# ATAC buses analysis

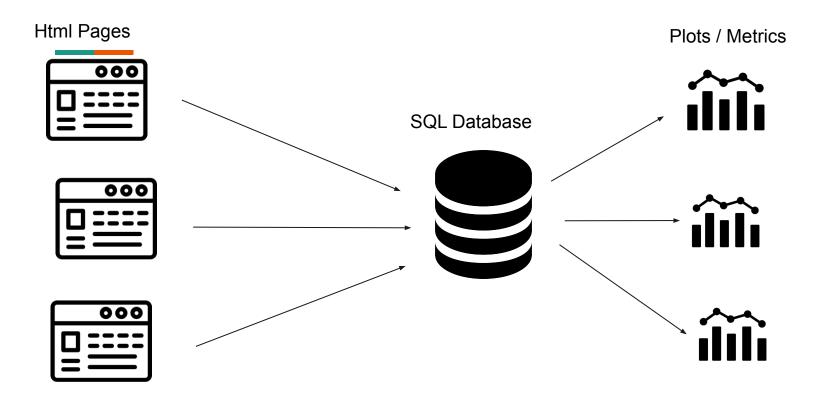
From 20 march to 30 june

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### The project:

- Dataset
  - Around 20 Gb of (compressed ) html pages crawled from atac website each 15 minutes
- Goal
  - Find Out if Atac's buses lateness is related to traffic
- Approach:
  - Parse raw data into a compact SQL database and then use it to retrieve metrics
- Problems:
  - Locate Useful data in the html pages
  - Data Size / Parsing Time
  - Data Granularity
  - Data Noise
  - Finding a meaningful metric

# **Approach**



### Locate data

# Daily scheduled departures

### Altri giorni

21:00

- » Martedi 31 marzo
  » Mercoledi 1 aprile
- » Giovedì 2 aprile
- » Venerdi 3 aprile
  » Sabato 4 aprile
- » Domenica 5 aprile
  » Lunedì 6 aprile

# Sequence of stops for a route

Nomentana/G. B. Vico

Nomentana/Monte Livata Nomentana/Monte Bianco

\* Nomentana/Poggio Fiorito

\* Nomentana/Palombarese

Nomentana/F.lli Maristi
 Nomentana/De La Riva
 Nomentana/Dante Da Maiano

\* Nomentana/Catacombe

\* Nomentana/Podere S. Alessandro

\* Nomentana/Cesarina

\* Nomentana/G.r.a.

\* Nomentana/Scuola Rurale \* Nomentana/Spaducci

\* Nomentana/Tosatti

\* Casal Boccone/Negri A.

\* Ojetti/Casal Boccone \* Ojetti/Aleramo

\* Ojetti/Pugliese \* Ojetti/Primoli

Ojetti/Tosatti Jonio/Talenti Jonio/Bandello

Stelvio Adamello

Carnaro

Monte Baldo

Tirreno/Sempione

Tirreno/Isole Eolie Conca D'oro (MB1)

\* Conca D'oro (MB1)

\* Conca D'oro (MB1)

**Bus Position** 

\* Nomentana/Monte Livata

\* Nomentana/Monte Bianco
B Nomentana/Poggio Fiorito

Nomentana/Palombarese

Nomentana/F.lli Maristi
Nomentana/De La Riva

\* Nomentana/Dante Da Maiano

\* Nomentana/Monte Livata
\* Nomentana/Monte Bianco
B Nomentana/Poggio Fiorito
\* Nomentana/Palombarese

