

## IM-UH 1511 Introduction to Digital Humanities

**HOMEWORK 9****Network of synonyms from NLTK WordNet****50 points totally**

```
In [1]: import math, random, pickle, collections, operator, string, community #!pip
import itertools as it
import pandas as pd
import numpy as np
import networkx as nx
from networkx import NetworkXNoPath
from networkx.drawing.nx_agraph import graphviz_layout
import pygraphviz
import matplotlib.pyplot as plt
import matplotlib as mpl
from nltk.corpus import wordnet as wn
import warnings
warnings.filterwarnings('ignore')
warnings.filterwarnings("ignore", category=RuntimeWarning)
```

```

In [2]: def g_diagnostics(G,st):

    print("The %s has %i nodes and %i edges \n" %(st,len(G.nodes()), len(G.

    if G.is_directed()==True:
        print("The %s is a directed graph" %st)
    else:
        print("The %s is an undirected graph" %st)
    if nx.is_weighted(G)==True:
        print("The %s graph is a weighted graph" %st)
    else:
        print("The %s graph is an unweighted graph" %st)
    if G.is_directed()==False:
        if nx.is_connected(G)==True:
            print("The %s is a connected graph" %st)
        else:
            print("The %s graph is a disconnected graph and it has %i conne
            giant = max(nx.connected_components(G), key=len)
            Glcc=G.subgraph(giant)
            print("The largest connected component of this graph has %i nod
    else:
        if nx.is_strongly_connected(G)==True:
            print("The %s is a strongly connected graph" %st)
        else:
            print("The %s graph is not strongly connected and it has %i str
            giant = max(nx.strongly_connected_components(G), key=len)
            Glcc=G.subgraph(giant)
            print("The largest strongly connected component of this graph h
        if nx.is_weakly_connected(G)==True:
            print("The %s is a weakly connected graph" %st)
        else:
            print("The %s graph is not weakly connected and it has %i weakl
            giantw = max(nx.weakly_connected_components(G), key=len)
            Glwcc=G.subgraph(giantw)
            print("The largest weakly connected component of this graph has
    print("The density of the %s is %.3f" %(st,nx.density(G)))
    print("The transitivity of the %s is %.3f" %(st,nx.transitivity(G)))
    if G.is_directed()==True:
        print("The reciprocity of the %s is %.3f" %(st,nx.reciprocity(G)))

def g_diameter(G,st):
    try:
        diameter=nx.diameter(G)
        print("The diameter of the largest strongly connected component of
    except Exception as e:
        print(e)

```

```
In [3]: def syn_ant(word):
        synonyms = []
        isynonyms=[]
        antonyms = []
        for syn in wn.synsets(word):
            for l in syn.lemmas():
                synonyms.append(l.name())
                if l.antonyms():
                    antonyms.append(l.antonyms()[0].name())
        synonyms = sorted(set([w for w in synonyms if w!=word and w.lower()!=word]))
        synonyms = sorted(set([w for w in synonyms if "ise" not in w and "isati
        antonyms = sorted(set([w for w in antonyms if w!=word and w.lower()!=word]))
        return (synonyms,antonyms) #isynonyms,
```

## The WordNet is lexical database in NLTK

There exist 117659 words in the WordNet database. However, some of them have empty sets of synonyms or antonyms.

```
In [4]: word = "good"
        sa=syn_ant(word)
        print("The synonyms of '%s' are: \n %s \n" %(word,sa[0]))
        print("The antonyms of '%s' are: \n %s" %(word,sa[1]))
```

The synonyms of 'good' are:

```
['adept', 'beneficial', 'commodity', 'dear', 'dependable', 'effective',
'estimable', 'expert', 'full', 'honest', 'honorable', 'in_effect', 'in_force',
'just', 'near', 'practiced', 'proficient', 'respectable', 'right',
'ripe', 'safe', 'salutary', 'secure', 'serious', 'skilful', 'skillful',
'sound', 'soundly', 'thoroughly', 'undecomposed', 'unspoiled', 'unspoil
t', 'upright', 'well']
```

The antonyms of 'good' are:

```
['bad', 'badness', 'evil', 'evilness', 'ill']
```

```
In [5]: word = "coronavirus"
        sa=syn_ant(word)
        print("The synonyms of '%s' are: \n %s \n" %(word,sa[0]))
        print("The antonyms of '%s' are: \n %s" %(word,sa[1]))
```

The synonyms of 'coronavirus' are:

```
[]
```

The antonyms of 'coronavirus' are:

```
[]
```

```
In [6]: word = "virus"
sa=syn_ant(word)
print("The synonyms of '%s' are: \n %s \n" %(word,sa[0]))
print("The antonyms of '%s' are: \n %s" %(word,sa[1]))
```

The synonyms of 'virus' are:  
[]

The antonyms of 'virus' are:  
[]

```
In [7]: word = 'ammoniac'
sa=syn_ant(word)
print("The synonyms of '%s' are: \n %s \n" %(word,sa[0]))
print("The antonyms of '%s' are: \n %s" %(word,sa[1]))
```

The synonyms of 'ammoniac' are:  
[]

