# **Discussion on 17.07.2018**

1. To calculate the quality of prediction:

## Example:

Current hour = 5

Next\_hour = 6

#### Step 1:

Calculate all stay points with IDs and their corresponding probabilities till next\_hour.  $g(I_1) = \{(id1,0.6), (id2,0.2), (id3,0.1), (id4,0.1)\}$ 

## Step 2:

Now predict based on training data for all these ids.

 $Id1_{t} = \{(id1,0.07), (id2,0.2), (id3,0.7), (id4,0.03)...)\}$ 

 $Id2_{t} = \{(id1,0.8), (id2,0.03), (id3,0.01), (id4,0.01)...\}$ 

 $Id3_t = \{(id1,0.7), (id2,0.01), (id3,0.2), (id4,0.01)...\}$ 

 $Id4_{+} = \{(id1,0.1), (id2,0.3), (id3,0.05), (id4,0.006)...)\}$ 

Weight all the ID predictions with their own probability.

 $Id1_{t} = \{(id1,0.07), (id2,0.2), (id3,0.7), (id4,0.03)...\} * 0.6$ 

 $Id2_{t} = \{(id1,0.8), (id2,0.03), (id3,0.01), (id4,0.01)...\} * 0.2$ 

 $Id3_t = \{(id1,0.7), (id2,0.01), (id3,0.2), (id4,0.01)...\} * 0.1$ 

 $Id4_{t} = \{(id1,0.1), (id2,0.3), (id3,0.05), (id4,0.006)...\} * 0.1$ 

And sum them up to get a single vector. Arrange these predictions in descending order and remove cumulative probabilities after a threshold  $\theta$  (example 0.8).

 $g(t_{t+1})' = \{(id1, 0.282), (id2,0.157), (id3,0.447), (id4,0216) ..\}$  This is the overall prediction.

## Step 3:

Calculate the stay points with IDs and their corresponding time of stay for next hour.  $g(I_{t+1}) = \{(id1_{t+1}, 0.9), (id2_{t+1}, 0.1)\}$ 

#### Step 4:

Now there are two cases:

A. User visits one of the predicted places

B. Calculate similarity for our predictions versus the ground truth

For Case A: It is enough if  $g(I_{t+1})$  has any location in  $g(I_{t+1})$ '.

For case B:

both the vectors.

Dot product of  $g(I_{t+1})$  and  $g(I_{t+1})$ ' together will all the missing locations in

Similarity =  $a(dot product) / b(magnitude of g(I_{t+1}))$ 

2. Try clustering on raw points to see if we have anything new. Each new cluster should be treated separately in each iteration.