

Finding the Best Threshold

May 4, 2021

1 Finding the Best Threshold

This notebook determines the best threshold/time window combinations for identifying episodic and chronic shelter users.

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```
[2]: %load_ext autoreload
      %autoreload 1
```

```
[3]: import numpy as np
      import pandas as pd
      import datetime, copy, imp
      import time
      import matplotlib.pyplot as plt
      import sys

      from tqdm.auto import tqdm, trange
      from tqdm.notebook import tqdm
      tqdm.pandas()

      %aimport di_data
      from di_data import *
      import KMeansCluster as km
```

1.0.1 Pre-Processing

```
[4]: dirStr = '~/data/plwh/'
```

```
[5]: def PreProcess():

    tblAll = pd.read_hdf(dirStr + 'UniversityExportAnonymized.hd5')

    tblAll = tblAll[tblAll.Date >= pd.to_datetime('2007-07-01')]

    print('Total Entries: {}'.format(len(tblAll.index)))
    print('Dates: ',min(tblAll.Date), ' to ',max(tblAll.Date))

    tbl = copy.deepcopy(tblAll[tblAll.EntryType == 'Sleep'][
↳['Date','ClientId' ]])

    # To address left censoring: Remove all clients with first sleep date
↳within a year of the 2008 data import.
    leftStart = tbl.Date.min()
    leftEnd = pd.to_datetime('2009-07-01')

    # To address right censoring: Remove all clients with first sleep date
↳within approximately 2 years of the end
    # of the data. Reasoning: We want to allow a 2 year window to give the
↳clients a chance to become chronic.
    rightStart = pd.to_datetime('2018-01-20')
    rightEnd = tbl.Date.max()

    nClientsAll = len(tbl.ClientId.unique())

    tbl = RemoveByStartDate(tbl,leftStart,leftEnd)
    nLeftRemoved = nClientsAll - len(tbl.ClientId.unique())

    tbl = RemoveByStartDate(tbl,rightStart,rightEnd)
    nRightRemoved = nClientsAll - nLeftRemoved - len(tbl.ClientId.unique())

    # Discard all data before September 1, 2008 due to import errors from
↳previous database.
    tbl = tbl.loc[tbl.Date >= pd.to_datetime('2008-09-01')]

    nClients = len(tbl.ClientId.unique())

    print('Total Clients: {:d}/{:d} ({:.1f}%) ({:d} removed left, {:d} removed
↳right)'
        .format(nClients,nClientsAll,100.0*nClients/
↳nClientsAll,nLeftRemoved,nRightRemoved))
```

```
return tbl
```

```
[6]: tbl = PreProcess()
```

Total Entries: 5431521

Dates: 2007-07-01 00:00:00 to 2020-01-20 00:00:00

Total Clients: 18398/34577 (53.2%) (12609 removed left, 3570 removed right)

```
[ ]:
```

1.0.2 Identify Episodes and Stays

```
[7]: def GenStays():  
      return tbl.groupby('ClientId').progress_apply(CalculateStaySequence)
```

```
[9]: stays = GenStays()
```

```
0%|          | 0/18398 [00:00<?, ?it/s]
```

```
[10]: def GenEpis():  
       return tbl.groupby('ClientId').progress_apply(CalculateEpisodeSequence)
```

```
[11]: epis = GenEpis()
```

```
0%|          | 0/18398 [00:00<?, ?it/s]
```

```
[12]: # Determine the total number of stays and episodes for each client and store in  
      ↪ seDat.
```

```
def JointStaysAndEpis():  
    nClients = len(tbl.ClientId.unique())  
    clientInds = stays.index.get_level_values(0).drop_duplicates()  
    seDat = pd.DataFrame({  
        'nSty': [0]*nClients, # Stores total stays.  
        'nEpi': [0]*nClients }, # Stores total episodes.  
        index=clientInds)  
    for id in tqdm(clientInds):  
        seDat.loc[id, 'nSty'] = stays.loc[id].Ind.max() # Number of stays is max  
        ↪ stay index.  
        seDat.loc[id, 'nEpi'] = epis.loc[id].Ind.max() # Number of episodes is  
        ↪ max episode index.  
  
    return seDat
```

```
[13]: seDat = JointStaysAndEpis()
```

```
0%|          | 0/18398 [00:00<?, ?it/s]
```

[]:

```
[14]: # Determine the number of new clients that register per month.
regDates = list(tbl.groupby('ClientId').Date.min()) # List of first dates that
↳each client appears.

