

# bicincittaProblems

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## 0.1 Problems in the Bicincitta data set from 2013

There are problems with the Bicincitta data that we need to address before loading the data into a reliable and proper data-base. We will point out these problems using examples, and measure their magnitude using systematic analysis, and then speculate about the source of these problems.

### 0.1.1 Data

We will load the data from JSONs provided to us by Bicincitta at the end of April 2015.

The resulting data is in the form of lists of dictionaries. We can take a peek at the keys in each of the four data types, by creating a data frame and displaying the first few rows.

a subnetwork is described by,

```
id
name
```

a station is described by,

```
name
longitude
subnetwork_name
latitude
id
subnetwork_id
```

a user is described by

```
subnetwork_id
gender
expires
postal_code
subnetwork_name
address
id
```

a transaction is described by,

```
direction
user_id
```

```

station_id
event_time
id

```

The resulting dictionaries have ids that are UTF-8 strings. We can change these to integers to make our work easier,

There are keys in a transaction that do not seem to correspond to the data, but refer to the time at which the data was loaded into the JSON provided to us. We will drop these variables, and change the *event\_time* to a time object. Following that we will sort the transactions by the event time

We also sort the subnetworks and stations by their *id*.

Stations and users have been assigned a *subnetwork\_id* in the data. We can add the subnetwork name to these data,

### 0.1.2 Who are the users?

The simplest question may be the fraction of females vs males,

Of all the users 60 percent are female and 39 percent males.

It would be interesting if 60% of the users were in fact female. However, as we will see later there seems to be a problem of user duplicacy biased towards females.

### 0.1.3 Subnetworks for stations and users

Are the *subnetwork\_ids* for stations and users sensible? The subnetworks that the stations fall in are

Subnetworks that have stations assigned to them

	id	name
0	1	La Cote
1	2	Agglo Fribourg
2	3	Bulle
3	4	Les Lacs-Romont
4	6	Chablais
5	7	Valais Central
6	8	Yverdon-les-Bains
7	9	Lausanne-Morges
8	10	Campus
9	11	Riviera
10	12	Lugano-Paradiso

We see that stations cover only 11 of the 18 subnetworks. Looking at the subnetworks with no stations,

Subnetworks without any assigned stations

	id	name
0	5	Bâle
1	13	PubliBike
2	14	Vevey
3	15	Morges
4	16	Ouchy
5	17	Paradiso
6	18	Cern

we can see why there are no stations corresponding to these subnetworks. Basel, and Cern because Bicincitta have not given us data for these regions. Vevey, Morges, Ouchy, and Paradiso include stations subsumed by other subnetworks. These networks may be a remnant from previous versions of the data. This leaves **PubliBike** unexplained. As it turns out, there are users that have been assigned the subnetwork PubliBike (*id* 13). In fact we see later that the users who have registered transactions in the data have been assigned only PubliBike, and no other subnetwork.

Number of users from subnetwork PubliBike 58927

Users that have been assigned subnetwork PubliBike compose 70% of the total users in the data. However we are not going to see PubliBike in any of their transactions as there are no stations for the subnetwork PubliBike! What about other subnetworks without stations?

subnetworks assigned to users

```
Out[607]:
```

	id	name
0	2	Agglo Fribourg
1	6	Chablais
2	7	Valais Central
3	8	Yverdon-les-Bains
4	9	Lausanne-Morges
5	10	Campus
6	12	Lugano-Paradiso
7	13	PubliBike
8	14	Vevey
9	16	Ouchy
10	17	Paradiso
11	18	Cern

subnets without any users

```
Out[608]:
```

	id	name
0	1	La Cote
1	3	Bulle
2	4	Les Lacs-Romont
3	5	Bâle
4	11	Riviera
5	15	Morges