Nomentana/F.lli Maristi Nomentana/De La Riva

Nomentana/Dante Da Maiano

Traffic State

### Locate data 2

### Route Id

### Bus Id

```
w <div class="stato4">

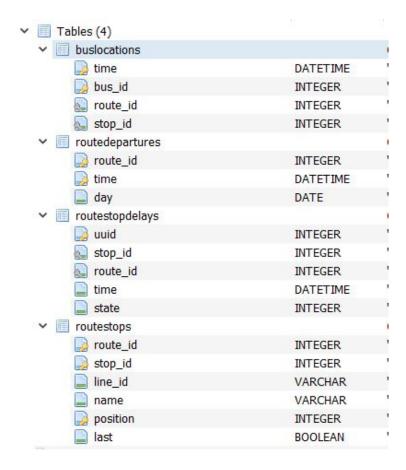
w <a class="nound" href="/paline/percorso/53671Pid veicolo=2148&nav=0&sessionid=None">

w <img class="dot" src="/paline/s/img/bus.png" alt="B" width="20" height="20">

::before

</img>
<//a>
```

# **Sqlite Database**



### **Data Size / Parsing Time**

#### Problem

- The dataset is quite heavy, extracting, moving or deleting the pages can take hours!
- Parsing the whole dataset can take days and one error may force to restart from the beginning

### My solution:

- Parse one bus line at a time, directly reading from the archive without extracting it with **winrar** library and parsing the html pages with **beautifulsoup4**
- Parsing all the days from one line takes 20-30 minutes this way
- Allows for short parsing / analysis / validation cycles

### **Data Granularity**

- As we can see 15 minutes is a long time, a bus can perform up to 10/20 stops in that time
- Buses appears on the first stop of the line several times before it actually departs
- Possible solution:
  - Interpolation
  - Using metrics that are meaningful with the granularity level we have

	line_id	route_id	bus_id	position	time(time)
1	88	53432	3376	0	07:03:11
2	88	53432	3376	14	07:18:19
3	88	53432	3376	20	07:26:02
4	88	53432	3376	29	07:33:27
5	88	53432	3376	42	07:40:47
6	88	53432	3376	0	08:33:07
7	88	53432	3376	0	08:40:35
8	88	53432	3376	0	08:48:00
9	88	53432	3376	4	08:55:42
10	88	53432	3376	23	09:10:36
11	88	53432	3376	31	09:18:16
12	88	53432	3376	42	09:25:53
13	88	53432	3376	0	10:26:42
14	88	53432	3376	0	10:34:12
15	88	53432	3376	0	10:41:38
16	88	53432	3376	8	10:49:18
17	88	53432	3376	16	10:56:48
18	88	53432	3376	25	11:04:13

### Possible metric 1 (Average Lateness from scheduled arrival)

$$M_{1, day} = \sum_{s \in stop(day)} \frac{|time(s) - closest\_sched\_time(s)|}{|stops(day)|}$$

- Idea: find the average difference between scheduled arrival time and actual arrival time
- Problems:
  - Finding a valid arrival schedule for each stop and each day
    - Atac doesn't seem to provide one
    - Third party may derive it from other metrics
  - Data only show when bus is close to a stop, even if it has not actually stopped
  - Data granularity is 15 minutes

Appears to be fragile

### Possible metric 2 (End to end average time)

$$M_{2, day} = \sum_{r \in runs(day)} \frac{arrival\_time(r) - departure\_time(r)}{|runs(day)|}$$

End to End time:

 $r \in runs(day)$ 

- Time required to travel from one end to the other of a route,
- Problems:
  - We need to find a bus on both ends
  - Dataset is somehow unbalanced under this point of view

	line_id	route_id	position	count(*)	count(distinct(date(buslocations.time)))	
1	88	53432	0	3036		53
2	88	53432	1	48		32
3	88	53432	2	73		39
4	88	53432	3	110		43
5	88	53432	4	211		51
6	88	53432	5	178		53
7	88	53432	6	204		50
8	88	53432	7	306		53
37	88	53432	36	67		41
38	88	53432	37	369		53
39	88	53432	38	113		43
40	88	53432	39	91		40
41	88	53432	40	229		52
42	88	53432	41	54		31
43	88	53432	42	109		45
44	88	53432	43	202	]	48

