Load Everything Here

```
[ ] L, 3 cells hidden
```

→ Read Data

```
df = pd.read_csv("Edgelist2019_2.csv")
df
```

	Source	Target	Weight
0	f419	l03115	0.0212
1	f419	103116	0.0224
2	f419	e039	0.1444
3	f419	e11649	0.0124
4	f419	e8342	0.0645
•••			
28920	c6921	c6922	0.7194
28921	t23342a	t23341a	0.6923
28922	t23342a	t23351a	0.5488
28923	t23342a	t23362a	0.7402
28924	t23341a	t23351a	0.7546

28925 rows × 3 columns

Convert to Graph and Visualize


```
%%time
graph=nx.convert_matrix.from_pandas_edgelist(df,source='Source', target='Target', edge_attr=N
graph.name = "Covid DisNet for Edgelist2019_2"
print(nx.info(graph))
print("-----")
```

```
Name: Covid DisNet for Edgelist2019 2
     Type: Graph
     Number of nodes: 2075
     Number of edges: 28925
     Average degree: 27.8795
     CPU times: user 70.4 ms, sys: 0 ns, total: 70.4 ms
     Wall time: 71.9 ms
degree centrality = nx.algorithms.centrality.degree centrality(graph)
first10pairs = {k: degree centrality[k] for k in sorted(degree centrality.keys())[:10]}
first10pairs
     {'a0472': 0.029411764705882353,
      'a0839': 0.00048216007714561236,
      'a084': 0.00048216007714561236,
      'a403': 0.00048216007714561236,
      'a4101': 0.0009643201542912247,
      'a4102': 0.0009643201542912247,
      'a4151': 0.00819672131147541,
      'a4159': 0.0014464802314368371,
      'a4181': 0.00048216007714561236,
      'a419': 0.13211186113789777}
eigenvector centrality = nx.algorithms.centrality.eigenvector centrality numpy(graph)
first10pairs = {k: eigenvector centrality[k] for k in sorted(eigenvector centrality.keys())[:
first10pairs
     {'a0472': 0.02891100865845509,
      'a0839': -5.5534461071812384e-18,
      'a084': 0.0005382745559326981,
      'a403': -4.0986084256558263e-19,
      'a4101': 0.0003591042805007767,
      'a4102': 0.0003591042805007741,
      'a4151': 0.005965411861105286,
      'a4159': 0.0007979795021039903,
      'a4181': 5.059927347811221e-07,
      'a419': 0.08380720406444628}
katz centrality = nx.algorithms.centrality.katz centrality numpy(graph)
first10pairs = {k: katz centrality[k] for k in sorted(katz centrality.keys())[:10]}
first10pairs
     {'a0472': 0.01463210426939267,
      'a0839': 0.00850252717177153,
      'a084': 0.009158457312284665,
      'a403': 0.00850252717177153,
      'a4101': 0.011119140974806818,
      'a4102': 0.011119140974806813,
      'a4151': 0.008266137920715882,
      'a4159': 0.013698653581779189,
```

```
'a4181': 0.00863492411871393,
'a410': 0.00863492411672003\
number_of_triangles = sum(nx.triangles(graph).values()) / 3
number_of_triangles

791705.0

nx.algorithms.cluster.transitivity(graph)

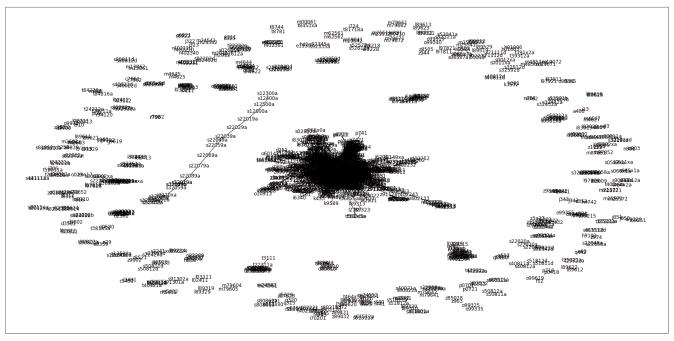
0.49345707598767147

print(nx.average_clustering(graph))
```

