

SABD Project

Urban Mobility Analysis

B. Agard & I. Joly

octobre 2020

Table of Content

SABD Project Description	2
Statement	2
Database description - mobility survey	3
Files list:	3
Topics and Data	3
List of proposed topics	4
1. Daily travel times	4
2. Mode of transport choice	4
3. Mobility behavior	4
4. Household motorization?	4
5. Engine Types	4
6. Frequence of use of urban public transport	5
7. Trips during rush hours	5
8. Traveled distance	5
Data Operations	6
Part One - Preprocessing	6
Database loading and exploration	6
Description of the database content	7
Variables Description and labels	7
Data Base Structure	10
New Variables Definition	11
Immobil person	11
Area number redefinition	11
Average travelled distances	11
Some variables definitions	12
Trip DataFrame: <i>Déplacements</i>	13
Person DataFrame <i>Individus</i>	14
Household DataFrame <i>Ménages</i>	15
Save of the 3 DataFrames	15
Part Two - Data processing	16
GIS Part	16
Import of shp with compatible coordinates	18
The shapefile encounters errors (holes) that are correcte with the tools <code>cleangeo()</code>	19

Map and Calculus of area indicators: CARTE NBD BTT	20
Merging with newESRI and data	21
Colored mapping	21
Map with labels	22
References	23
Appendix	24
Complete map of daily travel times	24

SABD Project Description

Statement

- This project is a **3 or 4** students group work.
- All groups work on the same database, with same expectations, but different angles and problematic.
- Precisely, each group has a specific **scientific question** (from the topics list at the end of this paper - topic assignment to group is randomly determined)
- Each topic points:
 1. A variable of interest Y
 2. A Set explicit variables to be **systematically included** in the analysis X and Z
 3. A topic's specific focus to be integrated in the analysis - variables describing this focus will have to be *identified / chosen* and **justified** by the students (new variable W)
- Each topic will be analyzed based on the following approach:
 1. **Present and describe** the dataset you are using (without giving priority to one variable over another): remarkable characteristics, properties of the different variables, underlying structures, creation of variables, factorization, etc.
 2. **Summary statistics**: Produce and draw conclusions from **Univariate and bivariate** study of each of the variables:
 - *variable to be explained*
 - *imposed variables* and their contribution to the explanation of the variable of interest
 - *relevant variable to be chosen for topic's specific focus*
 3. **Assumptions and Strategy Proposition**: Based on your *knowledge* of the topic, the **literature** (at least one relevant and well chosen paper is expected), and the former **summary statistics**:
 - *Define* and **justify** your strategy of analysis of the topic: what are the expected factors of the variable of interest?; what are your methodological choices (types of analysis, method, model,...)? what is your strategy (subsetting, algorithm, data handling,...)?
 4. The **specific focus** influence on the variable of interest will be examined. For that you will construct a new variable W that will synthesize a subset of other variables. You are asked to :
 - Select all pertinent variables related to the specific focus
 - Prepare all considered variables in order to create groups of users with similar specific focus situation
 - Cluster your dataset using those variables, clusters are stored in new variable W
 - Analyze your clusters relatively to the specific focus variables
 - Eventually iteratively improve your results reconsidering previous steps
 5. **Prediction** of variable of interest Y considering variables X , Z and W .
 - Select (with justifications) different prediction models
 - Make predictions
 - Evaluate the performance of each model.
 - Compare the different approaches

6. **Explanation** the relationships between the variable of interest and the explanatory variables you have identified
- *Discuss* the statistical significance (or equivalent notion) of the link between variables)
 - *Interpret* the link between variables in terms of mobility behaviors and analysis
 - *Conclude* on the assumptions you have drawn in the **point 3**.

You will produce a **10 pages maximum report** (eventually with few appendices) dealing with the following points (you will discuss them relatively to your topic):

For each point, specify if applicable if you worked on a subset of observations and / or if you deleted or created variables.

In addition, do not hesitate to collaborate with your comrades working on other variables, the issues being linked and the results of some can also be exploited by others.

The body of the report should be a synthesis of your analyzes, where you present the elements that you think are the most important. Concrete elements (values, tables, figures) should support your assertions, but do not give all the details of your analyzes in the body of the report (the appendices are made for this). For example, if you want to study the correlations of variables, indicate only the most significant in the report, and refer to the appendices for the complete matrix of correlations.

The report must be produced in Rmarkdown (or KnitR) format with all the elements allowing to **reproduce** your analysis, and returned by email (iragael.joly@grenoble-inp.fr or ijoly.research@gmail.com) on **31/01/2020** at 11:59 p.m. last deadline.

It is **imperative to ensure the reproducibility** of your work, both in the handling of the data and in the production of the report. So be sure to tell the user which packages to install, which intermediate files to run, etc.

Database description - mobility survey

Files list:

- Databases `allgreD.RData`, `allgreI.RData`, `allgreM.RData`
- Variables Dictionary `DICO_VAR_EMD_GRE2010.xls`
- File `EnonceTP_Eco3_2019.rmd` of this statement. This Rmd file creates files to be used in the project. It is advised to execute this file to initiate data handling and operations.
- GIS files: They are grouped in the folder: `FOND_MAP`
- GIS code are described: `Decoupage_simplifie.xlsx`

Topics and Data

Data are an extraction from mobility survey in Grenoble in 2010. Those data are available in R format: `allgreD.RData`, `allgreI.RData`, `allgreM.RData`

The data describe the trips made by members of Grenoble households in 2010 (Cerema (2013)). They provide data relating to individual and household characteristics (household structure, motorization, sex, age, professional status, etc.) and mobility practices (numbers of trips, travel times, locations of trips, modes used, reasons for trips, etc.).

The structure of the database is described in the following sections.

List of proposed topics

1. Daily travel times

Analyze the **daily travel times** (variable BTT).¹ Tips: The literature has used the DTT to define *extreme commuters*: people traveling more than 2 hours per day.

- Variable of interest Y : **btt** - Travel Time Budget
- Variable X : **jourdepl** - day of the trip
- Variable Z : **permis** - driving licence holder
- Specific focus W : influence of **housing** on your decision variable Y

2. Mode of transport choice

Analyze the 'car' choice of the mode of travel of individuals and identify the determinants of this choice among the individual or household characteristics and modes of transport.

- Variable of interest Y : **Voiture** (variable to be created according to `mode_depl_ag = 'VP'` or not)
- Variable X : **permis** - driving licence holder
- Variable Z : **age** - age
- Specific focus W : influence of **family structure** on your decision variable Y

3. Mobility behavior

Propose a characterization of the population of mobile and immobile people and identify the determinants of mobility

- Variable of interest Y : **immobil** (variable to be created according to `nbd = 0` or `nbd > 0` - `nbd` is the number of trips)
- Variable X : **age** - age
- Variable Z : **VP_DISP0** - number of available cars in the household
- Specific focus W : influence of **car ownership** on your decision variable Y

4. Household motorization?

What are the determinants of household motorization?

- Variable of interest Y : **motorisation** (variable to be created according to `VP_DISP0 = 0` or `VP_DISP0 > 0` - `VP_DISP0` is the number of available cars in household)
- Variable X : **taillemng** - household size
- Variable Z : **Permis_mng**: at least 1 driving licence in the household
- Specific focus W : influence of **sociodemographic household** on your decision variable Y

5. Engine Types

Study the Grenoble car fleet. What are the determinants of choosing a diesel engine? The question can focus on the 1st vehicle of the household.

- Variable of interest Y : **diesel** (variable to be created according to `ENERGIE1 = 3` or `ENERGIE1 <> 3` - `ENERGIE1` is the type of engine of the 1st car of the household)
- Variable X : **TYPE_HAB** - type of housing
- Variable Z : **taillemng** - household size
- Specific focus W : influence of **occupation** on your decision variable Y

¹French for TTB: Travel Time Budget is the sum of duration of all trips performed in a day by the individual

6. Frequency of use of urban public transport

Study the factors associated with frequent (daily) use of urban public transport (**frequ**tcu)

- Variable of interest Y : **frequ**tcu (variable to be created according to **frequ**tcu = 1 or not)
- Variable X : **permis** - driving licence holder
- Variable Z : **taillemng** - household size
- Specific focus W : influence of **car ownership** on your decision variable Y

7. Trips during rush hours

Study the factors associated with time of a trip during the rush hours (daily) use of urban public transport (**heured**ep and **mindep**)

- Variable of interest Y : **heured**ep and **mindep**
- Variable X : **motifor** - trip purpose at origin
- Variable Z : **D12** - traveled distance (distance as the crow flies)
- Specific focus W : influence of **household mobility equipment** on your decision variable Y

8. Traveled distance

Study the factors associated with traveled distance

- Variable of interest Y : **D13**
- Variable X : **motifor** - trip purpose at origin
- Variable Z : **TYPE_HAB** - type of housing
- Specific focus W : influence of **mobility accessibility and speed** on your decision variable Y

Data Operations

Data operations are in two parts.

1. First part describes data preprocessing (off the project) that were performed to create the dataframe. It shows you the operation, the scripts applied before the exercise. *You can not run these instructions*
2. Second part describes processing (on the project) that are to be performed to manage the data for the project. *You have to run these operations* and eventually modify them.

Part One - Preprocessing

Database loading and exploration

```
library(tidyverse)
```

Data loading This section describes former data manipulation that lead to the actual database.

The data are extract of half of the real household in the survey

```
# echo=TRUE, eval=FALSE
# Loading of the original database
load("allgre_TR.RData")
# Extracting half of the household for confidentiality issue
set.seed(123)
allgre$PB <- round( ave( sample( x= c(0,1), prob= c(0.5, 0.5), size =
  length(allgre$id_men),replace=T) , allgre$id_men, FUN= mean) )
table(allgre$PB)
allgre.PB <- allgre[allgre$PB ==1,]
# Saving
# Exclusion of non interesting variables
allgre.PB_V2 <- allgre.PB[, -c( 5 ,10 ,47 ,49 ,51 ,64 ,81 ,97 ,98 ,101 ,124:136, 139:148, 150, 151)]
save(allgre.PB_V2, file = "allgre.PB_V2.RData")
```

Here we load the actual database

```
# echo=TRUE, eval=FALSE
# Loading of the resulting DF
load("allgre.PB_V2.RData")
table(allgre.PB_V2$NBD)

##
##      0      1      2      3      4      5      6      7      8      9     10     11     12     13     14     15
## 1277    91 4023 1967 7037 3649 4063 2386 2063 1413  901  618  412  251  197   75
##      16     17     18     22
##    146     53     36     44

# Check of NO_DEPL - number of trips
table(allgre.PB_V2$NO_DEPL)

##
##      0      1      2      3      4      5      6      7      8      9     10     11     12     13     14     15
## 1277 7025 6795 4745 4152 2392 1665  990  648  394  237  144   91   55   36   21
##      16     17     18     19     20     21     22
##      16      7      4      2      2      2      2

# Check of id_depl - id of trips
table(is.na(allgre.PB_V2$id_depl))
```

```
##
## FALSE TRUE
## 29425 1277

# Creation of new id_depl
allgre.PB_V2$id_depl <- allgre.PB_V2$id_pers * 10 + allgre.PB_V2$NO_DEPL
table(is.na(allgre.PB_V2$id_depl))

##
## FALSE
## 30702

table(allgre.PB_V2$nbd)

##
##      0      1      2      3      4      5      6      7      8      9     10     11     12     13     14     15
## 1277  91 4023 1967 7037 3649 4063 2386 2063 1413  901  618  412  251  197   75
##    16    17    18    22
##   146    53    36    44
```

Description of the database content

```
Ncol <- 7;
v <- Ncol* (Nlin <- ceiling(length(names(allgre.PB_V2))/Ncol )) - length(names(allgre.PB_V2))
mat <- matrix(c(names(allgre.PB_V2), rep("",v)), ncol = Ncol,
              dimnames = list(1:Nlin,c("col1", "col2", "col3", "col4", "col5", "col6", "col7")))
# content
knitr::kable( mat, digits = 2, caption = "Table: Variable names in *allgre.PB*")
```

Table 1: Table: Variable names in *allgre.PB*

col1	col2	col3	col4	col5	col6	col7
tir	PUIS_VP1	TYPE_STAT3	permis	frequcu	minarr	durstat
NO_MEN	POSSES1	GENRE4	etabscol	frequtram	duree	autoroute
NO_PERS	LIEU_STAT1	ENERGIE4	OCCU1	frequurb	nbmodemec	abonpeage
NO_DEPL	TYPE_STAT1	AN_VP4	OCCU2	frequtransisere	prisecharge	id_men
zoneres.x	GENRE2	PUIS_VP4	csp	frequer	D12	id_pers
jourdepl	ENERGIE2	POSSES4	ABO_TC	situveil	D13	id_depl
TYPE_HAB	AN_VP2	LIEU_STAT4	VAL_ABO	zoneres.x.1	zoneres.y.1	id_traj
TYPE_OCU	PUIS_VP2	TYPE_STAT4	travdom	motifor	NO_TRAJ	nb_pers
Gare2	POSSES2	NB_velo	zonetrav	motoracc	TPS_MAP_DEP	nbd
Gare5	LIEU_STAT2	NB_2Rm	dispovp	zoneorig	mode	ntraj
telefon	TYPE_STAT2	COEF_MNG	PBM_STAT	heuredep	ZONE_D_TRAJ	btt
annuaire	GENRE3	zoneres.y	STAT_TRAV	mindep	ZONE_A_TRAJ	mode_V2
internet	ENERGIE3	sexe	fqvelo	motifdes	TPS_MAP_ARV	Couteff
VP_DISPO	AN_VP3	lien	FQ2R1	motdeacc	NUM_VEH	mode_depl_ag
GENRE1	PUIS_VP3	age	FQ2R2	nbarret	NB_OCCU	
ENERGIE1	POSSES3	TEL_PORT	fqvpcond	zonedest	LIEU_STAT	
AN_VP1	LIEU_STAT3	mail	fqvppass	heurearr	NAT_STAT	

Variables Description and labels

See variables dictionary (file: **DICO_VAR_EMD_GRE2010.xls**)

tir: Drawing number of the observation (often close to the residence zone number)
 NO_MEN: Household number
 NO_PERS: Number of the person in the household
 NO_DEPL: Number of the person's trip
 zonerex.x: Number of the area of residence (see correspondence file)
 jourdepl: Day of the move
 TYPE_HAB: Type of residence of the household
 TYPE_OCU: Type of occupation of the person
 Gare2: Department number of the reference sncf station
 Gare5: Postal code of the reference sncf station
 telefon: Availability of a telephone
 annuaire: Present in the telephone directory
 internet: Availability of an internet connection
 VP_DISPO: Number of private car available in the household
 GENRE1: Type of car for the 1st car of the household
 ENERGIE1: Fuel type of the 1st car in the household
 AN_VP1: Year of entry into service of the 1st household car
 PUIS_VP1: Power of the 1st car in the household
 POSSES1: Type of ownership of the 1st car in the household
 LIEU_STAT1: Parking place of the 1st car of the household
 TYPE_STAT1: Type of parking for the 1st car of the household
 GENRE2, ENERGIE2, AN_VP2, PUIS_VP2, POSSES2, LIEU_STAT2, TYPE_STAT2, GENRE3,
 ENERGIE3, AN_VP3, PUIS_VP3, POSSES3, LIEU_STAT3, TYPE_STAT3, GENRE4, ENERGIE4,
 AN_VP4, PUIS_VP4, POSSES4, LIEU_STAT4, TYPE_STAT4: idem for cars n ° 2, 3 and 4 of the
 household
 NB_velo: Number of bikes in the household
 NB_2Rm: Number of 2 motorized wheels in the household
 COEF_MNG: adjustment coefficient associated with the household
 zonerex.y: Residence zone number
 sexe: Gender
 lien: Link with the household reference person
 age: Age
 TEL_PORT: Possession of a cell phone
 mail: Possession of an email address
 permis: Possession of driving license
 etabscol: Last school attended
 OCCU1: Main occupation
 OCCU2: Other occupation
 csp: Socio-professional category (PCS)
 ABO_TC: Possession of a Public Transport subscription in general
 VAL_ABO: Validity of the TC subscription yesterday
 travdom: Work or home study
 zonetrav: Zone of the place of work or study (Main occupation)
 dispovp: Arrangement of a car in general (Travel home-work or study)
 PBM_STAT: Parking problems in general (at work or study)
 STAT_TRAV: Parking difficulties at work or study place
 fqvelo: Frequency of bicycle use
 FQ2R1: Frequency of use 2 wheels with motors (type 1)
 FQ2R2: Frequency of use 2 wheels with motors (type 2)
 fqvpcond: Frequency of use Car with driver
 fqvppass: Frequency of use Car as a passenger
 freqtcu: Frequency of use Urban network tag
 freqtram: Tramway use frequency only
 frequrb: Frequency of use Other urban network