Comparing the subnets for users to subnets with stations we see that there are **only 7 subnets** for which we have stations as well as users,

subnets with stations as well as users

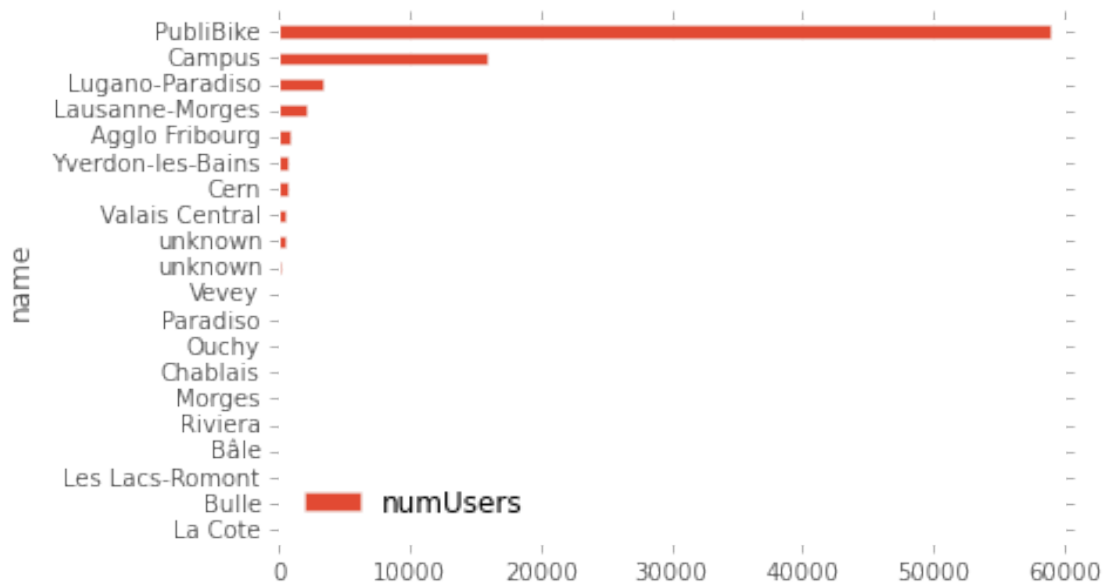
```
Out[609]:
```

	id	name
0	2	Agglo Fribourg
1	6	Chablais
2	7	Valais Central
3	8	Yverdon-les-Bains
4	9	Lausanne-Morges
5	10	Campus
6	12	Lugano-Paradiso

As a summary, let us tabulate the fraction of users in each of the subnets,

```
Out[611]:
```

	id	name	numUsers
0	1	La Cote	0
1	3	Bulle	0
2	4	Les Lacs-Romont	0
3	5	Bâle	0
4	11	Riviera	0
5	15	Morges	0
6	6	Chablais	3
7	16	Ouchy	3
8	17	Paradiso	3
9	14	Vevey	4
10	2011	unknown	152
11	2005	unknown	469
12	7	Valais Central	530
13	18	Cern	721
14	8	Yverdon-les-Bains	773
15	2	Agglo Fribourg	810
16	9	Lausanne-Morges	2183
17	12	Lugano-Paradiso	3425
18	10	Campus	15871
19	13	PubliBike	58927



So, most of the users are in PubliBike, which could create a problem as there are no stations associated to PubliBike. Lets first look at the transactions before we try to find a solution to this problem.

#### 0.1.4 Transaction users, stations, and subnetworks

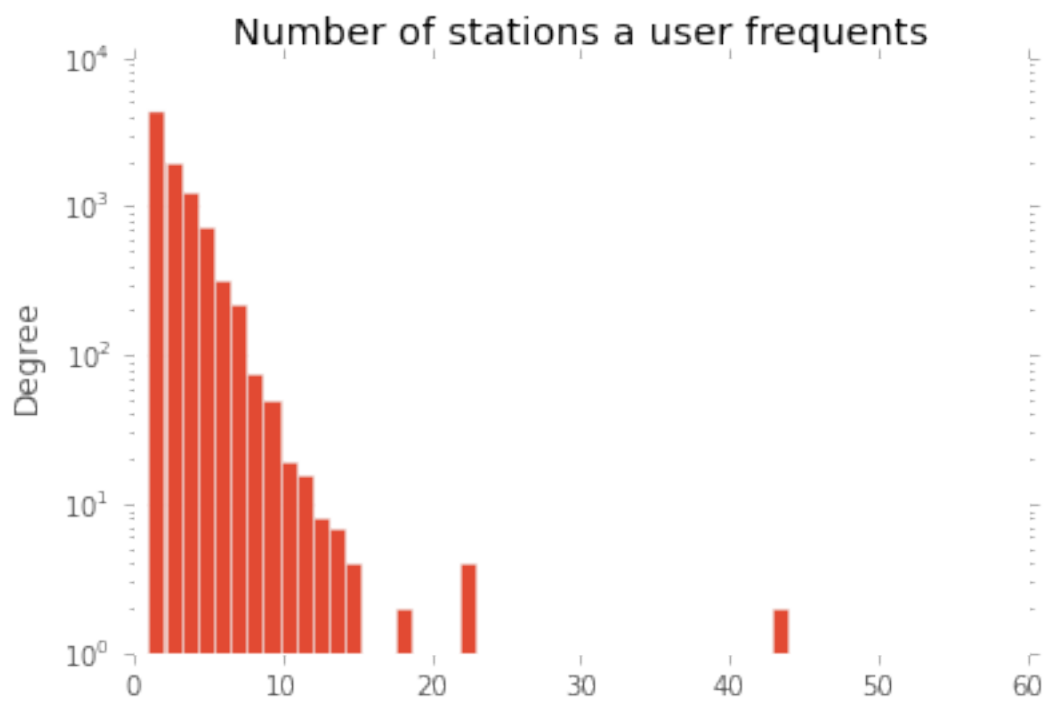
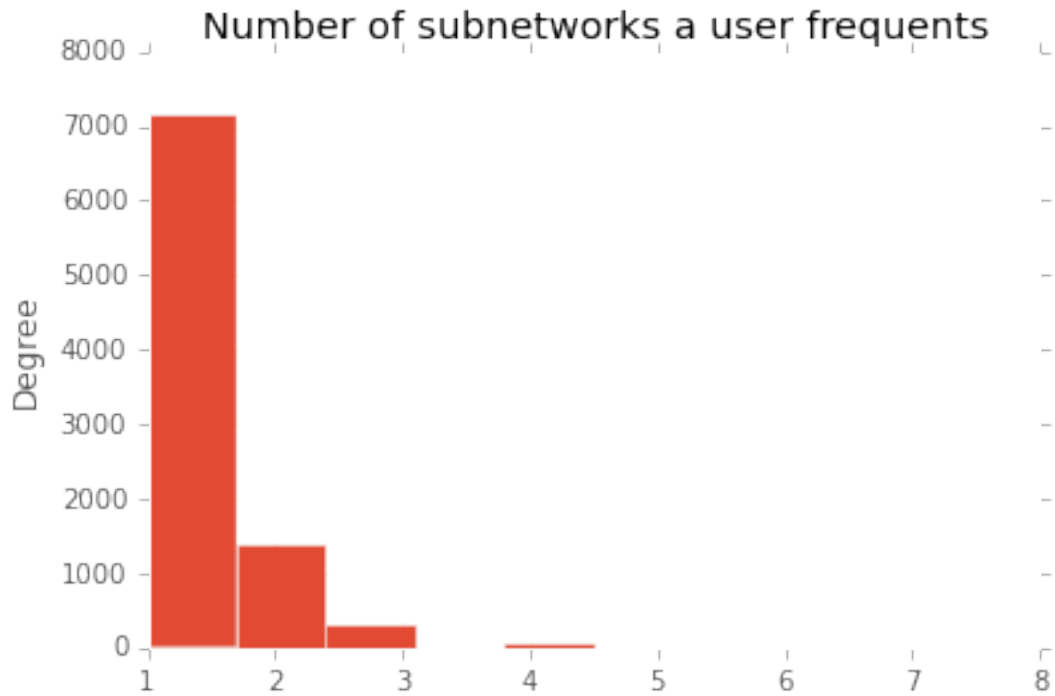
We begin by looking at how many users actually use the bike system ( have valid transactions)

fraction of users who have registered a transaction 0.10673152586

Only 10% users have registered a transaction! Are all the transactions of a user in the same subnetwork?

```
Out[800]:
```

	user_id	assigned_subnet	numStations	numSubnets	numTrxns
3462	108132	13	56	8	2180
4469	111523	13	46	7	2110
5399	84545	13	43	7	5134
3463	108133	13	43	6	3280
4673	112287	13	23	3	540

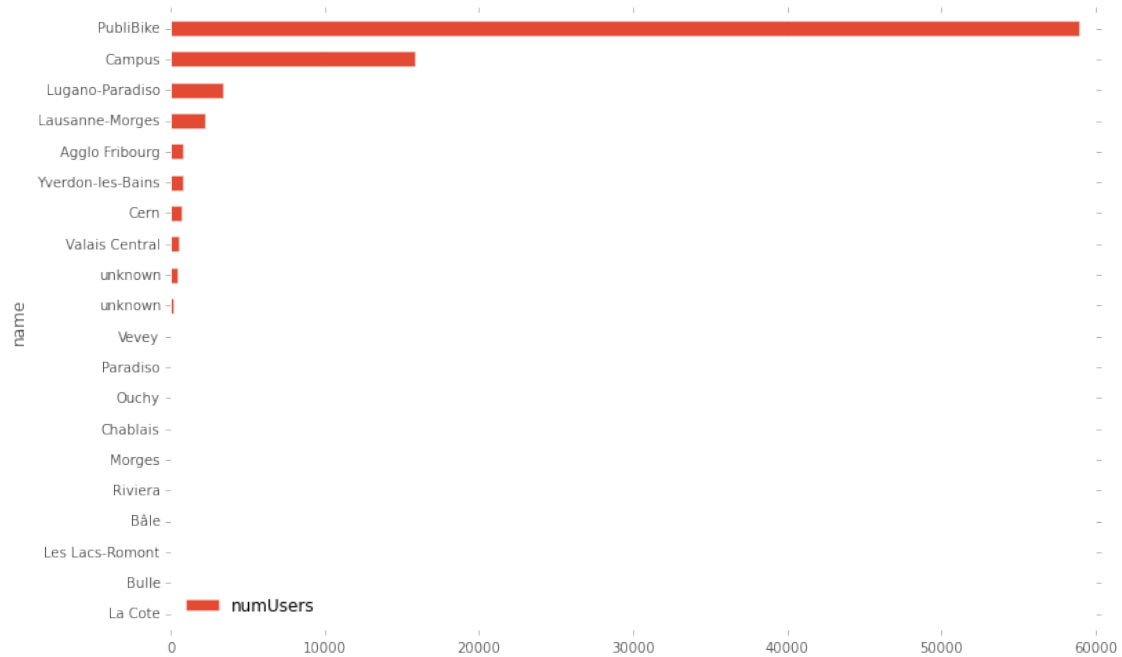


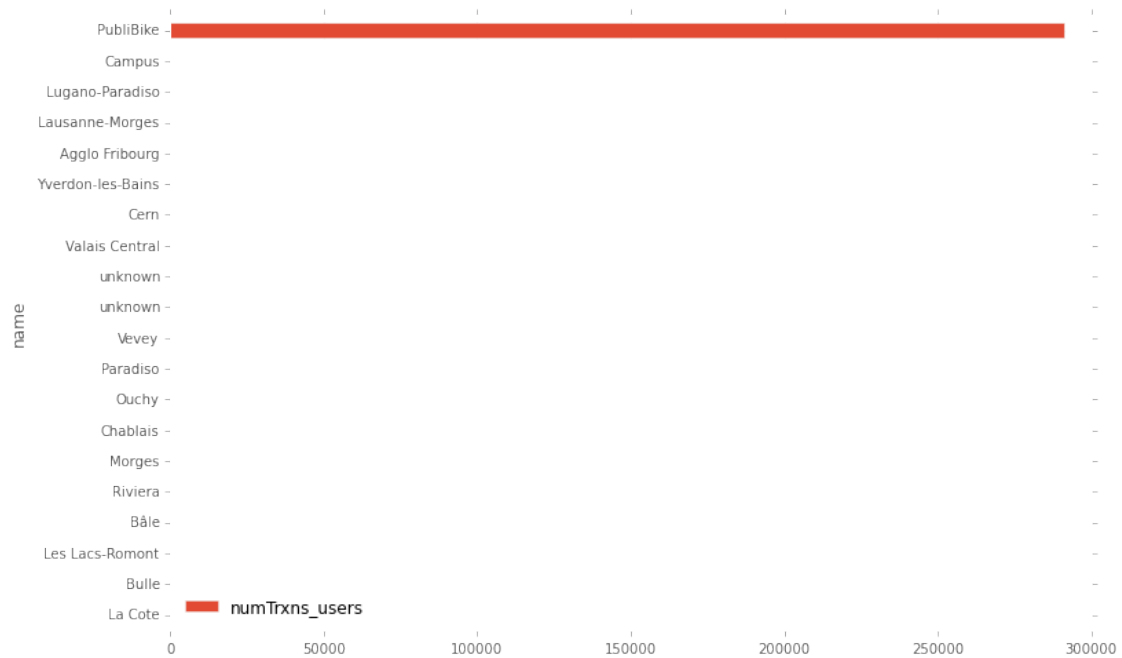
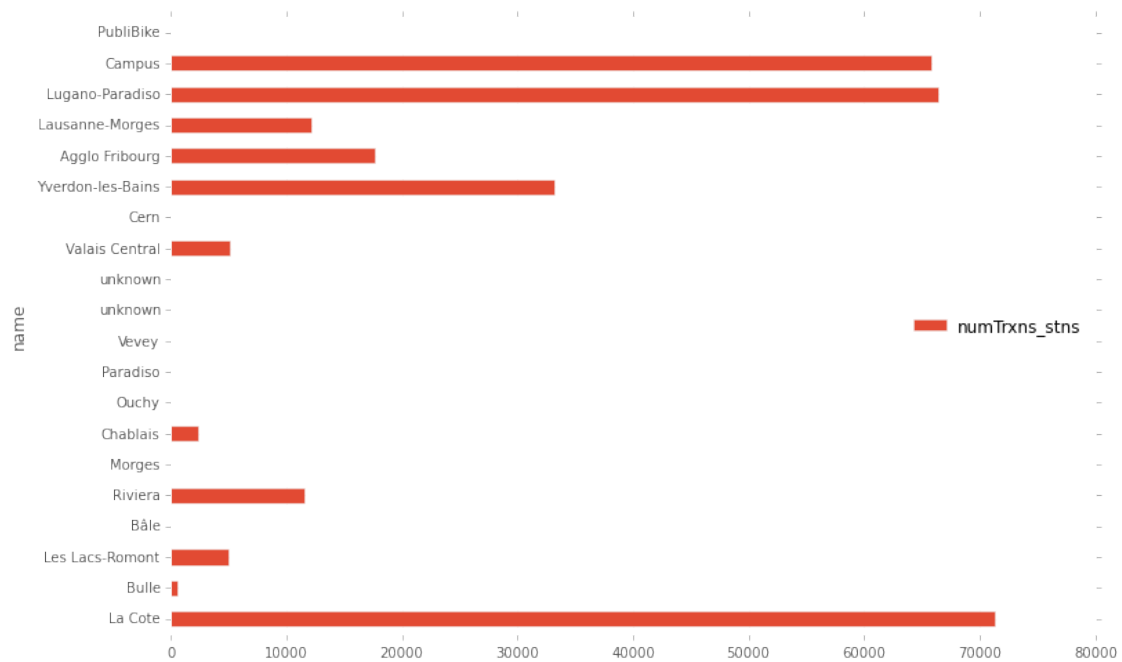
A simple question might guide us. How many transactions belong to Publibike users ( who have subnetwork\_id PubliBike) ?

The subnetworks of the users who make transactions make a set([13])

So, from the view point of the users, all the transactions are in the subnetwork PubliBike  
How many transactions in a subnetwork of the stations?

Now we can make a table for subnetworks, counting the number of trxn through user and station *subnetwork\_id*, in addition to the number of users.





```

Out[764]:
           id      name  numStations  numTrxns_stns  \
name
Vevey      14    Vevey             0             0
Cern       18     Cern             0             0
unknown   2005  unknown             0             0
unknown   2011  unknown             0             0

```



Paradiso	17	Paradiso	0	0
Ouchy	16	Ouchy	0	0
PubliBike	13	PubliBike	0	0
Morges	15	Morges	0	0
Bâle	5	Bâle	0	0