0.4677968262043043

whole graph plot

```
%%time
nx.draw_networkx(graph,
                #pos,
                with labels=True,
                node_size=30,
                node color="mistyrose",
                #edgelist=edges,
                #edge_color=weights,
                edge cmap=plt.cm.Accent,
                style="solid",
                width=1)
nx.draw_networkx(graph.subgraph('z20828'), font_size=16,node_size=120, node_color='red')
plt.subplots_adjust(left=1, bottom=3.2, right=4.8, top=6)
plt.show()
print("----")
print("Density:",nx.classes.function.density(graph))
```



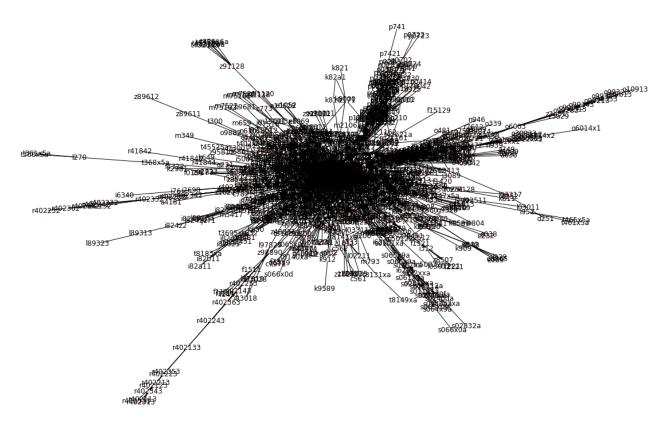
Density: 0.013442390584517433

CPU times: user 28.7 s, sys: 579 ms, total: 29.3 s

Wall time: 29 s

▼ partial graph plot

```
%%time
plt.figure(figsize=(16, 10))
gcc = max(nx.connected_components(graph), key=lambda x: len(x))
H = graph.subgraph(gcc)
nx.draw(H, node_size=30, node_color='mistyrose',with_labels=True,edge_cmap=plt.cm.Accent,styl
plt.subplots_adjust(left=1, bottom=3.2, right=4.8, top=6)
plt.show()
print("Density:",nx.classes.function.density(H))
print("-------")
```



Density: 0.024968072360538753

CPU times: user 19.1 s, sys: 431 ms, total: 19.5 s

Wall time: 19.6 s

▼ plot for z20828's neighbors

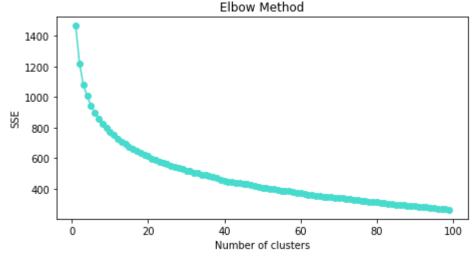
```
%%time
plt.figure(figsize=(16, 10))
Sub = nx.classes.function.induced_subgraph(graph,set(graph.neighbors(n="z20828")))
nx.draw_networkx(Sub, font_size=16,node_size=120, node_color='red')
print("-----")
print("Density:",nx.classes.function.density(Sub))
print("----")
```

```
Traceback (most recent call last)
       KeyError
       /usr/local/lib/python3.7/dist-packages/networkx/classes/graph.py in neighbors(self, n)
          1237
       -> 1238
                           return iter(self. adj[n])
          1239
                      except KeyError as e:
       KeyError: 'z20828'
       The above exception was the direct cause of the following exception:
       NetworkXError
                                                 Traceback (most recent call last)
                                          4 frames
       <decorator-gen-53> in time(self, line, cell, local_ns)
       <timed exec> in <module>()
▼ Fit node2vec
                           raise NetworkXFrror(f"The node {n} is not in the granh ") from e
       -> 1240
  vector_size = round(df.shape[0]**0.25)
  vector size
       13
       OF A DOLL OT A OIZ OVEDELOW
  setup = Node2Vec(graph,dimensions=vector_size, walk_length=5, num_walks=5)
  model = setup.fit(window=10, min count=1)
  print("----")
       Computing transition probabilities: 100%
                                       2075/2075 [00:44<00:00, 46.52it/s]
       Generating walks (CPU: 1): 0%
                                                 | 0/5 [00:00<?, ?it/s]
       Generating walks (CPU: 1): 100% | 5/5 [00:02<00:00, 1.91it/s]
       CPU times: user 48 s, sys: 484 ms, total: 48.5 s
       Wall time: 48.9 s
  %%time
  #vocab, vectors = model.wv.key to index, model.wv.get normed vectors()
  vocab, vectors = model.wv.vocab, model.wv.vectors
  # get node name and embedding vector index.
  name index = np.array([(v[0], v[1].index)) for v in vocab.items()]) #.index
  # init dataframe using embedding vectors and set index as node name
  node2vec output = pd.DataFrame(vectors[name index[:,1].astype(int)])
  node2vec_output.index = name_index[:,0]
       CPU times: user 5.17 ms, sys: 971 μs, total: 6.14 ms
       Wall time: 5.74 ms
```

→ K-means

▼ Find k

```
%%time
SSE = []
for i in range(1,100):
    kmeans = KMeans(n_clusters=i, init='k-means++', max_iter=100, n_init=50, random_state=42)
    kmeans.fit(node2vec_output)
    SSE.append(kmeans.inertia_)
plt.plot(range(1,100), SSE,"o-",color="#47DBCD")
plt.title('Elbow Method')
plt.xlabel('Number of clusters')
plt.ylabel('SSE')
plt.subplots_adjust(left=0.25, bottom=0.8, right=1.2, top=1.5)
plt.show()
```



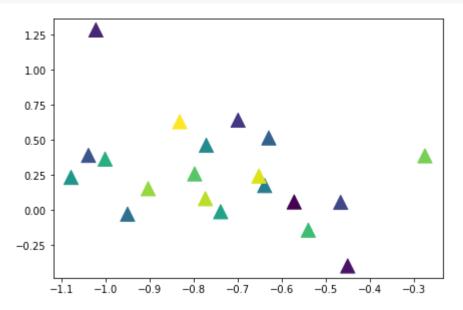
CPU times: user 7min 39s, sys: 5min 51s, total: 13min 30s Wall time: 7min 8s

▼ plot k-means clustering

```
n clusters=kmeans.n iter
https://colab.research.google.com/drive/1EjPfuPE3QoVIfZx4Qi2mF1pxR45U8raP#scrollTo=Y-XZ2eDxZIXd
```

```
kmeans = KMeans(n_clusters=n_clusters, init='k-means++', max_iter=1000, n_init=50, random_stakens.fit(node2vec_output)
```

```
t = np.arange(n_clusters)
plt.scatter(kmeans.cluster_centers_[:,0], kmeans.cluster_centers_[:,1], s=200, c=t,marker="^"
plt.subplots_adjust(left=0.1, bottom=0.1, right=1, top=1)
```



```
subsample=[]
for i in range(kmeans.n_clusters):
   temp = []
   temp=node2vec_output.iloc[kmeans.labels_==i,:]
   subsample.append(temp)
```

```
for list in range(len(subsample)):
    print("Group",list+1)
    print(subsample[list])
    print("-----")
    rows x 13 COLUMNS]
```

```
Group 20
                        1
                                  2
                                                 10
                                                           11
                                                                     12
189310 -0.993056 0.734035 -0.343363
                                      ... -0.365639
                                                     0.012149 0.749330
189320
       -1.015382 0.740550 -0.348918
                                      ... -0.330585
                                                     0.026440
                                                              0.759516
s32059a -1.191844 0.629160 -0.037903
                                      ... -0.271478
                                                     0.420447
                                                              0.703685
t391x2a -0.742426 0.455953 -0.467846
                                                              0.810865
                                      ... -0.174823
                                                     0.495758
t39312a -0.691739  0.492456 -0.419451
                                      ... -0.195004
                                                     0.468967
                                                              0.736829
s32029a -1.055378 0.560430 -0.100037
                                      ... -0.273258
                                                     0.442441
                                                              0.591646
```

```
s50812d -0.577900
                   0.797166 -0.047643
                                        ... -0.311124
                                                        0.413424
                                                                  0.663857
s50811d -0.565609
                   0.818055 -0.178556
                                        ... -0.359861
                                                        0.379919
                                                                  0.584093
s32049a -0.969280
                   0.455477 -0.049946
                                        ... -0.293215
                                                        0.240875
                                                                  0.495105
s32039a -0.989405
                   0.452811 -0.115008
                                        ... -0.333014
                                                        0.362007
                                                                  0.544861
s32019a -1.066514
                   0.608291 -0.185056
                                        ... -0.304432
                                                        0.406221
                                                                  0.556716
s22089a -0.931121
                   0.472039 -0.241528
                                        ... -0.252095
                                                        0.286668
                                                                  0.464168
s92322a -1.155321
                   0.710388 -0.406186
                                        ... -0.201485
                                                        0.041104
                                                                  0.026009
s92342a -1.202957
                   0.748560 -0.452954
                                        ... -0.146870
                                                        0.119967
                                                                  0.017911
s92332a -1.179161
                   0.736252 -0.464957
                                        ... -0.155866
                                                        0.115902
                                                                  0.023813
s50311a -0.628116
                   1.004994 -0.240101
                                        ... -0.042853
                                                        0.346823
                                                                  0.361430
s80212a -0.547678
                   0.744888 -0.193784
                                        ... -0.023248
                                                        0.243988
                                                                  0.441745
s22039a -0.765272
                   0.391081 -0.363202
                                        ... -0.153255 -0.145579
                                                                  0.388967
s22059a -0.763003
                   0.468817 -0.396569
                                        ... -0.161155 -0.140995
                                                                  0.361825
s22069a -0.818684
                   0.345560 -0.231986
                                        ... -0.134685
                                                        0.094852
                                                                  0.415746
s22079a -0.758011
                   0.408257 -0.255678
                                        ... -0.135622
                                                        0.176681
                                                                  0.477926
c7962
        -0.465184
                   0.314882 -0.283938
                                        ... -0.235773
                                                        0.285863
                                                                  0.640216
c7961
        -0.426870
                   0.311334 -0.234318
                                        ... -0.302480
