## **Textual Networks from Greek Texts**

```
In [85]:
# INPUT
import os
text file = os.path.expanduser('~/cltk data/greek/text/greek text first1kgreek pl
In [86]:
# READING THE INPUT FILE
with open(text_file) as f:
    text raw = f.read()
# CLEANING
from cltk.corpus.utils.formatter import tlg plaintext cleanup
text cleaned = tlg plaintext cleanup(text raw, rm punctuation=True, rm periods=Tr
# rm_periods=True returns no interpunction
# LEMMATIZATION
# does not work so good as I would like
from cltk.stem.lemma import LemmaReplacer
lemmatizer = LemmaReplacer('greek')
text_cleaned = text_cleaned.lower()
text lemmatized = lemmatizer.lemmatize(text cleaned, return string=True)
# STROPWORDS FILTERING
from nltk.tokenize.punkt import PunktLanguageVars
from cltk.stop.greek.stops import STOPS LIST
p = PunktLanguageVars()
text tokens = p.word tokenize(text lemmatized.lower())
text tok = [w for w in text tokens if not w in STOPS LIST]
text_str = " ".join(text_tok)
# WORD FREQUENCY DICTIONARY/COUNTER
from cltk.utils.frequency import Frequency
freq = Frequency()
text freq counter = freq.counter from str(text str)
just dial tok = text tok
```

## Word frequneces

just dial = text str

```
In [89]:
# WORD FREQUENCY DICTIONARY/COUNTER
from cltk.utils.frequency import Frequency
freq = Frequency()
just freq counter = freq.counter from str(just dial)
In [90]:
just_freq_counter.most_common(20)
Out[90]:
[('αὐτός', 967),
 ('σύ', 845),
 ('εἰμί', 830),
 ('ἔγώ', 703),
 ('oů', 570),
 ('αὐτὸν', 222),
 ('θεάομαι', 221),
 ('τὰς', 195),
 ('θεὸς', 175),
 ('γαῖα', 162),
 ('κύριος', 159),
 ('οὖν', 146),
 ('ἔθνος', 142),
 ('εἶπον', 141),
 ('φημί', 139),
 ('ἔχω', 139),
('ἀλλ'', 138),
 ('\pi\tilde{\alpha}C', 118),
In [ ]:
clem freq counter.most common(100)
```

## **Transforming format to NumPy array**

In [7]:

```
# CORPUS CREATION
texts = just_dial
In [8]:
# ANALYZING THE TEXT BY MEANS SKLEARN AND NUMPY
import numpy as np
```

```
import numpy as np
from sklearn.feature_extraction.text import CountVectorizer

vectorizer = CountVectorizer(texts)
dtm = vectorizer.fit_transform(texts)
vocab = vectorizer.get_feature_names()
```

```
dtm
Out[9]:
<2x20276 sparse matrix of type '<class 'numpy.int64'>'
        with 23077 stored elements in Compressed Sparse Row format>
In [10]:
dtm = dtm.toarray()
vocab = np.array(vocab)
Collocations
In [94]:
import nltk
from nltk.collocations import *
bigram_measures = nltk.collocations.BigramAssocMeasures()
trigram_measures = nltk.collocations.TrigramAssocMeasures()
In [95]:
# input
finder = BigramCollocationFinder.from words(just dial tok)
# finder = BigramCollocationFinder.from words(clem paed tok, window size = 5)
In [96]:
# filter
finder.apply_freq_filter(5)
# selection
best 20 = sorted(finder.nbest(bigram measures.raw freq, 20))
In [97]:
# to form a numpy array
import numpy as np
best_20 = np.array(best_20)
reversed_best_20 = np.flip(best_20, 1)
```

