

## Chapter 8 - Comparison of taxa and qualitative classification

November 10, 2020

```
[1]: import pandas as pd
import geopandas as gpd
import seaborn as sns
import matplotlib.pyplot as plt
import husl
from legendgram import legendgram
import mapclassify
from matplotlib_scalebar.scalebar import ScaleBar
from matplotlib.colors import ListedColormap
from random import shuffle
from tqdm import tqdm

[2]: clusters = pd.read_csv('/Users/martin/Dropbox/Academia/Data/Geo/Prague/
↳Clustering/complete data/200218_clusters_complete_n20.csv', index_col=0)

[3]: file = '/Users/martin/Dropbox/Academia/Contracts/UAP Prague/2020.01_Zakázka MF/
↳01_data/202004_Zakazka MF_predana data/20200421_ZakazkaMF_data_validacni.gdb'

[4]: import fiona
fiona.listlayers(file)

[4]: ['URK_LokalityStav_p', 'URK_SS_VyuzitiZakl_p', 'VyvojZastavby_p']

[5]: qual = gpd.read_file(file, layer='URK_LokalityStav_p')

[6]: buildings = gpd.read_file('/Users/martin/Dropbox/Academia/Data/Geo/Prague/
↳Clustering/geometry.gpkg', layer='buildings')

[7]: buildings['cent'] = buildings.centroid
buildings = buildings.set_geometry('cent')

[8]: buildings = buildings.to_crs(qual.crs)

[9]: joined = gpd.sjoin(buildings, qual, how='left')

[10]: joined = joined.merge(clusters, how='left', on='uID')
```

```
[11]: joined.head(2)
```

```
[11]:   uID                                geometry \
0    0  POLYGON ((-749844.585 -1052277.805, -749844.41...
1    1  POLYGON ((-744449.580 -1042702.590, -744447.15...

                                cent  index_right    ID  CISLO_TXT \
0  POINT (-749841.681 -1052279.951)         233  16170         617
1  POINT (-744432.289 -1042699.409)         702  10080         008

      NAZEV_STAV  ZASTAVENOST  VYUZITI_STAV  STRUKTURA_STAV \
0  Cementárna Radotín          1           2           8
1      Malá Strana             1           1           1

      MIRA_STABILITY_STAV  TYP_MESTA  KRAJINA  Shape_Length  Shape_Area \
0                5          1.0          1    2815.203269  239855.665606
1                4          1.0          1    5419.046731  692062.403477

      cluster
0          1.0
1          11.0
```

```
[11]: joined = joined.set_geometry('geometry')
```

## 0.1 analyse

```
[14]: import numpy as np

def show_values_on_bars(axes):
    def _show_on_single_plot(ax):
        for p in ax.patches:
            _x = p.get_x() + p.get_width() / 2
            _y = p.get_y() + p.get_height() + 0.02
            value = '{:.2f}'.format(p.get_height())
            ax.text(_x, _y, value, ha="center")

    if isinstance(axes, np.ndarray):
        for idx, ax in np.ndenumerate(axes):
            _show_on_single_plot(ax)
    else:
        _show_on_single_plot(axes)

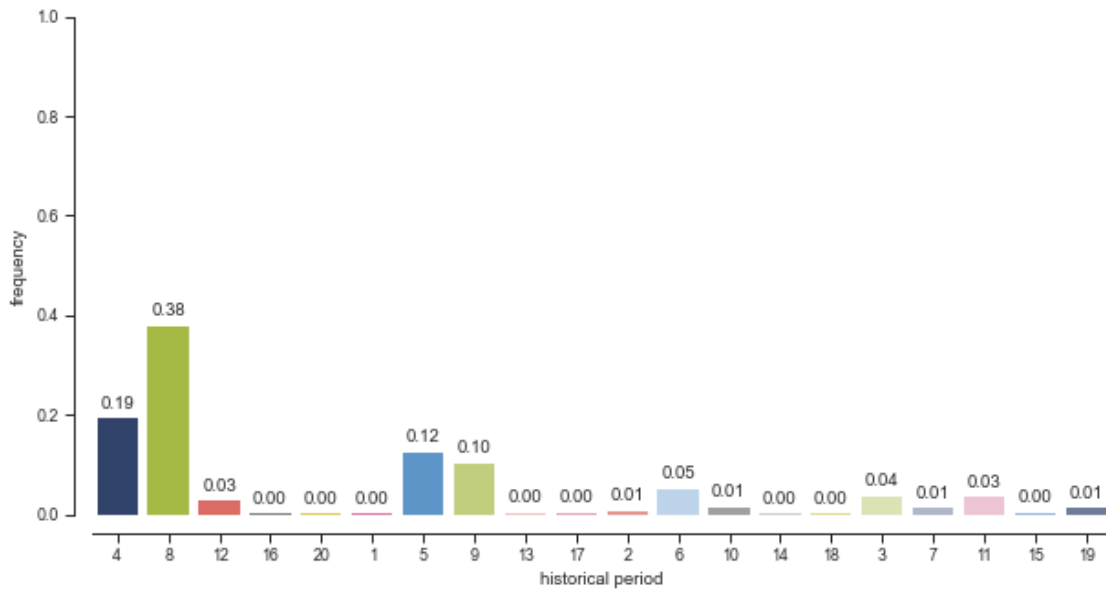
pal = cols
```

```
[53]: data = joined.loc[joined['cluster'].isin()]['STRUKTURA_STAV'].
      ↪value_counts(sort=False, normalize=True)
```

```

sns.set(context="paper", style="ticks", rc={'patch.force_edgecolor': False})
fig, ax = plt.subplots(figsize=(10, 5))
sns.barplot(ax=ax, x=data.index, y=data, order=data.index, palette=pal)
sns.despine(offset=10)
plt.ylabel('frequency')
plt.xlabel('historical period')
plt.ylim(0, 1)
show_values_on_bars(ax)

```



```

[14]: sample = joined.loc[joined['STRUKTURA_STAV'].isin([1, 2, 5, 6, 7, 8, 9])]

```

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[ ]: data = sample.loc[sample['cluster'].isin([11])]['STRUKTURA_STAV'].
    ↪value_counts(sort=False, normalize=True)

labels = ['organic', 'perimeter block', 'village', 'garden city', 'modernism',
    ↪'production', 'services']
sns.set(context="paper", style="ticks", rc={'patch.force_edgecolor': False})
fig, ax = plt.subplots(figsize=(10, 5))
sns.barplot(ax=ax, x=data.index, y=data, order=[1, 2, 5, 6, 7, 8, 9],
    ↪palette=pal)
sns.despine(offset=10)
plt.ylabel('frequency')
plt.xlabel('qualitative typology')
plt.ylim(0, 1)
ax.set_xticklabels(labels)
show_values_on_bars(ax)

```

```
[ ]: # save all clusters
for cl in range(20):
    data = sample.loc[sample['cluster'].isin([cl])]['STRUKTURA_STAV'].
    ↪value_counts(sort=False, normalize=True)

    fig, ax = plt.subplots(figsize=(10, 5))
    sns.barplot(ax=ax, x=data.index, y=data, order=[1, 2, 5, 6, 7, 8, 9],
    ↪palette=pal)
    sns.despine(offset=10)
    plt.ylabel('frequency')
    plt.xlabel('qualitative typology')
    plt.ylim(0, 1)
    ax.set_xticklabels(labels)
    show_values_on_bars(ax)
    for ext in ['pdf', 'png']:
        plt.savefig('figures/PRG_cluster_' + str(cl) + '_structure.' + ext,
    ↪bbox_inches='tight')
    plt.close()
```

```
[ ]: fig, ax = plt.subplots(2, 2, figsize=(14, 10))

data = sample.loc[sample['cluster'].isin([11])]['STRUKTURA_STAV'].
    ↪value_counts(sort=False, normalize=True)

sns.barplot(ax=ax[0, 0], x=data.index, y=data, order=[1, 2, 5, 6, 7, 8, 9],
    ↪palette=pal)
sns.despine(offset=10)
ax[0,0].set_ylabel('frequency')
ax[0,0].set_xlabel('qualitative typology')
ax[0,0].set_title('cluster 11')
ax[0,0].set_ylim(0, 1)
ax[0,0].set_xticklabels(labels)
show_values_on_bars(ax[0, 0])

data = sample.loc[sample['cluster'].isin([5])]['STRUKTURA_STAV'].
    ↪value_counts(sort=False, normalize=True)
sns.barplot(ax=ax[0, 1], x=data.index, y=data, order=[1, 2, 5, 6, 7, 8, 9],
    ↪palette=pal)
sns.despine(offset=10)
ax[0,1].set_ylabel('frequency')
ax[0,1].set_xlabel('qualitative typology')
ax[0,1].set_title('cluster 5')
ax[0,1].set_ylim(0, 1)
ax[0,1].set_xticklabels(labels)
show_values_on_bars(ax[0, 1])
```

```

data = sample.loc[sample['cluster'].isin([12])]['STRUKTURA_STAV'].
    ↳value_counts(sort=False, normalize=True)
sns.barplot(ax=ax[1, 0], x=data.index, y=data, order=[1, 2, 5, 6, 7, 8, 9],
    ↳palette=pal)
sns.despine(offset=10)
ax[1,0].set_ylabel('frequency')
ax[1,0].set_xlabel('qualitative typology')
ax[1,0].set_title('cluster 12')
ax[1,0].set_ylim(0, 1)
ax[1,0].set_xticklabels(labels)
show_values_on_bars(ax[1, 0])

data = sample.loc[sample['cluster'].isin([13])]['STRUKTURA_STAV'].
    ↳value_counts(sort=False, normalize=True)
sns.barplot(ax=ax[1, 1], x=data.index, y=data, order=[1, 2, 5, 6, 7, 8, 9],
    ↳palette=pal)
sns.despine(offset=10)
ax[1,1].set_ylabel('frequency')
ax[1,1].set_xlabel('qualitative typology')
ax[1,1].set_title('cluster 13')
ax[1,1].set_ylim(0, 1)
ax[1,1].set_xticklabels(labels)
show_values_on_bars(ax[1, 1])

plt.tight_layout()
plt.savefig('figures/PRG_cluster_structure_subplot.pdf')

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[ ]: fig, ax = plt.subplots(2, 2, figsize=(14, 10))

data = sample.loc[sample['cluster'].isin([11, 15, 5])]['STRUKTURA_STAV'].
    ↳value_counts(sort=False, normalize=True)

sns.barplot(ax=ax[0, 0], x=data.index, y=data, order=[1, 2, 5, 6, 7, 8, 9],
    ↳palette=pal)
sns.despine(offset=10)
ax[0,0].set_ylabel('frequency')
ax[0,0].set_xlabel('qualitative typology')
ax[0,0].set_title('compact city')
ax[0,0].set_ylim(0, 1)
ax[0,0].set_xticklabels(labels)
show_values_on_bars(ax[0, 0])

data = sample.loc[sample['cluster'].isin([3, 0, 8, 9, 13,
    ↳17])]['STRUKTURA_STAV'].value_counts(sort=False, normalize=True)
sns.barplot(ax=ax[0, 1], x=data.index, y=data, order=[1, 2, 5, 6, 7, 8, 9],
    ↳palette=pal)

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sns.despine(offset=10)
ax[0,1].set_ylabel('frequency')
ax[0,1].set_xlabel('qualitative typology')
ax[0,1].set_title('low-rise city')
ax[0,1].set_ylim(0, 1)
ax[0,1].set_xticklabels(labels)
show_values_on_bars(ax[0, 1])

data = sample.loc[sample['cluster'].isin([1, 19])]['STRUKTURA_STAV'].
    ↳value_counts(sort=False, normalize=True)
sns.barplot(ax=ax[1, 0], x=data.index, y=data, order=[1, 2, 5, 6, 7, 8, 9],
    ↳palette=pal)
sns.despine(offset=10)
ax[1,0].set_ylabel('frequency')
ax[1,0].set_xlabel('qualitative typology')
ax[1,0].set_title('industrial city')
ax[1,0].set_ylim(0, 1)
ax[1,0].set_xticklabels(labels)
show_values_on_bars(ax[1, 0])

data = sample.loc[sample['cluster'].isin([12, 14, 2, 10])]['STRUKTURA_STAV'].
    ↳value_counts(sort=False, normalize=True)
sns.barplot(ax=ax[1, 1], x=data.index, y=data, order=[1, 2, 5, 6, 7, 8, 9],
    ↳palette=pal)
sns.despine(offset=10)
ax[1,1].set_ylabel('frequency')
ax[1,1].set_xlabel('qualitative typology')
ax[1,1].set_title('heterogenous dense city branch')
ax[1,1].set_ylim(0, 1)
ax[1,1].set_xticklabels(labels)
show_values_on_bars(ax[1, 1])

plt.tight_layout()
plt.savefig('figures/PRG_branch_structure_subplot.pdf')

```

```

[12]: import scipy.stats as ss
import numpy as np

def cramers_v(x, y):
    confusion_matrix = pd.crosstab(x,y)
    chi2 = ss.chi2_contingency(confusion_matrix)[0]
    n = confusion_matrix.sum().sum()
    phi2 = chi2/n
    r,k = confusion_matrix.shape
    phi2corr = max(0, phi2-((k-1)*(r-1))/(n-1))
    rcorr = r-((r-1)**2)/(n-1)
    kcorr = k-((k-1)**2)/(n-1)

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return np.sqrt(phi2corr/min((kcorr-1),(rcorr-1)))
```

```
[15]: cramers_v(sample.cluster, sample.STRUKTURA_STAV)
```

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[15]: 0.6741832153043661
```

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[16]: confusion_matrix = pd.crosstab(sample.cluster, sample.STRUKTURA_STAV)  
chi, p, dof, exp = ss.chi2_contingency(confusion_matrix)
```

```
[20]: p
```

```
[20]: 0.0
```

```
[21]: dof
```

```
[21]: 114
```

```
[22]: chi
```

```
[22]: 325595.1995642433
```