```
In [40]:
```

```
import numpy as np import pandas as pd
```

## In [41]:

```
data = pd.read_csv("Kasus1.csv", delimiter=';',usecols=['text']) # read file csv nya dulu yang java heritage
data
```

### Out[41]:

### text

- **0** Apa yg bisa diharapkan dari politisi gaya pela...
- 1 Kalau sy jadi @DennyJA\_WORLD , Syahganda aka...
- 2 Di @PDemokrat banyak org2 yg punya akal sehat ...
- 3 Cypridophobia adalah fobia atau takut pada pel...
- 4 Yang RAMAI kerja waktu malam ni biasanya PELAC...

395 Letakan DiKepalak Sepatunya Klo Tidak Mau Koto...
396 Media pelacur lahir dari rahim yg tak jujur. h...
397 Jadi pelacur ko nanti dijawa nak. Jangan...
398 Kok jadi pelacur

399 /rlt/ no filter ini pake Samsung A50. Negara y...

400 rows × 1 columns

### In [42]:

1 pip install Sastrawi

Requirement already satisfied: Sastrawi in c:\users\user\anaconda3\lib\site-packages (1.0.1)Note: you may need to restart the kernel to use updated packages.

In [43]:

```
1
                        import Sastrawi
                        import re
                import re
from Sastrand.StephordRemover.StophordRemoverFactory
from Sastrand.Stemen-StemenFactory import StemenFactory
from Sastrand.Stemen-StemenFactory import StemenFactory
standard. Thanhad. "adi". "adai". 
                          from Sastrawi.StopWordRemover.StopWordRemoverFactory import StopWordRemoverFactory
                          from Sastrawi.Stemmer.StemmerFactory import StemmerFactory
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209
210
211
212
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216
                            "mgkin": "mungkin"}
217
218
219 processed_comments = []
220
221
           for sentence in data['text']:
222
                   # Remove all the special characters
                   processed\_comment = re.sub(r'\backslash W', '', str(sentence))
223
224
                   # Converting to Lowercase
225
226
                   processed_comment = processed_comment.lower()
227
228
                   #Remove number
                   processed_comment = re.sub(r'\d+', ' ', processed_comment)
229
230
                   # remove all sinale characters
231
                   processed_comment = re.sub(r'\s+[a-zA-Z]\s+', ' ', processed_comment)
232
233
234
                   #remove duplicate character
                   pattern=re.compile(r"(.)\1{1,}",re.DOTALL)
235
236
                   processed_comment=pattern.sub(r"\1",processed_comment)
237
238
                   #Corrected Slang words
239
                   words = processed comment.split()
                   rfrm=[slangs[word] if word in slangs else word for word in words]
processed_comment= " ".join(rfrm)
240
241
242
243
                   #remove stopword
244
                   factory = StopWordRemoverFactory()
245
                   f = open("stopwords-tala.txt", "r")
                   more_stopword = [] #menambahkan stopword
246
247
                    for line in f:
                            stripped_line = line.strip()
248
249
                            line_list = stripped_line.split()
                            more_stopword.append(line_list[0])
250
251
                   f.close()
252
                   stopwords = factory.get_stop_words() + more_stopword
                   temp = [t for t in re.findall(r'\b[a-z]+-?[a-z]+\b',processed_comment) if t not in stopwords]
253
254
                   processed_comment = ' '.join(temp)
255
256
                   #stemmina
257
                   stemmer = StemmerFactory().create_stemmer()
258
                   processed comment = stemmer.stem(processed comment)
259
260
                   #Subtitusing multiple spaces with single space
                   processed_comment = re.sub(r'\s+', ' ', processed_comment, flags=re.I)
261
262
                   processed comments.append(processed comment)
263
```

### In [44]:

```
1 processed comments
```

### Out[44]:

['harap politis gaya lacur pramagtis warna politik prinsip dlm politik milik integritas politis model suka rakyat', 'syahganda gugat syahganda mg mg lacur intlektual syahganda buka propaganda hasil survei deny bawaslu htps politik rmol co read syahganda buka propaganda hasil survei deny bawaslu'

