

Location Prediction Algorithm

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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

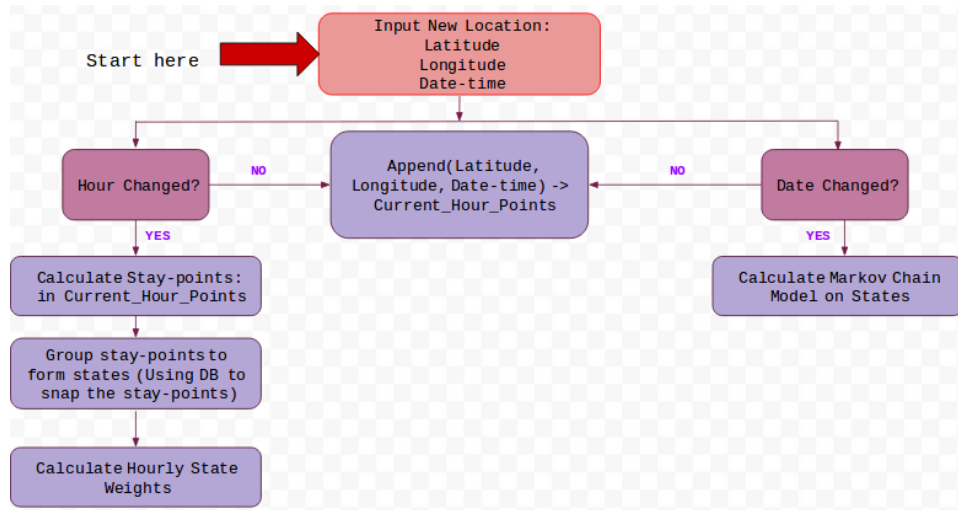


Figure 1: Algorithm Flow-chart

Hourly Matrix

STATES->	ST1	ST2
.....		
$t_4 = [$	0.6667	0
$t_5 = [$	0	0.6667
$t_6 = [$	0	0.9167
$t_7 = [$	0.6667	0
$t_8 = [$	0.4167	0.4167
$t_9 = [$	0	1
$t_{10} = [$	0	0.3333
.....		

Figure 2: Algorithm Flow-chart

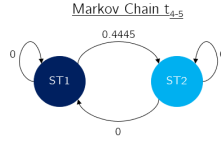


Figure 3: Algorithm Flow-chart

- State Hourly Weights
- $W_{ST1(4)} = (04:45 - 04:05) \text{ minutes}/60 = 40/60 = 0.6667$
 - $W_{ST2(4)} = 0$
 - $W_{ST1(5)} = 0$
 - $W_{ST2(5)} = (06:00 - 05:20) \text{ minutes}/60 = 40/60 = 0.6667$
 - $W_{ST1(6)} = 0$
 - $W_{ST2(6)} = (06:55 - 06:00) \text{ minutes}/60 = 55/60 = 0.9167$
 - $W_{ST1(7)} = (08:00 - 07:20) \text{ minutes}/60 = 40/60 = 0.6667$
 - $W_{ST2(7)} = 0$
 - $W_{ST1(8)} = (08:25 - 08:00) \text{ minutes}/60 = 25/60 = 0.4167$
 - $W_{ST2(8)} = (09:00 - 08:35) \text{ minutes}/60 = 25/60 = 0.4167$
 - $W_{ST1(9)} = 0$
 - $W_{ST2(9)} = (10:00 - 09:00) \text{ minutes}/60 = 60/60 = 1$
 - $W_{ST1(10)} = 0$
 - $W_{ST2(10)} = (10:20 - 10:00) \text{ minutes}/60 = 20/60 = 0.3333$
 -

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.

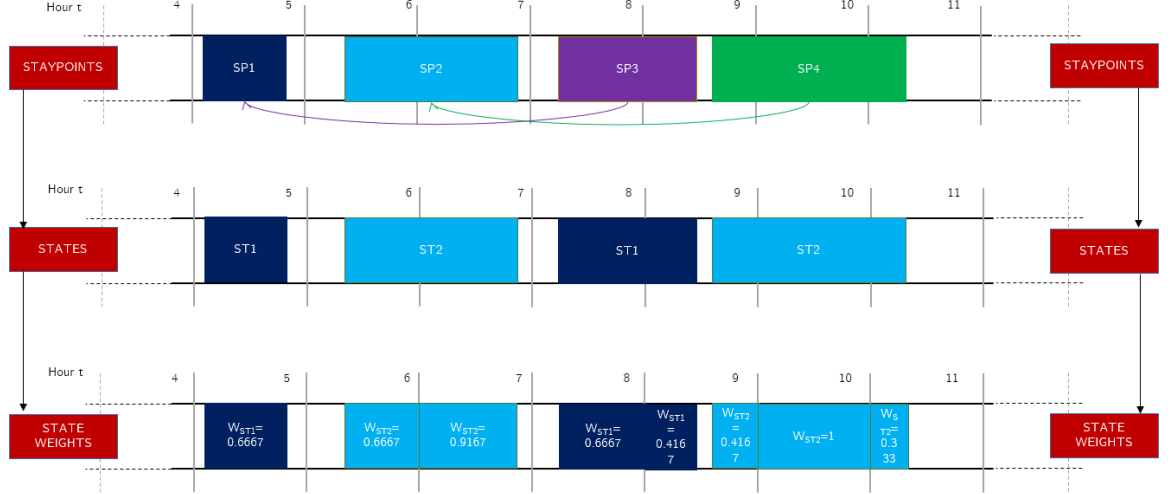


Figure 5: Algorithm Flow-chart

- **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

```
1: Read new latitude, new longitude, new datetime information and process
   these new points
2: while NewLocationDetected == True do
3:   Set newHour = datetime.hour
4:   Set newDate = datetime.date
5:   if newHour! = prevHour then
6:     prevHour  $\leftarrow$  newHour
7:     calculateLastHourStayPoints()
8:     formStates()
9:     calculateStateLastHourWeights()
10:  end if
11:  if newDate! = prevDate then
12:    prevDate  $\leftarrow$  newDate
13:    recalculateMarkovModel()
14:  end if
15: end while
```

Algorithm 2 calculateLastHourStayPoints() : Calculate last hour stay-points

```
1: Calculate stay-points
2: trackingThreshold: Maximum time distance between two points
3: thresholdDistance: Stay-point threshold distance covered
4: thresholdTime: Stay-point threshold time spent
5: for each LastHourPoint do
6:   if (point(i).datetime - point(i - 1).datetime) >= trackingThreshold
   then
7:     Add point(i), point(i-1) as stay-points
8:     recalculateStartEndStaypoint()
9:   end if
10:  if distance(point(i), cluster) <= thresholdDistance then
11:    add point i to cluster
12:    calculate Cluster Means
13:  else
14:    if (cluster! = empty) And duration(cluster) >= thresholdTime
    then
15:      Add this cluster to stay-points
16:      recalculateStartEndStaypoint()
17:    end if
18:  end if
19: end for
```

Algorithm 3 recalculateStartEndStaypoint() : Calculate start-end of staypoints

```
1: for eachStaypoint do
2:   Set  $distance \leftarrow distance(staypoint(i), staypoint(i + 1))$ 
3:   Set  $time \leftarrow timeDifference(staypoint(i), staypoint(i + 1))$ 
4:   Set  $AvgSpeed \leftarrow distance/time$ 
5:   Set  $AddTime \leftarrow \min(distance, thresholdDistance)/AvgSpeed$ 
6:   Set  $endTimeofStaypoint(i) \leftarrow endTimeStaypoint(i) + AddTime$ 
7:   Set  $startTimeofStaypoint(i + 1) \leftarrow startTimeStaypoint(i + 1) - AddTime$ 
8: end for
```

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

```
1: Form each unique stay-point as individual state
2: Now, within this loop, start combining the states
3:  $thresholdState$ : State distance threshold
4: for eachState do
5:   if  $distance(state(i), state(i + 1)) \leq thresholdDistance$  then
6:     calculate new state mean latitude, mean longitude
7:     if  $distance(allExitingState(i)Staypoints, NewMeanLatLong) \leq thresholdState$  then
8:       combine state i, i+1
9:       calculate State Means
10:    end if
11:  end if
12: end for
```

Algorithm 5 calculateStateLastHourWeights() : Calculate Hourly Weights of States

```
1: This creates a weights of all states from 0 Hrs to 24 Hrs for each date
2: for eachState do
3:   if ( $HourChanges$ ) Or ( $StateIDChanges$ ) then
4:     Calculate the start and end of the state i
5:   end if
6: end for
```

Algorithm 6 recalculateMarkovModel() : Recalculate the Markov Model

```
1: This algorithm creates the transition probabilities from state i to i+1 from hour h to h+1
2: for each  $Hth - hour$  from 0 - 24 do
3:   for each  $ith - state$  in  $state - hourly - weight$  do
4:      $state(i) \rightarrow state(i + 1)$  transition for  $H - hour = Matrix[state] * Matrix[State + 1]$ 
5:   end for
6: end for
```