The antonyms of 'ammoniac' are:  
[]

```
In [8]: word = 'conventionalization'
sa=syn_ant(word)
print("The synonyms of '%s' are: \n %s \n" %(word,sa[0]))
print("The antonyms of '%s' are: \n %s" %(word,sa[1]))
```

The synonyms of 'conventionalization' are:  
[]

The antonyms of 'conventionalization' are:  
[]

```
In [9]: word1 = "marriage"
sa1=syn_ant(word1)
word2 = "education"
sa2=syn_ant(word2)
word3 = "success"
sa3=syn_ant(word3)
print("The synonyms of '%s' are: \n %s" %(word1,sa1[0]))
print("The antonyms of '%s' are: \n %s \n" %(word1,sa1[1]))
print("The synonyms of '%s' are: \n %s" %(word2,sa2[0]))
print("The antonyms of '%s' are: \n %s \n" %(word2,sa2[1]))
print("The synonyms of '%s' are: \n %s" %(word3,sa3[0]))
print("The antonyms of '%s' are: \n %s \n" %(word3,sa3[1]))
```

The synonyms of 'marriage' are:

['man\_and\_wife', 'married\_couple', 'matrimony', 'spousal\_relationship', 'union', 'wedding', 'wedlock']

The antonyms of 'marriage' are:

[]

The synonyms of 'education' are:

['Department\_of\_Education', 'Education\_Department', 'breeding', 'didactics', 'instruction', 'pedagogy', 'teaching', 'training']

The antonyms of 'education' are:

[]

The synonyms of 'success' are:

['achiever', 'succeeder', 'winner']

The antonyms of 'success' are:

['failure', 'loser']

```
In [10]: # https://www.nltk.org/howto/wordnet.html

# synset1.path_similarity(synset2):
# Return a score denoting how similar two word senses are, based on the shortest path that connects the senses in the is-a (hypernym/hypnoym) taxonomy.
# The score is in the range 0 to 1.

# synset1.lch_similarity(synset2):
# Leacock-Chodorow Similarity: Return a score denoting how similar two word senses are, based on the shortest path that connects the senses (as above)
# and the maximum depth of the taxonomy in which the senses occur. The relationship is calculated as  $-\log(p/2d)$  where  $p$  is the shortest path length and  $d$  the taxonomy depth.

# synset1.wup_similarity(synset2):
# Wu-Palmer Similarity: Return a score denoting how similar two word senses are, based on the depth of the two senses in the taxonomy and the depth of their Least Common Subsumer (most specific ancestor node).

w1='good'
w2='dear'
w3='bad'
w1s=wn.synsets(w1)[0]
w2s=wn.synsets(w2)[0]
w3s=wn.synsets(w3)[0]
print("The path similarity between %s and %s is %s" % (w1,w2,w1s.path_similarity(w2s)))
print("The path similarity between %s and %s is %s \n" % (w1,w3,w1s.path_similarity(w3s)))
print("The lch similarity between %s and %s is %s" % (w1,w2,w1s.lch_similarity(w2s)))
print("The lch similarity between %s and %s is %s \n" % (w1,w3,w1s.lch_similarity(w3s)))
print("The wup similarity between %s and %s is %s" % (w1,w2,w1s.wup_similarity(w2s)))
print("The wup similarity between %s and %s is %s" % (w1,w3,w1s.wup_similarity(w3s)))
```

The path similarity between good and dear is 0.08333333333333333  
The path similarity between good and bad is 0.2

The lch similarity between good and dear is 1.1526795099383855  
The lch similarity between good and bad is 2.0281482472922856

The wup similarity between good and dear is 0.15384615384615385  
The wup similarity between good and bad is 0.6666666666666666

## Graph of References of Words to Synonyms

```
In [11]: # nos=10  #number of words

# wns=list(wn.all_synsets()) #117659
# syn_d={}
# for w in random.sample(wns,nos): #wns: #
#     w=w.lemmas()[0].name()
#     sl=syn_ant(w)
#     sl=syn_ant(w)[0]
#     sl=list(set(sl))
#     if len(sl)>0:
#         syn_d[w]=sl

# for k,v in syn_d.items():
#     print("The synonyms of '%s' are %s \n" %(k,v))
```

```
In [12]: nos=1500  #25000 #number of words

wns=list(wn.all_synsets())
syn_d={}
for w in random.sample(wns,nos): #wns: #
    w=w.lemmas()[0].name()
    sl=syn_ant(w)
    sl=syn_ant(w)[0]
    sl=list(set(sl))
    if len(sl)>0:
        syn_d[w]=sl

# for k,v in syn_d.items():
#     print("The synonyms of '%s' are %s \n" %(k,v))
```

```
In [13]: eds=[]
for k,v in syn_d.items():
    for vv in v:
        if vv in syn_d.keys():
            ks=wn.synsets(k)[0]
            vvs=wn.synsets(vv)[0]
            simi=ks.path_similarity(vvs)
            if simi==None:
                asimi=0
            else:
                asimi=1-ks.path_similarity(vvs)
            eds.append((k,vv,asimi))
print(len(eds),len(set(eds)))
eds
```

173 173

```
In [14]: G=nx.DiGraph()
G.add_weighted_edges_from(eds)

st="graph among %i words and their synonyms" %len(G)
g_diagnostics(G,st)
```

The graph among 129 words and their synonyms has 129 nodes and 173 edges

The graph among 129 words and their synonyms is a directed graph

The graph among 129 words and their synonyms graph is a weighted graph

The graph among 129 words and their synonyms graph is not strongly connected and it has 55 strongly connected components

The largest strongly connected component of this graph has 35 nodes and 70 edges

The graph among 129 words and their synonyms graph is not weakly connected and it has 36 weakly connected components

The largest weakly connected component of this graph has 49 nodes and 93 edges

The density of the graph among 129 words and their synonyms is 0.010

The transitivity of the graph among 129 words and their synonyms is 0.024

The reciprocity of the graph among 129 words and their synonyms is 0.867



```

In [15]: edge_width=[G[u][v]['weight'] for u,v in G.edges()]
edge_width=[w if type(w)==float else 0 for w in edge_width]
edge_width=[1*math.log(1.3+w) for w in edge_width]

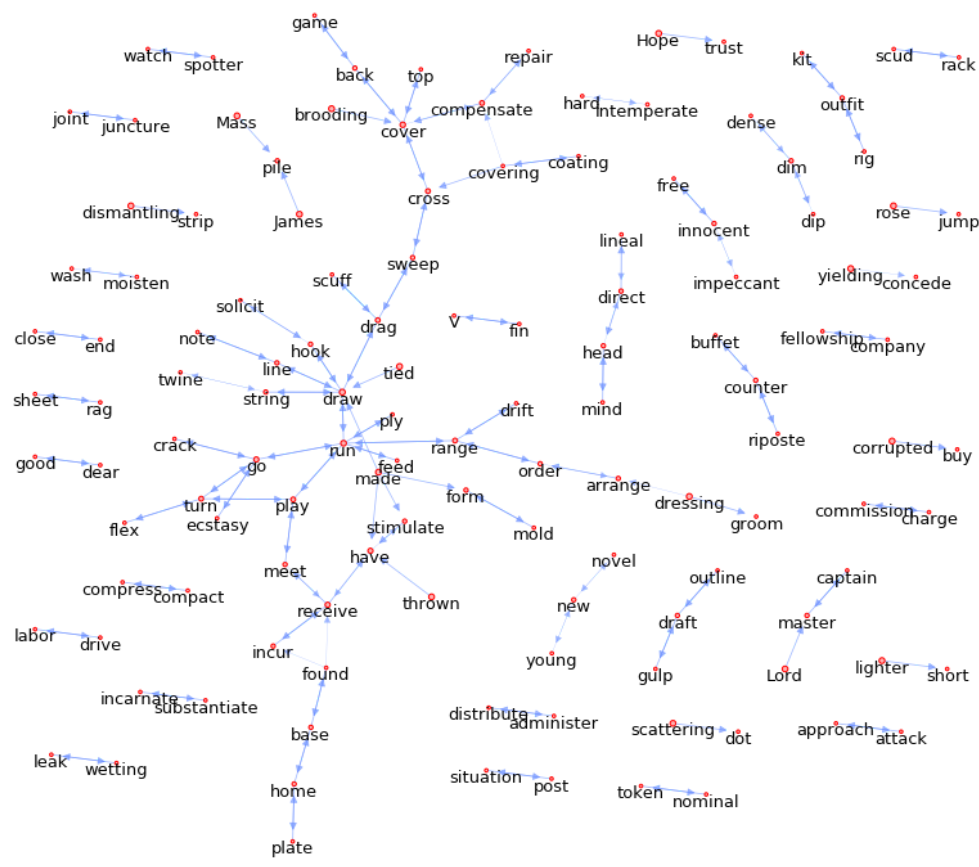
nsi=[]
for n in G.nodes():
    if G.in_degree(n)>0:
        nsi.append(10*math.log(1+G.in_degree(n)))
    else:
        nsi.append(20)
figsize=(17,13)
pos=graphviz_layout(G)

labels={}
for n in G.nodes():
    labels[n]=" "

node_color="#ffb3b3"
node_border_color="r"
edge_color="#668cff"
plt.figure(figsize=figsize);
nodes = nx.draw_networkx_nodes(G, pos, node_color=node_color,node_size=nsi)
nodes.set_edgecolor(node_border_color)
nx.draw_networkx_edges(G, pos,arrowsize=12, width=edge_width,edge_color=edge_color)
# nx.draw_networkx_labels(G,pos,labels=labels)
# nx.draw_networkx_edge_labels(G,pos,edge_labels=elabels);
plt.axis('off');
yoffset = {}
y_off = -10 # offset on the y axis
for k, v in pos.items():
    yoffset[k] = (v[0], v[1]+y_off)
nx.draw_networkx_labels(G, yoffset,font_size=13);
# st1="graph of %i words \n in a random sample of %i words and their synony
sst="The directed %s" %st
plt.title(sst,fontsize=20);
plt.margins(x=0.1, y=0)

