# CorrTest household personalIncome

June 16, 2021

### 1 Testing correlation between personal and household income

```
[1]: import pandas as pd
    from pandas import read_csv
    import numpy as np
    from sklearn.linear_model import LinearRegression
    from numpy import cov
    from scipy.stats import pearsonr
    from scipy.stats import spearmanr
    import matplotlib.pyplot as plt
    import seaborn as sn

pd.set_option('display.max_columns', None)
    pd.set_option('display.max_rows', None)
```

```
[75]: # Group D
     groupD=pd.read_csv("CompleteSet_GroupD.csv")
     groupD=groupD.drop("Unnamed: 0",axis=1)
     #Strip all leading whitespace in Area column
     groupD['Area'] = groupD['Area'].apply(lambda x: x.strip())
     #Remove total NZ row
     groupD = groupD.loc[(groupD['Area'] != "Total - New Zealand by Regional Council/
      ⇒SA2")]
     #Remove total regions
     groupD = groupD.loc[(groupD['ParentArea'] != "NewZealand")]
     #Keep only 2013 and 2018
     groupD = groupD.loc[(groupD['Year'] == 2013) | (groupD['Year']==2018)]
     groupD['TotInd_TotPeople'] = groupD['totStated_TotInd'] + groupD['notStated_TotInd']
     groupD['perc_less50k_TotInd']=groupD[["less5k_TotInd",_
      →"bet5k10k TotInd","bet10k20k TotInd","bet20k30k TotInd",'bet30k50k TotInd']].
```

```
groupD = groupD.drop(['less5k_TotInd', 'less5k_Wholesale', 'less5k_Retail',
                  'less5k_TransPostWare', 'bet5k10k_TotInd', __
'bet5k10k Retail', 'bet5k10k TransPostWare',
'bet10k20k_Wholesale', 'bet10k20k_Retail', __
'bet20k30k_TotInd', 'bet20k30k_Wholesale', _
'bet20k30k_TransPostWare', 'bet30k50k_TotInd', __
'bet30k50k_Retail', 'bet30k50k_TransPostWare', _
\hookrightarrow 'bet50k70k_TotInd',
                  'bet50k70k_Wholesale', 'bet50k70k_Retail', __
'greater70k_TotInd', 'greater70k_Wholesale', __
'greater70k_TransPostWare', 'totStated_TotInd', __
'totStated_Retail', 'totStated_TransPostWare', _
'notStated_Wholesale', 'notStated_Retail', __
→ 'notStated_TransPostWare'], axis=1)
print(groupD.shape)
```

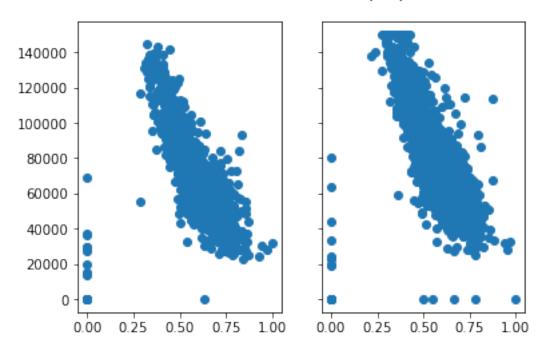
#### (4506, 5)

```
'totNotStated',], axis=1)
      print(groupG.shape)
     (4506, 4)
[76]: comboFrame= pd.merge(groupD, groupG, how="outer", on=["Area", ___
       →"ParentArea","Year"])
      comboFrame=comboFrame.fillna(0)
[77]: comboFrame.columns
[77]: Index(['Area', 'ParentArea', 'Year', 'TotInd_TotPeople', 'perc_less50k_TotInd',
             'MedInc'],
            dtype='object')
     1.0.1 Scatterplot
[78]: medInc_2013 = np.array(comboFrame.loc[(comboFrame['Year'] == 2013)].MedInc.
      →tolist())
      percLess50k 2013 = np.array(comboFrame.loc[(comboFrame['Year'] == 2013)].
       →perc_less50k_TotInd.tolist())
      medInc_2018 = np.array(comboFrame.loc[(comboFrame['Year'] == 2018)].MedInc.
       →tolist())
      percLess50k 2018 = np.array(comboFrame.loc[(comboFrame['Year'] == 2018)].
       →perc_less50k_TotInd.tolist())
[79]: fig, (ax1, ax2) = plt.subplots(1, 2,sharex=True,sharey=True)
      fig.suptitle('Median household income as fx of % of people less than 50k ')
```

[79]: <matplotlib.collections.PathCollection at 0x119760520>

ax1.scatter(percLess50k\_2013, medInc\_2013)
ax2.scatter(percLess50k\_2018, medInc\_2018)

## Median household income as fx of % of people less than 50k



#### 1.0.2 Correlation tests

```
[80]: # Covariance

print("2013")
    covariance = np.cov([percLess50k_2013], [medInc_2013])
    print(covariance)

# Pearson's correlation
    corrP, _ = pearsonr(percLess50k_2013, medInc_2013)
    print('Pearsons correlation: %.3f' % corrP)

# Spearman's correlation
    corrS, _ = spearmanr(percLess50k_2013, medInc_2013)
    print('Spearmans correlation: %.3f' % corrS)

print("2018")
    covariance = np.cov([percLess50k_2018], [medInc_2018])
    print(covariance)

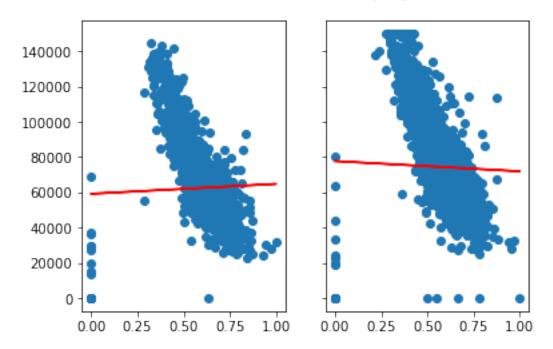
# Pearson's correlation
    corrP, _ = pearsonr(percLess50k_2018, medInc_2018)
```

```
print('Pearsons correlation: %.3f' % corrP)
      # Spearman's correlation
      corrS, _ = spearmanr(percLess50k_2018, medInc_2018)
      print('Spearmans correlation: %.3f' % corrS)
     2013
     [[2.87525099e-02 1.61383410e+02]
      [1.61383410e+02 6.43667861e+08]]
     Pearsons correlation: 0.038
     Spearmans correlation: -0.570
     2018
     [[ 2.67108039e-02 -1.54807622e+02]
      [-1.54807622e+02 9.31815329e+08]]
     Pearsons correlation: -0.031
     Spearmans correlation: -0.565
     1.0.3 Linear regression
     regressor - # Facilities; predictor - employee count
[81]: percLess50k_2013 = percLess50k_2013.reshape((-1, 1))
      percLess50k_2018 = percLess50k_2018.reshape((-1, 1))
[82]: model_2013 = LinearRegression().fit(percLess50k_2013, medInc_2013)
      model_2018 = LinearRegression().fit(percLess50k_2018, medInc_2018)
[83]: print("2013")
      r_sq = model_2013.score(percLess50k_2013, medInc_2013)
      print('coefficient of determination:', r_sq)
      print('intercept:', model_2013.intercept_)
      print('slope:', model_2013.coef_)
      print("2018")
      r_sq = model_2018.score(percLess50k_2018, medInc_2018)
      print('coefficient of determination:', r_sq)
      print('intercept:', model_2018.intercept_)
      print('slope:', model_2018.coef_)
     2013
     coefficient of determination: 0.0014072789383139384
     intercept: 59192.080801861244
     slope: [5612.84598349]
     2018
     coefficient of determination: 0.0009628704794407694
```

intercept: 77761.26683718424
slope: [-5795.69312162]

[84]: [<matplotlib.lines.Line2D at 0x119b92a90>]

## Median household income as fx of % of people less than 50k



[]: