## clustering\_hclust

June 11, 2018

```
In [1]: import pandas as pd
    import seaborn as sn
```

## 0.1 1. Preparing Data

Read data from a specified location

```
In [2]: from IPython.core.interactiveshell import InteractiveShell
       InteractiveShell.ast_node_interactivity = "all"
In [3]: raw_df = pd.read_csv( "/Users/Rahul/Documents/Datasets/Hclust_Beer data.csv",
                              sep = ',', na_values = ['', ''])
       raw_df.columns = raw_df.columns.str.lower().str.replace(' ', '_')
       raw_df.head()
Out [3]:
                    beer cal sod alc cost
          id
                              15 4.7 0.43
          1
               Budweiser 144
       1
           2
                 Schlitz 151
                                19 4.9 0.43
         3
               Lowenbrau 157
                                15 4.9 0.48
                                7 5.2 0.73
         4 Kronenbourg 170
                 Heineken 152
                                11 5.0 0.77
           5
```

## 0.2 2. Extract Features and Standardize

Two ways to extract the features:

- use pd.filter and pass the list of features to extract for scaling
- Use pd.drop and pass the list of features which need not be extracted

The feature can also be extracted by using dataframeName[[<name of features>]]

```
In [4]: #feature_df = raw_df[['cal', 'sod', 'alc', 'cost']]
    feature_df = raw_df.drop({'id','beer'}, axis =1)
    col_names = feature_df.columns
    #col_names

row_index = raw_df.iloc[:,1]
    #row_index
```

Use rename function, in case renaming of a specific column or index is required

```
In [6]: #feature_scaled_df.rename(index={'Budweiser':'Bud'}, inplace=True)
```

The referening of a row or column can be changed by using the below code chunk. Uncomment and change the values within iloc to understand how referencing works:

In [8]: feature\_scaled\_df

| Out[8]: |                | cal       | sod       | alc       | cost      |
|---------|----------------|-----------|-----------|-----------|-----------|
|         | beer           |           |           |           |           |
|         | Budweiser      | 0.393336  | 0.007795  | 0.351095  | -0.470541 |
|         | Schlitz        | 0.630693  | 0.631369  | 0.621168  | -0.470541 |
|         | Lowenbrau      | 0.834143  | 0.007795  | 0.621168  | -0.114071 |
|         | Kronenbourg    | 1.274950  | -1.239354 | 1.026278  | 1.668282  |
|         | Heineken       | 0.664602  | -0.615780 | 0.756205  | 1.953459  |
|         | Old Mil        | 0.427244  | 1.254943  | 0.216058  | -1.539953 |
|         | Augsburger     | 1.444491  | 1.410837  | 1.431387  | -0.684423 |
|         | Strohs         | 0.562877  | 1.878518  | 0.351095  | -0.541835 |
|         | Miller lite    | -1.132535 | -0.771673 | -0.189051 | -0.470541 |
|         | Bud light      | -0.657820 | -1.083460 | -0.999270 | -0.399247 |
|         | Coors          | 0.257703  | 0.475475  | 0.216058  | -0.399247 |
|         | Coors lite     | -1.030811 | 0.007795  | -0.459124 | -0.327953 |
|         | Michelob light | 0.088161  | -0.615780 | -0.324088 | 0.028518  |
|         | Becks          | 0.596785  | 0.631369  | 0.351095  | 1.882165  |
|         | Kirin          | 0.562877  | -1.395248 | 0.756205  | 2.096047  |
|         |                |           |           |           |           |

## 0.3 3. Cluster and Visualize

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Refer to http://docs.scipy.org/doc/scipy/reference/generated/scipy.cluster.hierarchy.linkage.html for reference to available methods, metric.

-2.183691 0.007795 -2.889782 -0.827012

0.393336 1.410837 0.621168 -0.470541

-2.048058 -1.395248 -2.079563 -0.256659

Schilitz lite -1.200352 -1.239354 -0.324088 -0.185365

Rather than standardizing the values above We could have set z\_score parameter inside the clustermap to 1 for standardizing the column values.

Out[9]: <seaborn.matrix.ClusterGrid at 0x116372748>