'pdemokrat org akal sehat maju bangsa orang begundal serta lacur politik grasak grusuk salah satu heran htps twiter com ferdinandhaean status',

'cypridophobia fobia takut lacur tular sakit kelamin',

'ramai kerja malam lacur siapa idea suh kerja malam tu takfahamtakapa',

'lacur intlektual kpucurangrakyatmelawan kpungejarquickcounthtps twiter com suwandaben status'

'lembaga lacur lsi denyja germo kejar target atas propinsi dlm opini publik simultan dg operasi ampo suara hina hidup me rekahtps twiter com status',

'bangun gada orang sukses kasur lacur selfreminder'

'abai lacur agama hidup nkri hidup presiden joko widodo',

sat bersih jenis kotor bangsa masive awal laknat lacur agama biadab ancam selamat nkri',

'sukses capai kasur lacur',

'harap politis gaya lacur pramagtis warna politik prinsip dlm politik milik integritas politis model suka rakyat',

'syahganda gugat syahganda mg mg lacur intlektual syahganda buka propaganda hasil survei deny bawaslu htps politik rmol co read syahganda buka propaganda hasil survei deny bawaslu'

'ndemokrat org akal sehat maju bangsa orang begundal serta lacur nolitik grasak grusuk salah satu beran btns twiter com

### LDA

### In [45]:

```
1 from sklearn.feature_extraction.text import CountVectorizer
2 tf_vectorizer = CountVectorizer(max_df=1.0, min_df=1)
3 tf = tf_vectorizer.fit_transform(processed_comments)
4 #hasil representasi
5 tf_terms = tf_vectorizer.get_feature_names()
6 print(tf_vectorizer.get_feature_names())
7 matrix = tf.toarray()
8 print(matrix)
```

['abai', 'abal', 'abdu', 'abg', 'abng', 'abu', 'abud', 'ada', 'adab', 'adik', 'adil', 'adrenalin', 'aduh', 'ae', 'agama', 'agenda', 'agt', 'ah', 'ahahaha', 'ahoax', 'ahok', 'air', 'ajah', 'ajak', 'ajakin', 'ajar', 'akal', 'akbar', 'akhirat', 'akomentar', 'aku', 'akubacadotcom', 'akun', 'akwkwk', 'ala', 'alah', 'alam', 'alas', 'alhamdullah', 'ali', 'alias', 'al im', 'alir', 'alohu', 'alus', 'aman', 'ambil', 'amca', 'amerika', 'amerin', 'ampas', 'amplop', 'amplop', 'ampou', 'a mpun', 'an', 'anak', 'analogi', 'anarkis', 'anatomi', 'ancam', 'anda', 'anda', 'andai', 'andik', 'andre', 'ane', 'aneh', 'angap', 'anget', 'angin', 'angkat', 'angota', 'angota', 'aniesbaswedan', 'anime', 'angin', 'argentina', 'argentina', 'argunsin', 'arow', 'arti', 'arus', 'as', 'asal', 'asal', 'asli', 'asyik', 'atas', 'atlet', 'atu', 'atur', 'audience', 'australi a', 'av', 'avatar', 'awal', 'ayah', 'ayam', 'ayo', 'ayu', 'azab', 'bab', 'babak', 'babi', 'baca', 'bacot', 'bada', 'bada n', 'badut', 'bae', 'bagak', 'bagus', 'bahan', 'bahas', 'bahasa', 'baim', 'baja', 'bajet', 'bajing', 'bal', 'balpsa', 'bangsa', 'bangsa', 'bangsa', 'banga', 'banga', 'bapa', 'bapak', 'baba', 'bapak', 'baba', 'baba', 'baba', 'baba', 'baba', 'baba', 'baba', 'baba', 'bapa', 'bapa', 'bapa', 'belanda', 'belanja', 'belanda', 'belanja', 'belanda', 'belanja', 'beliu', 'beludru', 'beroara', 'bercara', 'bersih', 'bersan', 'bersan', 'bersan', 'bersan', 'bersan', 'bersan', 'bersan', 'bersih', 'bersan', 'bersan', 'betan', 'biada', 'bhanahahaha', 'bhre', 'biadab', 'biang', 'bial', 'biaki', 'biali', 'biada', 'bledag', 'blowo', 'blowo',

### In [46]:

```
1 # panggil class LDA
2 from sklearn.decomposition import LatentDirichletAllocation as LDA
3 n_topics = 10 #untuk mendapatkan jumlah topik terbaik perlu trial
4 | lda = LDA(n_components=n_topics, learning_method='batch', random_state=0).fit(tf)
5 1da
```

LatentDirichletAllocation(random\_state=0)

0.00370386],

0.00454574],

```
In [47]:
```

```
[0.02000242, 0.02000153, 0.02000101, ..., 0.02000006, 0.02000044, 0.020000217], [0.05001691, 0.05001395, 0.05000139, ..., 0.0500008, 0.050000612, 0.54987628], [0.00526458, 0.00526548, 0.00526366, ..., 0.00526317, 0.95262346, 0.00526355]])

In [48]:

1 #Tampilkan nilai-nilai setiap fitur print(lda.components_)
```

```
... 0.1
[[2.1
             0.1
                         0.1
                                                     0.1
                                                                 0.1
 [0.1
             0.1
                         1.09999143 ... 0.1
                                                     0.1
                                                                 0.1
 [0.1
             0.1
                         1.09999925 ... 0.1
                                                     0.1
                                                                 0.1
                                     ... 0.1
 [0.1
             0.1
                                                     0.1
                                                                 0.1
                                     ... 0.1
 0.1
             0.1
                         0.1
                                                     0.1
                                                                 0.1
 [0.1
             0.1
                         1.09996271 ... 0.1
                                                     0.1
                                                                 0.10000178]]
```

[0.00454598, 0.00454641, 0.00454595, ..., 0.00454562, 0.00454585,

### In [49]:

```
#hasil Label topic model untuk setiap dokumen
topics = np.argmax(vsm_topics, axis=1)
topics
```

### Out[49]:

```
array([4, 8, 3, 1, 0, 3, 1, 9, 0, 6, 9, 4, 8, 3, 0, 3, 1, 0, 6, 1, 1, 7,
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                              5, 6, 4, 5, 5, 4,
       1, 9, 9, 5, 6, 9, 1, 6,
                                                 2, 6, 8, 4, 2,
       5, 9, 6, 6, 2, 8, 8, 6, 1, 5, 0, 6, 6, 2, 5, 0, 6, 4, 5, 1, 2, 1,
       0, 0, 0, 9, 0, 4, 1, 1, 6, 4, 0, 5, 1, 2, 1, 0, 0, 0, 4,
         3, 1, 8, 6, 4, 2, 2, 6, 4, 0, 9, 0, 9, 1, 5, 9, 4, 4,
         9, 5, 2, 1, 4, 5, 4, 9, 1, 4, 0, 6, 3, 5, 6, 3, 9, 6,
         6, 5, 4, 0, 4, 4, 0, 6, 9, 0, 4, 6, 9, 1, 9, 8, 2, 2, 6, 9, 1,
         9, 4, 2, 5, 5, 3, 5, 6, 5, 6, 1, 6, 5, 0, 5, 3, 4, 6, 4, 9, 2,
         5, 9, 1, 6, 6, 2, 5, 4, 6, 6, 4, 8, 0, 5, 6, 4, 9, 9, 7, 1, 4,
         4, 0, 2, 9, 3, 9, 1, 9, 8, 6, 6, 1, 6, 9, 6, 1, 9, 6,
         6, 9, 7, 9, 6, 9, 7, 8, 9, 1, 9, 6, 8, 6, 8, 4, 4, 2, 8, 2, 4,
         4, 0, 6, 7, 4, 3, 6, 1, 2, 1, 6, 6, 9, 3, 8, 8, 9, 4, 2, 0, 1,
         2, 2, 9, 8, 9, 0, 0, 2, 6, 1, 1, 6, 4, 4, 0, 4, 2, 4, 8, 5, 2,
       6, 1, 1, 8, 3, 4, 5, 8, 1, 1, 1, 1, 3, 0, 9, 1, 9, 1, 1, 5, 2, 5,
       9, 7, 1, 5, 8, 5, 1, 4, 0, 6, 9, 1, 9, 6, 9, 9, 9, 9, 6, 6, 6, 6,
       1, 4, 9, 8], dtype=int64)
```