# Resample the dates to determine how many new registrants we have per month.
nClients = len(tbl.ClientId.unique())
monthlyReg = pd.DataFrame([1]*nClients,index=regDates).groupby(pd.
↳Grouper(freq='M')).count()

print('New Clients per Month - Mean: %g, Median: %g' % (monthlyReg.
↳mean(),monthlyReg.median()))
```

New Clients per Month - Mean: 178.621, Median: 172

1.0.3 Perform Cluster Analysis

```
[15]: k = 3 # Established in previous work to yield the transitional, chronic and
↳episodic clusters.

cluster = km.KMeansCluster(k,seDat)
(mus,clstrAsgn,pTable) = cluster.Solve(40,0.005)

# The cluster indices representing chronic, episodic and transitional can
↳change from run to
# run due to how the cluster analysis is randomly initialized. This code
↳determines those indices.
chrId = np.argmax(mus[:,0]) # Chronic clients have the highest number of stays.
epiId = np.argmax(mus[:,1]) # Episodic clients have the highest number of
↳episodes.
tmpId = list( set([0,1,2]) - set([chrId,epiId]) )[0] # Transitional clients
↳are the group left over.
styMuInd = 0
epiMuInd = 1

pTable
```

```
[15]:   i  j      pVal
0  0  1  1.110223e-16
1  0  2  1.110223e-16
2  1  2  1.110223e-16
```

```
[16]: # Calculate statistics on the three clusters to compare to previously published
↳results.

tmpN = sum(clstrAsgn == tmpId) # Total number of clients in each group.
```

```

chrN = sum(clstrAsgn == chrId)
epiN = sum(clstrAsgn == epiId)

print('Trans (%d/%d, %.1f%%) > AvgStays: %g, AvgEpisodes: %g, '
      % (tmpN,nClients,100*tmpN/
      ↪nClients,mus[tmpId,styMuInd],mus[tmpId,epiMuInd]))
print('Epi (%d/%d, %.1f%%) > AvgStays: %g, AvgEpisodes: %g, '
      % (epiN,nClients,100*epiN/
      ↪nClients,mus[epiId,styMuInd],mus[epiId,epiMuInd]))
print('Chron (%d/%d, %.1f%%) > AvgStays: %g, AvgEpisodes: %g, '
      % (chrN,nClients,100*chrN/
      ↪nClients,mus[chrId,styMuInd],mus[chrId,epiMuInd]))

print('Max stays for single episode transitional client: ',
      seDat[(clstrAsgn == tmpId) & (seDat.nEpi == 1)].nSty.max())

```

```

Trans (15675/18398, 85.2%) > AvgStays: 30.3157, AvgEpisodes: 1.81684,
Epi (2184/18398, 11.9%) > AvgStays: 166.994, AvgEpisodes: 9.19048,
Chron (539/18398, 2.9%) > AvgStays: 1273.07, AvgEpisodes: 3.65863,
Max stays for single episode transitional client: 662

```

Demographics

```

[18]: def CalculateClientDemographics():
      return tbl.groupby('ClientId').progress_apply(ShelterGroupDemographics)

```

```

[19]: demog = CalculateClientDemographics()

```

```

0%|          | 0/18398 [00:00<?, ?it/s]

```

```

[ ]:

```

1.0.4 Threshold/Window Optimization

```

[20]: dayWinRange = [ 30, 90, 180, 365, 547 ]
      styThshFrac = np.array([ 0.5, 0.75, 0.9 ])
      epiThshRange = [ 2, 3, 4, 5 ]

```

