

fuzzyMM: Map Matching using a Fuzzy Logic
Based Map Matching Algorithm

Nikolai Gorte

March 12, 2015

1 Introduction

Map matching is the process of matching inaccurate GPS trajectories to a digital road network. `fuzzyMM` is a package that implements a fuzzy logic based map matching algorithm to solve this task. The algorithm consists of three major parts responsible for the identification of the links the vehicle is travelling on:

- Initial MapMatching Process (IMP)
- Subsequent MapMatching Process along a link (SMP-1)
- Subsequent MapMatching Process at a junction (SMP-2)

Each of these processes uses a Fuzzy Inference System (FIS) for the link identification.

A detailed description of the fuzzy logic map matching algorithm and the FIS can be found in Qudus (2006).

2 Usage

2.1 Map Matching

The GPS trajectory must be a `SpatialPointsDataFrame` as defined in `sp` or one of the `Track` classes defined in `trajectories`. It must include following data to be processed by the algorithm:

- GPS.Speed in km/h
- GPS.Bearing
- GPS.HDOP
- time as "POSIXct" or "POSIXlt"

Additionally a projection must be specified.

```
> names(track)

 [1] "id"                "time"              "GPS.Speed"
 [4] "O2.Lambda.Current" "Speed"            "GPS.HDOP"
 [7] "Short.Term.Fuel.Trim.1" "Rpm"              "Engine.Load"
[10] "CO2"               "Consumption"      "Intake.Pressure"
[13] "Throttle.Position" "GPS.Bearing"      "Intake.Temperature"
[16] "GPS.VDOP"          "Calculated.MAF"   "GPS.PDOP"
[19] "O2.Lambda.Current.ER" "GPS.Accuracy"     "GPS.Altitude"
[22] "Long.Term.Fuel.Trim.1"

> proj4string(track)

[1] "+proj=longlat +datum=WGS84 +no_defs +ellps=WGS84 +towgs84=0,0,0"
```

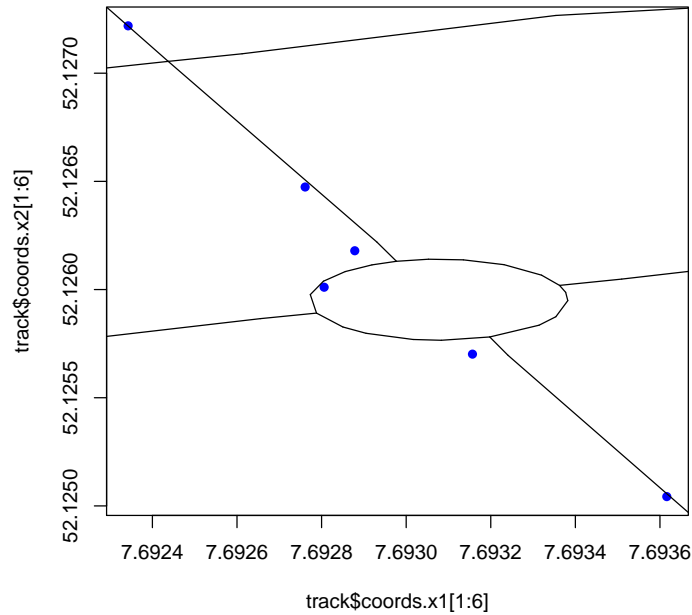


Figure 1: Inaccurate GPS track

As you can see in Figure 1 some of the points of our GPS track are not located on the roads.

Applying the `mm` function uses the fuzzy logic map matching algorithm to reconcile the GPS track with the digital road network.

```
> matched_track <- mm(track)
```

The result is an object of the same class as the input containing the the new map matched positions as can be seen in Figure 2.

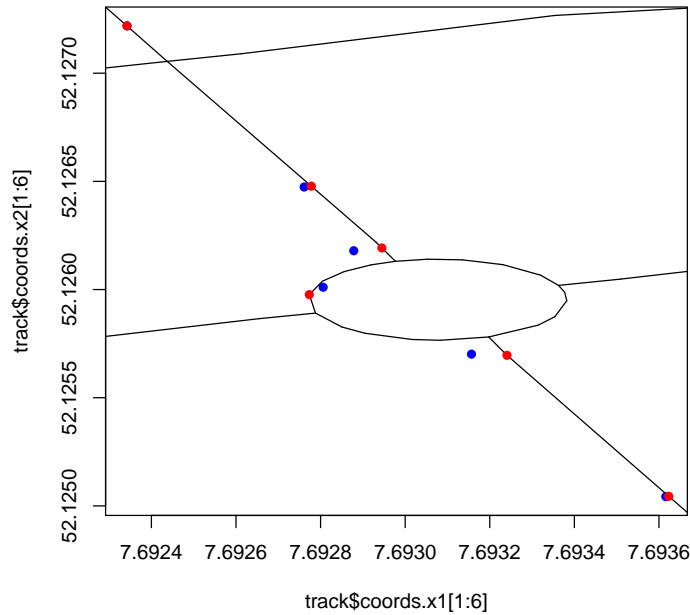


Figure 2: Result after Map Matching

2.2 Adjusting Membership Functions

It is possible to adjust the membership functions, used in the FIS, to your needs by changing the range of the fuzzy subsets of the input variables. Membership functions are used to fuzzify the input variables of the FIS, which means assigning them to values between 0 and 1. It is recommended to read Quddus (2006) to understand how input variables, fuzzy subsets and membership functions work together.

To see the current range of the fuzzy subsets use

```
> get_var_bounds()
```

	left_bounds	right_bounds	ID
speed_high	3	6	1
speed_low	2	4	2
speed_zero	0	2	3
HE_small	20	45	4
HE_large	25	60	5
PD_short	10	40	6
PD_long	20	50	7

HDOP_good	3	5	8
HDOP_bad	4	6	9
alpha_low	85	100	10
alpha_high	90	120	11
beta_low	85	100	12
beta_high	90	120	13
delta_dist_neg	-5	5	14
delta_dist_pos	-5	10	15
HI_small	10	20	16
HI_large	15	30	17
HI_180	150	200	18
connectivity_direct	0	1	19
connectivity_indirect	0	1	20
dist_err_small	5	15	21
dist_err_large	10	25	22

These bounds represent the x values at which the sigmoidal membership functions reach ≈ 0 or ≈ 1 respectively (e.g. "speed is high" ranges from 3 to 6 m/s).

You can change the bounds using `set_var_bounds(name, bounds)`.

```
> set_var_bounds("speed_high", c(4, 7))
```

When everything is set to your needs, update the membership functions and reinitialize the FIS.

```
> update_mf()
```

The parameters of the membership functions and all additional information regarding the FIS can be acquired by getting the FIS of IMP, SMP1 or SMP2.

```
> fis_imp <- get_fis("IMP")
> str(fis_imp)
```

List of 16

```
$ num.labels      : num [1, 1:4] 3 2 2 2
$ varout.mf       : NULL
$ rule            : chr [1:6, 1:17] "IF" "IF" "IF" "IF" ...
$ varinp.mf       : num [1:5, 1:9] 6 3.06 5.5 NA NA ...
..- attr(*, "dimnames")=List of 2
.. ..$ : NULL
.. ..$ : chr [1:9] "high" "low" "zero" "small" ...
$ range.data.ori  : num [1:2, 1:5] 0 50 0 360 0 60 0 20 0 100
$ type.model      : chr "TSK"
$ type.tnorm      : chr "MIN"
$ type.implication.func: chr "MIN"
$ type.mf         : chr "SIGMOID"
$ type.defuz      : NULL
$ type.snorm      : chr "MAX"
```

```

$ func.tsk          : num [1:6, 1] 50 10 50 10 100 10
$ method.type      : chr "MANUAL"
$ name             : chr "Sim-1"
$ colnames.var     : chr [1:5] "v" "HE" "PD" "HDOP" ...
$ class            :function (x)
- attr(*, "class")= chr "frbs"

```

```
> fis_imp$varinp.mf
```

	high	low	zero	small	large	short	long
[1,]	6.000000	6.000000	6.000000	6.00000000	6.00000000	6.00000000	6.00000000
[2,]	3.063413	-4.59512	-4.59512	-0.3676096	0.2625783	-0.3063413	0.3063413
[3,]	5.500000	3.000000	1.000000	32.50000000	42.50000000	25.00000000	35.00000000
[4,]	NA	NA	NA	NA	NA	NA	NA
[5,]	NA	NA	NA	NA	NA	NA	NA

	good	bad
[1,]	6.000000	6.000000
[2,]	-4.59512	4.59512
[3,]	4.000000	5.000000
[4,]	NA	NA
[5,]	NA	NA

It is also possible to plot the membership functions.

```
> plotMF(fis_imp)
```

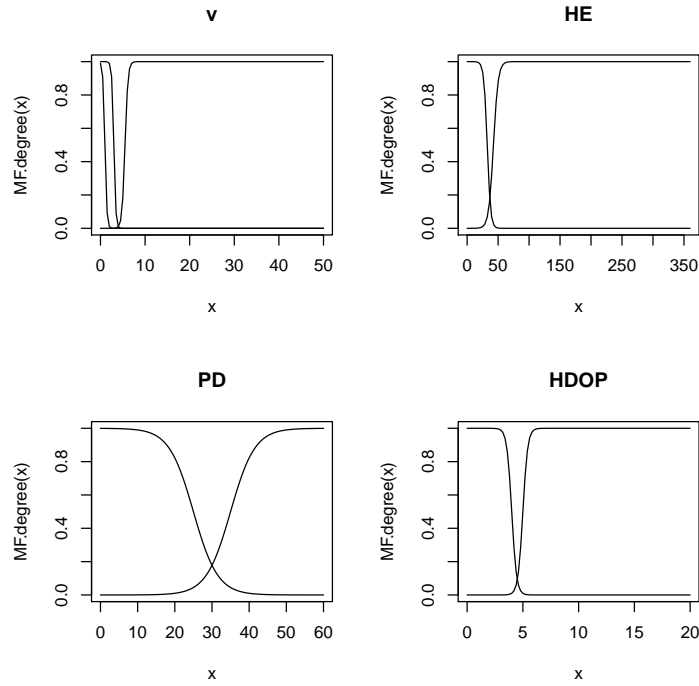


Figure 3: Membership Functions

References

Quddus, Mohammed A. 2006 (January). *High Integrity Map Matching Algorithms for Advanced Transport Telematics Applications*. Ph.D. thesis, Imperial College London, United Kingdom.