# Cleaning GPS data collected from Catoctin park hikers

In this porject I intend to clean a dataset collected between July 31st and August 20th from 121 GPS divices that were distributed among the hikers in Catoktin Mountain Park, MD. During this period, every hiker was given a GPS divice in the beginning of her hike at the visiting center and was asked to return the divice at the end of her hike, at the same location (i.e. visitors' center). The data collected from these GPS is not consistent as many GPS holders have occasionally turned off the divice or have drived long distances while holding these divices. Since we are looking into their hiking behaviors we are only interested in hiking trajectories and therefor will need to exclude undesired trajectories. Since the hikers have paused at the visitors' center for relatively long time, we also would like to exclude the data collected during that time since these data points significantly affect the average speed and time spent. Therefore we are looking at the following steps:

- Remove data points located at the visiting center
- Remove data points collected while driving by filtering high speed path segments
- Identify spots where the hiker turns of his/her diviceRemove path segments where the time interval or distance between two consecutive points is larger than one minute and 180 meters, respectively.
- Remove tracks with less than 50 points (shorter than roughly 10 minutes hikes)
- Remove extraneous points. These points include those points where more than one point has been collected from a certain location at a certain time
- Remove points that are not in the Catoctin Mountain Park Area.

## Accessing the data stored in shapefiles

```
In [1]:
        # open folders
        import os
        path = "C:/Users/sur216/Desktop/GPS Tracks_ALL"
        folders = os.listdir (path)[1:]
        dbf_path = []
        for n,f in enumerate(folders):
                if not "GPS " in f: continue
                n_path = path + "/"+f
                print f
                for j in os.listdir(n_path):
                     if "dbf" in j:dbf_path.append(str(n_path + "/" +j))
        GPS - Tracks 8-3-16
        GPS - Tracks 8-4-2016
        GPS Tracks - 7-31-16
        GPS Tracks - 8-1-16
        GPS Tracks - 8-10-2016
        GPS Tracks - 8-11-2016
        GPS Tracks - 8-13-2016
        GPS Tracks - 8-17-2016
        GPS Tracks - 8-19-2016
        GPS Tracks - 8-20-2017
        GPS Tracks - 8-5-2016
        GPS Tracks - 8-6-2016
        GPS Tracks - 8-7-2016
        GPS Tracks - 8-9-2016
        GPS Tracks- 8-14-2016
        GPS Tracks- 8-16-2016
```

We will next read through dbf files and record everything in python lists. As shown below, this data includes 22 columns.

```
In [2]:
        # open shapefiles and save the data into python lists
        from dbfread import DBF
        import pandas as pd
        records = []
        for n, i in enumerate(dbf path):
                 idr = i.split("/")[-1][:-4]
                 if not "pt" in idr:continue
                 for record in DBF(i):
                     record.update({"id":idr})
                     records.append(record)
        columns = []
        for n,r in enumerate(records):
             if n==1:
                 for j in r: columns.append(j)
        print columns
        [u'type', u'tident', u'ident', u'Latitude', u'Longitude', u'y_proj', u'x_pro
        j', u'comment', u'new_trk', u'new_seg', u'display', u'color', u'altitude',
         u'depth', u'temp', u'time', u'model', u'filename', u'ltime', u'desc', u'lin
        k', 'id']
In [3]:
        # convert the lists to Pandas DataFrame. We have 22 columns and 128930 rows
        recs = []
        for n,r in enumerate(records):
                 lst = []
                 for c in columns :
                     lst.append(r[c])
                 recs.append(1st)
        df = pd.DataFrame(recs)
        df.columns = columns
        print "table size :",df.shape
        table size : (128930, 22)
In [4]: | print len(df["id"].unique())
        42
```

## **Creating new ids**

As we can see this data includes 42 unique GPS ids, however, recall that these GPS divices have been assigned to different people at different dates. Therefore, we will need to create a new ID that considers both the date and the GPS divice. We will call this "new\_id". We can see that we have 146 unique IDs now.

```
In [5]: # create new id for each GPS track based on data and file name
    new_id = ["/".join(["-".join(list(i.split(" ")[0].split("/")[1:]))]+[j]) for
    i,j in zip(df['time'].tolist(),df['id'].tolist())]
In [6]: df['new_id'] = new_id
```

```
In [7]: print len(df["new_id"].unique())
146
```

#### **Excluding the points nearby the visitor center**

As discussed earlier, we wanted to exclude the points collected from the visitor center. To this end, I difined a 46X63 meters rectangle around the visitor center and excluded every point that was inside this square.

```
In [8]: # identifying those points that fall near the visitors' center
    df2 = df[(df['Latitude']<39.63435507) & (df['Latitude']>39.63378507) & (df['L
        ongitude']<-77.44983899) & (df['Longitude']>-77.45038231)]

In [9]: #df3 is the sets of points without those located at the visiting center
    df3 = pd.merge(df, df2, how='outer', indicator=True)
    df3 = df3[df3["_merge"]=="left_only"]
    df3 = df3.iloc[:,0:23]
```

## Calculating speed and distance

In the next step we will calculate the speed between every two consecutive points. This step is important as this will provide the basis for filtering the points. The distance between every two points has been calculated by latitude, longitude and altitude. Also, we will calculate and store the change in elevation, planar distance and time difference in separate columns.

```
In [10]: #convert to timestamp format
df3["time"] = pd.to_datetime(df3["time"])
```

```
In [11]:
         # calculate distances, time steps and speeddf4
          from geopy.distance import vincenty
          import math
          import datetime as dt
          import time
          tracks = df3['new_id'].unique()
          df4 = pd.DataFrame()
          for n,t in enumerate(tracks):
              q = df3[df3["new_id"]==t]
              q = q.sort(['time'], ascending=True)
             time = q["time"].tolist()
              lons = q["Longitude"].tolist()
              lats = q["Latitude"].tolist()
              alts = q["altitude"].tolist()
              lat_lng =zip(lats,lons)
              dist horiz = ["NA"]
              dist_vert = ["NA"]
              tot dist = ["NA"]
              time_dif = ["NA"]
              speed = ["NA"]
              for n,p in enumerate(lat_lng):
                  if not n==0:
                      y = vincenty(lat_lng[n], lat_lng[n-1]).meters
                      z = alts[n]-alts[n-1]
                      t = time[n]-time[n-1]
                      t = t.total_seconds()
                      d = math.sqrt(y**2+z**2)
                      dist_horiz.append(y)
                      dist vert.append(z)
                      tot_dist.append(d)
                      time dif.append(t)
                      if not t == 0:
                          speed.append(d/t)
                      else: speed.append("NA")
              q['dist_horiz'] = dist_horiz
              q['dist_vert'] = dist_vert
              q['time_dif'] = time_dif
              q['tot_dist'] = tot_dist
              q['speed'] = speed
              df4 = df4.append(q)
```

C:\Users\sur216\Anaconda2\lib\site-packages\ipykernel\\_\_main\_\_.py:10: FutureW
arning: sort(columns=....) is deprecated, use sort\_values(by=....)

```
In [222]: df4.to_csv("GPS_compiled.csv", index = False)
```

As we can see below, the time interval between concecutive points is not steady. Although the majority of points have been collected within 10 seconds intervals (104699 points). We can also see that some data points has been collected with 0 seconds intervals. If we print these points out we can see that the distance between these points are also 0. Therefore we will simply remove these from our dataframe.

defaultdict(<type 'int'>, {0: 3554, 1: 20, 2: 14, 3: 11, 4: 5, 5: 3, 6: 3, 7: 3, 8: 5, 9: 4944, 10: 104699, 11: 4961, 12: 11, 3085: 1, 14: 2, 15: 6, 16: 2, 17: 3, 530: 1, 19: 1, 20: 40, 21: 8, 22: 4, 23: 2, 24: 1, 29: 3, 30: 22, 31: 1, 32: 2, 10785: 1, 18: 1, 550: 1, 40: 14, 42: 2, 13319: 1, 44: 1, 7726: 1, 47: 1, 49: 1, 50: 11, 51: 2, 54: 1, 570: 1, 60: 15, 16959: 1, 64: 1, 65: 1, 68: 1, 69: 1, 70: 11, 1614: 1, 13: 2, 80: 4, 2132: 1, 100: 2, 90: 4, 9231: 1, 94: 1, 9828: 1, 619: 1, 108: 1, 109: 1, 110: 3, 111: 1, 119: 1, 9336: 1, 122: 1, 127: 2, 130: 2, 107: 1, 132: 1, 12808: 1, 4716: 1, 140: 1, 14479: 1, 660: 1, 150: 3, 1178: 1, 160: 2, 679: 1, 20649: 1, 170: 2, 1195: 1, 3954: 1, 180: 1, 187: 1, 189: 1, 1733: 1, 200: 1, 34: 1, 2256: 1, 120: 1, 2774: 1, 89 20: 1, 220: 2, 7391: 1, 229: 1, 3821: 1, 7037: 1, 16113: 1, 7412: 1, 250: 1, 1790: 1, 261: 1, 1801: 1, 1290: 1, 14722: 1, 270: 3, 795: 1, 5916: 1, 797: 1, 293: 1, 10538: 1, 300: 2, 3380: 1, 3893: 1, 311: 1, 314: 1, 318: 1, 1464 5: 1, 860: 1, 370: 1, 15736: 1, 893: 1, 2946: 1, 390: 1, 400: 1, 17498: 1, 41 2: 1, 2117: 1, 1955: 1, 420: 1, 9640: 1, 23985: 1, 13748: 1, 13754: 1, 4540: 1, 1469: 1, 963: 1, 460: 1, 2467: 1, 78: 1, 8271: 1, 8673: 1, 2021: 1, 1328 7: 1, 490: 1, 491: 1, 16366: 1, 4591: 1, 9713: 1, 500: 2})

```
In [13]: ls = [int(i) for i in df4[df4["tot_dist"].isin(["NA"])==False]['tot_dist'].un
ique()]
ls2 = [int(i) for i in df4[df4["tot_dist"].isin(["NA"])==False]['tot_dist']]
ls.sort()
from collections import defaultdict
fq= defaultdict( int )
for t in ls2:
    fq[t] += 1
print fq
```

defaultdict(<type 'int'>, {0: 23982, 1: 9085, 2: 5784, 3: 5469, 4: 5469, 5: 5 388, 6: 5867, 7: 6115, 8: 6704, 9: 6997, 10: 7103, 11: 6757, 12: 5697, 13: 45 46, 14: 3377, 15: 2302, 16: 1562, 17: 1097, 18: 716, 19: 500, 20: 430, 21: 32 2, 22: 286, 23: 206, 24: 174, 25: 150, 26: 113, 27: 117, 28: 120, 29: 78, 30: 70, 31: 62, 32: 70, 33: 39, 34: 47, 35: 60, 36: 28, 37: 34, 38: 38, 39: 33, 4 0: 27, 41: 31, 42: 33, 43: 27, 44: 37, 45: 26, 46: 25, 47: 25, 48: 15, 49: 3 3, 50: 22, 51: 29, 52: 23, 53: 18, 54: 14, 55: 20, 56: 14, 57: 15, 58: 14, 5 9: 13, 60: 10, 61: 16, 62: 16, 63: 13, 64: 12, 65: 21, 66: 17, 67: 16, 68: 9, 69: 9, 70: 9, 71: 11, 72: 11, 73: 13, 74: 8, 75: 9, 76: 1, 77: 8, 78: 4, 79: 9, 80: 4, 81: 7, 82: 7, 83: 9, 84: 5, 85: 8, 86: 3, 87: 7, 88: 5, 89: 9, 90: 8, 91: 8, 92: 8, 93: 8, 94: 11, 95: 15, 96: 8, 97: 11, 98: 8, 99: 8, 100: 13, 101: 8, 102: 9, 103: 9, 104: 9, 105: 9, 106: 9, 107: 8, 108: 8, 109: 3, 110: 11, 111: 11, 112: 12, 113: 7, 114: 6, 115: 8, 116: 6, 117: 4, 118: 6, 119: 3, 120: 7, 121: 11, 122: 9, 123: 7, 124: 9, 125: 8, 126: 9, 127: 6, 128: 5, 129: 12, 130: 5, 131: 5, 132: 15, 133: 11, 134: 7, 135: 5, 136: 6, 137: 5, 1 38: 7, 139: 6, 140: 9, 141: 10, 142: 11, 143: 11, 144: 10, 145: 11, 146: 5, 1 47: 9, 148: 7, 149: 6, 150: 3, 151: 9, 152: 3, 153: 7, 154: 8, 155: 5, 156: 9, 157: 5, 158: 4, 159: 6, 160: 12, 161: 8, 162: 5, 163: 3, 164: 7, 165: 5, 166: 1, 167: 5, 168: 6, 169: 3, 170: 6, 171: 6, 172: 7, 173: 5, 174: 3, 175: 7, 176: 4, 177: 2, 178: 8, 179: 3, 180: 4, 181: 1, 182: 3, 183: 5, 184: 4, 18 5: 2, 186: 3, 187: 4, 188: 1, 190: 1, 191: 1, 192: 2, 705: 1, 195: 2, 197: 1, 200: 2, 201: 1, 202: 3, 203: 1, 204: 1, 205: 2, 206: 1, 207: 1, 208: 1, 209: 1, 210: 1, 213: 2, 215: 1, 217: 1, 218: 1, 222: 1, 223: 1, 225: 1, 227: 3, 2 29: 1, 234: 1, 245: 1, 349: 1, 767: 1, 256: 1, 3843: 1, 262: 1, 271: 1, 1498: 1, 306: 1, 3385: 1, 1674: 1, 3390: 1, 3400: 1, 3408: 1, 3409: 1, 193: 1, 341 8: 1, 3420: 1, 3421: 1, 3422: 1, 3424: 1, 3432: 1, 361: 1, 3436: 1, 3645: 1, 3441: 1, 3442: 1, 3439: 1, 374: 1, 3453: 1, 3466: 1, 3467: 1, 3389: 1, 1092: 1, 411: 1, 3486: 1, 3488: 1, 3491: 1, 3497: 2, 3498: 1, 3500: 1, 3506: 1, 344 3: 1, 1974: 1, 3511: 1, 3515: 1, 3518: 1, 672: 1, 1558: 1, 673: 1, 3540: 1, 3 546: 1, 5098: 1, 668: 1, 1524: 1, 505: 1, 529: 1, 426: 1, 510: 1})

```
In [14]: #remove points with 0 time difference
df4 = df4[df4['time_dif'] !=0]
```

In [15]: df4.shape

Out[15]: (115105, 28)

#### filtering according to speed, time and total distance

We will now remove those tracks with less than 50 data points (less than 10 minutes hiking). After this step we will end up with 113 GPS tracks. We will also delet those points that show speeds higher than 3 m/s, the time difference between every two points is higher than 60 seconds and the distnce taken between tow points is higher than 180 meters.

```
In [16]:
         # walked for at least 10 mins
          111 = []
          df5 = pd.DataFrame()
          for n, t in enumerate(tracks):
              q = df4[df4["new_id"]==t]
              if not q.shape[0]<50: 111.append(q)</pre>
          for d in 111:
              if not d.empty: df5 = pd.concat([df5,d], ignore_index=True)
In [17]:
          df5 = df5[df5['speed']<3]
          df5 = df5[df5['time dif']<60]
          df5 = df5[df5['tot_dist']<180]</pre>
          df5.shape
Out[17]: (112809, 28)
In [18]: len(df5['new_id'].unique())
Out[18]: 113
In [24]:
          df5.to_csv("GPS_quasicleaned.csv", index = False)
```

## **Using Arcmap for filtering**

In the next step, we will use the arcmap software to reove the data points that are not close to the Catoctin park area. Occasionally, the GPS holders have drived to a nearby town or other areas without turning their GPS divices off. We are not interested in these data points and our cleaning process so far has not filtered these. To this end, we will use the Arcmap 10.4 software and manually remove these points. After this step, we will convert the DBF data to Pandas Dataframe:

```
In [19]:
          from dbfread import DBF
          import pandas as pd
          records = []
          dbf_path = "C:/Users/sur216/Desktop/GPS Tracks_ALL/all_compiled_projected.db
          for record in DBF(dbf_path):
                       records.append(record)
          print records[1]
          OrderedDict([(u'type', u'TRACK'), (u'tident', u'03-AUG-16-05-1'), (u'ident',
           u'T1'), (u'Latitude', 39.63843433), (u'Longitude', -77.44265922), (u'y_pro
          j', 39.63843433), (u'x_proj', -77.44265922), (u'comment', u''), (u'new_trk',
           u''), (u'new_seg', u''), (u'display', u''), (u'color', u''), (u'altitude', 4
          36.58), (u'depth', 0.0), (u'temp', 0.0), (u'time', datetime.date(2016, 8,
           3)), (u'model', u''), (u'filename', u''), (u'ltime', u''), (u'desc_', u'
           (u'link', u''), (u'id', u'05-1pt'), (u'new_id', u'08-03/05-1pt'), (u'dist_ho
          riz', 6.88132158169), (u'dist vert', 5.29), (u'time dif', 11.0), (u'tot dis
          t', 8.67967088723), (u'speed', 0.789060989748)])
In [107]:
          # save to csv after arcmap edits
          columns = []
          for n,r in enumerate(records):
               if n==1:
                   for j in r: columns.append(j)
          print columns
          recs = []
          for n,r in enumerate(records):
                   lst = []
                   for c in columns :
                       lst.append(r[c])
                   recs.append(lst)
          df = pd.DataFrame(recs)
          df.columns = columns
          print "table size :",df.shape
          [u'type', u'tident', u'ident', u'Latitude', u'Longitude', u'y_proj', u'x_pro
          j', u'comment', u'new_trk', u'new_seg', u'display', u'color', u'altitude',
           u'depth', u'temp', u'time', u'model', u'filename', u'ltime', u'desc_', u'lin
          k', u'id', u'new_id', u'dist_horiz', u'dist_vert', u'time_dif', u'tot_dist',
           u'speed']
```

## Aggregating the data

table size : (111748, 28)

In the next step we will aggregate the values that we are interested in. We would like to have the average speed, total distance and total time spent. We will also add the amount of vertical distance climbed up and down separately.

```
In [108]:
          sm = df.groupby(['new_id']).sum()
           mn = df.groupby(['new_id']).mean()
           mn_spd = mn.loc[:,['speed']]
           mn_spd['new_id']= mn_spd.index
           mn_vert = df.loc[:,['new_id','dist_vert']]
           clm_up = mn_vert[mn_vert['dist_vert']>0]
           clm_up = clm_up.groupby(['new_id']).sum()
           clm_dwn = mn_vert[mn_vert['dist_vert']<0]</pre>
           clm_dwn = clm_dwn.groupby(['new_id']).sum()
           clm_dwn['climb_down'] = clm_dwn['dist_vert']
           clm up['climb up'] = clm up['dist vert']
           del clm_up['dist_vert']
           del clm_dwn['dist_vert']
           clm_up['new_id'] = clm_up.index
           clm_dwn['new_id'] = clm_dwn.index
           mn_vert['dist_vert'] = mn_vert['dist_vert'].abs()
           mn_vert = mn_vert.groupby(['new_id']).mean()
           mn vert['mean elevation'] = mn vert['dist vert']
           mn_vert = mn_vert.loc[:,['mean_elevation']]
```

```
In [109]: up_dwn = clm_up.merge(clm_dwn)
    sm = sm.loc[:,['dist_horiz','time_dif','tot_dist']]
    sm['new_id']= sm.index
    df1 = up_dwn.merge(sm, on = 'new_id')
    df2 = mn_spd.merge(df1, on= 'new_id')
    df2.columns = ['average_speed','new_id','climb_up','climb_down','Total_planar_distance','Total_time','Total_distance']
```

,	new_id	average_speed	climb_up	climb_down	Total_planar_distance
0	07-31/02-1_pt	0.587001	1059.33	-1047.80	12975.771458
1	07-31/03-1_pt	0.813220	697.47	-680.64	7780.278455
2	07-31/06-1_pt	0.784326	445.62	-450.91	5254.870909
3	07-31/07-1_pt	0.763597	502.31	-413.39	4151.602668
4	07-31/08-1_pt	0.877671	474.88	-447.01	6094.750247
5	07-31/08-2_pt	0.307855	712.23	-661.28	5064.454156
6	07-31/09-1_pt	0.865830	1105.10	-1053.19	12523.573990
7	07-31/10-2_pt	0.824399	298.44	-274.88	4166.763060
8	07-31/11-1_pt	0.792242	723.90	-780.14	8188.272212
9	07-31/13-1_pt	0.764995	667.61	-800.27	8598.491833
	Total time To	otal distance			
0	22810.0	13374.081864			
1	9910.0	8056.900198			
2	6910.0	5416.952194			
3	5730.0	4383.761234			
4	7070.0	6221.389793			
5	17791.0	5473.632868			
6	14890.0	12910.774230			
7	5171.0	4261.997310			
8	10720.0	8489.355263			

In [106]: df2.to\_csv("GPS\_aggregated.csv", index = False)

8878.036376

9

11610.0