```

[21]: def OptimizeStayThreshold():
      styOpt = { d: { t: pd.DataFrame() for t in (styThshFrac*d).astype(int)} for d
      ↪d in dayWinRange }
      tBar = tqdm( total = len(styThshFrac)*len(dayWinRange) )
      for win in dayWinRange:
          for thsh in (win*styThshFrac).astype(int):
              styOpt[win][thsh] = stays.groupby('ClientId').
              ↪apply(TimeWinThresholdTest,posFlag='sty',negFlag='tmp',thresh=thsh,winSzDays=win)
              tBar.update()

```

```
tBar.close()
return styOpt
```

```
[22]: styOpt = OptimizeStayThreshold()
```

```
0%|          | 0/15 [00:00<?, ?it/s]
```

```
[23]: def OptimizeEpiThreshold():
```

```
    epiOpt = { d: { t: pd.DataFrame() for t in epiThshRange} for d in
→dayWinRange }
    tBar = tqdm( total = len(epiThshRange)*len(dayWinRange) )
    for win in dayWinRange:
        for thsh in epiThshRange:
            epiOpt[win][thsh] = epis.groupby('ClientId').
→apply(TimeWinThresholdTest,posFlag='epi',negFlag='tmp',thresh=thsh,winSzDays=win)
            tBar.update()

    tBar.close()
    return epiOpt
```

```
[24]: epiOpt = OptimizeEpiThreshold()
```

```
0%|          | 0/20 [00:00<?, ?it/s]
```

1.0.5 Evaluate Impact of Interventions

```
[25]: # Determines the number of shelter stays and days of shelter interaction saved
→if a client
# identified as chronic or episodic is placed in a house the day they're
→identified.
def CalcImpact(
    testList, # Client list after being processed by a shelter stay test.
    stays     # Stay table.
):

    staysSaved = 0
    tenureDaysSaved = 0

    # Loop through all clients that are not transitional.
    for clientId in testList[testList.Flag != 'tmp'].index:

        # Find the stays that occur after the client is identified
        saved = stays.loc[clientId][stays.loc[clientId].Date >= testList.
→loc[clientId].Date]
```

```

        staysSaved += saved.Ind.count()-1
        tenureDaysSaved += (saved.Date.iloc[-1]-testList.loc[clientId].Date).
→days

    nClientsIdentified = sum(testList.Flag != 'tmp')

    return np.array([
        nClientsIdentified, staysSaved, tenureDaysSaved,
        testList[testList.Flag!='tmp'].Time.sum(),
        testList[testList.Flag!='tmp'].Time.median()
    ])

```

```

[26]: def EvalThresholdImpacts(opt):

    keys1dim = list(opt.keys())
    nKeys = len(keys1dim) * len(opt[keys1dim[0]])
    impact = np.zeros((nKeys,7))

    iRow = 0
    tBar = tqdm(total=nKeys)
    for win in opt.keys():
        for thsh in opt[win].keys():
            impact[iRow,:] = np.concatenate( (np.
→array([win,thsh]),CalcImpact(opt[win][thsh],stays)) )
            iRow += 1
            tBar.update()

    tBar.close()
    return impact

```

```

[27]: impInd = { 'Thsh': 0, 'Win': 1, 'N': 2, 'TotStySv': 3, 'TotTnSv': 4, 'IdTTot':
→5, 'IdTMd': 6 }

```

```

[28]: impactEpi = EvalThresholdImpacts(opt=epiOpt)

```

```

0%|          | 0/20 [00:00<?, ?it/s]

```

```

[29]: impactSty = EvalThresholdImpacts(opt=styOpt)

```

```

0%|          | 0/15 [00:00<?, ?it/s]

```

```

[30]: def PrintImpact(tbl,typeFlg,nClients,nTop=10):
    iThsh=1; iWin=0; iN=2; iTotStySv=3; iTotTnSv=4; iIdTTot=5; iIdTMd=6

    print('Win/Thsh      n      StySvPr TenRdPr AvgSpd MdSpd')
    print('-----')

```