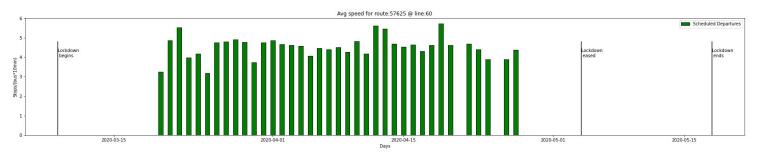
# Possible metric 2b (Average Speed stops/10min)

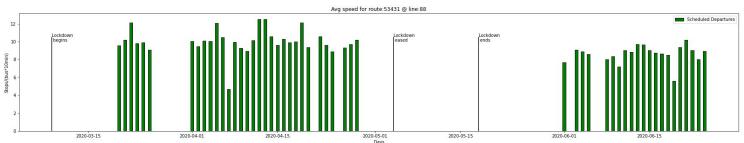
$$M_{2b, day} = \sum_{bus \in buses(day)} \sum_{s1,s2 \in succ\_stops(day, bus)} \frac{stop\_position(s2) - stop\_position(s1)}{time(s2) - time(s1)}$$

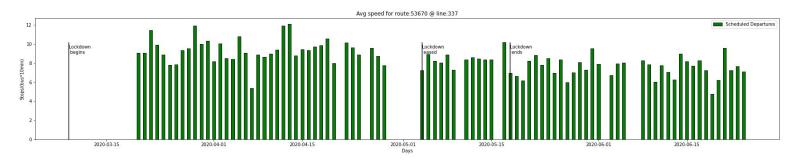
- Avg speed measured in stops
  - Basic Idea: a bus has to perform all the stops of a certain line after each departure, so we can basically take each successive stops and measure the average speed in each second (or each 10 minutes)

	line_id	route_id	bus_id	position	time(time)	
1	88	53432	3376	0	07:03:11	s1
2	88	53432	3376	14	07:18:19	s2
3	88	53432	3376	20	07:26:02	s3
4	88	53432	3376	29	07:33:27	Î
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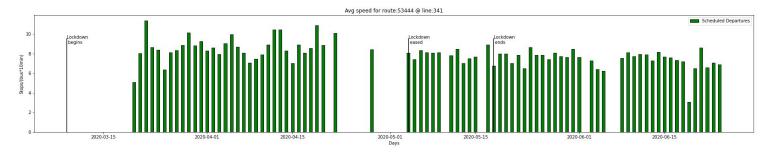
## Plots for 2b (Average Speed stops/10min)

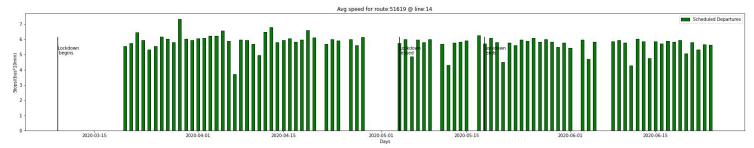






### Plots for 2b (Average Speed stops/10min)





#### Observations:

- Routes can change over time
- Some of them are shown as unavailable, others as not existing, for other there are just no pages for a certain day
- Doesn't seem to be meaningful

### Possible metric 3 (Bus Locations Heatmap)

Line 14 before lockdown

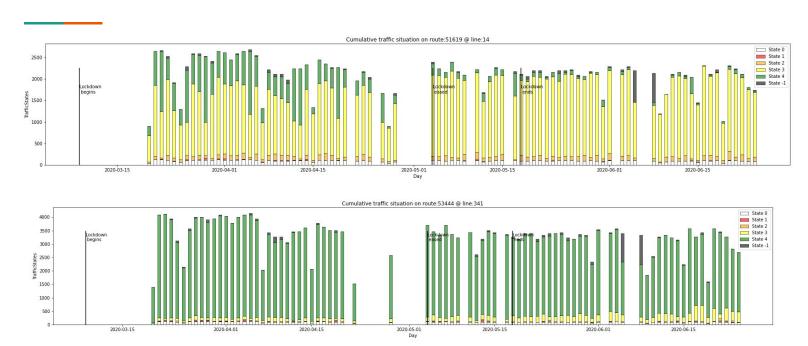


Line 14 after lockdown



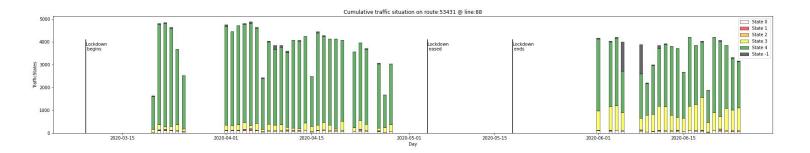
## Possible metric 4 (Daily cumulative traffic state)

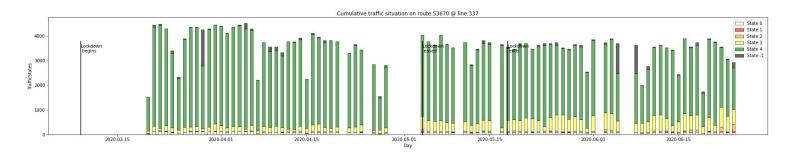
- Idea: using the state indicator from the html pages to retrieve the traffic conditions



- Observations:
  - This metric clearly shows the impact that lockdown had on traffic, and how much it impacted more the central lines compared to the periferic ones

## Possible metric 4 (Daily cumulative traffic state)

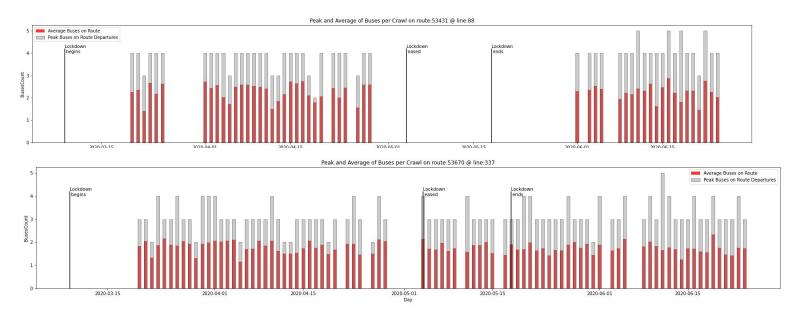


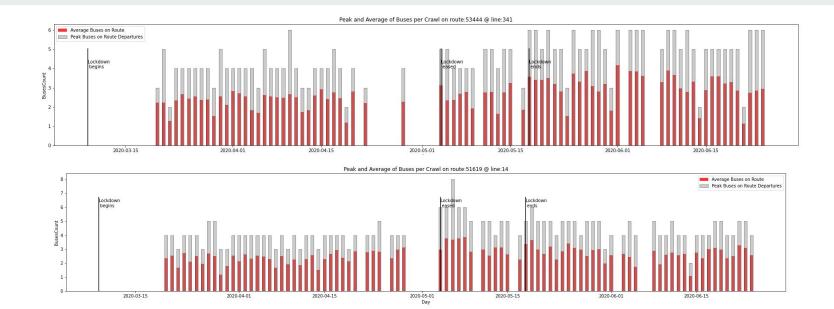


# Possible metric 5 (Average number of buses on route)

Idea: extract average/maximum number of buses we see on a certain route for a given day to have an esteem how many buses were assigned to a certain route

$$M_{4, day} = \sum_{p \in pages(day)} \frac{|buses(p)|}{|pages(day)|}$$





### Observation:

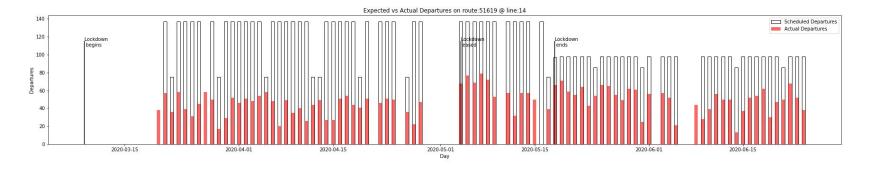
- The number of buses seen on one line at the same time generally increases after lockdown
- It can due either to the <u>traffic</u> or to an increase in <u>departures</u>, how to find out which one was ?

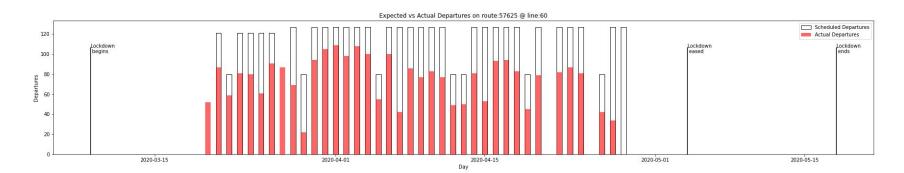
Possible metric 5 (Actual vs Expected Daily Departures)

- Does not suffer from granularity problem. (Unless a route is completed in less than 15 minutes)
- Easy to compute (once you figure out)
- Basically we can just scan the ordered list of **positions** and look for overflows

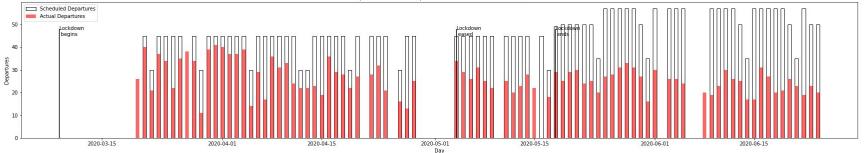
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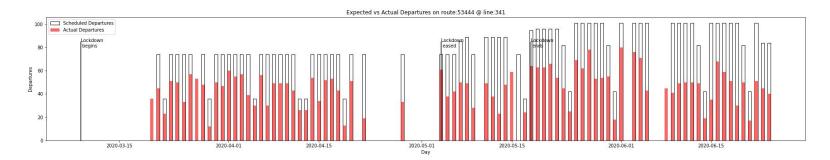
# Possible metric 6 (Actual vs Expected Daily Departures)







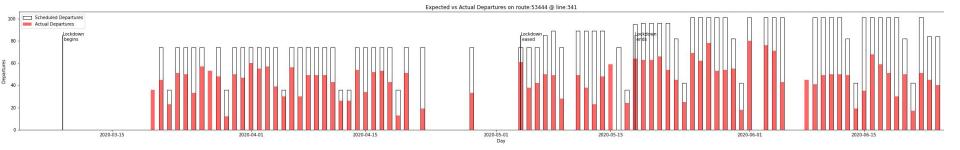




### Observations:

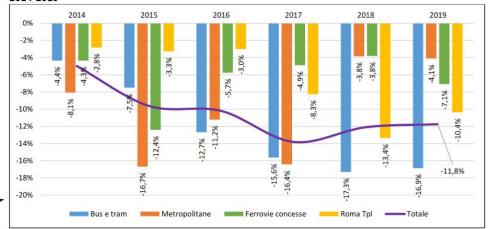
- This metric looks extremely interesting!
- Actual and Scheduled departures have signficative differences between different routes
- It looks like that promised changes in departures do not reflect real ones

### Final considerations



Atac states that they can guarantee more than 80 % of the departures in 2019, but this preliminary analysis show that this numbers may be inflated or that performance worsened this year

Graf. 6 - Differenza tra produzione effettuata e programmata per linea (% di vetture-km). Roma. Anni 2014-2019



Link to reference

Fonte: Elaborazioni Ufficio di Statistica di Roma Capitale su dati RSM / Atac

### **Further Improvements**

- Parse the whole database
- Extract global version of the above metrics
- Modify the scraper to parse the reason of page unavailability
- Investigate more on the difference between scheduled and actual departures

# Thanks for the attention