```

The directed graph among 129 words and their synonyms



```

In [16]: # nsi=[]
# for n in G.nodes():
#     if G.in_degree(n)>0:
#         nsi.append(10*math.log(1+G.in_degree(n)))
#     else:
#         nsi.append(20)
# figsize=(17,13)
# pos=graphviz_layout(G)

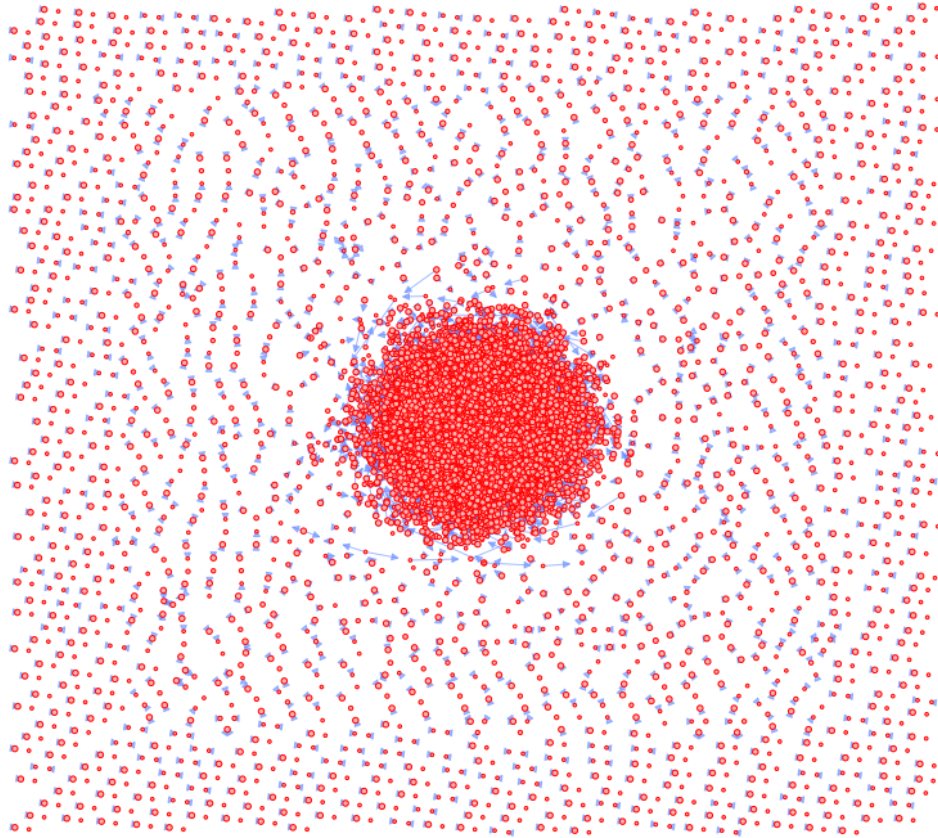
# labels={}
# for n in G.nodes():
#     labels[n]=" "

# node_color="#ffb3b3"
# node_border_color="r"
# edge_color="#668cff"
# plt.figure(figsize=figsize);
# nodes = nx.draw_networkx_nodes(G, pos, node_color=node_color,node_size=ns
# nodes.set_edgecolor(node_border_color)
# nx.draw_networkx_edges(G, pos,arrowsize=12, edge_color=edge_color,alpha=0
# nx.draw_networkx_labels(G,pos,labels=labels)
# # nx.draw_networkx_edge_labels(G,pos,edge_labels=elabels);
# plt.axis('off');
# yoffset = {}
# y_off = -20 # offset on the y axis
# for k, v in pos.items():
#     yoffset[k] = (v[0], v[1]+y_off)
# # nx.draw_networkx_labels(G, yoffset,font_size=13);
# # st1="graph of %i words \n in a random sample of %i words and their syno
# sst="The directed %s" %st
# plt.title(sst,fontsize=20);
# plt.margins(x=0.1, y=0)

```

```
In [17]: from IPython.display import Image  
Image(filename='allDirected.png',width=800, height=400)
```

Out[17]: The directed graph among 10741 words and their synonyms



```
In [18]: giant = max(nx.weakly_connected_components(G), key=len)
Glwcc=G.subgraph(giant)

st1="weakly connected component of the directed %s" %st
# graph of %i words \n in a random sample of %i words and their synonyms" %
g_diagnostics(Glwcc,st1)
```

The weakly connected component of the directed graph among 129 words and their synonyms has 49 nodes and 93 edges

The weakly connected component of the directed graph among 129 words and their synonyms is a directed graph

The weakly connected component of the directed graph among 129 words and their synonyms graph is a weighted graph

The weakly connected component of the directed graph among 129 words and their synonyms graph is not strongly connected and it has 10 strongly connected components

The largest strongly connected component of this graph has 35 nodes and 70 edges

The weakly connected component of the directed graph among 129 words and their synonyms is a weakly connected graph

The density of the weakly connected component of the directed graph among 129 words and their synonyms is 0.040

The transitivity of the weakly connected component of the directed graph among 129 words and their synonyms is 0.026

The reciprocity of the weakly connected component of the directed graph among 129 words and their synonyms is 0.860

```

In [19]: edge_width=[Glwcc[u][v]['weight'] for u,v in Glwcc.edges()]
edge_width=[w if type(w)==float else 0 for w in edge_width]
edge_width=[7*math.log(1.3+w) for w in edge_width]

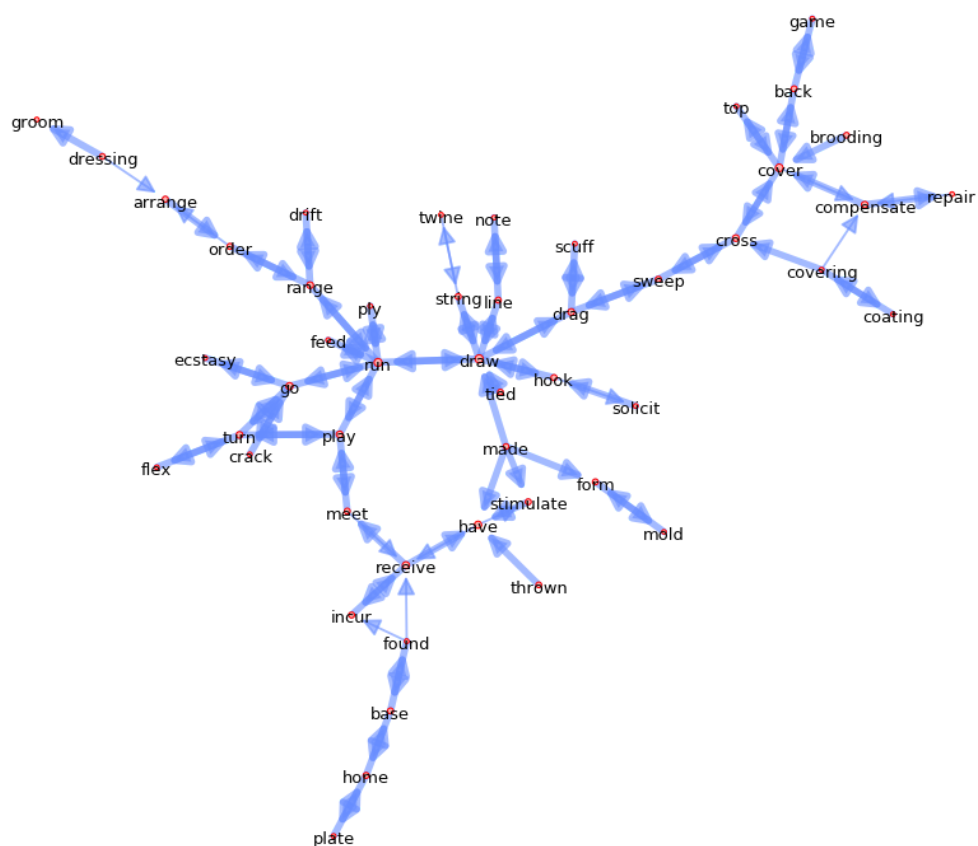
nsi=[]
for n in Glwcc.nodes():
    if Glwcc.in_degree(n)>0:
        nsi.append(20*math.log(1+Glwcc.in_degree(n)))
    else:
        nsi.append(20)
figsize=(17,13)
pos=graphviz_layout(Glwcc)

labels={}
for n in Glwcc.nodes():
    labels[n]=" "

node_color="#ffb3b3"
node_border_color="r"
edge_color="#668cff"
plt.figure(figsize=figsize);
nodes = nx.draw_networkx_nodes(Glwcc, pos, node_color=node_color,node_size=
nodes.set_edgecolor(node_border_color)
nx.draw_networkx_edges(Glwcc, pos,arrowsize=30,width=edge_width, edge_color
# nx.draw_networkx_labels(Glwcc,pos,labels=labels)
# nx.draw_networkx_edge_labels(G,pos,edge_labels=elabels);
plt.axis('off');
yoffset = {}
y_off = -4 # offset on the y axis
for k, v in pos.items():
    yoffset[k] = (v[0], v[1]+y_off)
nx.draw_networkx_labels(Glwcc, yoffset,font_size=13);
stl="The weakly connected component of the directed %s" %st
# graph of %i words \n in a random sample of %i words and their synonyms" %
sst=stl
plt.title(sst,fontsize=20);
plt.margins(x=0.1, y=0)