```

for rw in tbl[0:nTop]:
    print('%3d/%3d %4d,%4.1f%% %5.1f %5.1f %5.1f %5.1f'
          % (int(rw[iWin]), int(rw[iThsh]), int(rw[iN]), 100.0*rw[iN]/
↪nClients,
          rw[iTotStySv]/rw[iN], rw[iTotTnSv]/rw[iN], rw[iIdTTot]/rw[iN],
↪rw[iIdTmd] ) )

```

```

[31]: print('\nEpi Test (Tenure Reduction Ranking)')
order = ( -impactEpi[:,4] / impactEpi[:,2] ).argsort()
PrintImpact(impactEpi[order,:], 'Epi', nClients)

```

Epi Test (Tenure Reduction Ranking)

Win/Thsh	n	StySvPr	TenRdPr	AvgSpd	MdSpd
180/ 4	310, 1.7%	87.5	932.9	933.3	731.5
365/ 5	618, 3.4%	91.7	906.6	985.2	748.5
365/ 3	3476, 18.9%	99.0	872.8	602.2	313.0
547/ 5	1246, 6.8%	85.3	868.1	925.4	666.5
180/ 3	1903, 10.3%	85.3	864.7	712.0	460.0
547/ 4	2331, 12.7%	90.9	858.1	769.9	485.0
365/ 4	1581, 8.6%	86.3	857.0	816.9	569.0
547/ 3	4201, 22.8%	101.2	854.7	572.9	367.0
90/ 2	4277, 23.2%	93.6	829.7	477.7	181.0
90/ 3	275, 1.5%	78.3	829.3	871.9	663.0

```

[32]: print('Sty Test (Stays Saved Ranking)')
order = ( -impactSty[:,3] / impactSty[:,2] ).argsort()
PrintImpact(impactSty[order,:], 'Chr', nClients)

```

Sty Test (Stays Saved Ranking)

Win/Thsh	n	StySvPr	TenRdPr	AvgSpd	MdSpd
547/492	415, 2.3%	721.4	980.2	852.6	538.0
365/328	594, 3.2%	687.9	996.8	716.4	376.5
547/410	661, 3.6%	656.6	1000.3	787.7	496.0
365/273	904, 4.9%	619.5	1017.8	658.8	350.5
180/162	1075, 5.8%	588.4	1030.1	515.3	222.0
547/273	1138, 6.2%	555.7	1011.6	673.0	421.5
180/135	1583, 8.6%	504.4	1006.4	496.2	188.0
365/182	1536, 8.3%	499.7	1006.3	558.4	287.0
90/ 81	1815, 9.9%	471.2	979.7	419.3	137.0
180/ 90	2370, 12.9%	416.6	997.4	414.7	142.0

```

[33]: rpdChrThsh = 81
      rpdChrWin = 90
      rpdEpiThsh = 2

```



```
rpdEpiWin = 90
```

1.0.6 Compare Definitions

```
[34]: def CalcRapidChronicDefinition():  
        return stays.groupby('ClientId').  
        ↪progress_apply(TimeWinThresholdTest,posFlag='chr',negFlag='tmp',thresh=rpdChrThsh,winSzDays=
```

```
[36]: defRpdChr = CalcRapidChronicDefinition()
```

```
0%|          | 0/18398 [00:00<?, ?it/s]
```

```
[37]: def CalcRapidEpisodicDefinition():  
        return epis.groupby('ClientId').  
        ↪progress_apply(TimeWinThresholdTest,posFlag='epi',negFlag='tmp',thresh=rpdEpiThsh,winSzDays=
```

```
[38]: defRpdEpi = CalcRapidEpisodicDefinition()
```

```
0%|          | 0/18398 [00:00<?, ?it/s]
```

```
[39]: defRpd = ChooseEarliestTest(defRpdChr,defRpdEpi)
```

```
[40]: # Checks for clients are are 'continually homeless' for a threshold number of  
        ↪days.  
        # - We define this as having an episode of homelessness longer than the  
        ↪threshold.  
        def continually_homeless(tbl,flagStr,reqDuration):  
  
            gapVals = tbl.Date.diff().astype('timedelta64[D]') # Gaps are the  
            ↪difference between demog dates.  
  