Bulle	3	Bulle	2	566
Riviera	11	Riviera	5	11576
Valais Central	7	Valais Central	7	5077
Les Lacs-Romont	4	Les Lacs-Romont	9	4964
Yverdon-les-Bains	8	Yverdon-les-Bains	9	33227
Agglo Fribourg	2	Agglo Fribourg	10	17630
Chablais	6	Chablais	10	2377
Lausanne-Morges	9	Lausanne-Morges	11	12157
Lugano-Paradiso	12	Lugano-Paradiso	13	66415
La Cote	1	La Cote	13	71292
Campus	10	Campus	15	65853

	numTrxns_users	numUsers
name		
Vevey	0	4
Cern	0	721
unknown	0	469
unknown	0	152
Paradiso	0	3
Ouchy	0	3
PubliBike	291134	58927
Morges	0	0
Bâle	0	0
Bulle	0	0
Riviera	0	0
Valais Central	0	530
Les Lacs-Romont	0	0
Yverdon-les-Bains	0	773
Agglo Fribourg	0	810
Chablais	0	3
Lausanne-Morges	0	2183
Lugano-Paradiso	0	3425
La Cote	0	0
Campus	0	15871

### 0.1.5 User addresses

There are several problems associated with user addresses. We have already noticed, and fixed, that the provided addresses in the JSON have not been *unquoted* from their web encoding. Here we continue to explore other problems that may arise in the addresses.