```

The weakly connected component of the directed graph among 129 words and their synonyms



```

In [20]: # nsi=[]
# for n in Glscn.nodes():
#     if Glscn.in_degree(n)>0:
#         nsi.append(20*math.log(1+Glscn.in_degree(n)))
#     else:
#         nsi.append(20)
# figsize=(17,13)
# pos=graphviz_layout(Glscn)

# labels={}
# for n in Glscn.nodes():
#     labels[n]=" "

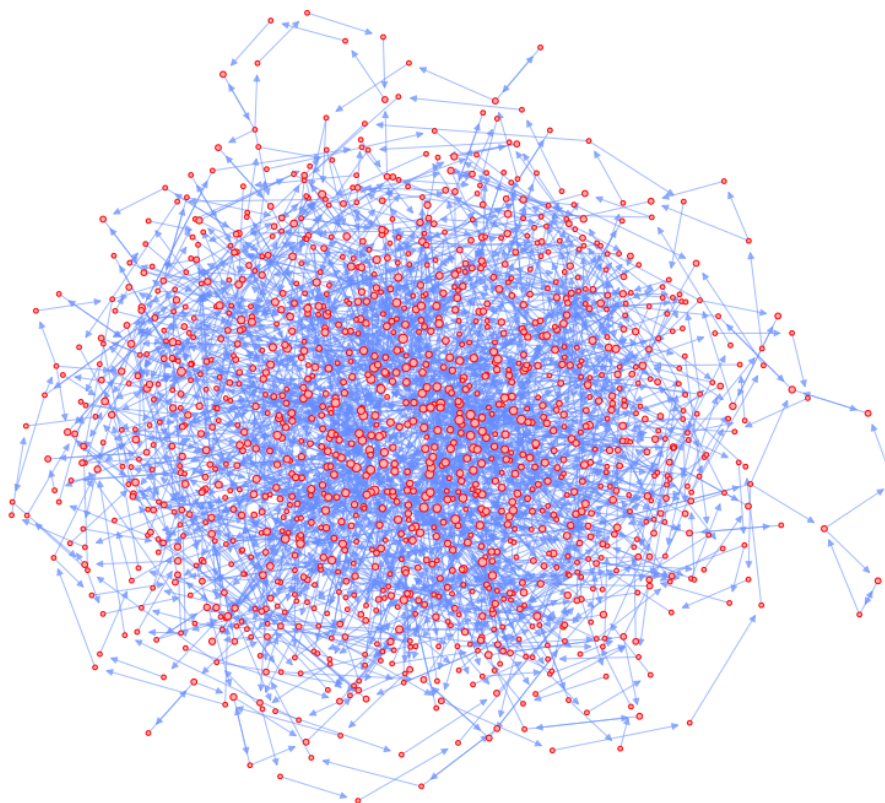
# node_color="#ffb3b3"
# node_border_color="r"
# edge_color="#668cff"
# plt.figure(figsize=figsize);
# nodes = nx.draw_networkx_nodes(Glscn, pos, node_color=node_color,node_size=100)
# nodes.set_edgecolor(node_border_color)
# nx.draw_networkx_edges(Glscn, pos,arrowsize=12, edge_color=edge_color,alpha=0.5)
# nx.draw_networkx_labels(Glscn,pos,labels=labels)
# # nx.draw_networkx_edge_labels(G,pos,edge_labels=elabels);
# plt.axis('off');
# yoffset = {}
# y_off = -10 # offset on the y axis
# for k, v in pos.items():
#     yoffset[k] = (v[0], v[1]+y_off)
# # nx.draw_networkx_labels(Glscn, yoffset,font_size=13);
# stl="The strongly connected component of the directed %s" %st
# # graph of %i words \n in a random sample of %i words and their synonyms"
# sst=stl
# plt.title(sst,fontsize=20);
# plt.margins(x=0.1, y=0)

```



```
In [21]: Image(filename='allDirectedLSCC.png',width=800, height=400)
```

Out[21]: The strongly connected component of the directed graph among 10741 words and their synonyms



```
In [22]: # for e in G.edges(data=True):  
#         print(e)
```

```
In [23]: reds=[]
for e in G.edges(data=True):
    if (e[1],e[0]) in G.edges():
        reds.append(e)
print(len(reds)) #,len(set(reds))
# for e in reds:
#     if (e[1],e[0]) in reds:
#         reds.remove(e)
# print(len(reds)) #,len(set(reds))
```

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```
In [24]: # for e in reds:
#         print(e)
```

```
In [25]: Gr=nx.Graph()
Gr.add_weighted_edges_from(reds)

stl="subgraph of reciprocating references among %i words and their synonyms
g_diagnostics(Gr,stl)
```

The subgraph of reciprocating references among 105 words and their synonyms has 105 nodes and 75 edges

The subgraph of reciprocating references among 105 words and their synonyms is an undirected graph

The subgraph of reciprocating references among 105 words and their synonyms graph is a weighted graph

The subgraph of reciprocating references among 105 words and their synonyms graph is a disconnected graph and it has 31 connected components

The largest connected component of this graph has 35 nodes and 35 edges

The density of the subgraph of reciprocating references among 105 words and their synonyms is 0.014

The transitivity of the subgraph of reciprocating references among 105 words and their synonyms is 0.000

```
In [26]: # for e in Gr.edges(data=True):
#         print(e)
```

```

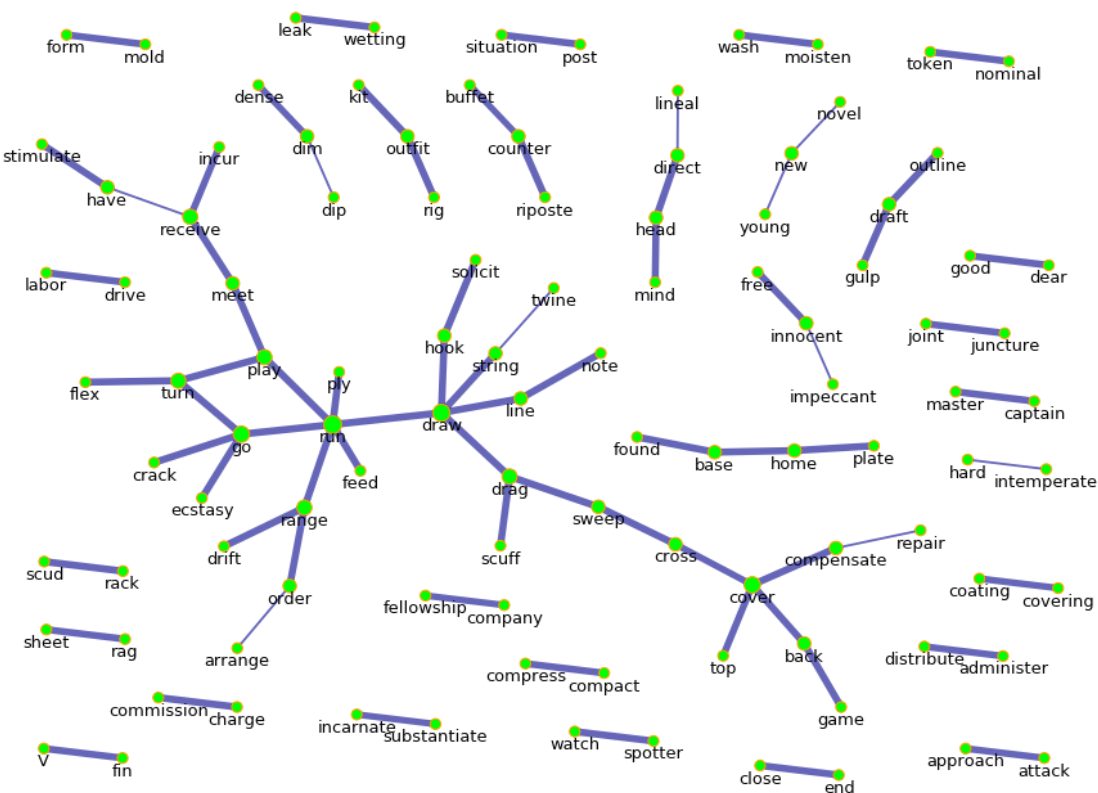
In [27]: edge_width=[Gr[u][v]['weight'] for u,v in Gr.edges()]
edge_width=[w if type(w)==float else 0 for w in edge_width]
edge_width=[7*math.log(1.3+w) for w in edge_width]

nsi=[]
for n in Gr.nodes():
    if Gr.degree(n)>0:
        nsi.append(100*math.log(1+Gr.degree(n)))
    else:
        nsi.append(20)
figsize=(17,13)
pos=graphviz_layout(Gr)

labels={}
for n in Gr.nodes():
    labels[n]=" "

node_color="lime"
node_border_color="orange"
edge_color="darkblue"
plt.figure(figsize=figsize);
nodes = nx.draw_networkx_nodes(Gr, pos, node_color=node_color,node_size=nsi)
nodes.set_edgecolor(node_border_color)
nx.draw_networkx_edges(Gr, pos,width=edge_width,edge_color=edge_color,alpha
# nx.draw_networkx_labels(G, pos)
# nx.draw_networkx_edge_labels(G,pos,edge_labels=elabels);
plt.axis('off');
yoffset = {}
y_off = -15 # offset on the y axis
for k, v in pos.items():
    yoffset[k] = (v[0], v[1]+y_off)
nx.draw_networkx_labels(Gr, yoffset,font_size=13);
# st1="subgraph of %i reciprocating references to synonyms \n in the %s" %(
sst="The undirected %s" %st1
plt.title(sst,fontsize=20);
plt.margins(x=0.1, y=0.1)

```



```

In [28]: # nsi=[]
# for n in Gr.nodes():
#     if Gr.degree(n)>0:
#         nsi.append(100*math.log(1+Gr.degree(n)))
#     else:
#         nsi.append(20)
# figsize=(17,13)
# pos=graphviz_layout(Gr)

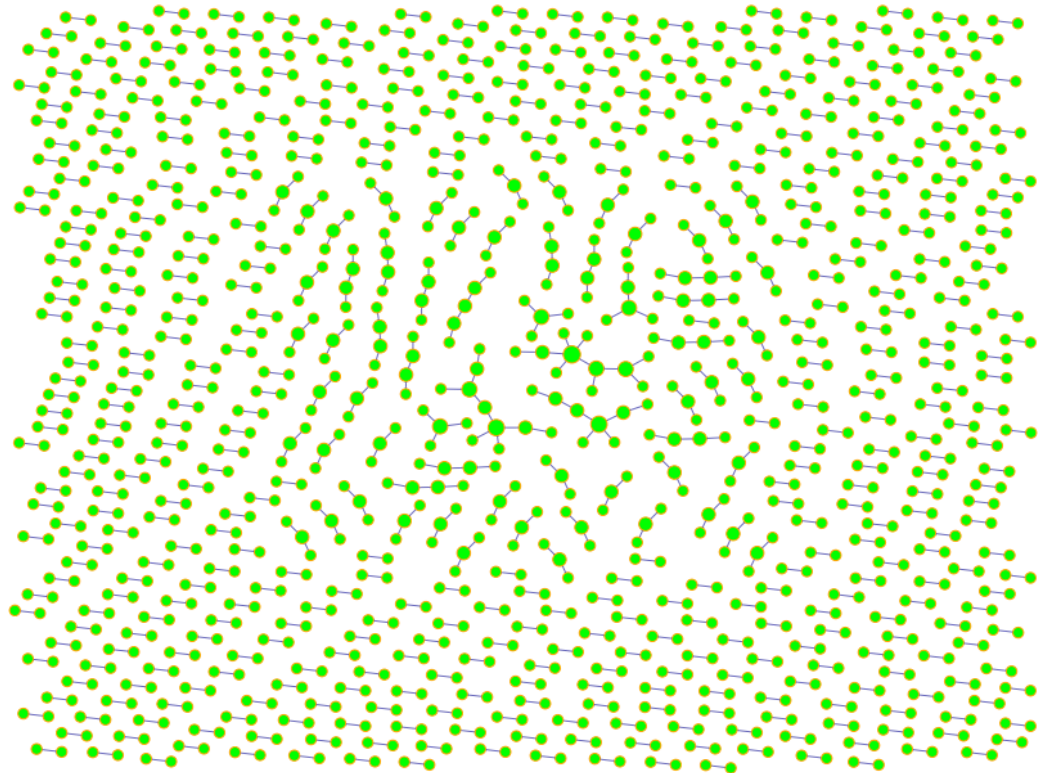
# labels={}
# for n in Glscn.nodes():
#     labels[n]=" "

