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In [5]:
        import numpy as np
        import pandas as pd
        import re
In [6]: def check_empty_bins(dtm,bins):
            # check empty bins
            bin_list = np.unique(dtm[bins].astype(str)).tolist()
            if 'nan' in bin_list:
                bin_list.remove('nan')
            binleft = set([re.match(r'\((.+),(.+)\)]', i).group(1) for i in bin_list]).di
            binright = set([re.match(r'\setminus((.+),(.+)\setminus]', i).group(2) for i in bin_list]).d
            if binleft != binright:
                bstbrks = sorted(list(map(float, ['-inf'] + list(binright) + ['inf'])))
                bstbrks.pop(-2)
                labels = ['[{},{})'.format(bstbrks[i], bstbrks[i + 1]) for i in range(le
                # print("The break points are modified into '[{}]'. There are empty bins
                # binning
                # dtm['bin'] = dtm['bin'].astype(str)
            # return
            return bstbrks
        #字符型或者唯一值较少的变量
        def psi1(data,psi data,var):
            a = data[var].value_counts().reset_index(drop=False)#.astype(str)
            a.rename(columns={var:'name','count':'开发'+var},inplace=True)
            b=psi_data[var].value_counts().reset_index(drop=False)#.astype(str)
            b.rename(columns={var:'name','count':'验证'+var},inplace=True)
            m=pd.merge(a,b,on='name',how='inner')
            m[var+'开发频率']=m['开发'+var]/sum(m['开发'+var])
            m[var+'验证频率']=m['验证'+var]/sum(m['验证'+var])
            m['psi']=(m[var+'开发频率']-m[var+'验证频率'])*np.log(m[var+'开发频率']/m[va
            psi_sum=sum(m['psi'])
            return psi_sum
        def psi2(data,psi data,var,brk):
            a = pd.cut(data[var], brk, right=False).value_counts().reset_index(drop=Fals
            a.rename(columns={var:'name','count':'开发'+var},inplace=True)
            b=pd.cut(psi_data[var], brk,right=False).value_counts().reset_index(drop=Fal
            b.rename(columns={var:'name','count':'验证'+var},inplace=True)
            m=pd.merge(a,b,on='name',how='inner')
            m[var+'开发频率']=m['开发'+var]/sum(m['开发'+var])
            m[var+'验证频率']=m['验证'+var]/sum(m['验证'+var])
            m['psi']=(m[var+'开发频率']-m[var+'验证频率'])*np.log(m[var+'开发频率']/m[va
            psi_sum=sum(m['psi'])
            return psi_sum
        def psi_hui(df,psi_data,target,n=5):
            chat=list(df.columns[df.dtypes == 'object'])
            name=df.columns.drop(target)
            psis=[]
            for i in name:
                X=df[i]
                Y=df[target]
                nuniq=X.nunique()
                if nuniq<=n:</pre>
                    chat.append(i)
                if i in chat:
                    psi=psi1(df,psi_data,i)
                    psis.append(psi)
                else:
                    d1=pd.DataFrame({"X":X,"Y":Y,"bin":pd.qcut(X,n,duplicates='drop')})
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d1['bin']=d1['bin'].astype(str)
                   brk=check_empty_bins(d1, 'bin')
                   psi=psi2(df,psi_data,i,brk)
                   psis.append(psi)
                print (i)
            d=pd.DataFrame({"name":name, "psi":psis})
            return d
In [7]: data = pd.read_csv('test_data.csv')
        print('Shape:',data.shape)
        data.head(10)
      Shape: (500, 7)
                              C
Out[7]:
                     В
                                        D
                                              Ε
                                                           target
        0 0.417022 87 0.154276
                                  0.383389 NaN 84.100880
                                                                0
        1 0.720324 57 0.758797
                                  0.769808 NaN 74.668182
                                                                1
        2 0.000114 24 0.197145 -0.105166 NaN
                                                  7.504475
                                                                0
        3 0.302333 55 0.442048
                                0.300465 NaN 46.824730
                                                                1
        4 0.146756 49 0.399363
                                                 45.993059
                                  0.096637 NaN
                                                                1
        5 0.092339 52 0.045981
                                 -0.022774 NaN
                                                50.068966
                                                                1
        6 0.186260 24 0.322268 -0.039216 NaN 37.597628
                                                                0
        7 0.345561 15 0.311033 0.412043 NaN
                                                 26.698194
          0.396767 67 0.393938 0.435124 NaN 56.782580
                                                                0
        9 0.538817 73 0.407933
                                  0.509187 NaN 71.328520
                                                                0
In [8]: psi_hui(data.iloc[0:250],data.iloc[250:500],'target')
      Α
      В
      C
      D
      Е
      F
Out[8]:
           name
                       psi
        0
              A 0.025587
        1
               B 0.038886
        2
              C 0.004393
        3
              D 0.037093
        4
               E 0.277259
               F 0.015742
        5
In [ ]:
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