We want to count the number of users at one address. Because the addresses have been provided as strings, we have to be able to aggregate all address strings that describe the same address. We have written a python function to do this task, which takes the address and postal-code strings to provide a combined string taking into account some empirical disambiguation criteria such as *Av*, *Ave*, for *Avenue*.

```
number of users with available address 21659
number of these addresses that are unique 15446
```

What fraction of unique addresses have multiple users?

0.230545124951

How many users at addresses with multiple users?

9774

which corresponds to a fraction of all users with available address,

0.451267371531

Multiple users at the same address could be actual multiple people, or multiple registrations by the same person, or a glitch in the data. We can consider as an example the address with the most multiplicity of 53,

address: via lambertenghi 1; 6900 , number of users: 53

We could say more about the multiple users at the same address if we look at their transactions. However as it turns out, we **do not have addresses for users who have registered transactions in the data**,

False

We can look at the subnetwork with addresses assigned to the *multi* users,

subnetworks with addresses assigned to multiple users set([2, 7, 8, 9, 10, 12, 18, 2005, 2011])

Some of the multi-user addresses have more than one subnetworks ( through the users at that address)

Subnetworks for users living at addresses with multiple registered users

There are as many as 37 users assigned to the same address that also have more than subnetwork assigned. Addresses with several users might represent problems of multiple subscription. For example, if we look at addresses with more than 10 users,

```
Out[806]:
```

	address	numFemales	numMales	numUsers	\
143	avenue des bains 9; 1007	37	0	37	
40	route cantonale 33; 1025	25	0	25	
105	avenue des bains 11; 1007	23	0	23	
48	place du tunnel 17; 1005	23	0	23	

	subnetworks
143	set([Lausanne-Morges, Campus])
40	set([Lausanne-Morges, Campus])
105	set([Lausanne-Morges, Campus])
48	set([Lausanne-Morges, Campus])

we see that the user is over-whelmingly females. However, a look at the lower end of such addresses seems alright,

```
Out[807]:
```

	address	numFemales	numMales	numUsers	\
3	poudrière 24; 1950	3	0	3	
61	avenue beaulieu 20; 1004	2	1	3	
23	avenue louis-ruchonnet 31; 1003	2	1	3	
104	eichenweg 12; 1718	2	1	3	
67	rue saint-rochemin 5; 1004	2	1	3	

	subnetworks
3	set([Campus, Valais Central])
61	set([Lausanne-Morges, Campus])
23	set([Lausanne-Morges, Campus])
104	set([Agglo Fribourg, Campus])
67	set([Lausanne-Morges, Campus])

These particular addresses appear sensible. There could be more than one person living at these addresses who have signed up with the bike system, albeit in different subnetworks. Or may be it is the same person with 2 different sign-ups in two different sub-networks. This raises the question: **How are users registered by the system? One individual = one signup? Or does a user need a sign-up for each subnetwork that she wants to use?** If it is the latter, then the provided *user\_ids* become less useful, because the same individual will appear as different users according to the *user\_ids*.

```
Out[808]:
```

	address	numFemales	numMales	numUsers
18	via lambertenghi 1; 6900	52	1	53
2288	chemin des falaises 3; 1005	52	0	52
1349	chemin des berges 12; 1022	41	0	41
2150	avenue des bains 9; 1007	37	0	37
332	via monte carmen 4; 6900	33	0	33
287	route cantonale 33; 1025	25	0	25
1444	via madonnetta 23; 6900	24	0	24
1649	place du tunnel 17; 1005	23	0	23
1826	avenue des bains 11; 1007	23	0	23
1997	rue de genève 76; 1004	22	0	22
1801	route cantonale 35; 1025	22	0	22
2534	via zurigo 1; 6900	20	1	21

```
Out[809]:
```

	address	numFemales	numMales	numUsers
0	bonne-espérance 28; 1006	1	0	1
1	37 route cantonnale; 1025	1	0	1
2	avenue de la dôle 4; 1005	1	0	1
3	abbesses 21; 2012	1	0	1
4	chemin de ponfilet 100; 1093	0	1	1