freqtransisere: Frequency of use of the Cars Transisère network
 freqter: TER use frequency
 situveil: Situation of the person the day before
 zonerex.x.1: Number of the area of residence
 motifor: Purpose at the origin
 motoracc: Purpose for the origin of the accompanied person
 zoneorig: Number of the area origin of the displacement
 heuredep: Start time (hour)
 mindep: Start minute
 motifdes: Purpose at destination
 motdeacc: Purpose at destination of the accompanied person
 nbarret: Number of stops in the trip
 zonedest: Destination area number
 heurearr: Arrival time (hour)
 minarr: Arrival minute
 duree: trip duration (declared)
 nbmodemec: Number of mechanized mode of transport used in the trip prisecharge: Transport cost are covered
 D12: traveled distance (distance as the crow flies)
 D13: traveled distance
 zonerex.y.1:
 NO_TRAJ:
 TPS_MAP_DEP: Walk time at origin
 mode: Transport mode
 ZONE_D_TRAJ: Area at origin of a stop
 ZONE_A_TRAJ: Area at destination of a stop
 TPS_MAP_ARV: Walk time at destination
 NUM_VEH: Number of the vehicle
 NB_OCCU: Number of occupant
 LIEU_STAT: Parking place
 NAT_STAT: Type of parking place
 durstat: Parking duration
 autoroute: Use of the highway
 abonpeage: Public transport pass holder
 id_men: Household id
 id_pers: Person id
 id_depl: Trips id
 id_traj: Stop id
 nb_pers: Number of persons
 nbd: Number of trips
 ntraj: Number of stops
 btt: Daily travel time
 mode_V2: transport mode version 2
 Couteff: Transport cost estimation
 mode_depl_ag: Transport mode aggregated version

Data Base Structure

This is a so-called ‘hierarchical’ database. It initially consists of 4 files:

1. File *Household*
2. File *Person*
3. File *Trip*
4. File *Route*

In surveys, trips are made up of a series of trips or stages made for a certain reason. For a trip, there may be several routes described, when the trip is made in several stages (for example: a bicycle trip from home to the station, a train trip, then a bicycle trip from the station instead of job). This is represented in the database provided by several lines for the same trip.

On the same hierarchical principle: an individual can make several trips during the day, and a household can be made up of several individuals.

The database provided is the ‘finest,’ since it describes all the household, individual and travel information for each trip. There are therefore as many lines as there are trips.

The structure is managed by the identifying numbers of households, individuals, trips and route:

- **id_men**: Unique Household identifier
- **id_pers**: Unique identifier Person
- **id_depl**: Unique identifier Displacement
- **id_traj**: Unique identifier of the route

The work requested in this project may require work, for example on an individual basis instead of trips. For this, it will be necessary to reduce the database (reduce the number of lines) to keep only information at the individual level (one line per individual).

The following script elements reduce the database to trip, person and household level.

They also give some guidance on how to aggregate the information available at a lower level.

New Variables Definition

Immobil person

```
allgre.PB_V2$UN <- as.numeric(allgre.PB_V2$nbrd !=0)
table(as.numeric(allgre.PB_V2$UN))
```

```
##
##      0      1
## 1277 29425
```

Area number redefinition

```
allgre.PB_V2$DEST <- round(allgre.PB_V2$zonedest/1000)
head(allgre.PB_V2$DEST)
```

```
## [1] 101 101 103 101 102 101
```

```
allgre.PB_V2$ORIG <- round(allgre.PB_V2$zoneorig/1000)
head(allgre.PB_V2$ORIG)
```

```
## [1] 101 101 101 103 101 102
```

```
table(allgre.PB_V2$ORIG)
```

```
##
## 101 102 103 104 105 106 107 108 109 110 111 112 113 114 115 116
## 813 593 486 314 289 580 172 242 257 377 181 446 433 195 380 357
## 117 118 119 120 121 122 123 124 125 126 127 128 129 130 131 132
## 192 243 560 731 139 767 511 120 249 270 334 568 221 241 499 385
## 133 134 135 136 137 138 139 140 141 142 143 201 202 203 204 205
## 172 350 481 144 256 139 306 241 180 160 222 377 249 333 231 255
## 301 302 303 304 305 306 307 308 309 310 311 312 313 314 401 402
## 189 273 253 297 403 293 231 166 205 273 183 200 314 250 238 212
## 403 501 502 503 504 505 506 507 508 509 510 511 512 513 514 515
## 312 1122 159 103 342 333 247 229 205 203 141 194 257 343 137 345
## 516 517 518 519 601 602 603 701 801 802 803 804 805 806 901 902
## 206 243 205 195 220 400 159 261 205 138 409 184 210 180 200 336
## 903 990
## 263 518
```

Average travelled distances

```
mean(allgre.PB_V2$D12, na.rm = T); mean(allgre.PB_V2$D13, na.rm = T)
```

```
## [1] 4629.925
```

```
## [1] 6.73701
```

```
summary(allgre.PB_V2$D12)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.     NA's
##         0      715    2078    4630    5402   71356    2243
```

```
# nombre de distance 'NA' par zone
```

```
table(allgre.PB_V2$zoneres.y[is.na(allgre.PB_V2$D12)==1])
```

```
##
```

```
## 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20
## 663 616 349 100 123 58 62 70 16 11 17 25 23 7 17 8 18 1 9 3
## 21 22 23 24 25 26 29 30 51 52
## 14 7 3 5 10 1 1 1 3 2
```

```
tapply(allgre.PB_V2$D12, as.factor(allgre.PB_V2$zoneres.y), mean, na.rm = TRUE)
```

```
## 1 2 3 4 5 6 7 8
## 4331.151 4383.808 4828.676 4884.900 4875.023 6120.861 5658.047 4439.104
## 9 10 11 12 13 14 15 16
## 8618.982 3655.540 7179.919 5220.047 7438.054 6879.554 10610.741 7656.462
## 17 18 19 20 21 22 23 24
## 4245.554 8160.481 3257.056 3302.265 5103.297 4204.639 1848.350 2472.000
## 25 26 27 29 30 51 52 59
## 4543.872 1817.467 5359.750 NaN 6926.115 5405.306 9644.750 2307.200
```

Some variables definitions

```
## GRENOBLE 2010 :
#####
table(allgre.PB_V2$OCCU1 )
```

```
##
## 1 2 3 4 5 6 7 8 9
## 11671 3168 238 1482 5404 1055 5402 1292 530
```

```
allgre.PB_V2$statut2 <- factor(allgre.PB_V2$OCCU1 ,
                              labels=c("TRAVAIL A TEMPS PLEIN", "TRAVAIL A TEMPS PARTIEL", "FORMATION STAGE", "E
table(allgre.PB_V2$statut2)
```

```
##
## TRAVAIL A TEMPS PLEIN TRAVAIL A TEMPS PARTIEL FORMATION STAGE
## 11671 3168 238
## ETUDIANT SCOLAIRE CHOMEUR
## 1482 5404 1055
## RETRAITE RESTE AU FOYER AUTRE
## 5402 1292 530
```

```
#####
levels(allgre.PB_V2$statut2) <- c("ACTIF", "ACTIF", "ACTIF", "ETUDIANT", "SCOLAIRE", "CHOMEUR", "RETRAIT
table(allgre.PB_V2$statut2)
```

```
##
## ACTIF ETUDIANT SCOLAIRE CHOMEUR RETRAITE
## 15077 1482 5404 1055 5402
## RESTE AU FOYER AUTRE
## 1292 530
```

```
#####
allgre.PB_V2$cspgroup <- trunc(allgre.PB_V2$csp / 10)
allgre.PB_V2$cspgroup[allgre.PB_V2$pcs %in% c(81,82)] <- 7
allgre.PB_V2$cspgroup[is.na(allgre.PB_V2$cspgroup)==T] <- 9
table(allgre.PB_V2$cspgroup)
```

```
##
## 1 2 3 4 5 6 8 9
## 399 1312 5161 5408 5930 3850 7936 706
```

Trip DataFrame: *Déplacements*

```
# BASE DEPLACEMENTS
## exemple de calcul : nombre de trajets par déplacement
allgre.PB_V2$nbt2 <- ave( allgre.PB_V2$NO_TRAJ, allgre.PB_V2$id_depl, FUN = max)
addmargins( table(allgre.PB_V2$nbt2) )
```

```
##
##      0      1      2      3      4      Sum
## 6743 20332 1691  628    12 29406
```

```
## Réduction au déplacement : ATTENTION On ejecte les immobiles
allgreD <- distinct(allgre.PB_V2, id_depl, .keep_all = T)
# ou avec subset(allgre.PB_V2, allgre.PB_V2$NO_TRAJ == allgre.PB_V2$ntraj )
# nb trajet
addmargins( table(allgreD$nbt2) )
```

```
##
##      0      1      2      3      4      Sum
## 6727 20093  840   208      3 27871
```

```
# Par mode
table(allgreD$mode_depl_ag)
```

```
##
## Autre  MAP  TCIU  TCU   VP
## 2038 6792  332 1791 16937
```

```
# durée
addmargins( table(allgre.PB_V2$duree))
```

```
##
##      1      2      3      4      5      6      7      8      9     10     11     12     13
## 225    708    562    105   6479    66    208    139    29   5744    14     70     57
## 14     15     16     17     18     19     20     21     22     23     24     25     26
## 24   4464     12     21     26     5   2439     2     30     31     14    982     7
## 27     28     29     30     32     33     34     35     36     37     38     39     40
## 12     14      6   2641      2      6      2   456      2     21     22     10   613
## 41     42     43     44     45     46     47     48     50     52     53     55     56
##  2      10     11      2    860      1      9      7   252      2      5    135      2
## 57     58     59     60     62     63     65     66     67     68     69     70     73
##  8      2      1    725      5      2     76      3      5      7      3    113      2
## 74     75     77     78     80     82     83     85     90     95     97     98    100
##  2    198      2      6     66      1      3     27    209     14      1      1     26
## 105    110    112    113    115    120    122    125    127    128    130    135    140
## 54     12      2      2     14     62      2      6      1      1      4      9     13
## 145    150    155    157    158    165    170    180    190    195    200    205    210
##  2      31      2      1      1      6      1     22      1      4      2      2     18
## 220    225    230    240    245    250    255    270    280    292    293    300    325
##  8      4      4     10      3      1      7     12      2      1      1      2      2
## 330    340    345    369    405    480    630    750    Sum
##  3      2      2      3      1      1      1      4 29425
```

```
table(is.na(allgreD$duree))
```

```
##
## FALSE TRUE
## 27890 1258

table(cut(allgreD$duree, breaks=seq(0, max(allgreD$duree, na.rm = T), 30)), useNA = "always")

##
## (0,30] (30,60] (60,90] (90,120] (120,150] (150,180] (180,210] (210,240]
## 24544 2529 510 144 55 30 23 19
## (240,270] (270,300] (300,330] (330,360] (360,390] (390,420] (420,450] (450,480]
## 15 5 4 4 1 1 0 1
## (480,510] (510,540] (540,570] (570,600] (600,630] (630,660] (660,690] (690,720]
## 0 0 0 0 1 0 0 0
## (720,750] <NA>
## 4 1258

# nb déplacements
table(allgre.PB_V2$nbd)

##
## 0 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15
## 1277 91 4023 1967 7037 3649 4063 2386 2063 1413 901 618 412 251 197 75
## 16 17 18 22
## 146 53 36 44

Person DataFrame Individus

#####
# BASE INDIVIDUS
## exemple de calcul : durée quotidienne des déplacements d'un individu
allgreD$btt2 <- ave( allgreD$duree, allgreD$id_pers, FUN = sum)
#addmargins( table(allgreD$btt2) )
## Reduction
allgreI <- distinct(allgreD, id_pers, .keep_all = T)
# subset(allgreD, allgreD$NO_DEPL == allgreD$nbd )
#addmargins( table(allgreI$btt2) )

summary(allgreI$btt); summary(allgreI$btt2)

## Min. 1st Qu. Median Mean 3rd Qu. Max. NA's
## 1.00 40.00 65.00 91.13 110.00 1110.00 1258

## Min. 1st Qu. Median Mean 3rd Qu. Max. NA's
## 1.00 38.00 65.00 80.38 105.00 870.00 1258

addmargins( table(allgreI$nbd) )

##
## 0 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15
## 1258 72 1766 616 1672 700 662 333 253 156 89 56 33 19 14 5
## 16 17 18 22 Sum
## 9 3 2 2 7720

table(is.na(allgreI$nbd))

##
## FALSE
## 7720
```

Remark: to calculate the daily travel time (BTT), elements `btt` and `btt2` differ due to preprocessing operations to calculate `btt` (correction of error in the declared time of trip)

Household DataFrame *Ménages*

```
#####
#   BASE MENAGE
## exemple de calcul : taille du ménage et permis-ménage
allgreI$taillemng <- ave(allgreI$NO_PERS, allgreI$id_men, FUN = max)
allgreI$Permis_mng <- ave(allgreI$permis, allgreI$id_men, FUN = max)
addmargins( table(allgreI$taillemng) )

##
##      1      2      3      4      5      6      7      8 Sum
## 887 2388 1295 1931  959  190   54   16 7720

## Reduction
allgreM <- distinct(allgreI, id_men, .keep_all = T)
      #subset(allgreI, allgreI$NO_PERS == allgreI$taillemng )
addmargins( table(allgreM$taillemng) )

##
##      1      2      3      4      5      6      7      8 Sum
## 887 1194  433  487  193   32    8    2 3236

addmargins( table(allgreM$Permis_mng) )

##
##      1      2      3 Sum
## 1730 1026  111 2867

table(allgreI$freqtcu)

##
##      1      2      3      4      5
## 625  545  620 1530 3869

table(allgreI$nbdb)

##
##      0      1      2      3      4      5      6      7      8      9      10      11      12      13      14      15
## 1258    72 1766  616 1672   700  662  333  253  156   89   56   33   19   14    5
##      16      17      18      22
##       9       3       2       2

table(allgreI$freqter)

##
##      1      2      3      4
## 6462  706   98    3
```

Save of the 3 DataFrames

```
allgreD <- allgreD[, c( c(1:4), c(75:105), 106:109, 112, 114:116, 117:119)]
allgreI <- allgreI[, c( c(1:3), c(46:74) , 106, 107, 111, 113, 117, 120, 121 )]
allgreM <- allgreM[,c( c(1,2), c(5:45), 106:109, 110, 124,125 )]
save(allgreD, file = "allgreD.RData")
```

```
save(allgreI, file = "allgreI.RData")
save(allgreM, file = "allgreM.RData")
```

Part Two - Data processing

You should start [here](#) You can download these data files and load them:

```
load("allgreD.RData")
load("allgreI.RData")
load("allgreM.RData")
```

GIS Part

```
# chargement des packages nécessaires
library(igraph)
```

```
## Warning: package 'igraph' was built under R version 3.6.3
##
## Attaching package: 'igraph'
##
## The following objects are masked from 'package:dplyr':
##
##   as_data_frame, groups, union
##
## The following objects are masked from 'package:purrr':
##
##   compose, simplify
##
## The following object is masked from 'package:tidyr':
##
##   crossing
##
## The following object is masked from 'package:tibble':
##
##   as_data_frame
##
## The following objects are masked from 'package:stats':
##
##   decompose, spectrum
##
## The following object is masked from 'package:base':
##
##   union
```

```
library(dismo)
```

```
## Loading required package: raster
## Warning: package 'raster' was built under R version 3.6.3
## Loading required package: sp
## Warning: package 'sp' was built under R version 3.6.3
##
## Attaching package: 'raster'
```



```

## The following object is masked from 'package:dplyr':
##
##   select
## The following object is masked from 'package:tidyr':
##
##   extract
library(maptools)

## Warning: package 'maptools' was built under R version 3.6.3
## Checking rgeos availability: TRUE
library(reshape2)      # transformation format long, format large

## Warning: package 'reshape2' was built under R version 3.6.3
##
## Attaching package: 'reshape2'
## The following object is masked from 'package:tidyr':
##
##   smiths
library(sp)             # objets spatiaux
library(rgdal)          # fonctions de la bibliothèque GDAL

## Warning: package 'rgdal' was built under R version 3.6.3
## rgdal: version: 1.5-12, (SVN revision 1018)
## Geospatial Data Abstraction Library extensions to R successfully loaded
## Loaded GDAL runtime: GDAL 3.0.4, released 2020/01/28
## Path to GDAL shared files: C:/Users/joly.GAEL/Documents/R/win-library/3.6/rgdal/gdal
## GDAL binary built with GEOS: TRUE
## Loaded PROJ runtime: Rel. 6.3.1, February 10th, 2020, [PJ_VERSION: 631]
## Path to PROJ shared files: C:/Users/joly.GAEL/Documents/R/win-library/3.6/rgdal/proj
## Linking to sp version:1.4-2
## To mute warnings of possible GDAL/OSR exportToProj4() degradation,
## use options("rgdal_show_exportToProj4_warnings"="none") before loading rgdal.
library(ggplot2)        # fonctions graphiques
#library(ggthemes)      # thèmes pour ggplot
library(grid)           # fonction arrow
library(cartography)    # cartographie thématique

## Warning: package 'cartography' was built under R version 3.6.3
library(RColorBrewer)    # palettes de couleurs de C. Brewer
library(dplyr)
library(RgoogleMaps)

## Warning: package 'RgoogleMaps' was built under R version 3.6.3
library(PBSmapping)

## Warning: package 'PBSmapping' was built under R version 3.6.3
##
## -----
## PBS Mapping 2.72.1 -- Copyright (C) 2003-2020 Fisheries and Oceans Canada

