

# Location Prediction Algorithm

Shashank Sharma

August 2018

## 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

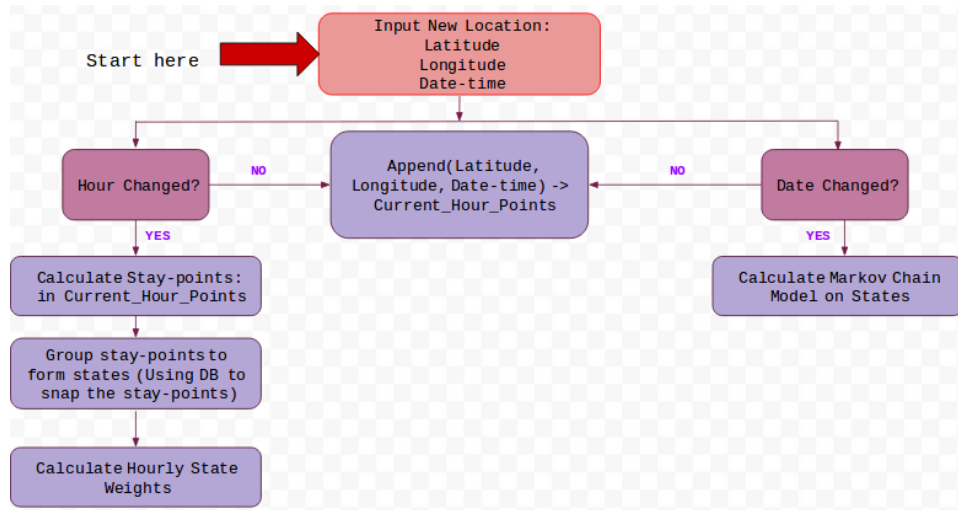


Figure 1: Algorithm Flow-chart

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

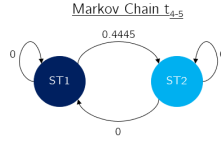


Figure 3: Algorithm Flow-chart

- 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.

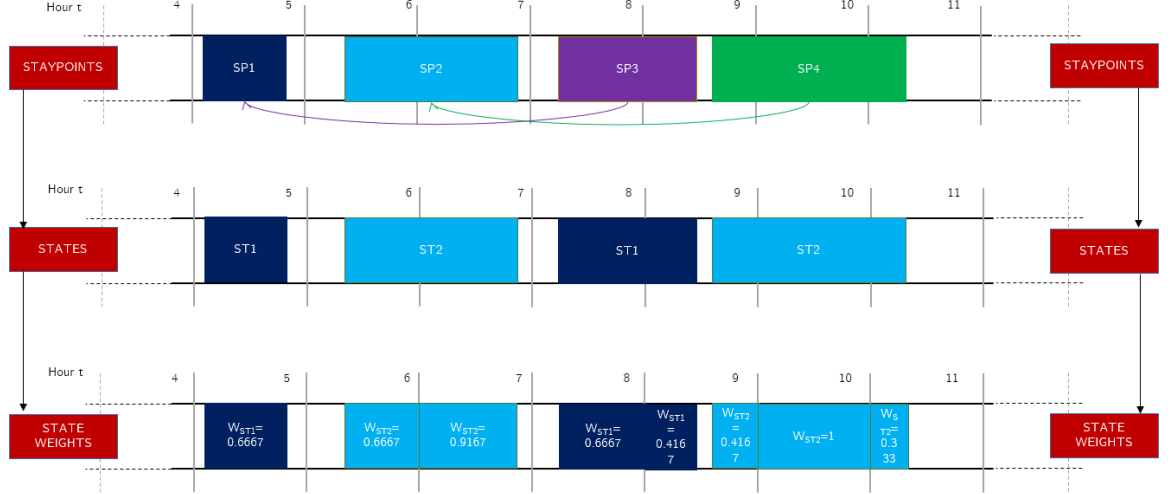


Figure 5: Algorithm Flow-chart

- **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** 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**
- 3:      $sp \leftarrow lst\_hr\_pts_i, lst\_hr\_pts_{i+1}$
- 4:   **else if**  $distanceBtw(lst\_hr\_pts_i, cluster) \leq th\_d$  **then**
- 5:      $cluster \leftarrow lst\_hr\_pts_i$
- 6:     Update Cluster Mean
- 7:   **else if** ( $cluster \neq empty$ ) And  $duration(cluster) \geq th\_t$  **then**
- 8:      $sp \leftarrow cluster$
- 9:   **end if**
- 10: **end for**

---



---

**Algorithm 2** 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:    $d \leftarrow distanceBtw(sp_i, sp_{i+1})$
- 3:    $t \leftarrow timeDiff(sp_i, sp_{i+1})$
- 4:    $s \leftarrow d/t$
- 5:   **if**  $s \neq 0$  **then**
- 6:      $delta\_t \leftarrow \min(d, th\_d)/s$
- 7:   **else**
- 8:      $delta\_t \leftarrow t/2$
- 9:   **end if**
- 10:   Update end\_time  $d_e + delta\_t/2$  for  $sp_i$
- 11:   Update start\_time  $d_s - delta\_t/2$  for  $sp_{i+1}$
- 12: **end for**

---

---

**Algorithm 3** formStates() : Form states from stay-points

---

**Input:** Stay-point  $sp_{n+1}$  , States  $st$   
**Output:** States  $st$

```

1: for each  $st_j$  in  $st$  do
2:    $is\_sp_{n+1\_added} \leftarrow False$ 
3:   if  $distanceBtw(sp_{n+1}, st_j) \leq th\_d$  then
4:     Add  $sp_{n+1}$  to  $st$ 
5:      $newMean \leftarrow mean(st)$ 
6:     if  $distanceBtw(All\ sp's\ of\ st_j, newMean) \leq th\_d$  then
7:       Add  $sp_{n+1}$  to  $st_j$ 
8:        $is\_sp_{n+1\_added} \leftarrow True$ 
9:       Break
10:    else
11:      Remove  $sp_{n+1}$  from  $st$ 
12:    end if
13:  end if
14: end for
15: if  $is\_sp_{n+1\_added} = False$  then
16:   Add new state in  $st$ 
17: end if

```

---



---

**Algorithm 4** markovModel() : Create the Markov Model with transition probabilities)

---

**Input:** State Weights  $w$   
**Output:** Markov Chain Model  $mc$

```

1: for each  $h - hour$  from 0to23 do
2:   if  $h! = 23$  then
3:      $trn\_mat_{h+1} \leftarrow w^{h^{[T]}} * w^{h+1}$ 
4:      $mc_{h+1} \leftarrow eachCell(trn\_mat_{h+1})/rowSum(trn\_mat_{h+1})$ 
5:   else
6:      $trn\_mat_0 \leftarrow w^{23^{[T]}} * w^0$ 
7:      $mc_0 \leftarrow eachCell(trn\_mat_0)/rowSum(trn\_mat_0)$ 
8:   end if
9: end for

```

---