## In [50]:

```
#mencetak word fitur dengan nilai tertinggi pada setiap topik
n_top_words = 10 # jumlah fitur tertinggi yang kita tentukan
topic_words = {}
for topic, comp in enumerate(lda.components_):
    word_idx = np.argsort(comp)[::-1][:n_top_words]
    # store the words most relevant to the topic
    topic_words[topic] = [tf_vectorizer.get_feature_names()[i]+' '+str(comp[i]) for i in word_idx]
```

### In [51]:

```
for topic, words in topic_words.items():
    print('Topic: %d' % topic)
    print(' %s' % ', '.join(words))
```

Topic: 6

lacur 36.858302877433104, ahok 14.100063547015367, com 10.33124212517124, twiter 9.3115701625257, negeri 9.10005257926472 3, kerja 7.100019285676232, agama 6.100051317894472, presiden 6.100040681345306, tu 6.100037092189047, laki 6.1000116111161 855

Topic: 1

lacur 34.17682018061304, status 20.95321637287863, twiter 20.72725394922268, com 20.633781250824796, htps 16.6273719501033 8, leceh 9.002413895605685, tutup 8.100028846486598, tusuk 8.100019253365828, ras 8.099903939505213, verbal 7.9899674768977

Topic: 2

lacur 23.100425806660525, ras 11.788535040417313, orang 9.83517213284242, politik 6.100118994339808, iya 6.10002453701038, suka 5.944528899123657, dasar 5.100115092764825, jaya 5.0999999995441625, gitu 5.09999999923475, maf 4.10008184079854
Topic: 3

lacur 12.784085463794048, com 7.100041782956527, twiter 6.100040680309208, orang 5.100091145426324, status 5.1000283347639 49, ga 4.100110571668511, ras 4.099959526783606, gak 3.957519079901731, politik 3.1000609901631715, abdu 3.1000392953393807 Topic: 4

lacur 43.823716325696914, ras 16.100238531340743, politik 16.055280738731454, gak 7.98385730774608, anjing 6.1000224106218 83, politis 5.100032068752101, khianat 4.100031667465745, kayak 3.9825075811719395, rakyat 3.755852239722175, yah 3.1000448 193071923

Topic: 5

lacur 41.35840756780545, politik 10.100082105401173, com 10.10001285224174, twiter 9.100020097764915, https 9.1000192583666 63, leceh 8.100048691293653, verbal 7.100032011533481, status 7.0999990467894145, nama 6.479139947659218, anjing 6.10002202 4478947

Topic: 6

lacur 56.97048436460543, agama 10.10004876004643, ras 9.411608944711196, mesum 8.099994396168306, twiter 8.09995653113680 4, com 8.099930792909472, media 7.100021220351084, orang 6.364847741523899, jual 6.100053142647907, demokrasi 6.10002990096 1755

Topic: 7

mesum 2.1001448956667024, org 2.100064838133289, skrg 2.100011354295569, duduk 2.100005646965668, iseng 2.09999999803583 4, gegara 2.0999999997168217, hasil 1.100040421259895, nya 1.1000272434582734, cina 1.100025188701681, parah 1.100023937424 0526

Topic: 8

lacur 10.488518025497777, ras 8.100149221777906, syahganda 8.099999999695108, buka 5.100057243928322, orang 5.100054285997 409, emang 5.1000509453611125, kpop 5.09999999964672, gak 4.9446389126115395, https://doi.org/10.1001277366085 69

Topic: 9

lacur 56.33942820903878, nya 10.8824002150346, orang 9.053454869108535, mesum 8.519297213112779, gak 7.100095266646303, ot ak 7.040117200387565, iya 6.889353722015582, hina 5.81791365053057, beda 5.100014331414758, juang 5.100014231682571

## **LSA**

## In [52]:

- 1 from sklearn.feature\_extraction.text import TfidfVectorizer
- vektor = TfidfVectorizer(max\_features = 1000)

```
In [53]:
```

```
#menghitung tf-idf dengan TfidfTransformer
 1
    vektor_dt = vektor.fit_transform(data['text'])
 2
 3 print(vektor dt)
 4 print(vektor_dt.shape)
  (0, 762)
                0.15272779359201613
  (0, 242)
                0.19192313773966022
  (0, 36)
                0.14177393727307744
  (0, 110)
                0.18255764560035231
  (0, 607)
                0.19192313773966022
  (0, 601)
                0.19192313773966022
  (0, 817)
                0.14679231387154673
  (0, 190)
                0.1752932055260729
                0.19192313773966022
  (0, 359)
  (0, 578)
                0.19192313773966022
  (0, 925)
                0.29358462774309346
                0.1496249699362145
  (0, 802)
  (0, 805)
                0.12857401747832858
  (0, 134)
                0.18255764560035231
                0.1643393492071342
  (0, 245)
  (0, 742)
                0.19192313773966022
  (0, 756)
                0.13545318669050188
  (0, 733)
                0.19192313773966022
  (0, 979)
                0.18255764560035231
  (0, 389)
                0.15999223366629556
  (0, 941)
                0.24078494856409796
  (0, 737)
                0.19192313773966022
                0.04390090115685355
  (0, 695)
  (0, 295)
                0.19192313773966022
  (0, 734)
                0.3286786984142684
  (397, 626)
                0.5441070786149599
  (397, 382)
                0.3955651449392686
  (397, 475)
                0.4969607890072219
  (397, 623)
                0.4427114345470066
  (397, 376)
                0.2983654185284531
  (397, 695)
                0.1244601946296941
  (398, 477)
                0.8143305631184415
  (398, 376)
                0.5356650091115894
  (398, 695)
                0.22344738079620147
  (399, 792)
                0.35686995865224147
  (399, 155)
                0.2859150865896594
  (399, 712)
                0.3339049400532245
                0.2716809628530919
  (399, 629)
  (399, 674)
                0.2657134797002912
                0.2657134797002912
  (399, 726)
  (399, 358)
                0.16915875560193464
  (399, 851)
                0.23242200425197623
  (399, 552)
                0.26031523455388494
  (399, 992)
                0.1830994138898145
  (399, 187)
                0.1642305438990429
  (399, 960)
                0.16713572966240264
  (399, 923)
                0.20739716938742805
  (399, 96)
                0.24665612798854372
  (399, 217)
                0.3403960541036873
  (399, 695)
                0.07637811647778488
(400, 1000)
In [54]:
 1 idf = vektor.idf_
    dd = dict(zip(vektor.get_feature_names(), idf))
 3 1 = sorted(dd, key = (dd).get)
 4 print(l[0], l[-1])
 5 print(dd['pelacur'])
 6 print(dd['wilde'])
pelacur wilde
1.3485145296633316
6.300814246746624
In [55]:
 1 from sklearn.decomposition import TruncatedSVD
    lsa_model = TruncatedSVD(n_components=10, algorithm='randomized', n_iter=10, random_state=42)
 3 lsa_top = lsa_model.fit_transform(vektor_dt)
In [56]:
 1 # jumlah dokumen * jumlah topik
 2 print(lsa_top.shape)
(400, 10)
```

```
(400, 10,
```

```
In [57]:
```

```
1 # top Lsa
 2 print(lsa_top)
[[ 1.48659449e-01 -9.07423341e-02 -9.32955306e-02 ... -1.12854271e-01
   5.71319799e-02 -1.41289657e-01]
 [ 1.17667791e-01 -6.98253277e-03 -9.08301957e-03 ... 1.49051278e-01
   4.12538022e-02 3.94359639e-03]
 [\ 2.63552696e-01\ -1.22096489e-01\ \ 1.58809497e-01\ \dots\ -1.89116206e-02
   8.87302284e-02 -8.13867092e-02]
 [ 1.43796548e-01 -6.33206813e-03 -9.86829494e-02 ... 4.92812579e-02
   8.67508362e-02 1.10058605e-01]
 [ 2.17107931e-01 -1.82523631e-04 -1.11976866e-01 ... 8.87682837e-02
  -2.13490196e-02 7.28173096e-02]
 [ 2.65659717e-01 -1.95614557e-01 3.84057286e-02 ... 1.87547711e-01
  -9.76421356e-02 2.07392673e-02]]
In [58]:
 1 # memunculkan nilai lsa setiap topik
 2 1 = 1sa_top[0]
 3 print('Topik - Topik :')
 4 for i, topic in enumerate(1):
        print('Topic ', i, ' : ', topic*100)
Topik - Topik :
Topic 0 : 14.865944900587117
Topic 1 : -9.074233411497108
Topic 2 : -9.329553056265286
Topic 3 : -1.5971514002380256
Topic 4 : -0.6461870061277563
Topic 5 : -5.2107829719236065
Topic 6 : -18.1891684898388
Topic 7 : -11.28542705524751
Topic 8 : 5.713197985199807
Topic 9 : -14.128965680771529
In [59]:
 1 #memunculkan jumlah kata-kata dalam setiap topik
 2 print(lsa_model.components_.shape)
 3 print(lsa_model.components_)
(10, 1000)
\hbox{\tt [[ 0.00725228 \ 0.00192549 \ 0.00399414 \dots \ 0.00162078 \ 0.0292714 ]}
   0.00869898]
 [ 0.00018748 -0.00018537 -0.004088 ... -0.00342359 0.03029904
  -0.00146251]
 [-0.00159705 -0.00032499 -0.00182139 ... -0.0034003
  -0.00775767]
 [-0.01669732 \quad 0.0085105 \quad -0.00188842 \ \dots \ -0.01050738 \ -0.02939351
   0.00095881]
 [-0.01419099 0.00259324 0.006313 ... 0.00109504 -0.01150681
   0.0032488 ]
 0.02292166]]
```

In [60]:

```
1 # word / kata paling penting dalam setiap topik
    vocab = vektor.get_feature_names()
for i, comp in enumerate(lsa_model.components_):
 3
 4
        vocab_comp = zip(vocab, comp)
        sorted_words = sorted(vocab_comp, key = lambda x:x[1], reverse=True)[:10]
print('Topic '+str(i)+': ')
 5
 6
 7
        for t in sorted_words:
             print(t[0], end=', ')
 8
 9
        print('\n')
pelacur, politik, com, twitter, yg, ini, status, di, dan, https,
Topic 1:
politik, pelacur, negri, la, yusrilnorakdankasarmainnya, tutup01tusuk02, jabatan, cinta, para, yim,
twitter, com, status, https, politik, pic, tutup01tusuk02, kpucurangrakyatmelawan, kpungejarquickcounthttps, suwandaben,
Topic 3:
politik, rasis, mesum, aja, ini, gak, lu, yg, ya, otak,
mesum, aku, otak, ini, juga, babi, izinkan, cabul, ulama, gak,
Topic 5:
aku, izinkan, menjadi, tuhan, politik, ini, dan, binatang, jalang, filosofi,
rasis, pelacur, aja, aku, mereka, kamu, tu, mesum, gak, kau,
Topic 7:
di, ahok, mesum, kecuali, kasur, jadi, negeri, tutup01tusuk02, otak, pic,
kau, tu, politik, kata, orang, yang, la, kecuali, hina, tidak,
Topic 9:
kau, ini, anjing, muka, eh, tu, jadi, la, nak, pki,
```

# Kesimpulan

- Terdapat beberapa perbedaan hasil Topic Modelling menggunakan LDA dan LSA
- Topik pertama dari kedua metode LDA maupun LSA sama dan menghasilkan output "pelacur" sebagai kata teratas
- Sedangkan di metode LDA, kata teratas yang dominan dari 10 topik menghasilkan "pelacur