Taller 9

Métodos Computacionales para Políticas Públicas - URosario

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Instrucciones:

- Guarde una copia de este *Jupyter Notebook* en su computador, idealmente en una carpeta destinada al material del curso.
- Modifique el nombre del archivo del notebook, agregando al final un guión inferior y su nombre y apellido, separados estos últimos por otro guión inferior. Por ejemplo, mi notebook se llamaría: mcpp_taller9_santiago_matallana
- Marque el notebook con su nombre y e-mail en el bloque verde arriba. Reemplace el texto "
 [Su nombre acá]" con su nombre y apellido. Similar para su e-mail.
- Desarrolle la totalidad del taller sobre este *notebook*, insertando las celdas que sea necesario debajo de cada pregunta. Haga buen uso de las celdas para código y de las celdas tipo *markdown* según el caso.
- Recuerde salvar periódicamente sus avances.
- Cuando termine el taller:
 - 1. Descárguelo en PDF. Si tiene algún problema con la conversión, descárguelo en HTML.
 - 2. Suba todos los archivos a su repositorio en GitHub, en una carpeta destinada exclusivamente para este taller, antes de la fecha y hora límites.

```
• Chapter 1: 22, 26, 28
           • Chapter 2: 2, 4, 11
In [107]: %matplotlib inline
          import matplotlib.pyplot as plt
          plt.rcParams["figure.figsize"] = [18.0, 8.0]
 In [2]: import nltk
 In [3]: from nltk.book import *
          *** Introductory Examples for the NLTK Book ***
          Loading text1, ..., text9 and sent1, ..., sent9
          Type the name of the text or sentence to view it.
          Type: 'texts()' or 'sents()' to list the materials.
          text1: Moby Dick by Herman Melville 1851
          text2: Sense and Sensibility by Jane Austen 1811
          text3: The Book of Genesis
          text4: Inaugural Address Corpus
          text5: Chat Corpus
          text6: Monty Python and the Holy Grail
          text7: Wall Street Journal
          text8: Personals Corpus
          text9: The Man Who Was Thursday by G . K . Chesterton 1908
          22. Find all the four-letter words in the Chat Corpus (text5). With the
          help of a frequency distribution (FreqDist), show these words in
          decreasing order of frequency.
```

NLTK Book (http://www.nltk.org/book/), Exercises:

```
In [4]: four_letter = [w for w in text5 if len(w) == 4]
         four_letter
Out[4]: ['left',
          'with',
          'this',
          'name',
          'PART',
          'well',
          'NICK',
          'name',
          'U121',
          'golf',
          'clap',
          'JOIN',
          'that',
          'nice',
          'JOIN',
          'PART',
          'golf',
          'clap',
          'fuck',
          'U121',
          'PART',
          'PART',
          'clap',
          'your',
          'PART',
          'dont',
          'even',
          'know',
          'what',
          'that',
          'that',
          'chat',
          'JOIN',
          'drew',
          'cast',
          'PART',
```

'sexy', 'U115', 'JOIN', 'PART', 'drew', 'girl', 'with', 'legs', 'hope', 'draw', 'PART', 'head', 'legs', 'JOIN', 'JOIN', 'good', 'JOIN', 'PART', 'take', 'have', 'docs', 'Slip', 'away', 'Fade', 'away', 'Days', 'away', 'feel', 'have', 'back', 'U115', 'U129', 'U115', 'chat', 'with', 'PART', 'JOIN', 'JOIN', 'fast',

```
'U116',
'bowl',
'bong',
'JOIN',
'well',
'glad',
'hard',
'from',
'here',
'back',
'PART',
'PART',
'JOIN',
'U121',
'name',
'hard',
'very',
'fire',
'from',
'here',
'JOIN',
'PART',
'itch',
'JOIN',
'U133',
'ogan',
'male',
'JOIN',
'JOIN',
'show',
'will',
'talk',
'PART',
'haha',
'opps',
'JOIN',
'PART',
'U115',
'nice',
```

```
'warm',
'guys',
'with',
'cams',
'play',
'sits',
'JOIN',
'JOIN',
'guyz',
'chat',
'U126',
'PART',
'chat',
'PART',
'gooo',
'sure',
'U126',
'JOIN',
'what',
'feel',
'like',
'room',
'yeee',
'JOIN',
'want',
'pics',
'look',
'U139',
'PART',
'PART',
'JOIN',
'here',
'JOIN',
'PART',
'JOIN',
'U139',
'PART',
'JOIN',
'U138',
```

'U139', 'make', 'U139', 'that', 'U126', 'late', 'lmao', 'ahah', 'PART', 'U121', 'U121', 'does', 'like', 'that', 'guys', 'male', 'JOIN', 'U139', 'well', 'what', 'yeah', 'know', 'U136', 'hell', 'with', 'U139', 'U101', 'like', 'when', 'plan', 'PART', 'JOIN', 'gold', 'jeep', 'make', 'sure', 'nice', 'ring', 'U115',

```
'isnt',
'that',
'U136',
'hell',
'have',
'have',
'doin',
'U139',
'U121',
'many',
'Just',
'fine',
'that',
'like',
'PART',
'hiya',
'room',
'lmao',
'doin',
'Deep',
'Show',
'that',
'love',
'that',
'Turn',
'take',
'Hand',
'just',
'even',
'look',
'hang',
'PART',
'that',
'such',
'word',
'U141',
'hear',
'!!!!!',
'PART',
```

'JOIN', 'PART', 'deaf', 'here', 'dont', 'U115', 'U115', 'hugs', 'chat', 'with', 'baby', 'Only', 'U121', 'U121', 'PART', 'have', 'away', 'from', 'U121', 'what', 'read', 'here', 'with', 'JOIN', 'read', 'have', 'here', 'JOIN', 'want', 'chat', 'talk', 'U121', 'JOIN', 'U121', 'VBox', 'PART', 'take', 'that',

```
'JOIN',
'PART',
'hate',
'when',
'U121',
'U115',
'lmao',
'PART',
'your',
'know',
'what',
'your',
'what',
'JOIN',
'love',
'more',
'than',
'ELSE',
'serg',
'well',
'most',
'love',
'JOIN',
'know',
'that',
'what',
'lmao',
'well',
'have',
'eyes',
'lmao',
'know',
'JOIN',
'girl',
'jerk',
'kids',
'guys',
'type',
'much',
```

'shut', 'fuck', 'girl', 'nice', 'shut', 'fuck', 'PART', 'dont', 'want', 'JOIN', 'want', 'U115', 'what', 'miss', 'much', 'work', 'nice', 'U116', 'PART', 'PART', 'heyy', 'U148', 'hate', 'boys', 'JOIN', 'U148', 'hate', 'what', 'PART', 'hate', 'U121', 'fuck', 'your', 'ugly', 'JOIN', 'bein', 'PART', 'What', 'U115',

'whys', 'that', 'deep', 'U121', 'what', 'JOIN', 'tape', 'Your', 'sexs', 'best', 'phil', 'said', 'ugly', 'PART', 'date', 'feel', 'your', 'they', 'form', 'PART', 'sits', 'JOIN', 'sits', 'with', 'hmph', 'hate', 'does', 'that', 'mean', 'want', 'room', 'this', 'been', 'PART', 'JOIN', 'U115', 'U116', 'your', 'here',

```
'talk',
'wait',
'that',
'perv',
'lets',
'hope',
'U121',
'PART',
'U115',
'!!!!',
'lets',
'chat',
'JOIN',
'rule',
'land',
'wont',
'then',
'find',
'need',
'this',
'HUGE',
'perv',
'that',
'deal',
'????',
'JOIN',
'shit',
'hell',
'lmao',
'PART',
'hell',
'JOIN',
'here',
'guys',
'have',
'U121',
'JOIN',
'U155',
'only',
```

'Poor', 'U121', 'love', 'pick', 'much', 'that', 'PART', 'PART', 'sits', 'with', 'U121', 'nads', 'JOIN', 'from', 'pick', 'your', 'pick', 'your', 'nose', 'pick', 'your', 'nose', 'JOIN', 'face', 'with', 'PART', 'U115', 'OWWW', 'PART', 'JOIN', 'U116', 'PART', 'does', 'want', 'talk', 'head', 'gags', 'even', 'U121',

```
'neck',
'Meep',
'U115',
'LAst',
'time',
'that',
'wash',
'your',
'dude',
'gets',
'JOIN',
'U121',
'dang',
'just',
"pm's",
'that',
'1.99',
· . . . . <sup>'</sup> ,
'yeah',
'nice',
'neck',
'U115',
'like',
'shut',
'free',
'JOIN',
'goes',
'wash',
'lmao',
'Lies',
'lmao',
'U115',
'lick',
'very',
'lmao',
'U115',
'ummm',
'U109',
'dont',
```