                                                        0.332442
                                                                  0.670176
h4742
        -0.948616
                   0.726801 -0.271734
                                        ... -0.460077
                                                        0.239947
                                                                  0.303764
d352
        -1.352407
                   1.220397 -0.595024
                                        ... -0.714845
                                                        0.402737
                                                                  0.286710
                   1.205491 -0.542173
j342
        -1.350485
                                        ... -0.708781
                                                        0.458047
                                                                  0.247367
j343
        -0.925696
                   0.713223 -0.297658
                                        ... -0.500147
                                                        0.232816
                                                                  0.267582
s22049a -0.764399
                   0.358083 -0.271678
                                        ... -0.148444 -0.089374
                                                                  0.385005
c7b8
        -0.441223
                   0.718800 -0.332277
                                             0.045817 -0.002821
                                                                  0.772620
c7a8
        -0.454384
                   0.675361 -0.292155
                                        ... 0.040029
                                                        0.099672
                                                                  0.765976
                   0.343687 -0.824926
b372
        -0.765097
                                        ... -0.311394
                                                        0.277953
                                                                  0.591798
        -0.793705
                   0.339783 -0.800079
                                        ... -0.196104
1304
                                                        0.305985
                                                                  0.488225
189529
        -1.048253 -0.017454 -0.629358
                                        ... -0.199440
                                                        0.564745 -0.084565
                   0.675923 -0.470203
                                        ... -0.702325
                                                        0.010078
t59811a -1.185262
                                                                  0.223276
i705
        -1.204738
                   0.665935 -0.501996
                                        ... -0.761729
                                                        0.039683
                                                                  0.302483
s82202b -0.589536
                   0.739573 -0.216320
                                        ... -0.019602
                                                        0.646297
                                                                  0.411887
s82402b -0.602419
                   0.744510 -0.157587
                                        ... 0.009142
                                                        0.659131
                                                                  0.475003
s80211a -0.574273
                   1.004498 -0.267113
                                        ... -0.014799
                                                        0.369407
                                                                  0.415824
s0081xa -0.562279
                   0.599095 -0.152111
                                        ... -0.075801
                                                        0.189345
                                                                  0.441407
m19072
        -1.036905
                   0.910544 -0.037598
                                        ... 0.282082
                                                        0.247060
                                                                  0.582505
m19071
        -0.968061
                   0.908698 -0.112139
                                             0.287499
                                                        0.301793
                                                                  0.659054
s12490a -0.811155
                   0.600155 -0.532949
                                        ... -0.411080
                                                        0.435647 -0.019122
s12590a -0.939699
                   0.849238 -0.825214
                                        ... -0.574542
                                                        0.634803 -0.264070
s12690a -0.804431
                   0.603982 -0.499284
                                        ... -0.414211
                                                        0.414069 -0.023284
n803
        -0.941909
                   0.453790 -0.759537
                                        ... -0.058291
                                                        0.471336
                                                                  0.539394
n801
        -0.982363
                   0.421095 -0.755633
                                        ... 0.008651
                                                        0.458387
                                                                  0.633568
s22030a -0.517664
                   0.980204 -0.998884
                                        ... -0.607953
                                                        0.304184
                                                                  0.172500
s22020a -0.495499
                   0.556402 -0.541966
                                        ... -0.447460
                                                        0.183472
                                                                  0.301029
s22040a -0.536504
                   0.684664 -0.770988
                                        ... -0.501595
                                                        0.219759
                                                                  0.262707
r402111 -0.739676
                   0.570173 -0.630862
                                        ... -0.097161
                                                        0.072572 -0.125812
r402311 -0.669979
                   0.547992 -0.582326
                                        ... -0.163980
                                                        0.110357 -0.064199
r402211 -0.694935
                   0.546233 -0.570987
                                        ... -0.131386
                                                        0.136321 -0.085218
```