            # Give each episode of shelter demoges a unique index number.  
            gapInd = (gapVals >= episodeGap).astype('int').cumsum()  
  
            for iGap in range(max(gapInd)+1):  
                startDate = tbl[gapInd==iGap].Date.min()  
                endDate = tbl[gapInd==iGap].Date.max()  
                curDuration = (endDate - startDate).days + 1  
  
                if curDuration > reqDuration:  
                    # Find first date the client's episode exceeded the threshold.  
                    idDate = tbl[(tbl.Date - startDate).dt.days >= reqDuration].Date.  
                    ↪min()  
  
                    epiStartDate = tbl[gapInd==iGap].Date.iloc[0]  
                    overMaxStays = (tbl.Date - epiStartDate).dt.days >= reqDuration
```

```

        return pd.Series({
            'Flag': flagStr, # Flag indicating test was satisfied.
            'Date': idDate, # Date client satisfied the test.
            'Time': (idDate - startDate).days # How long it took to
→satisfy the test.
        })

# Returned if client doesn't satisfy the test.
return pd.Series({
    'Flag': 'tmp',
    'Date': pd.NaT,
    'Time': np.nan
})

# Function test:
#continually_homeless(stays.loc[21910], 'cnt', 365)

```

```

[41]: # Helper function that prints out demog statistics for a cohort of clients.
def print_stats(tblStr, tbl, fields):
    print('--- %s ---' % (tblStr))
    nEntry = len(tbl.index)
    print( 'Clients in cohort: %d/%d (%.1f%%)' % (nEntry, nClients, 100*nEntry/
→nClients))
    print( 'Size of 10%% of cohort: %d/%d (%.1f%%)' % (nEntry*0.
→1, nClients, 100*nEntry*0.1/nClients))
    for field in fields:
        print('%s:' % (field))
        nEntry = sum(~np.isnan(tbl[field]))

        print(' Avg: %.1f, Med: %.1f, 10thPct: %.1f, 90thPct: %.1f'
              %(tbl[field].mean(), tbl[field].median(),
                tbl[field].sort_values().iloc[int(nEntry*0.1)],
                tbl[field].sort_values().iloc[int(nEntry*0.9)]))

```

```

[42]: def CalcAlbertaDefinition():
    cnt = stays.groupby('ClientId').
→progress_apply(continually_homeless, flagStr='cnt', reqDuration=365)
    epi = epis.groupby('ClientId').
→progress_apply(TimeWinThresholdTest, posFlag='epi', negFlag='tmp', thresh=4, winSzDays=365)
    return ChooseEarliestTest(cnt, epi)

```

```

[43]: defGoa = CalcAlbertaDefinition()

```

```

0%|          | 0/18398 [00:00<?, ?it/s]
0%|          | 0/18398 [00:00<?, ?it/s]

```

```
[44]: def CalcCanadaDefinition():
    sty = stays.groupby('ClientId').
    ↳progress_apply(TimeWinThresholdTest,posFlag='sty',negFlag='tmp',thresh=180,winSzDays=365)
    epi = stays.groupby('ClientId').
    ↳progress_apply(TimeWinThresholdTest,posFlag='sty',negFlag='tmp',thresh=546,winSzDays=1095)
    return ChooseEarliestTest(sty,epi)

[45]: defGoc = CalcCanadaDefinition()

0%|          | 0/18398 [00:00<?, ?it/s]
0%|          | 0/18398 [00:00<?, ?it/s]
```

Group Demography

```
[46]: demogGoaTmp = demog[defGoa.Flag == 'tmp']
demogGoaFlg = demog[defGoa.Flag != 'tmp']
demogGocTmp = demog[defGoc.Flag == 'tmp']
demogGocFlg = demog[defGoc.Flag != 'tmp']

demogRapidTmp = demog[defRpd.Flag == 'tmp']
demogRapidFlgChr = demog[defRpdChr.Flag == 'chr']
demogRapidFlgEpi = demog[defRpdEpi.Flag == 'epi']
```

Stats for Clients Identified by a Definition

```
[47]: fields = [ 'Tenure', 'UsagePct', 'AvgGapLen', 'TotalStays', 'TotalEpisodes' ]
fields = [ 'TotalStays', 'TotalEpisodes', 'Tenure', 'UsagePct' ]

print_stats('Government of Canada Chronic Definition',demogGocFlg,fields)
print_stats('Government of Alberta Chronic Definition',demogGoaFlg,fields)
print_stats('RAPID Chronic Definition',demogRapidFlgChr,fields)
print_stats('RAPID Episodic Definition',demogRapidFlgEpi,fields)
```

```
--- Government of Canada Chronic Definition ---
Clients in cohort: 1549/18398 (8.4%)
Size of 10% of cohort: 154/18398 (0.8%)
TotalStays:
  Avg: 702.7, Med: 522.0, 10thPct: 235.0, 90thPct: 1430.0
TotalEpisodes:
  Avg: 4.4, Med: 3.0, 10thPct: 1.0, 90thPct: 9.0
Tenure:
  Avg: 1564.1, Med: 1439.0, 10thPct: 404.0, 90thPct: 2940.0
UsagePct:
  Avg: 53.0, Med: 48.6, 10thPct: 17.6, 90thPct: 95.8
--- Government of Alberta Chronic Definition ---
Clients in cohort: 2443/18398 (13.3%)
Size of 10% of cohort: 244/18398 (1.3%)
TotalStays:
```

```

    Avg: 438.6, Med: 216.0, 10thPct: 19.0, 90thPct: 1128.0
TotalEpisodes:
    Avg: 7.4, Med: 7.0, 10thPct: 2.0, 90thPct: 14.0
Tenure:
    Avg: 1672.2, Med: 1592.0, 10thPct: 549.0, 90thPct: 2974.0
UsagePct:
    Avg: 28.3, Med: 14.1, 10thPct: 1.7, 90thPct: 84.1
--- RAPID Chronic Definition ---
Clients in cohort: 1815/18398 (9.9%)
Size of 10% of cohort: 181/18398 (1.0%)
TotalStays:
    Avg: 601.0, Med: 411.0, 10thPct: 131.0, 90thPct: 1329.0
TotalEpisodes:
    Avg: 4.0, Med: 3.0, 10thPct: 1.0, 90thPct: 9.0
Tenure:
    Avg: 1399.9, Med: 1241.0, 10thPct: 226.0, 90thPct: 2837.0
UsagePct:
    Avg: 53.1, Med: 49.1, 10thPct: 13.7, 90thPct: 96.2
--- RAPID Episodic Definition ---
Clients in cohort: 4277/18398 (23.2%)
Size of 10% of cohort: 427/18398 (2.3%)
TotalStays:
    Avg: 136.9, Med: 39.0, 10thPct: 4.0, 90thPct: 376.0
TotalEpisodes:
    Avg: 6.3, Med: 5.0, 10thPct: 2.0, 90thPct: 12.0
Tenure:
    Avg: 1308.4, Med: 1149.0, 10thPct: 90.0, 90thPct: 2763.0
UsagePct:
    Avg: 11.2, Med: 5.2, 10thPct: 0.7, 90thPct: 30.4

```

Impact Comparison

```

[48]: def CalcDefinitionImpact(definition):
      return CalcImpact(definition,stays)

[49]: impactGoa = CalcDefinitionImpact(definition=defGoa)

[50]: impactGoc = CalcDefinitionImpact(definition=defGoc)

[51]: impactRpd = CalcDefinitionImpact(definition=defRpd)

[52]: def print_impact_table(imp,titleStr):
      nClients = 18346
      print("%s & %d (%.1f\%) & %.1f & %.1f & %.1f & %.1f \\\ \"
            % (titleStr, imp[0], 100.0*imp[0]/nClients, imp[1]/imp[0], imp[2]/
            ↪imp[0], imp[3]/imp[0], imp[4]))

```

```
[53]: print_impact_table(impactGoa,'Government of Alberta')
      print_impact_table(impactGoc,'Government of Canada')
      print_impact_table(impactRpd,'RAPID')
```

```
Government of Alberta & 2443 (13.3\%) & 264.8 & 914.2 & 609.8 & 365.0 \\
Government of Canada & 1549 (8.4\%) & 498.4 & 1007.9 & 555.2 & 285.0 \\
RAPID & 5507 (30.0\%) & 194.8 & 874.4 & 402.2 & 98.0 \\
```

Program Load

```
[54]: idRpd = defRpd[defRpd.Flag != 'tmp'].groupby(pd.Grouper(key='Date',freq='M')).
      ↪Flag.count()
```

```
[55]: print('Mean ID: %g, Median ID: %g' % ( idRpd.mean(), idRpd.median() ))
```

```
Mean ID: 44.4113, Median ID: 46
```

```
[ ]:
```

Stats for Transitional Clients not Identified by Definitions

```
[56]: fields = [ 'Tenure', 'UsagePct', 'AvgGapLen', 'TotalStays', 'TotalEpisodes' ]
      fields = [ 'TotalStays', 'TotalEpisodes', 'Tenure', 'UsagePct' ]
      print_stats('RAPID Transitional Clients',demogRapidTmp,fields)
```

```
--- RAPID Transitional Clients ---
Clients in cohort: 12891/18398 (70.1%)
Size of 10% of cohort: 1289/18398 (7.0%)
TotalStays:
  Avg: 14.9, Med: 2.0, 10thPct: 1.0, 90thPct: 40.0
TotalEpisodes:
  Avg: 1.6, Med: 1.0, 10thPct: 1.0, 90thPct: 3.0
Tenure:
  Avg: 382.7, Med: 7.0, 10thPct: 1.0, 90thPct: 1450.0
UsagePct:
  Avg: 59.6, Med: 93.5, 10thPct: 0.4, 90thPct: 100.0
```

Transitional Client 90th Percentile Investigation

- These are the group of clients in the program delivery transitional population that are in the top 90th percentile in terms of the length of their shelter tenure.
- There are lots of these folks (more than the chronic population), they interact with shelter over a very long period and are likely flying under the radar of most support programs.

```
[57]: tenure90thValue = 1289 # Taken from the output provided above.
      demog90th = demogRapidTmp[demogRapidTmp.Tenure > tenure90thValue]
```

```
[58]: fields = [ 'TotalStays', 'TotalEpisodes', 'Tenure', 'UsagePct', 'AvgGapLen' ]
      print_stats('Transitional Client 90th Percentile',demog90th,fields)
```

```
--- Transitional Client 90th Percentile ---
Clients in cohort: 1518/18398 (8.3%)
Size of 10% of cohort: 151/18398 (0.8%)
TotalStays:
  Avg: 35.8, Med: 7.5, 10thPct: 2.0, 90thPct: 85.0
TotalEpisodes:
  Avg: 3.3, Med: 3.0, 10thPct: 2.0, 90thPct: 5.0
Tenure:
  Avg: 2073.6, Med: 1935.5, 10thPct: 1399.0, 90thPct: 2999.0
UsagePct:
  Avg: 1.8, Med: 0.4, 10thPct: 0.1, 90thPct: 4.4
AvgGapLen:
  Avg: 635.8, Med: 302.4, 10thPct: 23.0, 90thPct: 1747.0
```

[]: