavanja u *file* u prethodnoj laboratorijskoj vježbi.

Ispisati cijeli *dataset* ili prvi red podataka nakon učitavanja.

```
In [80]: from ast import literal_eval
import pandas as pd
ds = pd.read_csv('preprocessed_text.csv')
ds = ds[['tokens', 'no_of_occurences']]
ds['tokens'] = ds['tokens'].apply(literal_eval)
ds['no_of_occurences'] = ds['no_of_occurences'].apply(literal_eval)
```

```
In [81]: print(ds)
print(ds['tokens'][0], ds['no_of_occurences'][0])
```

```

                                tokens
no_of_occurences
0      [new, york, yield, certif, deposit, offer, maj...  [2, 2, 3, 1, 3, 1, 1, 3, 1, 1,
2, 2, 1, 4, 1, ...
1      [the, wall, street, journal, onlineth, morn, b...  [3, 1, 1, 1, 1, 1, 1, 1, 2, 1,
1, 2, 1, 1, 1, ...
2      [washington, in, effort, achiev, bank, reform,...  [1, 1, 2, 1, 14, 1, 3, 1, 1, 3,
1, 2, 3, 1, 4,...
3      [the, statist, enorm, cost, employe, abus, wel...  [3, 1, 1, 2, 6, 2, 1, 1, 3, 1,
1, 2, 1, 1, 1, ...
4      [new, york, indecis, mark, dollar, tone, trade...  [4, 4, 1, 1, 4, 1, 1, 1, 1, 1,
1, 1, 2, 2, 2, ...
...
...
7995 [secretari, commerc, charl, w, sawyer, said, y...  [1, 1, 1, 1, 1, 6, 1, 1, 1, 1,
1, 1, 1, 1, 1, ...
7996 [us, stock, inch, last, week, overcom, concern...  [3, 1, 1, 1, 2, 1, 1, 2, 1, 1,
1, 1, 1, 1, 1, ...
7997 [ben, s, bernank, clear, key, hurdl, thursday,...  [1, 1, 9, 1, 1, 1, 1, 6, 1, 3,
1, 1, 4, 1, 7, ...
7998 [the, white, hous, push, contract, mani, feder...  [1, 1, 1, 1, 1, 1, 2, 1, 1, 3,
2, 1, 1, 3, 1, ...
7999 [new, york, april, 17automobil, stock, put, be...  [1, 1, 2, 1, 2, 1, 1, 1, 1, 1,
1, 1, 1, 2, 1, ...

[8000 rows x 2 columns]
['new', 'york', 'yield', 'certif', 'deposit', 'offer', 'major', 'bank', 'drop', 'tenth',
'percentag', 'point', 'latest', 'week', 'reflect', 'overall', 'declin', 'shortterm', 'int
erest', 'rateson', 'smalldenomin', 'consum', 'cd', 'sold', 'directli', 'averag', 'sixmon
th', 'fell', '549', '562', 'end', 'yesterday', 'accord', '18bank', 'survey', 'banxquot',
'money', 'market', 'wilmington', 'del', 'inform', 'serviceon', 'threemonth', 'sank', '52
9', '542', 'two', 'citibank', 'corest', 'pennsylvania', 'pay', 'less', '5', 'cdsdeclin',
'somewhat', 'smaller', 'fiveyear', 'eas', '737', '745', 'saidyield', 'treasuri', 'bill',
'monday', 'auction', 'plummet', 'fifth', 'previou', '546', '563', 'respect'] [2, 2, 3,
1, 3, 1, 1, 3, 1, 1, 2, 2, 1, 4, 1, 1, 1, 1, 1, 2, 3, 2, 2, 1, 2, 2, 1, 1, 1, 1,
2, 1, 2, 4, 1, 1, 1, 1, 1, 3, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,
1, 1, 1, 1, 1, 1, 1, 1, 1]
```

Zadatak 2. - VSM (Vector Space Model)

U okviru ovog zadatka potrebno je izvršiti predstavljanje prethodno učitanih podataka koristeći model vektorskog prostora.

Zadatak:

Za prvih 10 redova podataka iz *dataseta* formirati VSM prikaz za sve jedinstvene riječi. Potrebno je razmatrati sve podatke zajedno tj. kreirati samo jedan model za svih 10 redova podataka iz *dataseta*.

Kako su podaci za svaki red već prikazani u obliku tokena, potrebno je izvršiti njihovo spajanje u jedinstvenu listu tokena, a zatim primijeniti VSM kako bi se za svaki red podataka dobio *one-hot* vektor podataka.

```
In [82]: docs = sum(ds['tokens'][0:10], [])
```

```
In [83]: from numpy import array
from numpy import argmax
from sklearn.preprocessing import OneHotEncoder
from sklearn.preprocessing import LabelEncoder
values = array(docs)
encoded = LabelEncoder().fit_transform(values)
print(encoded)
encoded = encoded.reshape(len(encoded), 1)
onehot = OneHotEncoder(sparse=False).fit_transform(encoded)
print(onehot)
```

```
[581 945 944 ... 35 941 34]
[[0. 0. 0. ... 0. 0. 0.]
 [0. 0. 0. ... 1. 0. 0.]
 [0. 0. 0. ... 0. 0. 0.]
 ...
 [0. 0. 0. ... 0. 0. 0.]
 [0. 0. 0. ... 0. 0. 0.]
 [0. 0. 0. ... 0. 0. 0.]
```

```
In [84]: d = {}
for i in range(0, len(values)):
    d[values[i]] = onehot_encoded[i]
for i in range(0, 10):
    s = []
    for j in ds['tokens'][i]:
        s.append(onehot[numpy.where(values==j)])
    print(s)
    print()
```

Streaming output truncated to the last 5000 lines.

```
0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0.,
0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0.,
0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0.,
0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0.,
0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0.,
0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0.,
0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0.,
0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0.,
0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0.,
0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0.,
0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0.,
0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0.,
0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0.,
0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0.,
```

$\theta., \theta., \theta.,$

$\theta., \theta., \theta.,$

$\theta., \theta., \theta.,$

$\theta., \theta., \theta.,$

$\theta., \theta., \theta.,$

$$0., 0., 0., 0., 0., 0.,$$

$\theta., \theta., \theta.,$

$\theta., \theta., \theta.,$

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```

0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0.,
0., 0., 0., 0.]], array([[0., 0., 0., ..., 0., 0., 0.],
[0., 0., 0., ..., 0., 0., 0.],
[0., 0., 0., ..., 0., 0., 0.],
[0., 0., 0., ..., 0., 0., 0.],
[0., 0., 0., ..., 0., 0., 0.],
[0., 0., 0., ..., 0., 0., 0.])), array([[0., 0., 0., 0., 0., 0.,
, 0., 0., 0., 0.,

```

$\theta., \theta., \theta.,$

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$\theta., \theta., \theta.,$

$\theta., \theta., \theta.,$

$0., 0., 0.,$

$\theta., \theta., \theta.,$

$\theta., \theta., \theta.,$

$\theta., \theta., \theta.,$

$$\theta_1, \theta_2, \theta_3, \theta_4, \theta_5, \theta_6,$$

$\theta., \theta., \theta.,$

$\theta., \theta., \theta.,$

$\theta., \theta., \theta.,$

[illegible]

```
0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0.,
0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0.,
0., 0., 0., 0.])), array([[0., 0., 0., ..., 0., 0., 0.],
[0., 0., 0., ..., 0., 0., 0.],
[0., 0., 0., ..., 0., 0., 0.],
[0., 0., 0., ..., 0., 0., 0.],
[0., 0., 0., ..., 0., 0., 0.],
[0., 0., 0., ..., 0., 0., 0.])), array([[0., 0., 0., 0., 0., 0.,
0., 0., 0., 0., 0.,
0., 0., 0., 0., 0.,
```

```
0., 0., 0., 0.])), array([[0., 0., 0., ..., 0., 0., 0.],
```

[illegible]

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$$\theta., \theta., \theta., \theta., \theta., \theta.,$$

$\theta., \theta., \theta.,$

$\theta., \theta., \theta.,$

$\theta., \theta., \theta.,$

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$\theta., \theta., \theta.,$

$\theta., \theta., \theta.,$

$\theta., \theta., \theta.,$

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$\theta., \theta., \theta.,$

$$0., 0., 0., 0., 0., 0.,$$

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$\theta., \theta., \theta.,$

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0., 0., 0.,

$\theta_1, \theta_2, \theta_3,$

$$0., 0., 0.,$$


```
[0., 0., 0., 0.])), array([[0., 0., 0., ..., 0., 0., 0.],  
[0., 0., 0., ..., 0., 0., 0.])), array([[0., 0., 0., ..., 0., 0., 0.],  
[0., 0., 0., ..., 0., 0., 0.])), array([[0., 0., 0., ..., 0., 0., 0.],  
[0., 0., 0., ..., 0., 0., 0.],  
[0., 0., 0., ..., 0., 0., 0.])), array([[0., 0., 0., ..., 0., 0., 0.],  
[0., 0., 0., ..., 0., 0., 0.],  
[0., 0., 0., ..., 0., 0., 0.])), array([[0., 0., 0., ..., 0., 0., 0.],
```

[illegible]

```
0., 0., 0., 0.])), array([[0., 0., 0., ..., 0., 0., 0.],
```

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$\theta., \theta., \theta.,$

$\theta., \theta., \theta.,$

$\theta., \theta., \theta.,$

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$\theta., \theta., \theta.,$

$\theta., \theta., \theta.,$

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$\theta., \theta., \theta.,$

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$\theta., \theta., \theta.,$

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$\theta., \theta., \theta.,$

$\theta., \theta., \theta.,$

$\theta., \theta., \theta.,$

[illegible]

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$$\theta., \theta., \theta., \theta., \theta., \theta.,$$

$\theta., \theta., \theta.,$

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[illegible]

$\theta., \theta., \theta.,$

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$\theta., \theta., \theta.,$

0., 0., 0., 0., 0., 0.,

$\theta., \theta., \theta.,$

$\theta., \theta., \theta.,$

$$0., 0., 0., 0., 0., 0.,$$

```
0., 0., 0., 0.]]])]
```

Izvršiti analizu dobivenih rezultata sa ciljem određivanja one metode koja je najbolja za primjenu nad datim *datasetom*.

Da li je ovakav prikaz podataka intuitivan? Da li je lako razumjeti šta svaki red podataka sadrži? Da li postoje neki nedostaci ovakvog pristupa (npr. da li bi izvršavanje za veliki broj redova dugo trajalo i sl.)? Detaljno objasniti sve odgovore.

- Prikaz podataka jeste intuitivan
- Lahko je razumjeti šta svaki red podataka sadrži jer se svaka riječ kodira na jednostavan način (1 na mjestu za tu riječ u rječniku)
- Nedostatak jeste to što bi izvršavanje za veliki broj redova i veliki rječnik dugo trajalo te se prostor ne koristi efikasno (samo jedna 1, a ostalo nule)

Zadatak 3. - BoW (*Bag-of-Words*)

U okviru ovog zadatka potrebno je izvršiti predstavljanje prethodno učitanih podataka koristeći model prikaza vreće riječi.

Zadatak:

Za prvih 10 redova podataka iz *dataseta* formirati BoW prikaz za sve jedinstvene riječi. Potrebno je razmatrati sve podatke zajedno tj. kreirati samo jedan model za svih 10 redova podataka iz *dataseta*.

Kako su podaci za svaki red prikazani u obliku tokena, a BoW metoda očekuje rečenice, potrebno je za svaki red podataka formirati string koji će se sastojati od svih riječi iz tog reda podataka. Kako je parcijalna rekonstrukcija moguća zbog očuvanog redoslijeda jedinstvenih riječi, potrebno je formirati string dodajući razmak između svake jedinstvene riječi, a riječi koje se ponavljaju više od jednom u datom redu podataka dodati na kraj rečenice.

Tako će se npr. za listu tokena i ponavljanja `['natural', 'language', 'processing', 'is'], [1, 2, 1, 1]` dobiti rezultni string `"natural language processing is language"`.

Nakon formiranja ovakvih stringova, potrebno je primijeniti BoW kako bi se za svaki red podataka dobio BoW vektor reprezentacije podataka.

```
In [99]: res = ds['tokens'][0:10]
no_of_occurences = ds['no_of_occurences'][0:10]
for i in range(0, 10):
    for j in range(0, len(no_of_occurences[i])):
        res[i] = res[i] + (no_of_occurences[i][j] - 1) * [res[i][j]]
```

```
In [86]: print(len(res[0]))
```

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```
In [102... from sklearn.feature_extraction.text import CountVectorizer
cv = CountVectorizer()
l = []
for x in res:
    l.append(" ".join(x))
bow_rep = cv.fit_transform(l)
for i in range(0, 10):
    print(cv.transform([l[i]]).toarray())
    print()
```

```
[[ 0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  1  0  0
  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0
  1  0  0  1  1  1  1  1  0  0  0  0  1  1  0  0  0  0  0  0  0  0  0  4
  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0
  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0
  0  0  0  0  0  0  0  1  0  4  0  0  0  0  7 10  0  0  0  0  0  0  0
  0  0  0  0  0  1  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0
  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0
  0  0  0  0  0  4  1  0  0  1  0  0  0  0  1  0  0  0  0  0  0  0  0
  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  7  0  0  0  0
  1  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0
  0  1  0  0  0  0  0  1  0  0  0  0  0  7  0  0  0  0  0  0  0  1  0  0
  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  1  0  0  0  0
  0  0  0  0  0  0  1  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0
  0  0  1  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0
  0  0  0  0  0  0  0  0  0  0  0  0  0  0  1  0  1  0  0  0  0  0  0
  1  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0
  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0
  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0
  0  0  0  0  0  0  0  0  0  1  0  0  1  0  0  0  0  0  0  0  0  0  0
  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  1  0
  0  0  0  0  0  0  1  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0
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```

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```

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0 0 0 0 0 0 0 0 0 0 0 0 16 0 0 0 0 0 0 0 1 1 0 0
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0 0 0 0 4 0 0]]
```

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```

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0 0 0 0 0 0 0 0]]

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0 0 0 0 0 0 1]]

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```

[illegible]

```

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```

```

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```



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```

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0 0 0 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
4 0 4 1 7 1 0]]

```

Izvršiti analizu dobivenih rezultata sa ciljem određivanja one metode koja je najbolja za primjenu nad datim *datasetom*.

Da li je ovakav prikaz podataka intuitivan? Da li je lako razumjeti šta svaki red podataka sadrži? Da li postoje neki nedostaci ovakvog pristupa (npr. da li bi izvršavanje za veliki broj redova dugo trajalo i sl.)? Detaljno objasniti sve odgovore. Najveću pažnju obratiti na korak dodavanja riječi koje se ponavljaju više puta na kraj stringa i kako bi se taj problem mogao riješiti.

Dodatno izvršiti usporedbu sa prethodno prikazanim metodama prikaza podataka (VSM).

- prikaz podataka je intuitivan (svaki broj je broj ponavljanja neke riječi)
- lahko je razumjeti šta svaki red sadrži (iz istog razloga kao prethodna stavka)
- nedostatak je što zanemaruje redoslijed riječi te je spor za primjenu u ovom slučaju zbog toga što se riječi trebaju dodati u jedan string. To bi se moglo riješiti kad bi se jednostavno poslao broj ponavljanja kao parametar.

Zadatak 4. - TF-IDF (*Term Frequency - Inverse Document Frequency*)

U okviru ovog zadatka potrebno je izvršiti predstavljanje prethodno učitanih podataka koristeći model prikaza frekventnosti riječi.

Zadatak:

Za prvih 10 redova podataka iz *dataseta* formirati TF-IDF prikaz za sve jedinstvene riječi. Potrebno je razmatrati sve podatke zajedno tj. kreirati samo jedan model za svih 10 redova podataka iz *dataseta*.

Kao i BoW, i TF-IDF očekuje stringove, pa je podatke kreirane u okviru prošlog zadatka potrebno ponovo iskoristiti kako bi se formirao ovaj prikaz podataka.

```

In [103... from sklearn.feature_extraction.text import TfidfVectorizer
l = []
for w in res:
    l.append(" ".join(w))

```

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0.02337096]]					

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0.04954872	0.	0.	0.	0.	0.
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0.	0.	0.	0.	0.	0.
0.	0.06663066	0.	0.	0.	0.
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0.03517093	0.0298985	0.	0.4572221	0.24619651	0.03517093
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.03517093
0.	0.	0.	0.	0.	0.
0.	0.	0.03517093	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.02615766	0.	0.
0.	0.	0.	0.	0.	0.
0.03517093	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.0298985	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.03517093	0.
0.	0.	0.	0.	0.03517093	0.
0.	0.	0.	0.	0.	0.0298985
0.	0.02325603	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.24619651	0.	0.	0.	0.	0.

0.	0.	0.	0.	0.	0.
0.	0.09302413	0.	0.	0.	0.
0.	0.	0.	0.03517093	0.	0.03517093
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.03517093	0.	0.
0.02615766	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.0298985
0.	0.	0.	0.	0.	0.
0.	0.0298985	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.02615766	0.	0.	0.	0.
0.	0.	0.	0.03517093	0.	0.
0.	0.	0.	0.	0.	0.
0.0298985	0.	0.03517093	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.03517093	0.	0.	0.	0.03517093
0.03517093	0.0298985	0.03517093	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.02615766	0.	0.	0.	0.03517093	0.
0.03517093	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.0298985	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.03517093	0.
0.	0.03517093	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.03517093	0.	0.	0.	0.
0.	0.	0.03517093	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.02615766	0.	0.	0.	0.03517093	0.
0.	0.	0.	0.	0.	0.
0.03517093	0.	0.	0.	0.	0.
0.	0.03517093	0.	0.	0.	0.24619651
0.03517093	0.	0.	0.	0.	0.
0.18310361	0.03517093	0.	0.	0.	0.03517093
0.	0.	0.02615766	0.	0.	0.
0.	0.	0.	0.	0.0298985	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.26157658	0.0298985	0.	0.	0.
0.	0.03517093	0.	0.	0.	0.
0.	0.03517093	0.	0.	0.	0.
0.	0.	0.	0.	0.03517093	0.
0.14068372	0.	0.	0.	0.	0.03517093
0.	0.0298985	0.	0.	0.	0.
0.	0.01888075	0.03517093	0.	0.	0.
0.	0.03517093	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.03517093	0.	0.	0.	0.	0.
0.	0.	0.	0.02615766	0.	0.
0.	0.	0.	0.03517093	0.03517093	0.
0.	0.	0.	0.03517093	0.09302413	0.
0.	0.	0.	0.	0.	0.
0.03517093	0.	0.	0.	0.	0.
0.03517093	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.

0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.02615766	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.]]				
[[0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.03773083	0.03773083
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.03773083	0.03773083
0.	0.	0.	0.	0.	0.03773083
0.03773083	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.03773083	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.15092333	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.03773083	0.	0.	0.	0.	0.
0.03773083	0.	0.	0.	0.	0.
0.	0.03207466	0.	0.	0.	0.
0.	0.	0.	0.11224613	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.37730833	0.	0.	0.	0.	0.
0.	0.03773083	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.02806153	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.12829862	0.	0.	0.
0.	0.03207466	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.15092333	0.	0.	0.	0.	0.
0.03773083	0.	0.	0.	0.	0.
0.	0.11224613	0.	0.	0.	0.
0.	0.	0.	0.	0.03773083	0.02806153
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.03773083	0.	0.	0.	0.	0.
0.	0.	0.	0.53316913	0.	0.
0.	0.	0.	0.	0.	0.02025498
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.03207466	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.03207466	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.03773083

0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.03207466
0.03773083	0.	0.	0.	0.	0.
0.	0.03773083	0.11224613	0.	0.	0.
0.	0.	0.	0.	0.02240536	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.03773083	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.03773083	0.	0.03773083	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.03773083	0.	0.03773083
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.22452259	0.	0.	0.03773083	0.	0.
0.	0.03207466	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.11224613	0.03207466	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.03773083	0.	0.03773083	0.	0.
0.	0.	0.03773083	0.	0.	0.
0.02806153	0.02494871	0.	0.	0.	0.
0.	0.12829862	0.02494871	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.22405355	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
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0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.03773083	0.	0.	0.	0.
0.	0.03773083	0.	0.	0.	0.
0.19643073	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.03773083
0.12829862	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.03773083
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.02806153	0.	0.	0.	0.
0.02806153	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.

0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.03207466	0.
0.	0.	0.	0.	0.	0.
0.	0.03773083	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.03773083
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.20254976	0.	0.	0.	0.
0.03773083	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.03773083	0.	0.02025498	0.	0.	0.
0.	0.03773083	0.	0.	0.	0.
0.	0.	0.	0.28061533	0.	0.
0.03773083	0.	0.	0.	0.	0.
0.15092333	0.03773083	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.11224613	0.	0.	0.
0.	0.	0.	0.	0.	0.03773083
0.	0.	0.	0.	0.	0.
0.03773083	0.	0.	0.	0.	0.
0.	0.	0.	0.03773083	0.03773083	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.12829862	0.	0.11224613	0.03207466	0.174641	0.03773083
0.]]				

Izvršiti analizu dobivenih rezultata sa ciljem određivanja one metode koja je najbolja za primjenu nad datim *datasetom*.

Da li je ovakav prikaz podataka intuitivan? Da li je lako razumjeti šta svaki red podataka sadrži? Da li postoje neki nedostaci ovakvog pristupa (npr. da li bi izvršavanje za veliki broj redova dugo trajalo i sl.)? Detaljno objasniti sve odgovore. Najveću pažnju obratiti na korak dodavanja riječi koje se ponavljaju više puta na kraj stringa i kako bi se taj problem mogao riješiti.

Dodatno izvršiti usporedbu sa prethodno prikazanim metodama prikaza podataka (VSM, BoW).

- prikaz podataka nije intuitivan
- nije lahko razumjeti šta svaki red podataka sadrži
- nedostatak pristupa je isto kao u prethodnom slučaju (moraju se riječi dodavati u niz onoliko puta koliko se ponavljaju)
- pristup je bolji od prethodnih jer uzima u obzir broj ponavljanja neke riječi u dokumentu u odnosu na broj ponavljanja te riječi u ostalim dokumentima

Zadatak 5. - CBoW (*Continuos Bag of Words*)

U okviru ovog zadatka potrebno je izvršiti predstavljanje prethodno učitanih podataka koristeći model kontinualne vreće riječi.

Zadatak:

Za prvih 10 redova podataka iz *dataseta* formirati CBoW prikaz za sve jedinstvene riječi. Potrebno je razmatrati sve podatke zajedno tj. kreirati samo jedan model za svih 10 redova podataka iz *dataseta*.

Potrebno je prvo istrenirati *word2vec* model koristeći *common_texts* testni *dataset*. Nakon toga potrebno je formirati 10-dimenzionalni vektor za svaku jedinstvenu riječ.

Za ukupno 5 riječi iz prvih 10 redova podataka prikazati koje su najbližije riječi pronađene u *common_texts* testnom *datasetu*.

```
In [104... from gensim.test.utils import common_texts
from gensim.models import Word2Vec
cmodel = Word2Vec(common_texts, size=10, window=5, min_count=1, workers=4)
# sačuvaj model
cmodel.save("tempmodel.w2v")
prikazano = []
for i in range(0, 10):
    for word in ds['tokens'][i]:
        if word in cmodel.wv.vocab:
            if word in prikazano:
                continue
            prikazano.append(word)
            print(word)
            print(cmodel.wv[word])
            print("Najsličnija riječ za riječ", word, "je", cmodel.wv.most_similar(word))
            print()
```

```
survey
[-0.0244414 -0.02775992  0.04930634  0.02116048 -0.01538798 -0.03491447
 -0.0309788  0.04308886  0.00632437 -0.00297378]
Najsličnija riječ za riječ survey je [('graph', 0.14713113009929657)]
```

```
time
[ 0.04951971 -0.04842655 -0.01076342  0.04877225 -0.03298072  0.0388741
 -0.00707307 -0.01232153 -0.03449285  0.02710408]
Najsličnija riječ za riječ time je [('response', 0.7158101797103882)]
```

```
human
[-0.0225958  0.0050643 -0.00627972 -0.03471392  0.00747667  0.03553994
 0.03244582  0.0244666 -0.03306053 -0.03149919]
Najsličnija riječ za riječ human je [('minors', 0.4442402124404907)]
```

```
system
[-0.03520911  0.00092805  0.02971833 -0.04406856  0.02822959 -0.01849993
 0.04826899 -0.04344392 -0.00955474 -0.02274698]
Najsličnija riječ za riječ system je [('eps', 0.7224090099334717)]
```

```
In [105... import numpy as np
uniques = []
for token in ds['tokens']:
    for rijec in token:
        uniques.append(rijec)
```

```
uniques = list(set(prikazano))
dctnry = {}
for word in uniques:
    if word in cmodel.wv.vocab:
        dctnry[word] = cmodel.wv[word]
    else:
        dctnry[word] = np.array(10 * [0])
```

Izvršiti analizu dobivenih rezultata sa ciljem određivanja one metode koja je najbolja za primjenu nad datim *datasetom*.

Da li je ovakav prikaz podataka intuitivan? Da li je lako razumjeti šta svaki red podataka sadrži? Da li postoje neki nedostaci ovakvog pristupa (npr. da li bi izvršavanje za veliki broj redova dugo trajalo i sl.)? Detaljno objasniti sve odgovore. Najveću pažnju obratiti na najsličnije riječi koje su pronađene i da li su one zaista ispravno određene.

Dodatno izvršiti usporedbu sa prethodno prikazanim metodama prikaza podataka (VSM, BoW, TF-IDF).

- ovaj prikaz podataka nije intuitivan
- stoga nije lagan ni za razumjeti
- nedostatak jeste upravo to što nije lahko za razumjeti te bi izvršavanje trajala duže za veliki broj redova
- najsličnije riječi nisu slične (recimo par (survey, graph) nije ni na koji način sličan)
- za razliku od prethodnih metoda VSM, BoW i TF-IDF nemamo ideju šta tačno predstavlja rezultat dok je kod prethodnih metoda to intuitivno jasno
- također nedostatak je to što ne može raditi sa riječima koje ne postoje u rječniku

Zadatak 6. - BERT (*Bidirectional Encoder Representations from Transformers*)

U okviru ovog zadatka potrebno je izvršiti predstavljanje prethodno učitanih podataka koristeći model transformacije i kodiranja riječi u dva smjera.

Zadatak:

Formirati BERT matricu za prvih 100 redova podataka *dataseta*, odnosno tokens kolone. Voditi računa o tome da korištenje BERT metode zahtijeva korištenje dosta memorije, u kom slučaju je neophodno još smanjiti broj redova (npr. na 50).

Za vršenje BERT transformacije koristiti [sljedeći link](#). Potrebno je izmijeniti samo jedan korak, zbog specifičnosti transformacije nad datim podacima zbog čega se broj tenzora povećava.

Lambda funkcija za tokenizaciju treba biti definisana na sljedeći način:

```
lambda x: tokenizer.encode(x, padding=True,
truncation=True,max_length=50, add_special_tokens = True
```

U suprotnom, zbog varijabilne dužine tenzora zbog razdvajanja jedne riječi na dvije rezultujuću matricu neće biti moguće ekstrahovati.

Prikazati rezultujuću matricu podataka.

```
In [106... import numpy as np
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LogisticRegression
from sklearn.model_selection import GridSearchCV
from sklearn.model_selection import cross_val_score
import torch
import transformers as ppb
import warnings
import numpy
warnings.filterwarnings('ignore')

data = ds['tokens'][0:100]
model_class, tokenizer_class, pretrained_weights = (ppb.BertModel, ppb.BertTokenizer, '
tokenizer = tokenizer_class.from_pretrained(pretrained_weights)
model = model_class.from_pretrained(pretrained_weights)
tokenized = data.apply((lambda x: tokenizer.encode(x, padding=True, truncation=True, max
max_len = 0
for i in tokenized.values:
    if len(i) > max_len:
        max_len = len(i)
padded = np.array([i + [0]*(max_len-len(i)) for i in tokenized.values])
np.array(padded).shape
attention_mask = np.where(padded != 0, 1, 0)
attention_mask.shape
input_ids = torch.tensor(padded)
attention_mask = torch.tensor(attention_mask)
with torch.no_grad():
    last_hidden_states = model(input_ids, attention_mask=attention_mask)
features = last_hidden_states[0][:,0,:].numpy()
print(features)
```

```
[[ 0.01521892  0.14486612  0.06561588 ... -0.31224737  0.4081184
  0.05731893]
 [-0.10328158  0.16096936  0.05731817 ... -0.13732474  0.24288437
  0.10751716]
 [-0.13013381 -0.04330372  0.34306824 ... -0.28022444  0.3998893
  0.23977862]
 ...
 [-0.11114109  0.0085783  0.1235089 ... -0.32736745  0.36346015
  0.04649123]
 [-0.2772286  0.1184981  0.08896639 ... -0.2920881  0.5145193
  0.15718897]
 [-0.02775527  0.01898104  0.14675032 ... -0.09862834  0.24124363
  0.25877112]]
```

Izvršiti analizu dobivenih rezultata sa ciljem određivanja one metode koja je najbolja za primjenu nad datim *datasetom*.

Da li je ovakav prikaz podataka intuitivan? Da li je lako razumjeti šta svaki red podataka sadrži? Da li postoje neki nedostaci ovakvog pristupa (npr. da li bi izvršavanje za veliki broj redova dugo trajalo i sl.)? Detaljno objasniti sve odgovore. Najveću pažnju obratiti na korištenje tenzora i izgled rezultujućih podataka.

Dodatno izvršiti usporedbu sa prethodno prikazanim metodama prikaza podataka (VSM, BoW, TF-IDF, CBoW).

- prikaz nije intuitivan
- nije lahko razumjeti šta svaki red podataka sadrži
- i za ovih 100 redova izvršavanje je relativno sporo tako da je za očekivati da bi bilo sporije za više redova

Zadatak 7. - Spremanje *dataseta*

U okviru ovog zadatka potrebno je sačuvati preprocesirane podatke kako bi se omogućilo njihovo ponovno učitavanje za dalje korake analize i primjene metoda mašinskog učenja.

Zadatak:

Izvršiti prethodno definisane korake predstavljanja teksta putem CBoW modela nad svim jedinstvenim riječima iz cijelog *dataseta*.

Ovo zapravo znači da je potrebno izvršiti ekstrakciju svih jedinstvenih riječi iz *dataseta*, a zatim primijeniti prethodno istrenirani `word2vec` model kako bi se za svaku riječ dobio jedinstveni 10-dimenzionalni vektor.

```
In [107...  
uniques = []  
for i in ds['tokens']:  
    for rijec in i:  
        uniques.append(rijec)  
uniques = list(set(prikazano))  
dctnry = {}  
for word in uniques:  
    if word in cmodel.wv.vocab:  
        dctnry[word] = cmodel.wv[word]  
    else:  
        dctnry[word] = np.array(10 * [0])
```

Zadatak:

Formirati novi preprocesirani *dataset* koristeći jedinstvene riječi i njihove CBoW reprezentacije.

Za svaki red podataka iz učitano *dataseta* potrebno je formirati novi red podataka koji se sastoji od CBoW reprezentacija svake pojedinačne riječi. U proizvoljnom redoslijedu potrebno je dodati sva ponavljanja svih riječi.

Npr. ukoliko je preprocesirani red podataka `['new', 'york']` uz ponavljanje riječi 'new' 3 puta, potrebno je da novi red podataka bude `[10D cbow za riječ 'new', 10D cbow za riječ 'york', 10D cbow za riječ 'new', 10D cbow za riječ 'new']`, gdje je '10D cbow za riječ' 10-dimenzionalna numerička reprezentacija CBoW metodom za datu riječ. Dozvoljeno je umetanje 10D vektora za riječi koje se ponavljaju na bilo koje mjesto u redu podataka, jer to više ne mijenja rezultat.

```
In [108... data = ds['tokens']
no_of_occurences = ds['no_of_occurences']
for i in range(0, len(data)):
    for j in range(0, len(no_of_occurences[i])):
        data[i] = data[i] + (no_of_occurences[i][j] - 1) * [data[i][j]]
```

```
In [109... res = []
for dictionary in data:
    row = []
    for word in dictionary:
        try:
            row.append(dctnry[word])
        except:
            row.append(np.array(10 * [0]))
    res.append(row)
```

Zadatak:

Sačuvati novu formu *datasetsa* u novi *file* proizvoljnog naziva.

U novi *file* trebaju se sačuvati sljedeće kolone:

`original_tokens` , kolona koja sadrži tekstualni prikaz svakog reda podataka iz *datasetsa*,

`vector_CBoW` , kolona koja sadrži vektor za svaki pojedinačni red podataka kreiran koristeći CBoW metodu.

Zbog velikog zauzeća memorije i relativno loših performansi pri usporedbi s ostalim korištenim metodama, VSM, BoW i TF-IDF metode neće biti korištene u daljim koracima klasifikacije podataka.

Voditi računa da će se u sljedećoj laboratorijskoj vježbi vršiti učitavanje ove forme *datasetsa* zbog čega je neophodno što jednostavnije moći izvršiti njegovo ponovno učitavanje (u numeričke varijable). Najbolje je da se CBoW vektor sačuva u formi broj-razmak, gdje je razmak delimiter.

```
In [110... dictionary = {'original_tokens': data, 'vector_CBoW': res}
pd.DataFrame(data=dictionary).to_csv("ds.csv")
```