```

```
##
## PBS Mapping comes with ABSOLUTELY NO WARRANTY;
## for details see the file COPYING.
## This is free software, and you are welcome to redistribute
## it under certain conditions, as outlined in the above file.
##
## A complete user guide 'PBSmapping-UG.pdf' is located at
## C:/Users/joly.GAEL/Documents/R/win-library/3.6/PBSmapping/doc/PBSmapping-UG.pdf
##
## Packaged on 2019-03-14
## Pacific Biological Station, Nanaimo
##
## All available PBS packages can be found at
## https://github.com/pbs-software
##
## To see demos, type '.PBSfigs()'.
## -----
```

```
library(rgeos)
```

```
## Warning: package 'rgeos' was built under R version 3.6.3
## rgeos version: 0.5-3, (SVN revision 634)
## GEOS runtime version: 3.8.0-CAPI-1.13.1
## Linking to sp version: 1.4-2
## Polygon checking: TRUE
```

```
library(cleangeo)
```

```
## Warning: package 'cleangeo' was built under R version 3.6.3
```

```
library(devtools)
```

```
## Warning: package 'devtools' was built under R version 3.6.3
## Loading required package: usethis
## Warning: package 'usethis' was built under R version 3.6.3
```

```
library(lwgeom)
```

```
## Warning: package 'lwgeom' was built under R version 3.6.3
## Linking to liblwgeom 3.0.0beta1 r16016, GEOS 3.8.0, PROJ 6.3.1
```

```
Sectir2010<-readOGR("FOND_MAP",layer="Sectir2010")
plot(Sectir2010)
# Affichage system de coordonnees
pathToShp <- "FOND_MAP"
Sectir2010@proj4string
ogrInfo(dsn = pathToShp,layer="Sectir2010")
```

Import of shp with compatible coordinates

```
newESRI <- readOGR(dsn = "G:\\MyDATA\\TEst_Reunion_Distance\\FOND_MAP", layer = "NewESRI")
```

```
## OGR data source with driver: ESRI Shapefile
## Source: "G:\\MyDATA\\TEst_Reunion_Distance\\FOND_MAP", layer: "NewESRI"
## with 97 features
```

```
## It has 5 fields
## Integer64 fields read as strings:  cod_sectir
#ACADIA$data <- DF_Zone
plot(newESRI)
```



```
#plot(ACADIA)
```

The shapefile encounters errors (holes) that are correcte with the tools `cleangeo()`

This cleaning is needed *before* merging, which will create many polygons

```
#####
# https://gis.stackexchange.com/questions/113964/fixing-orphaned-holes-in-r
require(maptools)
# mysp <- readShapePoly("C:\\Users\\ijoly.INRA\\Desktop\\TEst_Reunion_Distance\\FOND_MAP\\Sectir2010.shp")
plot(newESRI, border= "lightgray")
```



```
require(devtools)
## to detect potential geometry issues
# devtools::install_github("eblondel/cleangeo")
require(cleangeo)
# exploration of sh
report <- clgeo_CollectionReport(newESRI)
clgeo_SummaryReport(report)
# cleaning
mysp.clean <- clgeo_Clean(newESRI)

## Warning in proj4string(sp): CRS object has comment, which is lost in output

# check
report.clean <- clgeo_CollectionReport(mysp.clean)
clgeo_SummaryReport(report.clean)
# geometry validity
require(rgeos)
sapply(slot(mysp.clean, "polygons"), function(x){
  gIsValid(SpatialPolygons(Srl = list(x)))
})
```

Map and Calculus of area indicators: CARTE NBD BTT

```
DF <- allgreI[, c("nbd", "btt", "tir", "zoneres.y", "id_pers")]
# Calcul base individus des indicateurs individuels
DF_Indiv <- na.omit(DF)
```

```

addmargins(table(allgreI$zoneres.y))
DF_Indiv$BTT_moy <- ave(DF_Indiv$btt,DF_Indiv$tir, FUN = mean)
DF_Indiv$BTT_med <- ave(na.omit(DF_Indiv$btt),DF_Indiv$tir, FUN = median)
DF_Indiv$nbd_med <- ave(DF_Indiv$nbd,DF_Indiv$tir, FUN = median)
DF_Indiv$nbd_moy <- ave(DF_Indiv$nbd,DF_Indiv$tir, FUN = mean)
DF_Indiv$Maxind <- ave(DF_Indiv$id_pers,DF_Indiv$tir, FUN = max)
# Réduction à la base Zone (avec le max(ind) par TIR)
DF_Zone <- subset(DF_Indiv, DF_Indiv$id_pers == DF_Indiv$Maxind, select =c(zoneres.y,BTT_med, BTT_moy,

```

Merging with newESRI and data

```

library(sp) # the trick is that this package must be loaded!
# JOINTURE avec option duplicateGEOM=T pour avoir autant de polygons que d'observations dans data
DF_Indiv$cod_sectir <- DF_Indiv$tir
newESRI_DF <- merge(mysp.clean, DF_Indiv, by = "cod_sectir", duplicateGeoms = TRUE)
PolyG <- SpatialPolygons(mysp.clean@polygons,proj4string=mysp.clean@proj4string)

```

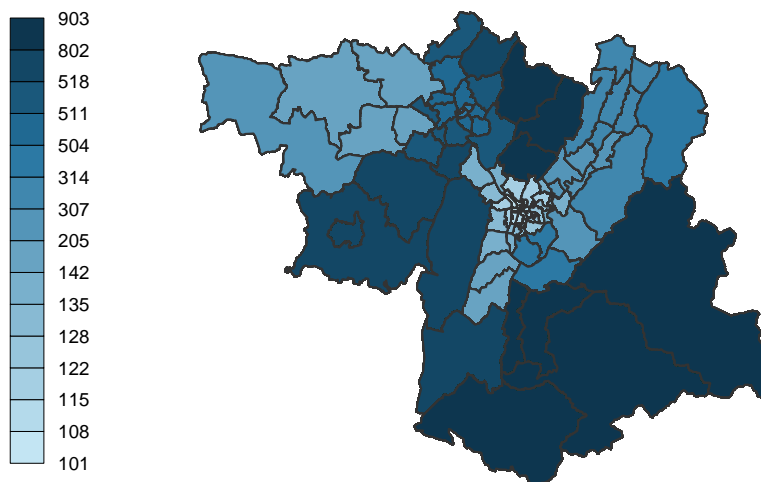
Colored mapping

```

choroLayer(spdf = newESRI_DF, df = newESRI_DF@data, var = "tir")
title("Mapping of the area number - not really useful")

```

Mapping of the area number – not really useful



Map with labels

It may need to install lwgeom

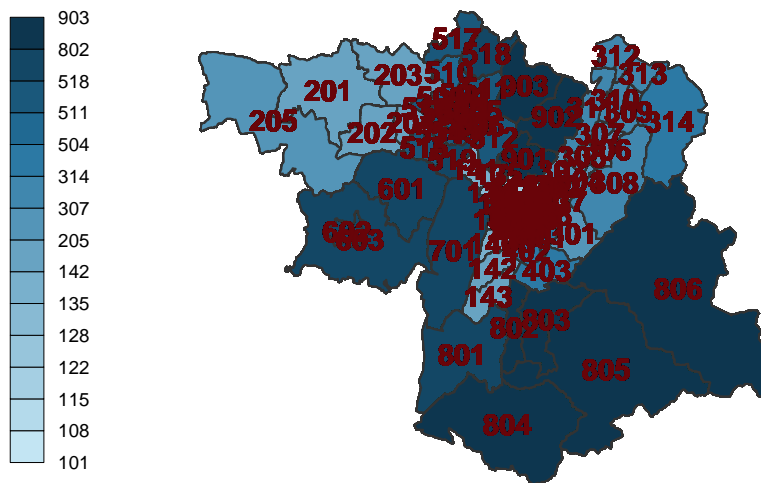
```
choroLayer(spdf = newESRI_DF, df = newESRI_DF@data, var = "tir")
# Label plot of the Mean
labelLayer(spdf = newESRI_DF, df = newESRI_DF@data, txt = "tir", col = "#690409", cex = 0.9, font = 2, >

## Warning in st_centroid.sfc(x = sf::st_geometry(x), of_largest_polygon =
## max(sf::st_is(sf::st_as_sf(x), : st_centroid does not give correct centroids for
## longitude/latitude data

## Warning: package 'sf' was built under R version 3.6.3

## Linking to GEOS 3.8.0, GDAL 3.0.4, PROJ 6.3.1

## Warning in text.default(x = cc[, 1], y = cc[, 2], labels = words, cex = cex, :
## "add" n'est pas un paramètre graphique
```



References

- Cerema. 2013. *Enquêtes ménages déplacements « Standard Certu »*. Cerema (ex-Certu). <https://books.google.fr/books?id=nhtpmAEACAAJ>.
- Chang, Winston. 2013. *R Graphics Cookbook*. O'Reilly Media, Inc.
- Harrell, F. E. 2013. *Regression Modeling Strategies: With Applications to Linear Models, Logistic Regression, and Survival Analysis*. Springer Series in Statistics. Springer New York. <https://books.google.fr/books?id=7D0mBQAAQBAJ>.
- Munzner, Tamara. 2015. *Visualization Analysis and Design*. CRC Press, Routledge.
- Xie, Yihui. 2015. *Dynamic Documents with R and Knitr*. 2nd ed. Boca Raton, Florida: Chapman; Hall/CRC. <https://yihui.name/knitr/>.
- . 2016. *Bookdown: Authoring Books and Technical Documents with R Markdown*. Boca Raton, Florida: Chapman; Hall/CRC. <https://github.com/rstudio/bookdown>.

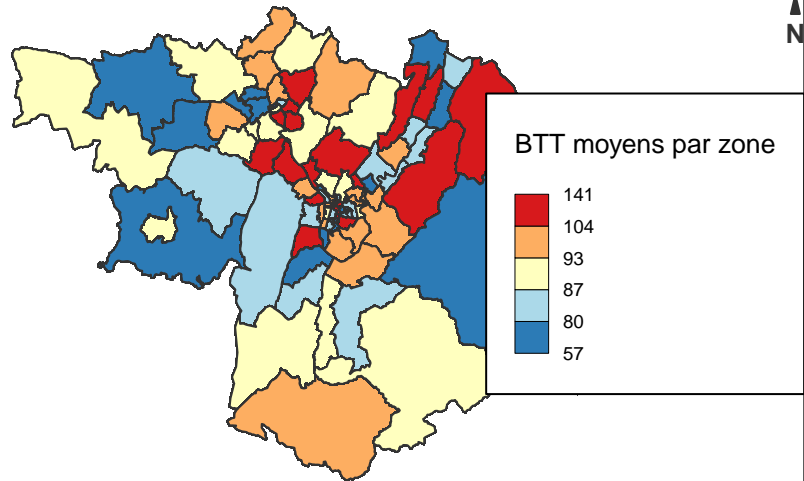
Appendix

Complete map of daily travel times

The script may need to be run out of MARKDOWN

```
Nclasse <- 5
##### Carte colorée
choroLayer(spdf = newESRI_DF,
           df = DF_Zone,
           spdfid = "cod_sectir",
           dfid = "tir",
           var = "BTT_moy",
           nclass = Nclasse,
           col = rev(brewer.pal(n = Nclasse, "RdYlBu")),
           legend.pos = "right", legend.frame = TRUE,
           legend.title.txt = "BTT moyens par zone")
##### Ajout de label : nbd par zone
# Label creation
DF_Zone$lab <- paste(round(DF_Zone$nbd_moy,1), sep = "")
# Label plot of the Mean
#labelLayer(spdf = newESRI_DF, df = DF_Zone, txt = "lab", col = "#690409", cex = 0.9, font = 2, add=T)
##### Ajout d'éléments de présentations : Layout plot
layoutLayer(title = "BTT moyens par zone",
            sources = "EMD Grenoble 2010",
            author = "I. Joly",
            scale = 0,
            north = T,
            frame = TRUE,
            col = "black",
            coltitle = "white")
```


BTT moyens par zone



EMD Grenoble 2010
I. Joly