In [9]:

```
In [98]:
bgs = nltk.bigrams(clem paed tok)
fdist = nltk.FreqDist(bgs)
for k,v in fdist.items():
    print(k,v)
('κεφάλαια', 'πρώτου') 1
('πρώτου', 'λόγου') 1
('λόγου', 'ἐπαγγέλλεται') 1
('ἐπαγγέλλεται', 'παιδαγωγός') 2
('παιδαγωγός', 'τὰς') 1
('τὰς', 'ἁμαρτίας') 10
('ἁμαρτίας', 'ἔγώ') 2
('ἔγώ', 'παιδαγωγὸς') 7
('παιδαγωγὸς', 'ἐπιστατέω') 2
('έπιστατέω', 'γ') 1
('Υ', 'φιλάνθρωπος') 1
('φιλάνθρωπος', 'παιδαγωγός') 2
('παιδαγωγός', 'δ') 1
('\delta', ' \dot{\epsilon}\pi'') 1
('ἐπ'', 'ἴσος') 9
('ἴσος', 'ἀνδρόω') 2
('ἀνδρόω', 'γυνή') 4
('γυνή', 'λόγος') 2
('λόγος', 'παιδαγωγός') 3
In [139]:
finder = BigramCollocationFinder.from words(just dial tok, window size = 3)
pairs list = sorted(finder.ngram fd.items(), key=lambda t: (-t[1], t[0]))
In [140]:
pairs list
Out[140]:
[(('oὐ', 'εἰμί'), 74),
 (('σύ', 'σύ'), 50),
 (('αὐτός', 'εἰμί'), 48),
(('αὐτός', 'αὐτός'), 47),
 (('εἰμί', 'αὐτός'), 39),
 (('σύ', 'εἰμί'), 39),
 (('\vec{\epsilon}\gamma\acute{\omega}', '\vec{\epsilon}\gamma\acute{\omega}'), 39),
 (('εἰμί', 'σύ'), 36),
 (('εἰμί', 'ἐγώ'), 34),
 (('σύ', 'οὐ'), 32),
 (('αὐτός', 'ού'), 31),
 (('ἔγώ', 'οὐ'), 31),
 (('ἐγώ', 'εἰμί'), 28),
 (('εἰμί', 'εἰμί'), 27),
 (('εἰμί', 'οὖ'), 26),
 (('σύ', 'ἐγώ'), 26),
(('ἐγώ', 'σύ'), 26),
 (('εἰμί', 'τούσων'), 25),
```

```
import numpy as np
pairs_array = np.array(pairs_list)

Short version (300 pairs)

In [193]:
pairs_array_short = pairs_array[:300]

In [199]:
# changing the data structure

def column(matrix, i):
    return [row[i] for row in matrix]
pairs_only = column(pairs_array_short, 0)
```

def column(matrix, i):

In [200]:

In [201]:

Out[201]:

counts combined

array([['οὐ', 'εἰμί', '74'],

['σύ', 'σύ', '50'],

['σύ', 'εἰμί', '39'], ['ἐγώ', 'ἐγώ', '39'], ['εἰμί', 'σύ', '36'], ['εἰμί', 'ἐγώ', '34'], ['σύ', 'οὐ', '32'],

['αὐτός', 'οὐ', '31'], ['ἐγώ', 'οὐ', '31'], ['ἐγώ', 'εἰμί', '28'],

['εἰμί', 'εἰμί', '27'], ['εἰμί', 'οὐ', '26'],

['σύ', 'ἐγώ', '26'], ['ἐγώ', 'σύ', '26'],

['εἰμί', 'τρύφων', '25'],

['αὐτός', 'εἰμί', '48'], ['αὐτός', 'αὐτός', '47'], ['εἰμί', 'αὐτός', '39'],

return [row[i] for row in matrix]
counts\_only = column(pairs\_array\_short, 1)

counts\_combined = np.insert(pairs\_only, 2, counts\_only, axis=1)

pairs\_only = np.array(pairs\_only)
counts\_only = np.array(counts\_only)

## **NetworkX**

```
In [202]:
import networkx as nx
G = nx.Graph()
In [203]:
G.clear()
G.add_weighted_edges_from(counts_combined)
In [204]:
G.number of nodes()
Out[204]:
164
In [205]:
G.number_of_edges()
Out[205]:
254
In [206]:
list(G.edges())
Out[206]:
[('οὐ', 'εἰμί'),
 ('οὐ', 'σύ'),
 ('οὐ', 'αὐτός'),
 ('οὐ', 'ἔγώ'),
       'ἀλλ''),
 ('oὐ',
 ('οὐ', 'μόνον'),
 ('οὐ', 'οἳ'),
 ('οὐ', 'ἔθνος'),
 ('οὐ', 'φημί'),
 ('οὐ', 'ἔχω'),
 ('oὐ',
       ' εἰπον' ) ,
 ('οὐ',
       'θεάομαι'),
 ('oὐ', 'oὐ'),
 ('οὐ', 'γιγνώσκω'),
 ('ού', 'ἄνθρωπος'),
 ('οὐ', 'αὐτόν'),
 ('οὐ', 'γαῖα'),
 ('οὐ', 'κύριος'),
```

```
nx.degree_centrality(G)
Out[207]:
{'αὐτός': 0.3067484662576687,
 'αἰών': 0.03680981595092025,
 'αἰῶνος·': 0.006134969325153374,
 'αἴρω': 0.006134969325153374,
 '\alpha i\mu\alpha': 0.006134969325153374,
 'αὐτὸν': 0.03680981595092025,
 'αὐτόν': 0.012269938650306749,
 'αὐτῷ・': 0.006134969325153374,
 'βασιλεὺς': 0.012269938650306749,
 'γαῖα': 0.03067484662576687,
 'γαστρὶ': 0.024539877300613498,
 'γεννάω': 0.006134969325153374,
 'γιγνώσκω': 0.024539877300613498,
 'γραφάς': 0.006134969325153374,
 'γένει': 0.012269938650306749,
 'γένους': 0.018404907975460124,
 \delta(\delta\omega\mu): 0.006134969325153374,
 'δαμασκός': 0.024539877300613498,
In [208]:
G
Out[208]:
<networkx.classes.graph.Graph at 0x120a70550>
In [209]:
nx.write_gexf(G, "test.gexf")
In [ ]:
```

In [207]: