

test1_versione2

June 29, 2020

#Clone directory

```
[1]: !git clone https://github.com/pm-lab-polito/pm_networks
```

```
Cloning into 'pm_networks'...
remote: Enumerating objects: 23, done.
remote: Counting objects: 100% (23/23), done.
remote: Compressing objects: 100% (14/14), done.
remote: Total 23 (delta 8), reused 22 (delta 7), pack-reused 0
Unpacking objects: 100% (23/23), done.
```

```
[2]: import networkx as nx
import matplotlib.pyplot as plt
import pandas as pd
from networkx import community
import community as community_louvain
```

#Importa i networks

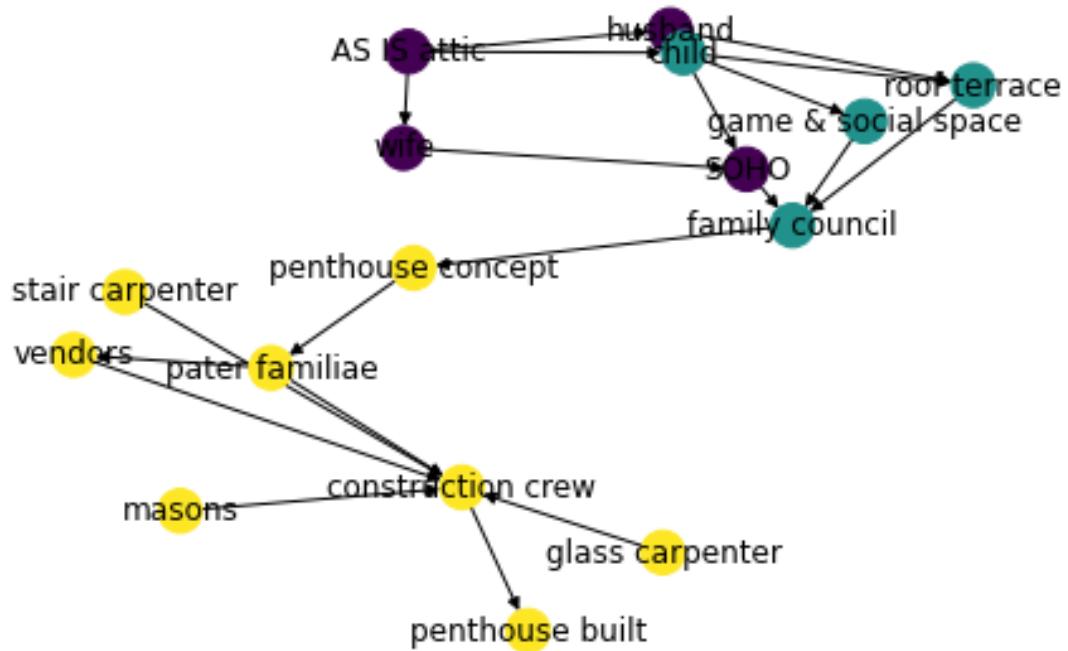
```
[3]: G0 = nx.read_gml('pm_networks/graphml/penthouse_00.gml')
G1 = nx.read_gml('pm_networks/graphml/penthouse_01.gml')
G2 = nx.read_gml('pm_networks/graphml/penthouse_02.gml')
list_net = [G0,G1,G2]
```

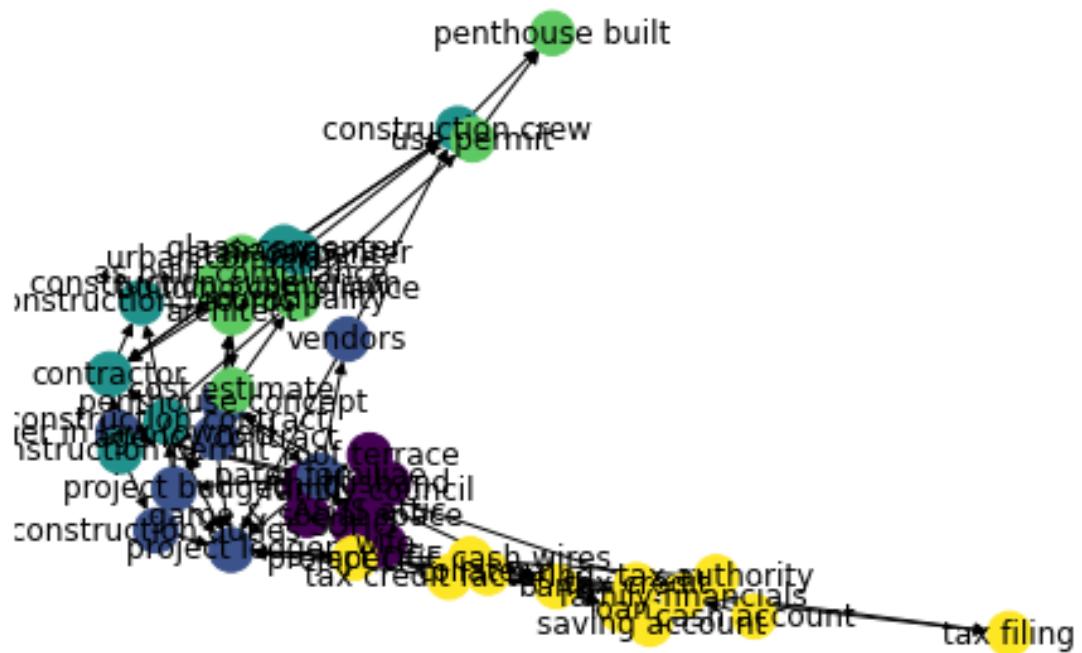
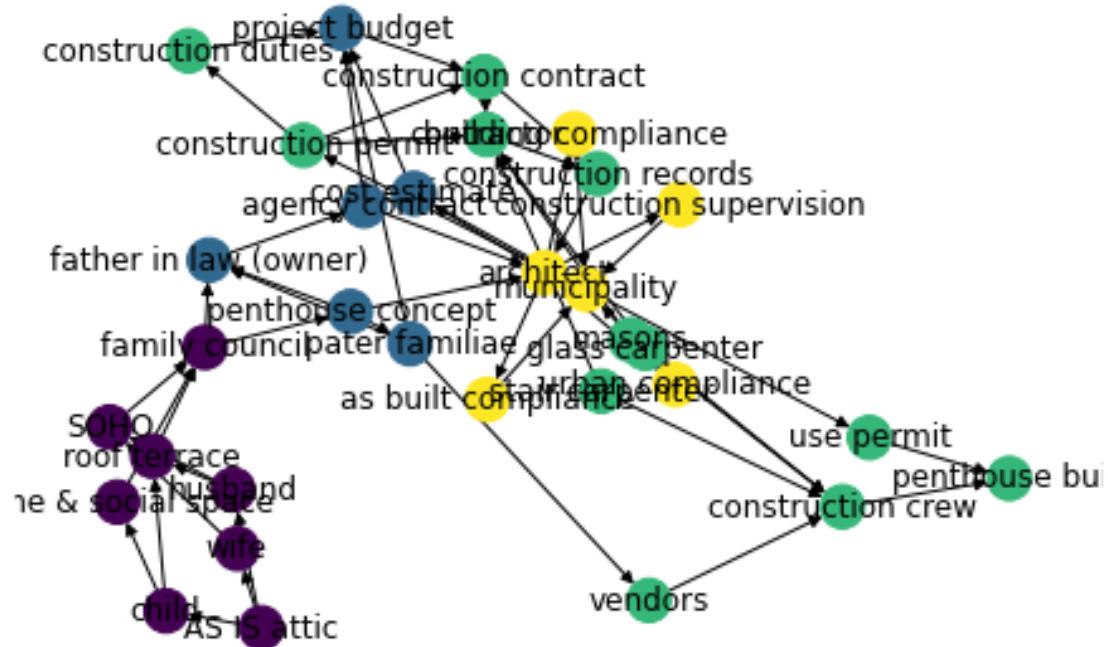
1 Creiamo le comunità con l'algoritmo python louvain (massimizza la modularità delle comunità)

```
[4]: #le communities del louvain funzionano su undirected networks, quindi le ↵ trasformiamo in undirected momentaneamente
H0 = G0.to_undirected()
partition0 = community_louvain.best_partition(H0)
H1 = G1.to_undirected()
partition1 = community_louvain.best_partition(H1)
H2 = G2.to_undirected()
partition2 = community_louvain.best_partition(H2)
lista_partizioni =[partition0 ,partition1,partition2]
```

#Plotta i networks con i colori che corrispondono alla community

```
[5]: for i in range (len(list_net)):
    nx.draw(list_net[i], node_color = list(lista_partizioni[i].values()), with_labels=True)
    plt.show()
```





```
#Calcola le metriche del network
```

```
[6]: list_density = []
list_SH_path = []
list_node_conn = []
list_node_trans = []
```

```
[7]: for i in list_net:
    Den = nx.density(i)
    sh = nx.average_shortest_path_length(i)
    conn = nx.average_node_connectivity(i)
    #trans = nx.transitivity(1) # only for undirected
    list_density.append(Den)
    list_SH_path.append(sh)
    list_node_conn.append(conn)
    #list_node_trans.append(trans)
```

```
[8]: report = pd.DataFrame(list_density, columns=['Density'])
report['Shortest_path'] = list_SH_path
report['Node_conn']= list_node_conn
```

```
[9]: report
```

	Density	Shortest_path	Node_conn
0	0.083333	0.895833	0.337500
1	0.053427	1.940524	0.549395
2	0.040909	1.625253	0.447980

```
[10]: report_delta = report/report.shift(1)-1
```

```
[11]: report_delta
```

	Density	Shortest_path	Node_conn
0	NaN	NaN	NaN
1	-0.358871	1.166167	0.627838
2	-0.234305	-0.162467	-0.184595

2 Calcola le metriche relative ai singoli nodi (per ora degree e betweeness centrality)

Degree

```
[12]: # calcoliamo la degree per ogni nodo
list_metr_nodes_degree = []
A = 0
for i in list_net:
```

```

colonna = ("G"+str(A)+"_Degree")
nodi = pd.DataFrame()
nodi = pd.DataFrame((nx.degree(i)), columns = ['Nodes', colonna])
nodi = nodi.set_index(['Nodes'], drop=True)
nodi[colonna]=nodi[colonna]/nodi[colonna].sum() #normalizziamo il grado

list_metr_nodes_degree.append(nodi)
A = A+1

```

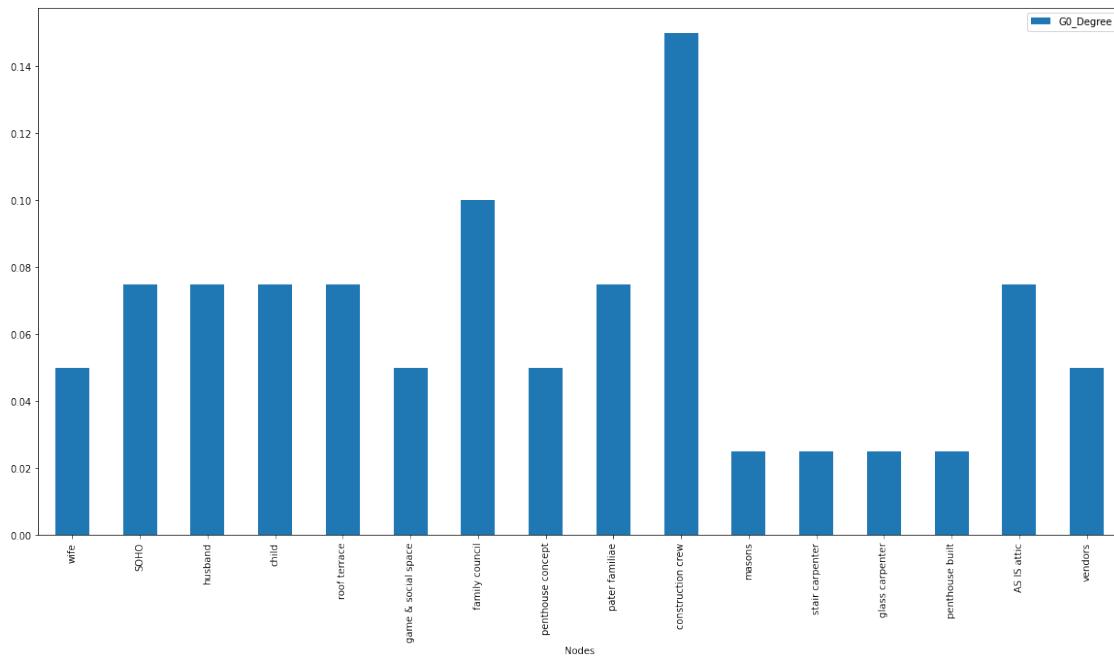
a

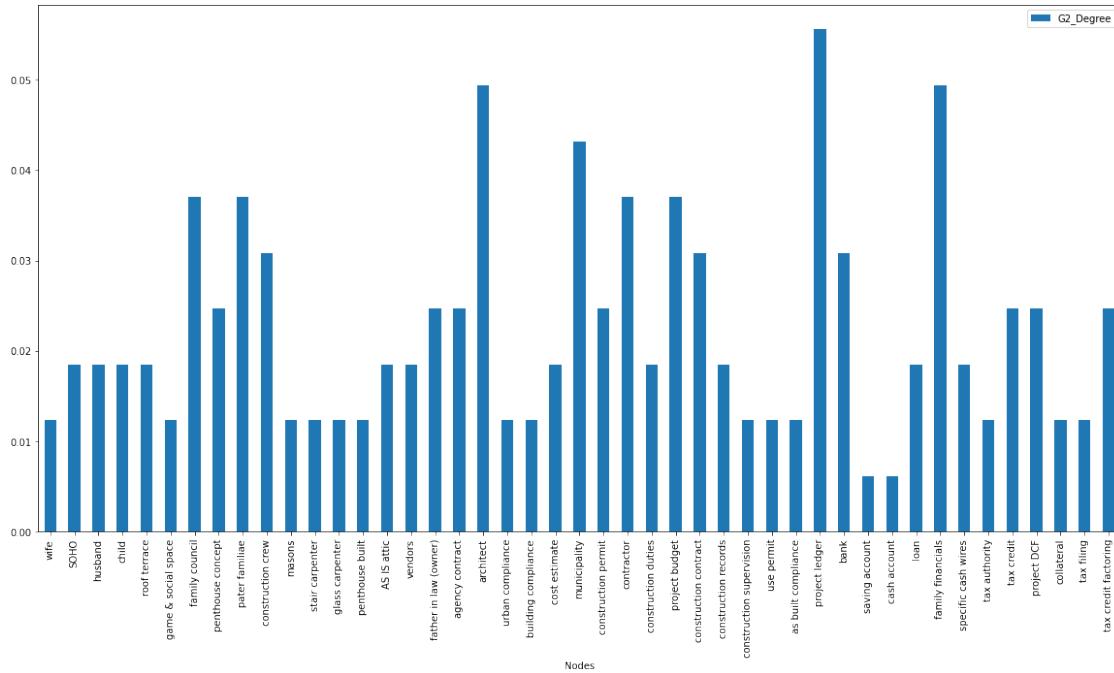
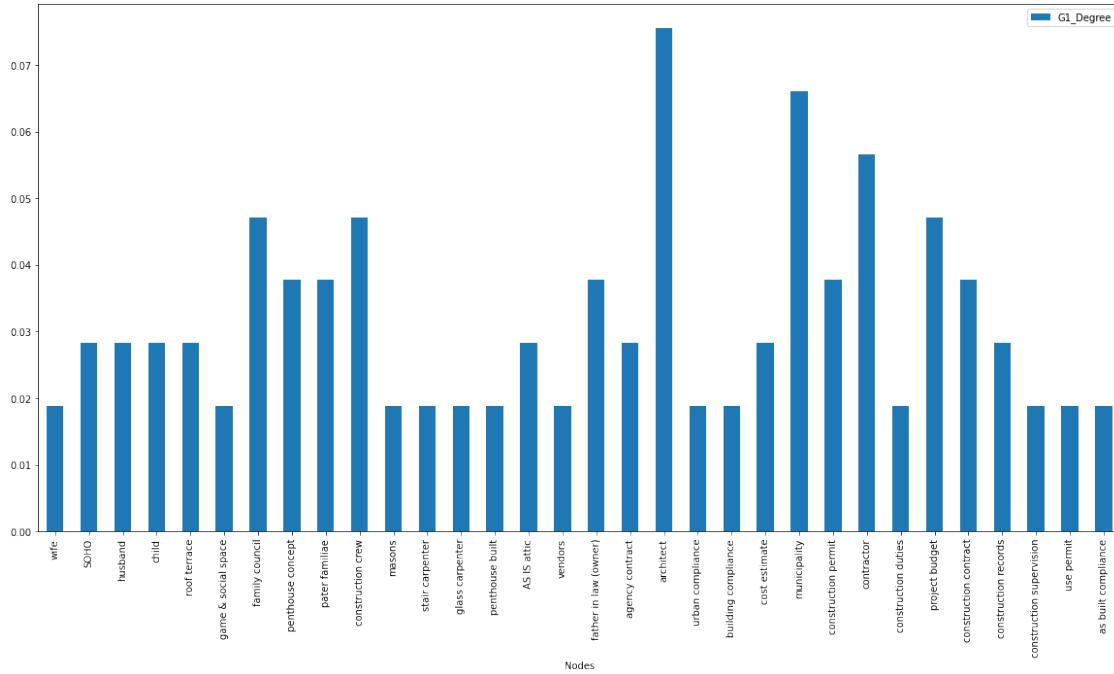
[13]: *#plottiamo la degree dei singoli nodi*

```

for i in range (3):
    list_metr_nodes_degree[i].plot(kind='bar', figsize=(20,10))

```

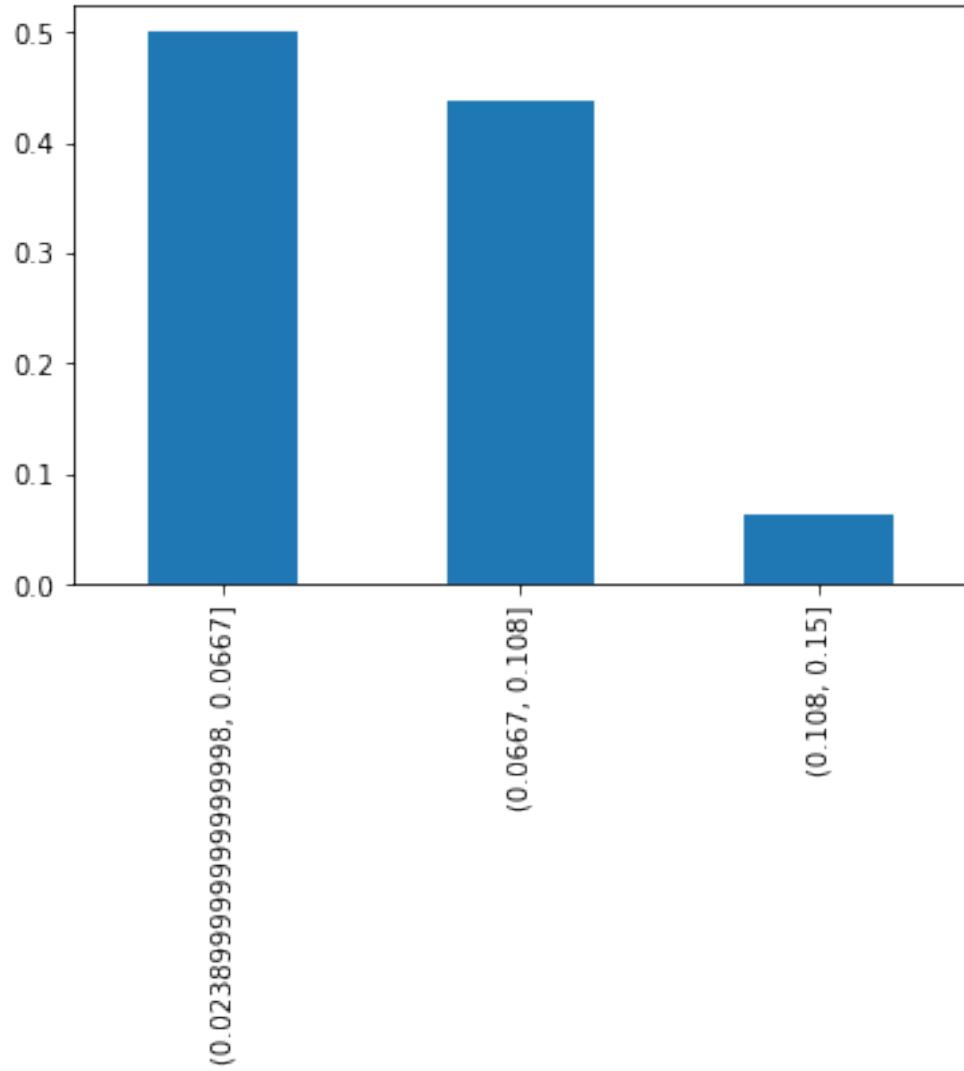


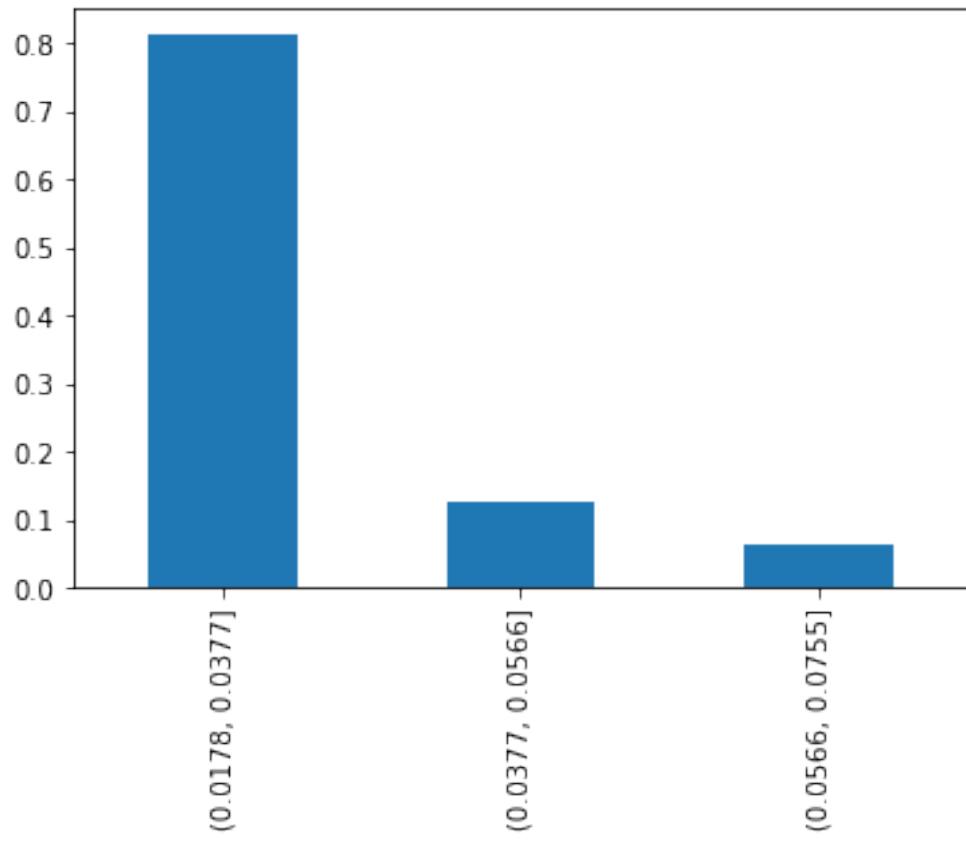


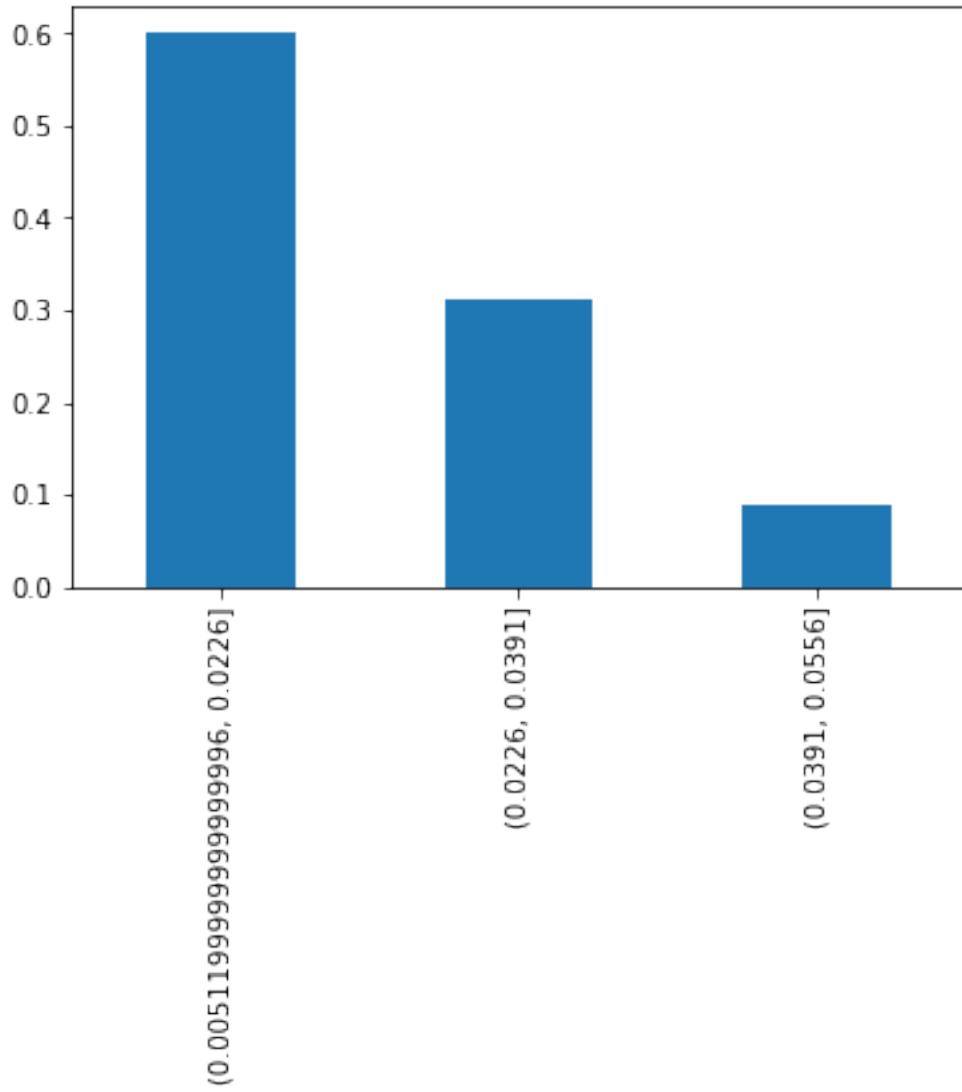
Plottiamo la distribuzione normalizzata della degree

[14]: #plottiamo la distribuzione normalizzata con istogrammi

```
for i in range (3):
    colonna = ("G"+str(i)+"_Degree")
    list_metr_nodes_degree[i][colonna].value_counts(normalize = True, bins = 3).
    ↪plot(kind='bar')
    plt.show()
```







Confrontiamo la degree dei nodi PERMANENTI nelle 3 fasi del progetto

```
[15]: #uniamo i DF per vedere come cambia la degree del nodo nelle tre fasi
mappa_degree = list_metr_nodes_degree[0]
for i in range (1,3):

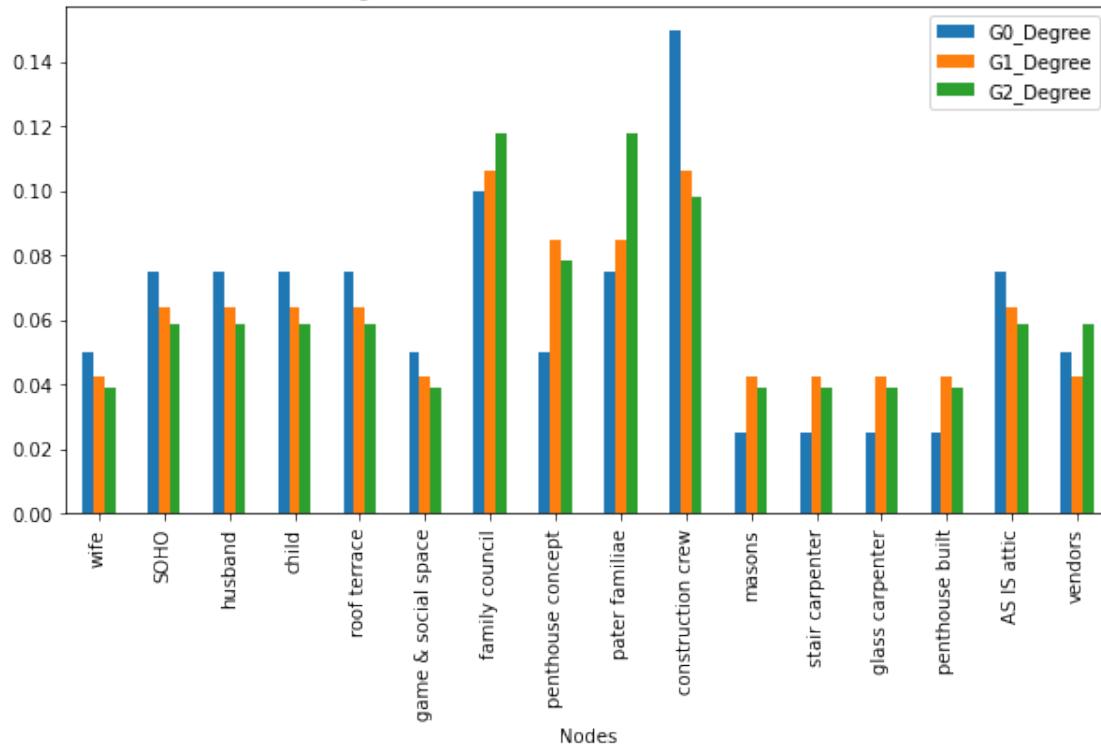
    mappa_degree = mappa_degree.join(list_metr_nodes_degree[i])

mappa_degree = mappa_degree/mappa_degree.sum()

mappa_degree.plot(kind='bar', figsize=(10,5))
plt.title("confronto degree nodi nella 3 fasi normalizzata (solo nodi costanti)")
```

[15]: Text(0.5, 1.0, 'confronto degree nodi nella 3 fasi normalizzata (solo nodi costanti)')

confronto degree nodi nella 3 fasi normalizzata (solo nodi costanti)

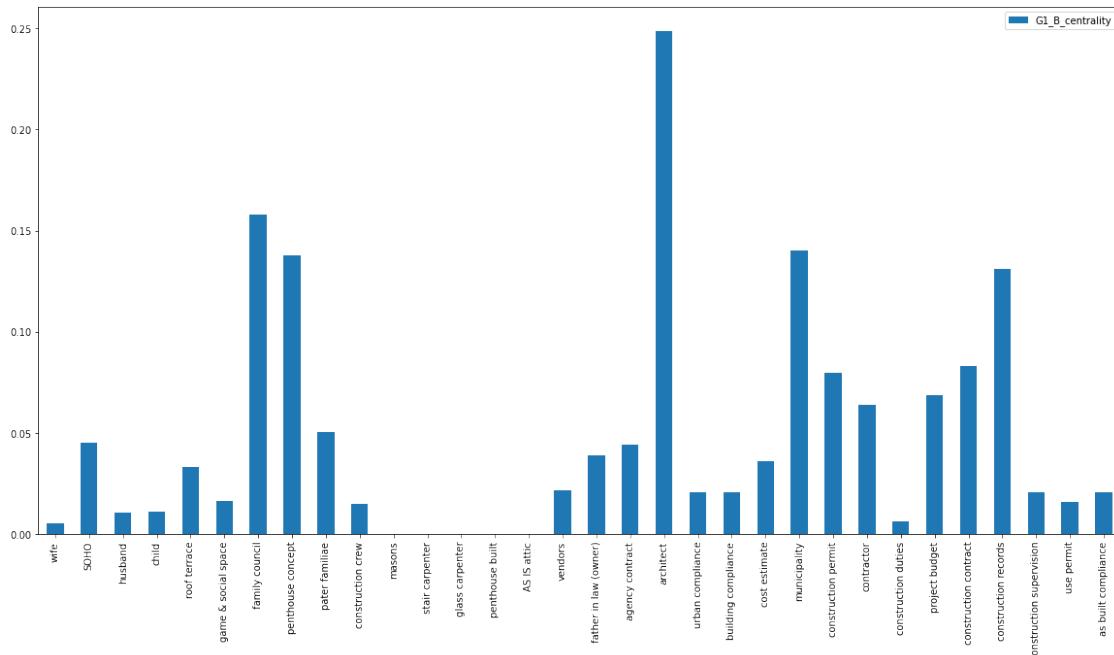
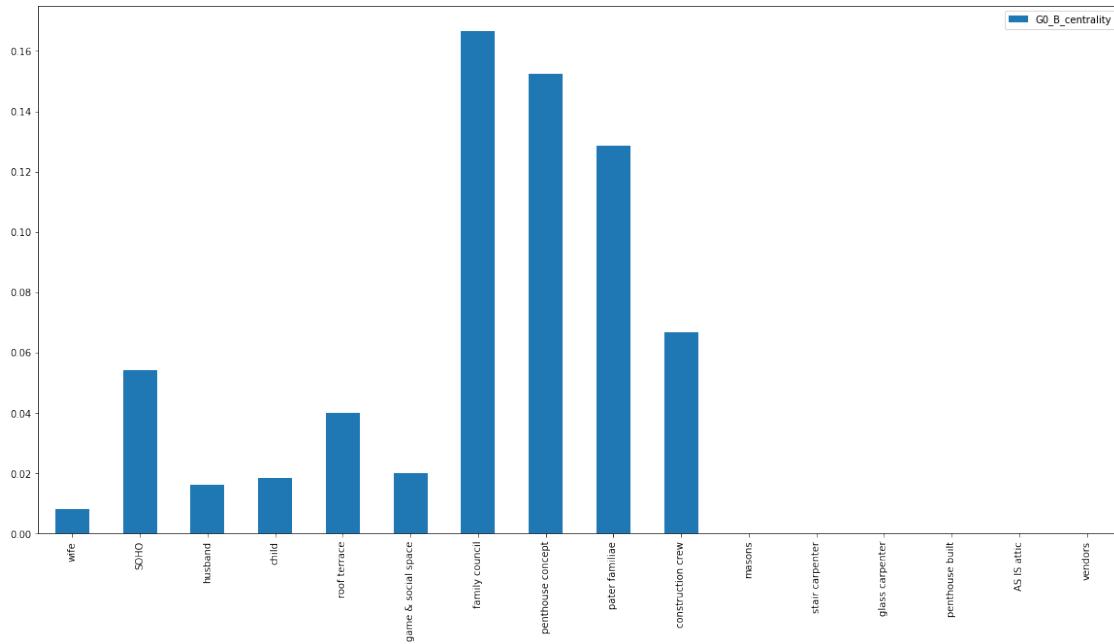


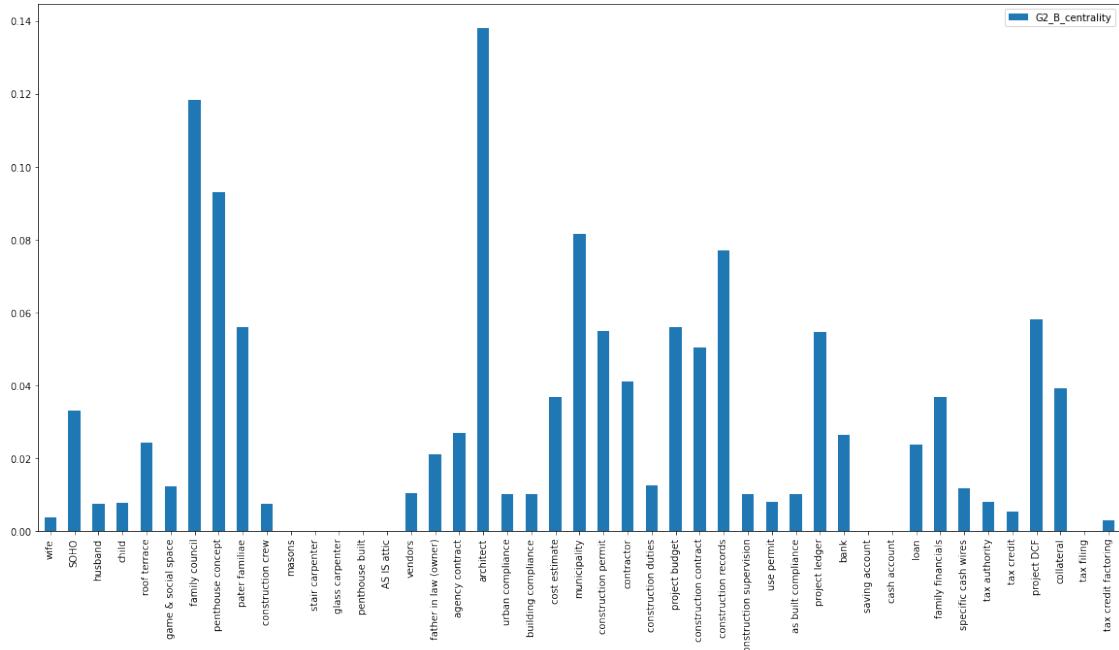
Centrality

```
[16]: # calcoliamo la centraility per ogni nodo
list_metr_nodes_centrality = []
A = 0
for i in list_net:

    colonna = ("G"+str(A)+"_B_centrality")
    nodi = pd.DataFrame(nx.betweenness_centrality(i).values(), index= nx.
    ↪betweenness_centrality(i).keys(), columns=[colonna])
    list_metr_nodes_centrality.append(nodi)
    A = A+1
```

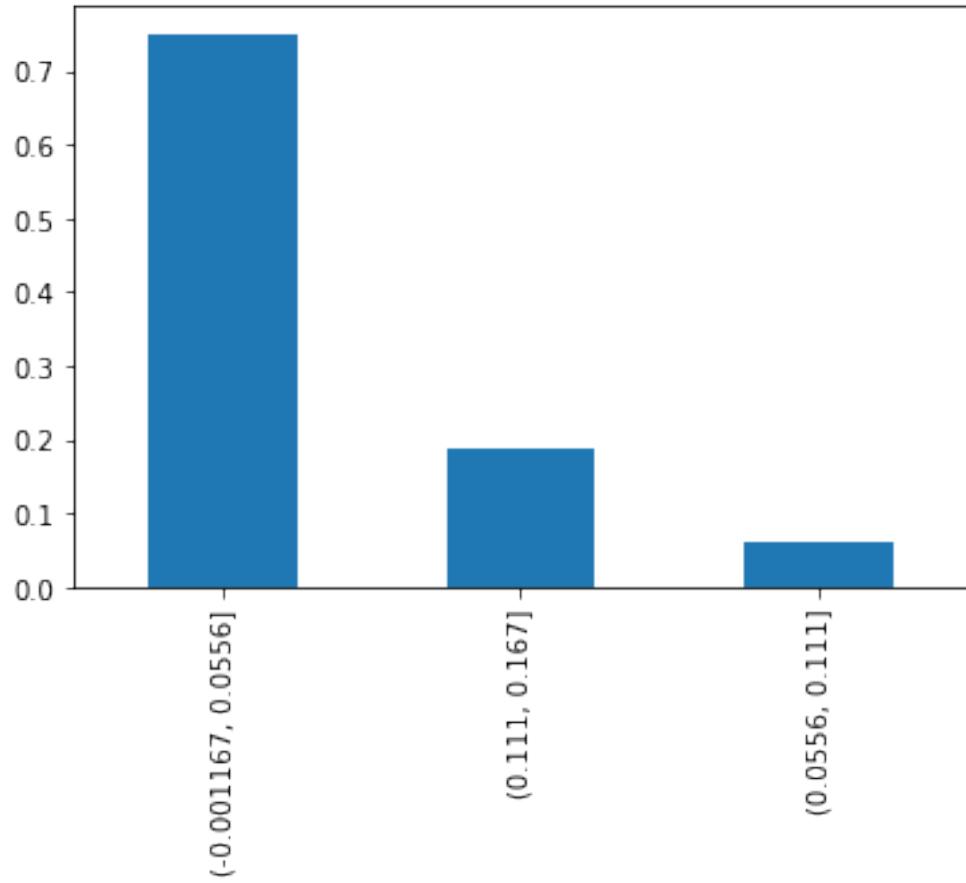
```
[17]: #plottiamo la centrality dei singoli nodi
for i in range (3):
    list_metr_nodes_centrality[i].plot(kind='bar', figsize=(20,10))
```

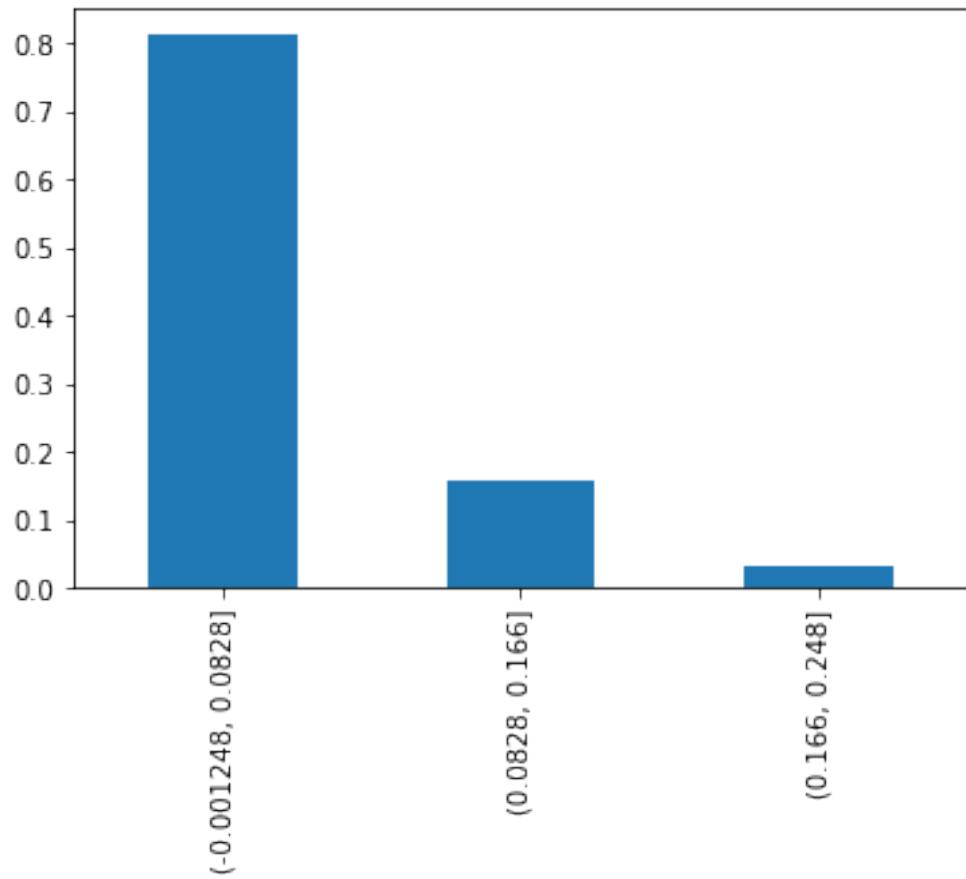


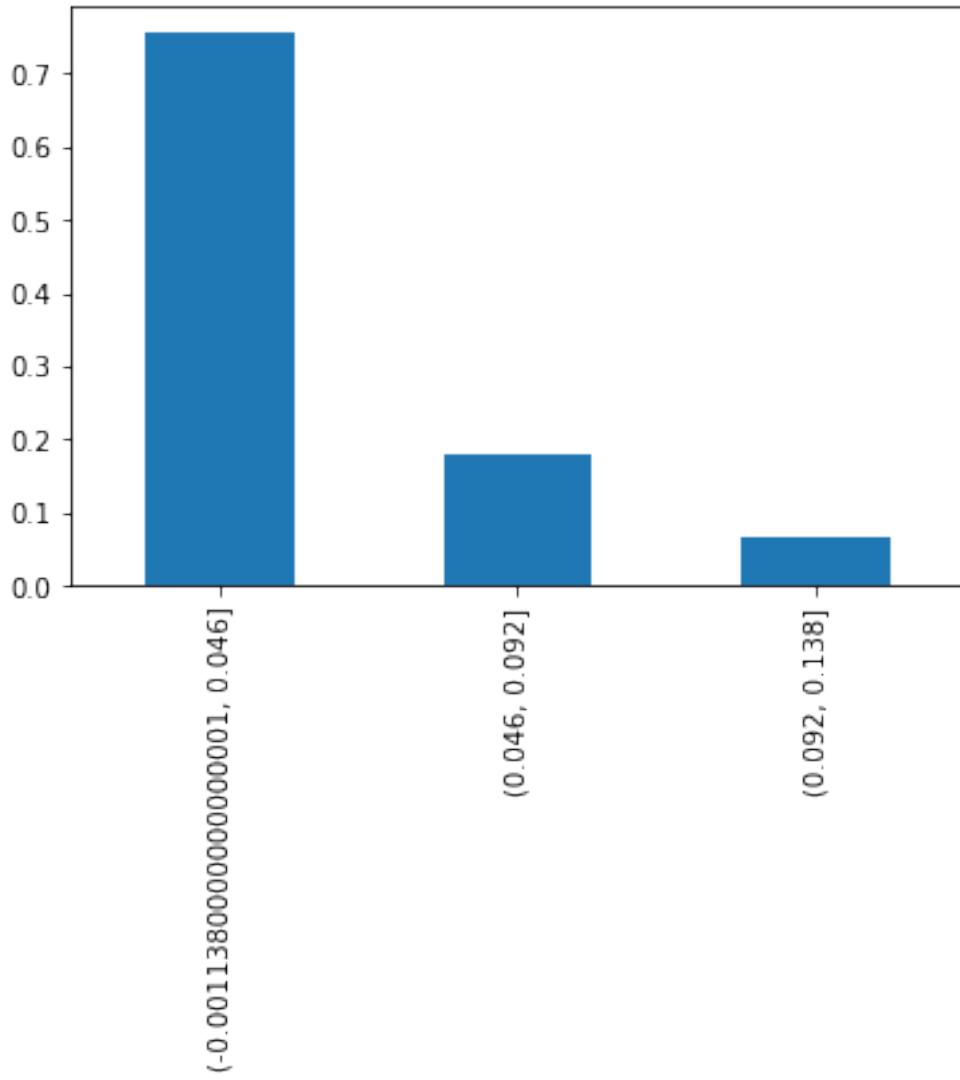


Plottiamo la distribuzione normalizzata della Centrality

```
[18]: #plottiamo la distribuzione normalizzata con istogrammi
for i in range (3):
    colonna = ("G"+str(i)+"_B_centrality")
    list_metr_nodes_centrality[i][colonna].value_counts(normalize = True, bins = 3).plot(kind='bar')
    plt.show()
```







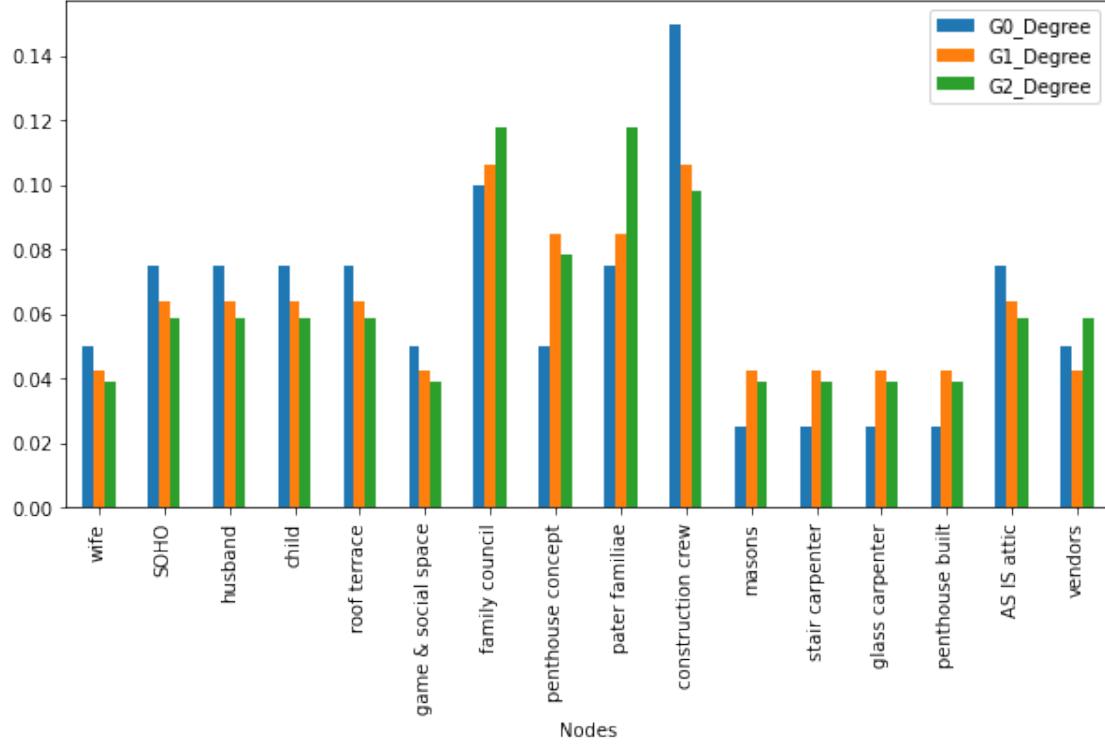
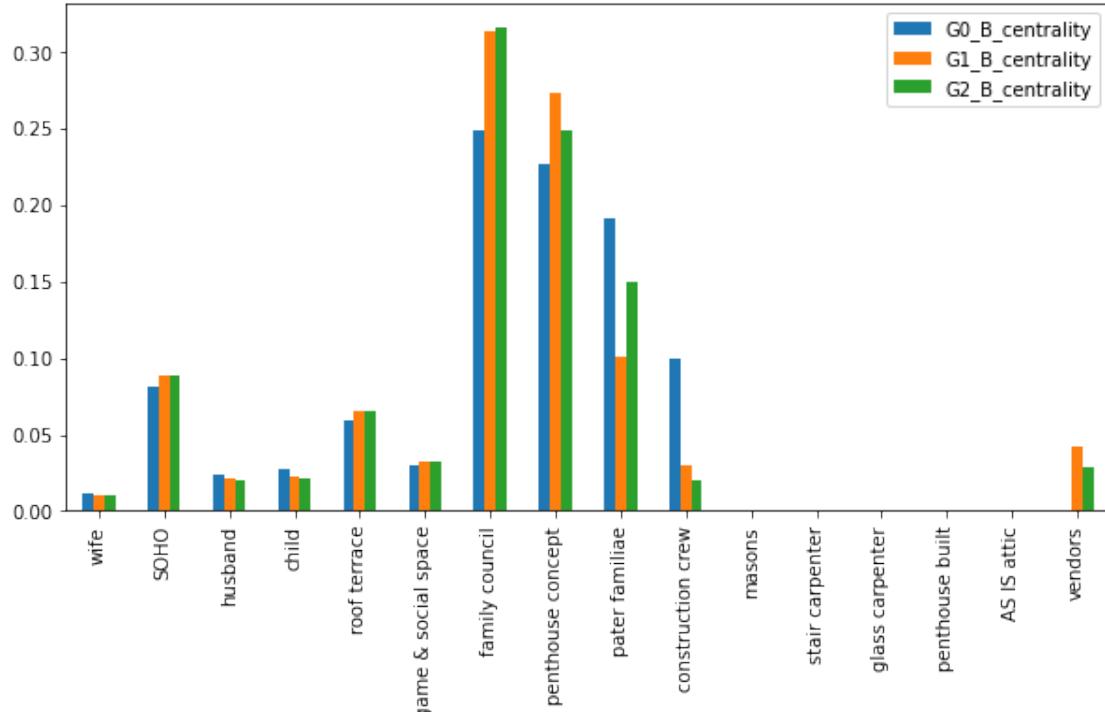
Confrontiamo la centrality dei nodi PERMANENTI nelle 3 fasi del progetto

```
[19]: #uniamo i DF per vedere come cambia la centralità del nodo nelle tre fasi
mappa_centrality = list_metr_nodes_centrality[0]
for i in range (1,3):
    mappa_centrality = mappa_centrality.join(list_metr_nodes_centrality[i])

mappa_centrality = mappa_centrality/mappa_centrality.sum()

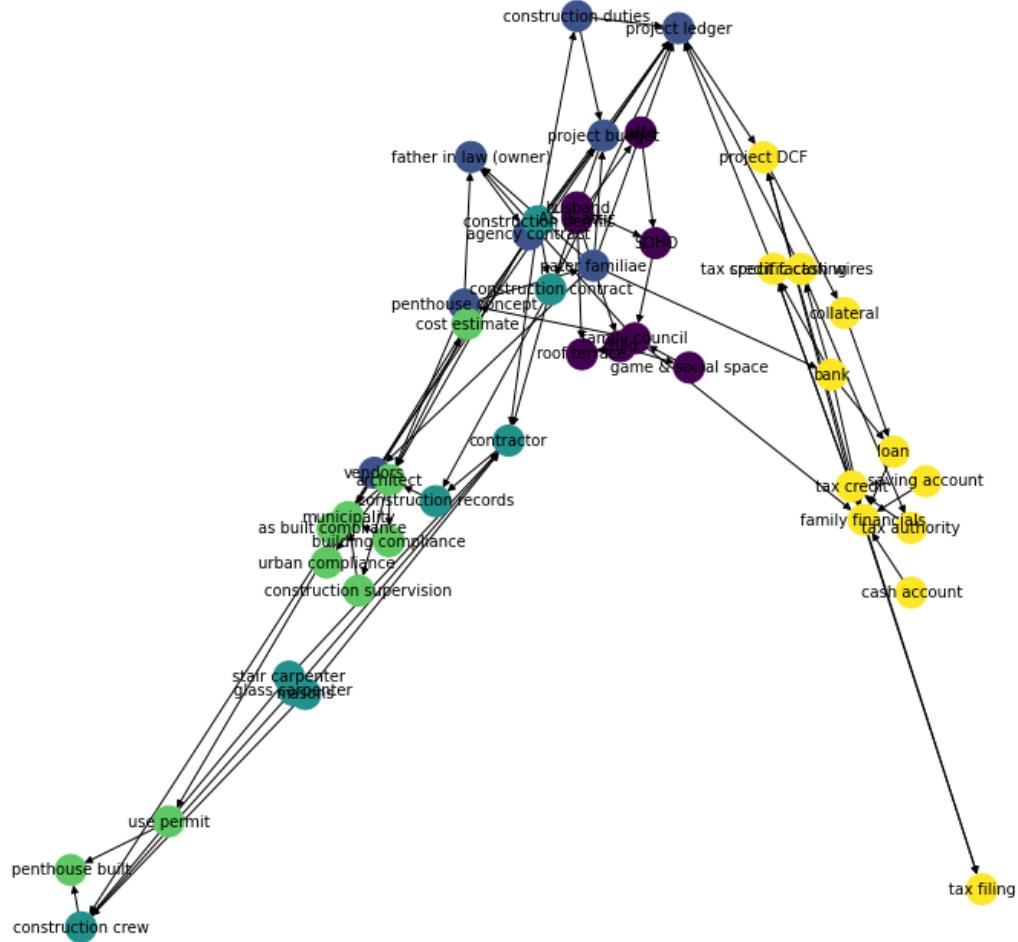
mappa_centrality.plot(kind='bar', figsize = (10,5))
mappa_degree.plot(kind='bar', figsize=(10,5))
```

[19]: <matplotlib.axes._subplots.AxesSubplot at 0x7fd739c9af0>



#disegna più grande per leggibilità

```
[20]: plt.figure(figsize=(10,10))
nx.draw(G2, node_color = list(lista_partizioni[2].values()), with_labels=True,
        font_size=10, node_size=400)
```



```
[20]:
```