[52 rows x 13 columns]

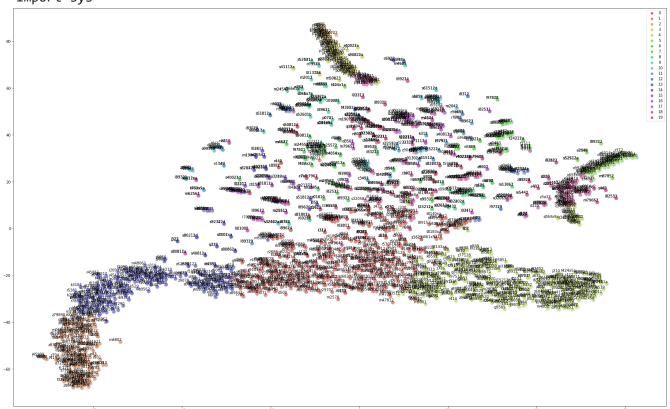
▼ T-SNE

```
def tsne plot(model):
```

```
"Creates and TSNE model and plots it"
labels = []
tokens = []
for word in model.wv.vocab:
    tokens.append(model[word])
    labels.append(word)
tsne model = TSNE(perplexity=30, n components=2, learning rate=10, init='random', n iter=
new_values = tsne_model.fit_transform(tokens)
x = []
y = []
for value in new values:
    x.append(value[0])
    y.append(value[1])
plt.figure(figsize=(32, 20))
sns.scatterplot(
    x=x, y=y,
    hue= kmeans.labels_,
    palette=sns.color_palette("hls", len(set(kmeans.labels_))),
    legend="full",
    alpha=0.7,
    s=120
    )
for i in range(len(x)):
  plt.annotate(labels[i],
                 xy=(x[i], y[i]),
                 xytext=(3, 1),
                 textcoords='offset points',
                 ha='right',
                 va='bottom')
plt.show()
```

```
%%time
tsne_plot(model)
```

/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:7: DeprecationWarning: Cal import sys



CPU times: user 2min 40s, sys: 1.94 s, total: 2min 42s

Wall time: 1min 27s

✓ 1s completed at 12:07 AM