'dead', 'more', 'than', 'call', 'just', 'case', 'dead', 'good', 'neck', 'talk', 'what', 'ummm', 'else', 'wont', 'bite', 'U115', 'yeah', 'wait', 'yeah', 'PART', 'your', 'want', 'have', 'sexy', 'bite', 'lmao', 'call', 'have', 'free', 'call', 'mins', 'JOIN', 'nite', 'lool', 'know', 'that', 'kina', 'give', 'away',

'then', 'room', 'call', 'yeah', 'U155', 'PART', 'U115', 'more', 'U115', 'guys', 'baby', 'U109', 'fuck', 'case', 'know', 'were', 'girl', 'JOIN', 'baby', 'what', 'U109', 'guys', 'chat', 'have', 'have', 'sext', 'piff', 'dont', 'talk', 'read', 'dang', 'lazy', 'dont', 'read', 'PART', 'JOIN', 'mean', 'fine', ¹ ¹ ,

```
'busy',
'work',
'okay',
'dont',
'talk',
'calm',
'down',
'busy',
'busy',
'want',
'chat',
'arms',
'kids',
'name',
'PART',
'sits',
'down',
'eats',
'JOIN',
'hugs',
'want',
'U121',
'near',
'just',
'PART',
'JOIN',
'PART',
'hell',
'yeah',
'U115',
'near',
'near',
'good',
'smax',
'JOIN',
'haha',
'only',
'>:->',
'near',
```

'PART', 'piff', 'VVil', 'JOIN', 'free', 'wont', 'cold', 'U121', 'cell', 'runs', 'thru', 'back', 'hair', 'eyes', 'neck', 'yeah', 'caps', 'PART', 'JOIN', 'PART', 'PART', 'U165', 'jump', 'U165', 'baby', 'here', 'over', 'your', 'good', 'PART', 'that', 'left', 'room', 'este', 'U115', 'will', 'PART', 'U121', 'U165',

```
'lmao',
'PART',
'PART',
'very',
'guys',
'wana',
'chat',
'chik',
'from',
'mean',
'chat',
'well',
'PART',
'that',
'dont',
'shit',
'U165',
'U165',
'left',
'room',
'with',
'your',
'JOIN',
'U115',
'what',
'list',
'wish',
'cmon',
'U128',
'nice',
'JOIN',
'list',
'PART',
'list',
'U115',
'good',
'lmao',
'U128',
'hehe',
```

```
'hows',
'bout',
'good',
'hear',
'U165',
'have',
'JOIN',
'good',
'PART',
'JOIN',
'wats',
'they',
'PART',
'piff',
'aint',
'know',
'shut',
'much',
'good',
'PART',
'JOIN',
'JOIN',
'yeah',
'lost',
'like',
'same',
'well',
'work',
'what',
'Boyz',
'rock',
'what',
'hehe',
'went',
'back',
'some',
'then',
'came',
'back',
```

```
'home',
'PART',
'what',
'they',
'coat',
'nice',
'read',
'many',
'nice',
'hehe',
'lmao',
'even',
'JOIN',
'well',
'talk',
'nite',
'what',
'very',
'time',
'What',
'kind',
· . . . . <sup>'</sup> ,
'nite',
'PART',
'Eyes',
'Dawn',
'last',
'song',
'LIVE',
'cool',
'good',
'nite',
'mauh',
'nite',
'mike',
'keep',
'must',
'girl',
'seem',
```

'pick', 'else', 'take', 'your', 'your', 'JOIN', 'lmao', 'just', 'days', 'late', 'with', 'room', 'JOIN', 'PART', 'good', 'ques', 'lmao', 'JOIN', 'that', 'like', 'dont', 'quit', 'what', 'your', '4.20', 'PART', 'like', 'mine', 'over', 'cali', 'good', 'this', 'year', 'NICK', 'whoa', 'have', 'have', 'have',

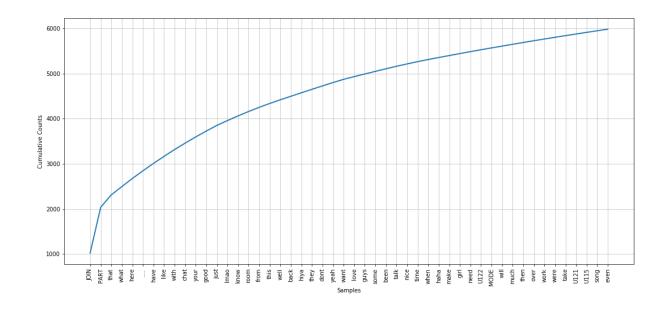
'what', 'boys', 'gosh', 'that', 'ruff', 'what', 'PART', 'hell', 'rock', 'roll', 'PART', 'with', 'like', 'that', 'nope', · · · · · · , 'rest', 'rock', 'roll', '....', ¹ ¹ , 'sing', 'from', 'kids', '....', 'mame', 'nada', 'cali', 'here', 'that', 'cool', 'kids', 'cool', 'JOIN', 'with', 'from', 'said', 'alot', 'JOIN',

```
'year',
'band',
'JOIN',
'NICK',
'cool',
'nice',
'here',
'hair',
'hard',
'what',
'type',
'does',
'your',
'band',
'play',
'hair',
'yeah',
'what',
'doin',
'....',
'with',
'hand',
'what',
'JOIN',
'room',
'....',
'sexy',
'PART',
'dumb',
'they',
'wont',
'chat',
'with',
'lmao',
'damn',
'what',
'orgy',
'orgy',
'lmao',
```

```
'what',
'word',
'some',
'easy',
'PART',
'JOIN',
'back',
'orgy',
'lmao',
'PART',
'lets',
'play',
'room',
'PART',
'JOIN',
'JOIN',
'were',
'damn',
'PART',
'good',
'call',
'what',
'like',
'just',
'late',
'date',
'know',
'push',
'PART',
'lose',
'name',
'shit',
'head',
'long',
'time',
'said',
'shit',
'lost',
'baby',
```

'then', 'JOIN', 'PART', 'JOIN', 'What', 'sure', 'JOIN', 'sure', 'lmao', 'this', 'room', '....', '....', 'your', '....', 'lmao', 'that', 'yeah', 'have', 'many', 'lmao', 'sexy', 'stay', 'keep', 'lmao', 'with', 'hair', 'like', 'that', '....', 'down', 'door', 'prob', '....', 'hair', 'lmao', 'JOIN', 'sexy', 'just',

```
'this',
            'life',
            'just',
            'PART',
            'room',
            'with',
            'hair',
            'like'.
            'what',
            'from',
            'here'.
            'said',
            'wild',
            '....',
            'even',
            'from',
            'JOIN',
            'cool',
            'sexy',
            'JOIN',
            'sexy',
            'only',
            'none',
            'sexy',
            'whew',
            'sexy',
            'hell',
            'have',
            ...]
  In [8]: fdist5 = FreqDist(four letter)
           fdist5
  Out[8]: FreqDist({'JOIN': 1021, 'PART': 1016, 'that': 274, 'what': 183, 'here':
           181, '....': 170, 'have': 164, 'like': 156, 'with': 152, 'chat': 142,
           ...})
In [108]: fdist5.plot(50, cumulative=True)
```



26. What does the following Python code do?

sum(len(w) for w in text1)

Can you use it to work out the average word length of a text?

El código está sumando el tamaño de cada palabra, por lo que al final va a dar el tamaño total de la suma de todas las palabras. Una vez ejecutado el código, se puede encontrar el tamaño promedio de la palabra de un texto si se divide por el número total de palabras.

Una aclaración importante, es que esta operación cuenta tokens, no únicamente palabras. Para tener el promedio de palabras únicamente, se debe utiliza la función set.

```
In [11]: total_length = sum(len(w) for w in text1)
total_length
Out[11]: 999044
```

```
In [22]: total words = len(text1)
         total words
Out[22]: 260819
In [68]: avg wrd ln = (sum(len(w) for w in text1))/(len(text1))
         avg wrd ln
Out[68]: 3.830411128023649
In [23]: def avg word lenght(text):
             total length = sum(len(w) for w in text)
             total words = len(text)
              avg word = (total length)/(total words)
              return avg word
In [24]: avg word lenght(text1)
Out[24]: 3.830411128023649
In [25]: avg word lenght(text2)
Out[25]: 3.881371136350794
         En promedio, el tamaño de las palabras del texto1 son de alrededor de 4 caracteres (3.8)
         28. Define a function percent(word, text) that calculates how often a
         given word occurs in a text, and expresses the result as a
         percentage.
In [72]: def percent(word,text):
             fdist = FreqDist(text)
             w = fdist [word]
             l = len(text)
             word percentage = 100 * w / l
```

```
return word percentage
In [73]: percent("whale", text1)
Out[73]: 0.3473673313677301
In [74]: percent("God", text3)
Out[74]: 0.5160396747386292
         2. Use the corpus module to explore austen-persuasion.txt. How
         many word tokens does this book have? How many word types?
In [75]: import nltk
         nltk.corpus.gutenberg.fileids()
Out[75]: ['austen-emma.txt',
          'austen-persuasion.txt',
          'austen-sense.txt',
          'bible-kjv.txt',
          'blake-poems.txt',
          'bryant-stories.txt',
          'burgess-busterbrown.txt',
          'carroll-alice.txt',
          'chesterton-ball.txt',
          'chesterton-brown.txt',
          'chesterton-thursday.txt',
          'edgeworth-parents.txt',
          'melville-moby dick.txt',
          'milton-paradise.txt',
          'shakespeare-caesar.txt',
          'shakespeare-hamlet.txt',
          'shakespeare-macbeth.txt',
          'whitman-leaves.txt'l
In [77]: from nltk.corpus import gutenberg
```

```
qutenberg.fileids()
Out[77]: ['austen-emma.txt',
          'austen-persuasion.txt',
          'austen-sense.txt',
          'bible-kjv.txt',
          'blake-poems.txt',
          'bryant-stories.txt',
          'burgess-busterbrown.txt',
          'carroll-alice.txt',
          'chesterton-ball.txt'.
          'chesterton-brown.txt',
          'chesterton-thursday.txt',
          'edgeworth-parents.txt',
          'melville-moby dick.txt',
          'milton-paradise.txt',
          'shakespeare-caesar.txt',
          'shakespeare-hamlet.txt',
          'shakespeare-macbeth.txt',
          'whitman-leaves.txt'l
In [80]: austen p = gutenberg.words('austen-persuasion.txt')
         #word tokens
         len(austen p)
Out[80]: 98171
In [81]: #word types
         len(set(austen p))
Out[81]: 6132
         4. Read in the texts of the State of the Union addresses, using the
         state union corpus reader. Count occurrences of men, women, and
         people in each document. What has happened to the usage of these
         words over time?
```

Create PDF in your applications with the Pdfcrowd HTML to PDF API

```
In [84]: from nltk.corpus import state union
          state union.fileids()
Out[84]: ['1945-Truman.txt',
           '1946-Truman.txt',
           '1947-Truman.txt',
           '1948-Truman.txt',
           '1949-Truman.txt',
           '1950-Truman.txt',
           '1951-Truman.txt',
           '1953-Eisenhower.txt',
           '1954-Eisenhower.txt',
           '1955-Eisenhower.txt'.
           '1956-Eisenhower.txt'.
           '1957-Eisenhower.txt',
           '1958-Eisenhower.txt',
           '1959-Eisenhower.txt'.
           '1960-Eisenhower.txt',
           '1961-Kennedy.txt',
           '1962-Kennedy.txt',
           '1963-Johnson.txt',
           '1963-Kennedy.txt',
           '1964-Johnson.txt',
           '1965-Johnson-1.txt',
           '1965-Johnson-2.txt',
           '1966-Johnson.txt',
           '1967-Johnson.txt'.
           '1968-Johnson.txt',
           '1969-Johnson.txt',
           '1970-Nixon.txt',
           '1971-Nixon.txt',
           '1972-Nixon.txt'.
           '1973-Nixon.txt'.
           '1974-Nixon.txt',
           '1975-Ford.txt',
           '1976-Ford.txt',
           '1977-Ford.txt',
           '1978-Carter.txt',
           '1979-Carter.txt',
           '1980-Carter.txt',
```

```
'1981-Reagan.txt',
            '1982-Reagan.txt',
            '1983-Reagan.txt',
           '1984-Reagan.txt',
           '1985-Reagan.txt',
           '1986-Reagan.txt',
            '1987-Reagan.txt',
            '1988-Reagan.txt',
           '1989-Bush.txt',
           '1990-Bush.txt',
            '1991-Bush-1.txt',
           '1991-Bush-2.txt',
            '1992-Bush.txt',
           '1993-Clinton.txt',
            '1994-Clinton.txt',
           '1995-Clinton.txt',
           '1996-Clinton.txt',
            '1997-Clinton.txt',
           '1998-Clinton.txt',
            '1999-Clinton.txt',
            '2000-Clinton.txt',
           '2001-GWBush-1.txt',
           '2001-GWBush-2.txt',
           '2002-GWBush.txt',
           '2003-GWBush.txt',
            '2004-GWBush.txt',
           '2005-GWBush.txt',
           '2006-GWBush.txt']
In [92]: len(state union.fileids())
Out[92]: 65
In [109]: cfd = nltk.ConditionalFreqDist(
               (interest, fileid[:4])
              for fileid in state_union.fileids()
              for w in state union.words(fileid)
              for interest in ['men', 'women', 'people']
```

```
if w.lower() == interest)
cfd.plot()

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pag
```

11. Investigate the table of modal distributions and look for other patterns. Try to explain them in terms of your own impressionistic understanding of the different genres. Can you find other closed classes of words that exhibit significant differences across different genres?

```
'lore',
           'mystery',
           'news',
           'religion',
           'reviews',
           'romance',
           'science fiction']
In [102]: cfd = nltk.ConditionalFregDist(
                     (genre, word)
                     for genre in brown.categories()
                     for word in brown.words(categories=genre))
          genres = ['news', 'religion', 'hobbies', 'science fiction', 'romance',
          'humor'l
          modals = ['can', 'could', 'may', 'might', 'must', 'will']
          cfd.tabulate(conditions=genres, samples=modals)
                            can could
                                       may might must will
                             93
                                                    50
                                                          389
                                   86
                                         66
                                               38
                     news
                                   59
                             82
                                        78
                                              12
                                                    54
                                                          71
                 religion
                  hobbies
                            268
                                  58
                                        131
                                              22
                                                    83
                                                         264
                                  49
                                              12
                                                    8
          science fiction
                             16
                                                          16
                                              51
                                                    45
                                                          43
                             74
                                 193
                                         11
                  romance
                    humor
                             16
                                  30
                                               8
                                                          13
In [103]: cfd = nltk.ConditionalFregDist(
              (genre, word)
              for genre in brown.categories()
              for word in brown.words(categories=genre))
          genres = ['news', 'religion', 'hobbies', 'science fiction', 'romance',
          'humor', 'editorial', 'belles lettres', 'government']
          pronouns = ['I', 'you', 'he', 'she', 'it', 'we', 'they']
          cfd.tabulate(conditions=genres, samples=pronouns)
                             I you
                                     he she
                                              it
                                                    we they
                               55 451
                                         42
                                              363
                                                    77 205
                     news 179
                 religion 155 100 137
                                          10
                                              264 176 115
                  hobbies 154
                               383 155
                                          21
                                              476 100 177
                           98
                                81 139
                                           36 129
                                                         53
          science fiction
                                                    30
```

```
romance 951 456 702 496 573 78 168 humor 239 131 146 58 162 32 70 editorial 201 83 268 41 386 167 148 belles_lettres 845 188 1174 178 1059 398 488 government 97 74 120 0 218 112 92
```

```
In [106]:
    cfd = nltk.ConditionalFreqDist(
        (genre, word)
        for genre in brown.categories()
        for word in brown.words(categories=genre))
    genres = ['news', 'religion', 'hobbies', 'science_fiction', 'romance',
        'humor', 'editorial', 'belles_lettres', 'government']
    others = ['love', 'family', 'entertainment', 'important', "society"]
    cfd.tabulate(conditions=genres, samples=others)
```

	love	family	entertainment	important
society				
news	3	41	7	13
12				
religion	13	20	2	17
10				
hobbies	6	25	2	40
science_fiction	3	1	0	2
0				
romance	32	23	0	4
1				
humor	4	6	1	7
2				
editorial	13	14	0	16
2				
belles_lettres	68	54	2	61
78				
government	1	10	7	44
3				