# node_color="lime"
# node_border_color="orange"
# edge_color="darkblue"
# plt.figure(figsize=figsize);
# nodes = nx.draw_networkx_nodes(Gr, pos, node_color=node_color,node_size=n
# nodes.set_edgecolor(node_border_color)
# nx.draw_networkx_edges(Gr, pos,edge_color=edge_color,alpha=0.6)
# # nx.draw_networkx_labels(G, pos)
# # nx.draw_networkx_edge_labels(G,pos,edge_labels=elabels);
# plt.axis('off');
# yoffset = {}
# y_off = -15 # offset on the y axis
# for k, v in pos.items():
#     yoffset[k] = (v[0], v[1]+y_off)
# # nx.draw_networkx_labels(Gr, yoffset,font_size=13);
# # st1="subgraph of %i reciprocating references to synonyms \n in the %s"
# sst="The undirected %s" %st1
# plt.title(sst,fontsize=20);
# plt.margins(x=0.1, y=0.1)

```

```
In [29]: Image(filename='allReci.png',width=800, height=400)
```

Out[29]: The undirected subgraph of reciprocating references among 1186 words and their synonyms



```
In [30]: giant = max(nx.connected_components(Gr), key=len)
Gr1scc=Gr.subgraph(giant)

st2="largest connected component of the undirected %s" %st1
# graph of reciprocating synonyms of %i words \n in a random sample of %i w
g_diagnostics(Gr1scc,st2)
```

The largest connected component of the undirected subgraph of reciprocating references among 105 words and their synonyms has 35 nodes and 35 edges

The largest connected component of the undirected subgraph of reciprocating references among 105 words and their synonyms is an undirected graph  
The largest connected component of the undirected subgraph of reciprocating references among 105 words and their synonyms graph is a weighted graph

The largest connected component of the undirected subgraph of reciprocating references among 105 words and their synonyms is a connected graph  
The density of the largest connected component of the undirected subgraph of reciprocating references among 105 words and their synonyms is 0.059  
The transitivity of the largest connected component of the undirected subgraph of reciprocating references among 105 words and their synonyms is 0.000

```

In [31]: edge_width=[Grlscc[u][v]['weight'] for u,v in Grlscc.edges()]
edge_width=[w if type(w)==float else 0 for w in edge_width]
edge_width=[7*math.log(1.3+w) for w in edge_width]

nsi=[]
for n in Grlscc.nodes():
    if Grlscc.degree(n)>0:
        nsi.append(100*math.log(1+Grlscc.degree(n)))
    else:
        nsi.append(20)
figsize=(17,13)
pos=graphviz_layout(Grlscc)

node_color="lime"
node_border_color="orange"
edge_color="darkblue"
plt.figure(figsize=figsize);
nodes = nx.draw_networkx_nodes(Grlscc, pos, node_color=node_color,node_size
nodes.set_edgecolor(node_border_color)
nx.draw_networkx_edges(Grlscc, pos,width=edge_width,edge_color=edge_color,a
# nx.draw_networkx_labels(G, pos)
# nx.draw_networkx_edge_labels(G,pos,edge_labels=elabels);
plt.axis('off');
yoffset = {}
y_off = -4 # offset on the y axis
for k, v in pos.items():
    yoffset[k] = (v[0], v[1]+y_off)
nx.draw_networkx_labels(Grlscc, yoffset,font_size=13);
# st3="The largest connected component of the undirected %s" %st1
st3="The largest connected component of the \n undirected subgraph of recip
# subgraph of \n %i reciprocating references to synonyms \n in the %s" %(le
sst=st3
plt.title(sst,fontsize=20);
plt.margins(x=0.1, y=0.1)

```



A network graph visualization showing word relationships. The central node is 'run'. Other nodes include 'go', 'play', 'draw', 'range', 'order', 'drift', 'arrange', 'note', 'line', 'hook', 'solicit', 'stuff', 'sweep', 'cross', 'cover', 'back', 'game', 'compensate', 'repair', 'twine', 'string', 'drag', 'feed', 'ply', 'ecstasy', 'crack', 'flex', 'turn', 'meet', 'receive', 'incur', 'have', 'stimulate', 'order', and 'arrange'. Edges connect the central node to its immediate neighbors and further out to other related words.