Location Prediction Algorithm

Shashank Sharma

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1 Introduction

This algorithm is designed to predict human locations in a real world scenario. The GPS data is taken as input and the processed using the below algorithm. The Algorithm has several steps:

- Detect stay-points (also detect start or end of the trajectory)
- Group stay-points to form states
- Calculate hourly weights for the states
- Apply Markov chain for the data available

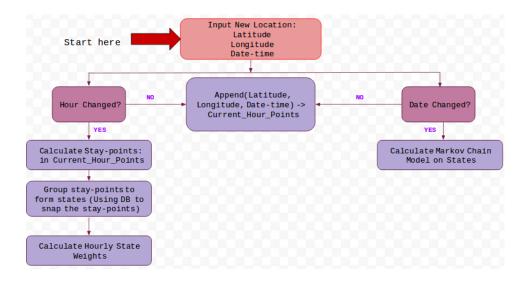


Figure 1: Algorithm Flow-chart

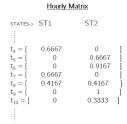


Figure 2: Algorithm Flow-chart



Figure 3: Algorithm Flow-chart

State Hourly Weights

 $\begin{array}{l} \bullet \quad W_{ST1(4)} = (04.45 - 04.05). minutes/60 = 40/60 = 0.6667 \\ \bullet \quad W_{ST2(4)} = 0 \\ \bullet \quad W_{ST1(5)} = 0 \\ \bullet \quad W_{ST2(5)} = (06.00 - 05.20). minutes/60 = 40/60 = 0.6667 \\ \bullet \quad W_{ST2(6)} = (06.55 - 06.00). minutes/60 = 55/60 = 0.9167 \\ \bullet \quad W_{ST1(7)} = (08.00 - 07.20). minutes/60 = 40/60 = 0.6667 \\ \bullet \quad W_{ST2(7)} = 0 \\ \bullet \quad W_{ST2(7)} = 0 \\ \bullet \quad W_{ST2(8)} = (08.25 - 08.00). minutes/60 = 25/60 = 0.4167 \\ \bullet \quad W_{ST2(8)} = (09.00 - 08.35). minutes/60 = 25/60 = 0.4167 \\ \bullet \quad W_{ST2(9)} = 0 \\ \bullet \quad W_{ST2(9)} = (10.00 - 09.00). minutes/60 = 60/60 = 1 \\ \bullet \quad W_{ST2(10)} = 0 \\ \bullet \quad W_{ST2(10)} = (10.20 - 10.00). minutes/60 = 20/60 = 0.3333 \\ \dots \end{array}$

Figure 4: Algorithm Flow-chart

2 Definitions

• Stay-points: Stay-points are any points which are stayed by the user in user trajectories or it is the start or the end of the trajectory. For example, if user start at his home, the home itself is a stay-point. Now he move towards work, but he visit a cafe in between for breakfast. The cafe is also a stay-point and then he finishes his trajectory at work, where work is again a stay-point. The places like cafe in this case is identified using distance and time based clustering. For example, a set of points within 200m with total duration of stay greater than 20 minutes can be regarded as a stay-point within the trajectory.

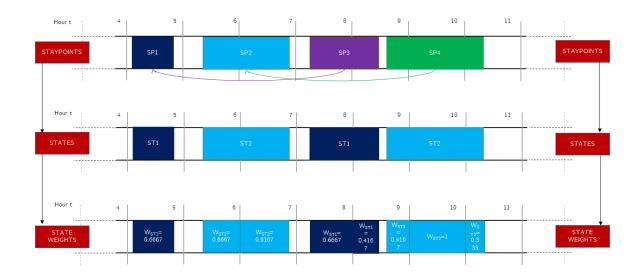


Figure 5: Algorithm Flow-chart

• State: A state is formed using a group of stay-points. This is done using a distance threshold for states. All the stay-points within this threshold distance are grouped together as a single state. This is called snapping stay-points to the states. The mean of all location latitudes and longitudes from stay-points within a state are stored per state. Finally Markov Chain model is applied to the states. Note: A new stay-point is only added to the state if after calculating the mean of the new state, all the existing stay-points still stay within the distance threshold from this mean. This is done to avoid drifting problem while aggregating the stay-points into states.

3 Algorithms

Algorithm 1 Read new location and process

```
Input: GPS Points P/x, y, d/
    Output: Markov Chain mc
1: for each P[x, y, d] do
2:
       newHour \leftarrow d_h
       newDate \leftarrow d_d
3:
       if newHour! = prevHour then
4:
           prevHour \leftarrow newHour
5:
           extractStayPoints()
6:
 7:
           formStates()
           calculateStateWeights()
8:
           if newDate! = prevDate then
9:
               prevDate \leftarrow newDate
10:
               updateMarkovModel()
11:
           end if
12:
           clear lst\_hr\_pts[x, y, d]
13:
       else
14:
           lst\_hr\_pts[x, y, d] \leftarrow P[x, y, z]
15:
       end if
16:
17: end for
```

Algorithm 2 extractStayPoints(): Calculate stay-points

```
Input: Last hour GPS points lst\_hr\_pts[x, y, d]
    Output: Stay-points sp
1: for each lst\_hr\_pts[x, y, d]_i in lst\_hr\_pts[x, y, d] do
 2:
       if d_i - d_{i+1} > th-tck then
           sp \leftarrow lst\_hr\_pts_{i}, lst\_hr\_pts_{i+1}
3:
           adjustStartEndStaypoint()
4:
           continue
5:
       end if
6:
       if distanceBtw(lst\_hr\_pts_i, cluster) \le th\_d then
7:
8:
           cluster \leftarrow lst\_hr\_pts_i
           calculate New Cluster Mean
9:
       else
10:
           if (cluster! = empty) And duration(cluster) >= th_t then
11:
               sp \leftarrow cluster
12:
               adjustStartEndStaypoint()
13:
           end if
14:
       end if
15:
16: end for
```

Algorithm 3 adjustStartEndStaypoint(): Adjust start-end time of stay-points

```
Input: Stay-points sp
   Output: Stay-points with adjusted start and end points sp
1: for each sp_i in sp do
2:
      distance \leftarrow distanceBtw(sp_i, sp_{i+1})
3:
      time \leftarrow timeDifference(sp_i, sp_{i+1})
      avgSpeed \leftarrow distance/time
4:
      delta_t \leftarrow min(distance, thresholdDistance)/AvgSpeed
5:
      Update endTime d_e + delta_t/2 for sp_i
6:
7:
      Update startTime d_s-delta_t/2 for sp_{i+1}
8: end for
```

Algorithm 4 formStates(): Form states from stay-points

```
Input: Stay-points sp
    Output: States st
 1: for each sp_i in sp do
2:
       for each sp_i in sp do
          if distanceBtw(sp_i, sp_j) \le th_d then
3:
              Calculate new mean latitude and longitude for st with sp_i
 4:
              newMean \leftarrow mean(st)
5:
6:
              if distanceBtw(st, newMean) \le th_d then
                  Combine sp_i and sp_j in st
7:
8:
              else
                  Form new st from sp_i
 9:
              end if
10:
          end if
11:
12:
       end for
13: end for
```

Algorithm 5 calculateStateWeights(): Calculate Hourly Weights of States

```
Input: States st
Output: States Weights w

1: This creates a weights of all states from 0 Hrs to 24 Hrs for each date

2: for each st_i in st do

3: w_i \leftarrow minutesContribution(st_i)/60

4: end for
```

Algorithm 6 updateMarkovModel() : Recalculate the Markov Model(This algorithm creates the transition probabilities from state i to i+1 from hour j to j+1)

```
Input: State Weights w
Output: Markov Chain Model mc

1: for each h - hour from 0to24 do

2: for each w_i in w do

3: for each w_j in w do

4: mc \leftarrow w_i[h] * w_j[h+1]

5: end for

6: end for

7: end for
```