



Characterizing and Modeling Mobile Networks User Traffic at Millisecond Level

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[Developing the
Science of Networks]

- Mobile traffic data often unavailable due to various reasons
- Lack of mobile datasets limits research advances
- We fill this gap making available a large dataset



| YEAR | OPERATOR | ZONE | CARRIER FREQUENCY (MHz) | ID | TOTAL TIME (H) | TOTAL USERS | TRAFFIC VOLUME (GB) | DATASET SIZE (GB) |
|------|----------|------|----------------------------|-----|----------------|-------------|---------------------|-------------------|
| 2020 | A | I | 796 | BS1 | 247 | 72 321 636 | 388 691 | 11 |
| 2020 | B | I | 1815 | BS2 | 227 | 23 508 594 | 242 406 | 3 |
| 2020 | A | I | 2650 | BS3 | 74 | 10 493 540 | 18 403 | 0.24 |
| 2021 | B | II | 816 | BS4 | 163 | 14 813 731 | 579 376 | 12 |
| 2021 | C | II | 1835 | BS5 | 313 | 59 463 421 | 2 378 256 | 27 |
| 2021 | A | II | 2650 | BS6 | 353 | 33 650 085 | 1 308 064 | 11 |

Motivation and Contribution

- In-depth characterization of millisecond-level mobile traffic
- We make open:
 - Large dataset from multiple Base Stations (BSs)
 - Real LTE control traffic data at millisecond level
 - Valuable for research (especially data-driven/AI driven optimizations)

At COVID time!



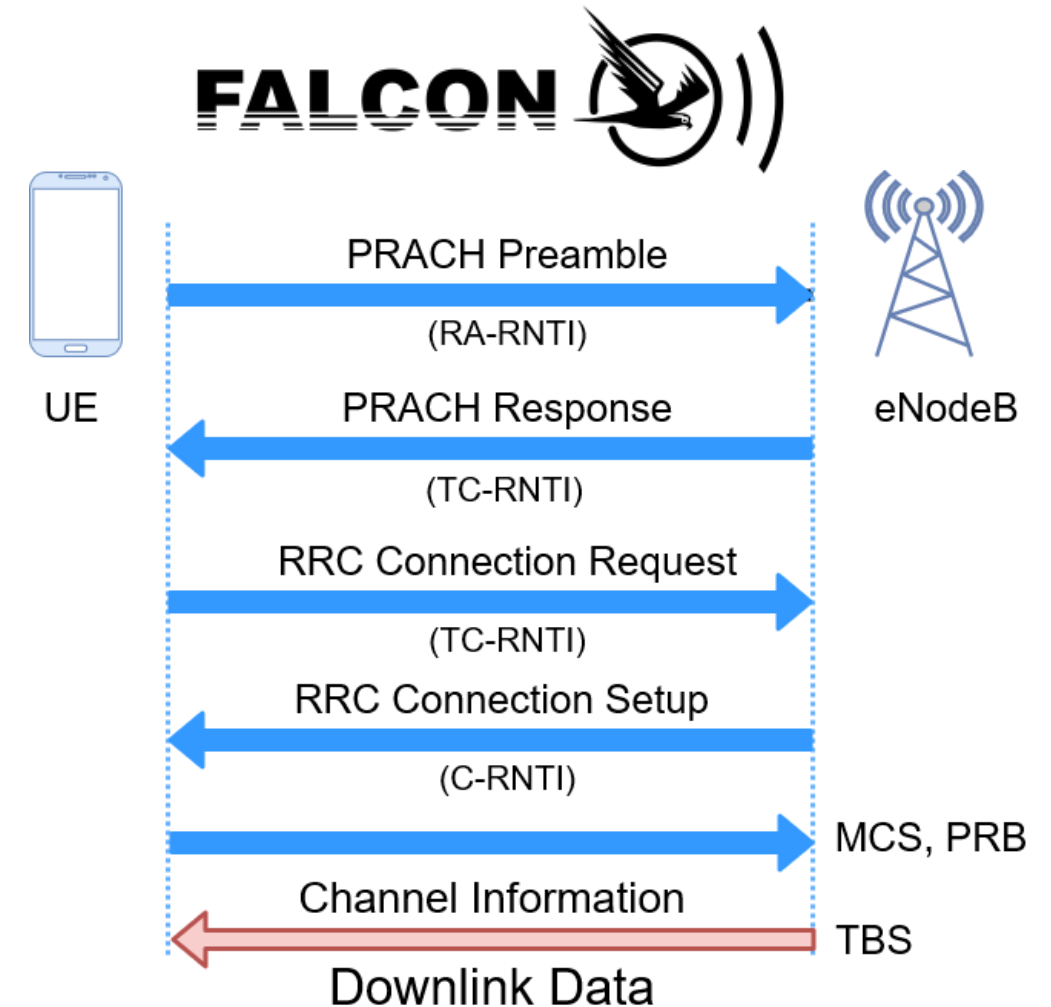
The Decoded Data

- **Information:**

- Temporary user ID (RNTI) associated with the user
- Frame ID containing traffic allocation for each C-RNTI
- Associated transport block size (TBS)
- Transmission details: Modulation and Coding Scheme (MCS), and utilized Physical Resource Blocks (PRB)

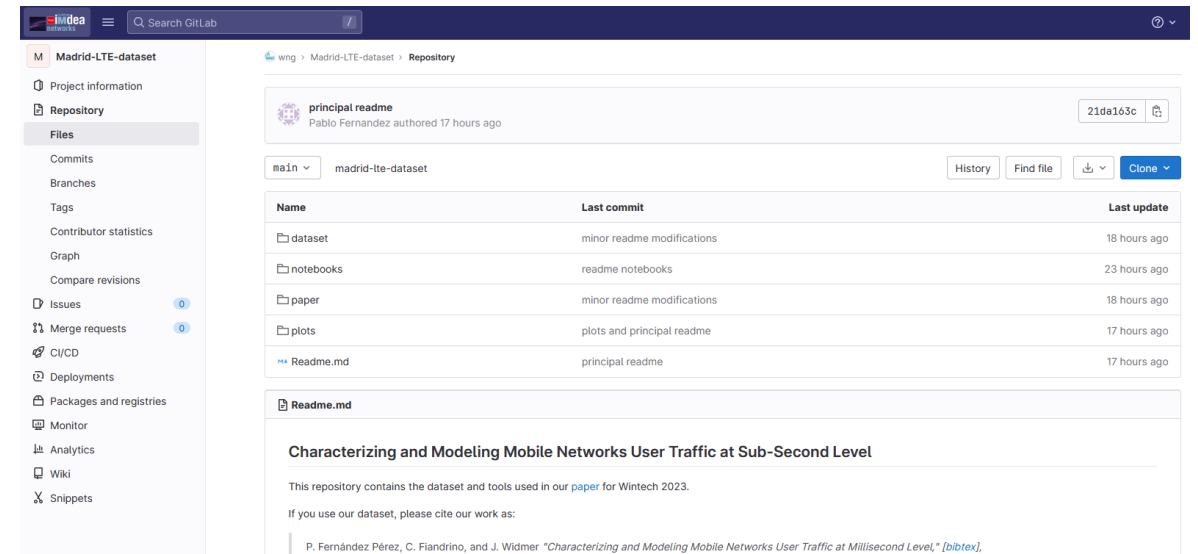
- **User identification:**

- Analyze inter-transmission times between identical RNTIs.
- Set a time threshold of 10 s for user lifetime.



Our Dataset in a Nutshell

- Repository: data and code
- Raw dataset 1 ms granularity
- Minimal processing (RNTI scrambling)
- Also stored 1 s granularity processed data



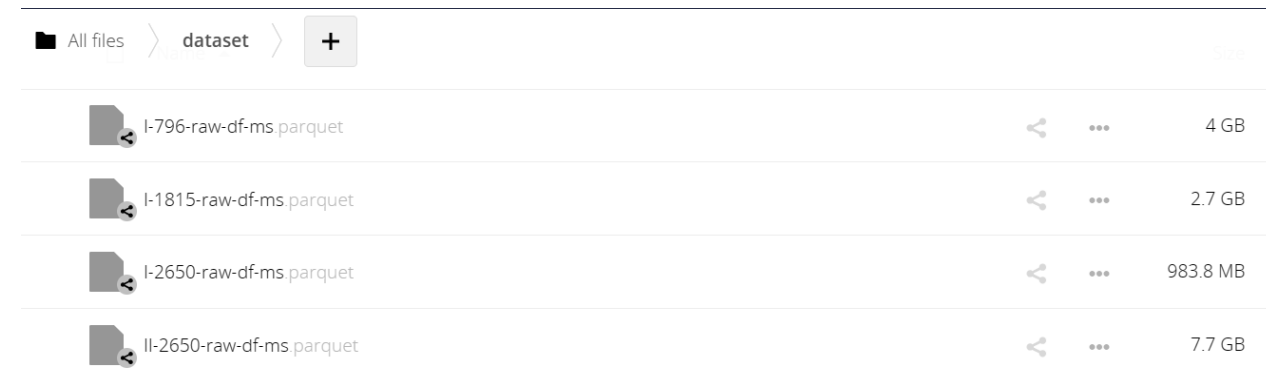
The screenshot shows the GitLab interface for the repository 'Madrid-LTE-dataset'. The main content is the 'principal readme' file, which includes the following text:

Characterizing and Modeling Mobile Networks User Traffic at Sub-Second Level

This repository contains the dataset and tools used in our [paper](#) for Wintech 2023.

If you use our dataset, please cite our work as:

P. Fernández Pérez, C. Fiandrino, and J. Widmer "Characterizing and Modeling Mobile Networks User Traffic at Millisecond Level," [bibtext].

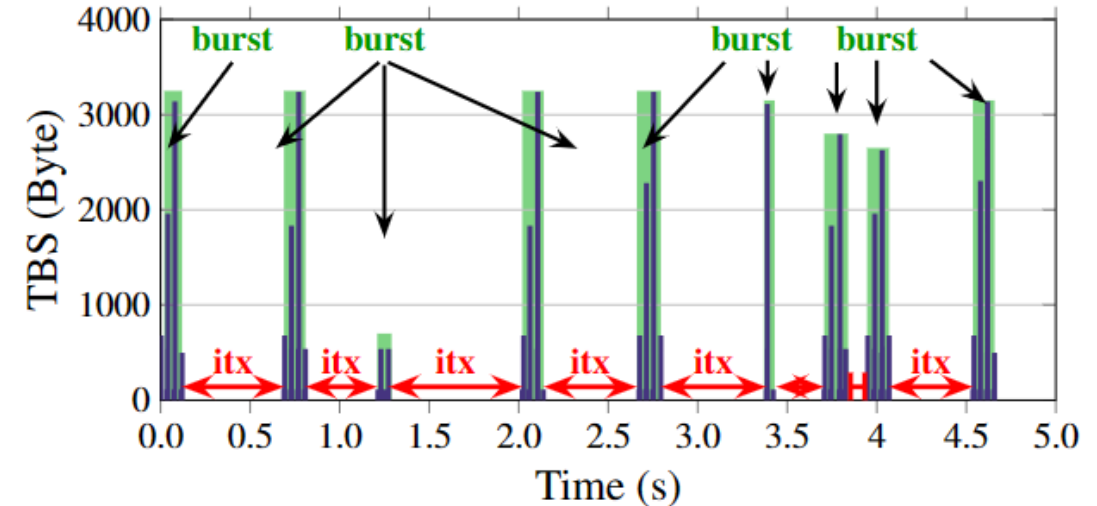
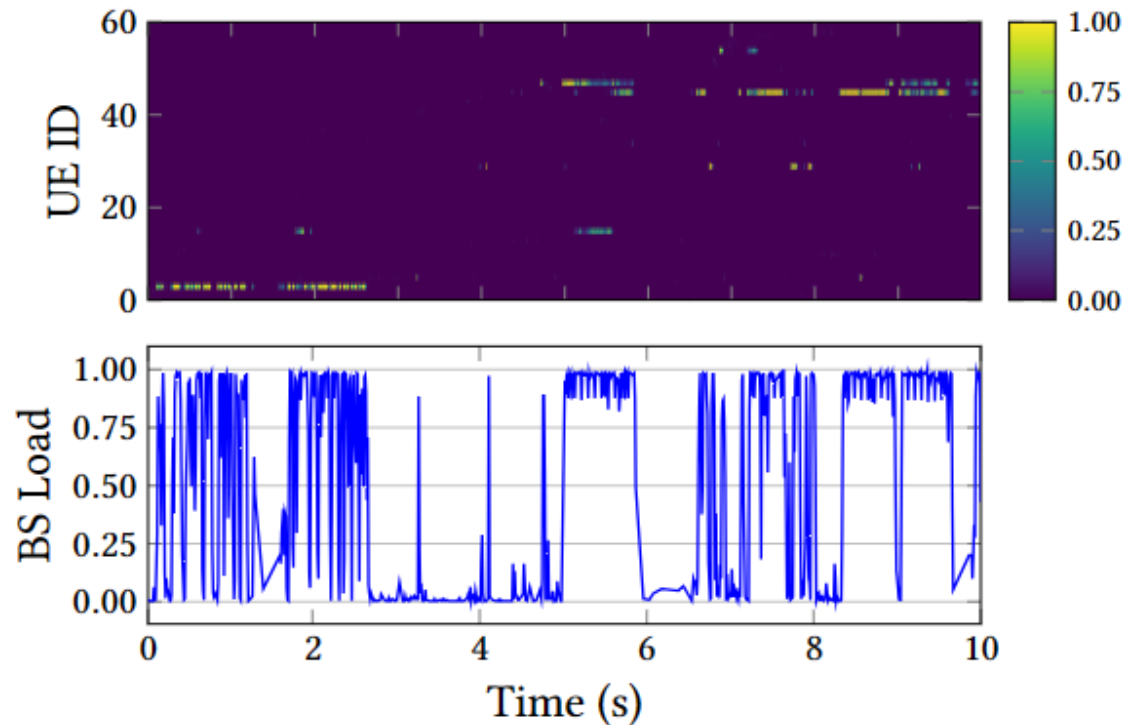


The screenshot shows the file browser for the 'dataset' directory. It lists four files with their names and sizes:

| File Name | Size |
|---------------------------|----------|
| l-796-raw-df-ms.parquet | 4 GB |
| l-1815-raw-df-ms.parquet | 2.7 GB |
| l-2650-raw-df-ms.parquet | 983.8 MB |
| ll-2650-raw-df-ms.parquet | 7.7 GB |

<https://git2.networks.imdea.org/wng/madrid-lte-dataset>

Looking traffic at BS level and at user level



- Burst: consecutive non-zero TBs
- Intertransmission times (itx): time elapsed between consecutive TBs

Results overview

Traffic at BS level

- User-Traffic distribution.
- Self similarity and long-range dependence.

Traffic at user level

- 2d-histograms of:
 - t_{bs}, n° occurrences, n° users.
 - itx, n° occurrences, n° users.

Number of RRC
connected users

- Bimodal distribution.

Traffic at BS level

Traffic at BS level

- User-Traffic distribution.
- Self similarity and long-range dependence.

Traffic at user level

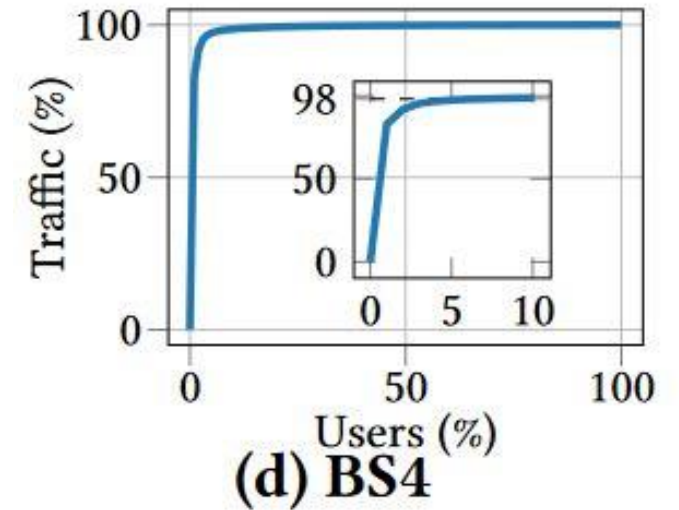
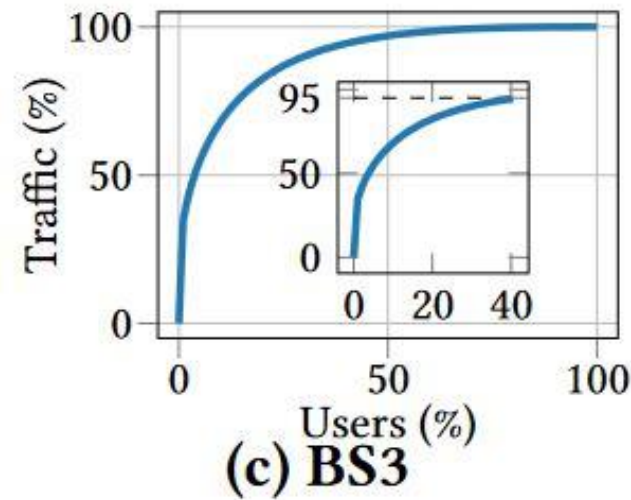
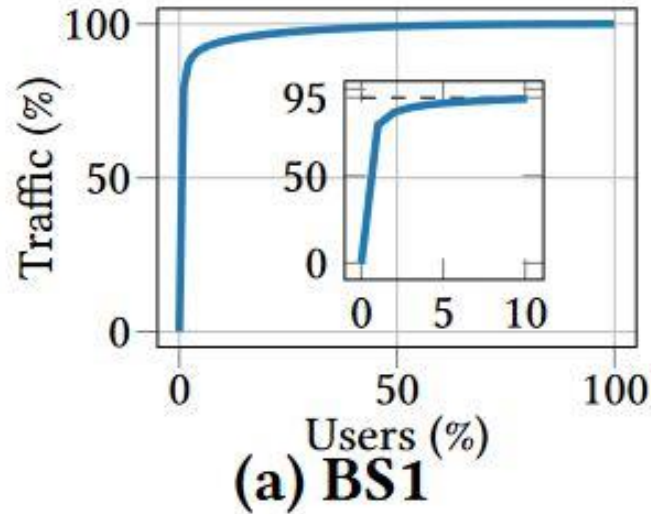
- 2d-histograms of:
 - t_{bs}, n° occurrences, n° users.
 - it_{x}, n° occurrences, n° users.

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- Bimodal distribution.

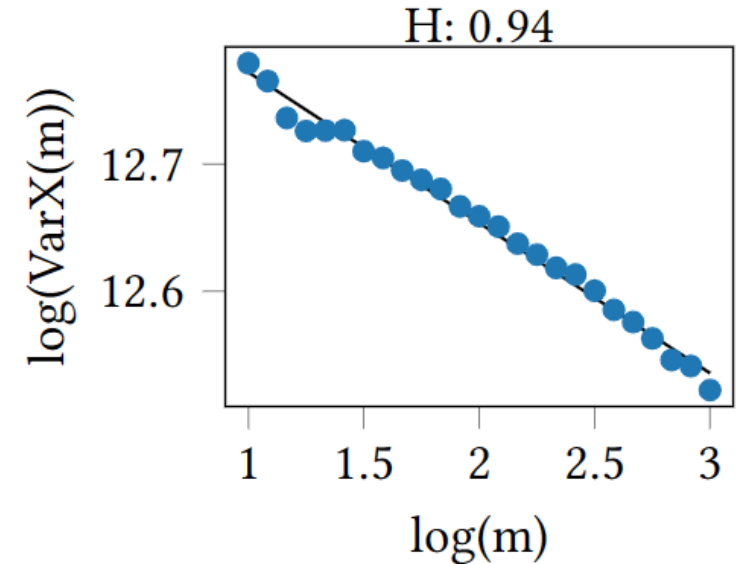
Traffic at BS level: User distribution

- Top 10% of the users consume 90% of the traffic
- Only BS3 has a slightly different distribution where top 30% of the users consume 90% of the traffic



Traffic at BS level: Hurst Parameter

- Long range dependence (LRD):
the sum of the complete sequence of
the autocorrelation function is infinite
- Self-similarity: statistical similarity
across different scales (Hurst parameter H)



$$x_{at} \sim a^H x_t$$

- Self-similarity more evident for downlink than uplink traffic

Traffic at User Level

Traffic at bs level

- User-Traffic distribution.
- Self similarity and long-range dependence.

Traffic at user level

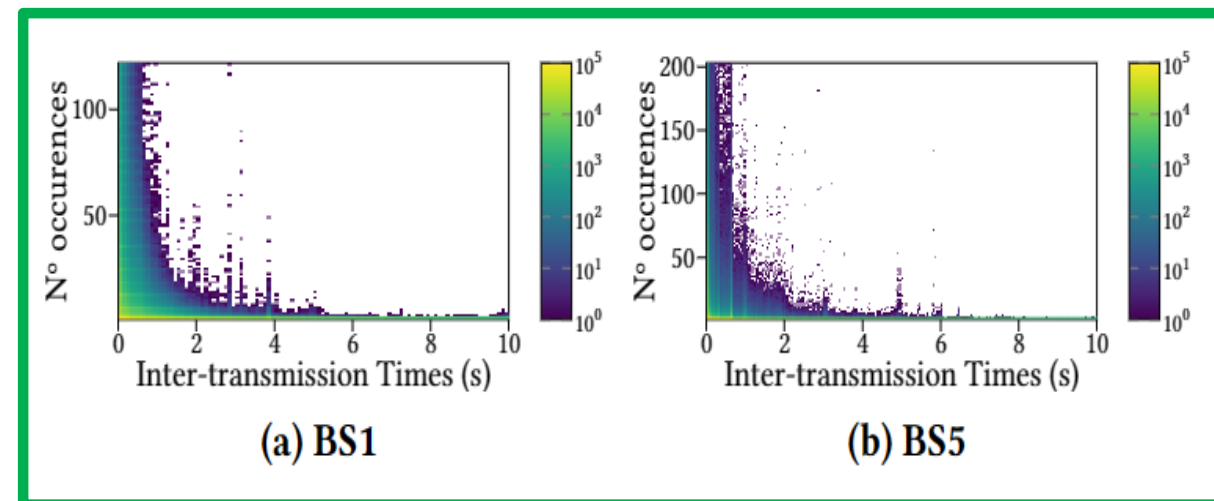
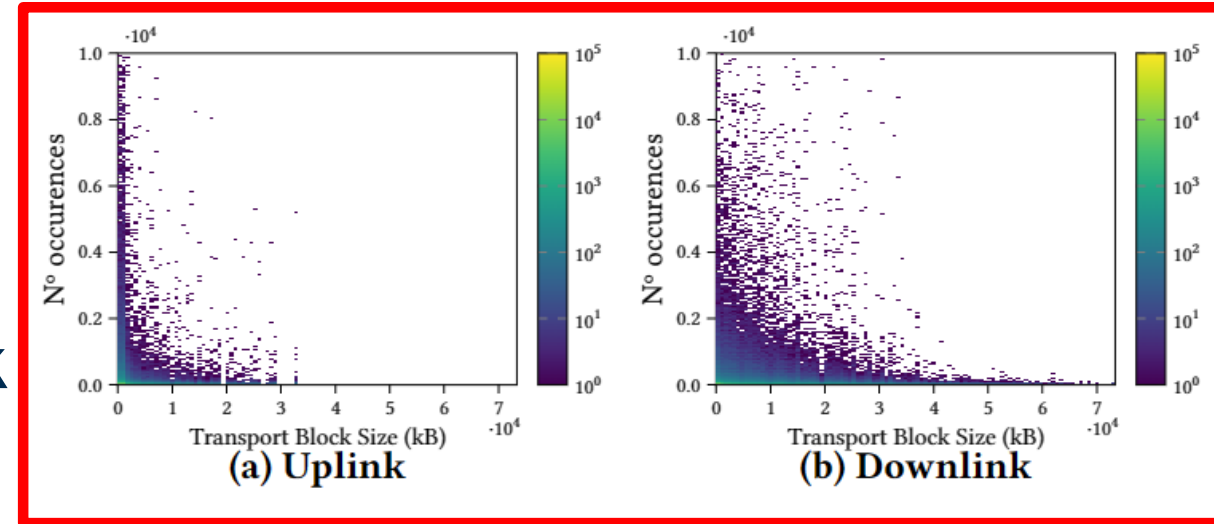
- 2d-histograms of:
 - t_{bs}, n° occurrences, n° users.
 - i_{tx}, n° occurrences, n° users.

Number of RRC
connected users

- Bimodal distribution.

Analysis of User Level Traffic

- **Transport block size** and **inter-transmission times (itx)** analysis.
- Separate analysis from uplink and downlink traffic.
- Differences are attributed to traffic nature and resource allocation policies



Number of RRC connected users

Traffic at bs level

- User-Traffic distribution.
- Self similarity and long-range dependence.

Traffic at user level

- 2d-histograms of:
 - t_{bs}, n° occurrences, n° users.
 - itx, n° occurrences, n° users.

Number of RRC
connected users

- Bimodal distribution.

Number of RRC Connected Users

- These users are potentially active (in either uplink or downlink)
- Number of RRC connected users follows a bimodal distribution
- We find all BSs following a bimodal distribution with significant shape differences

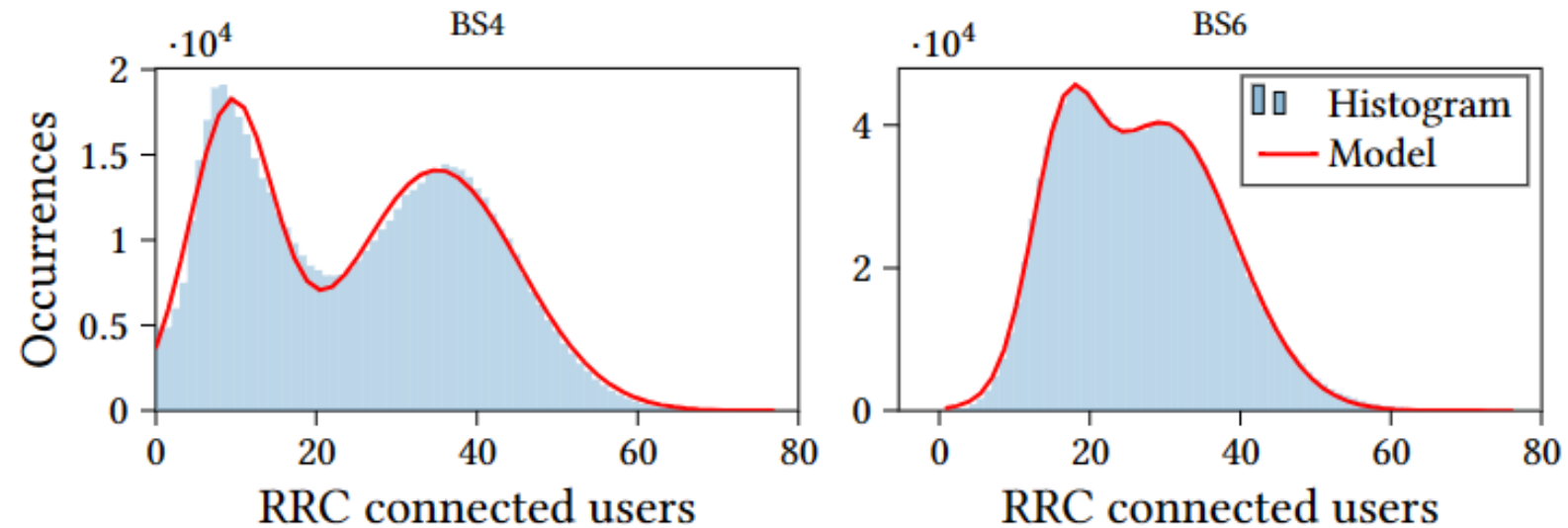


Figure 10: Bi-modal distribution

Take-Home Messages

- Large real-world LTE traffic dataset at ms level granularity
- Enables data-driven research as well as AI-based network optimization
- Important for research reproducibility



<https://git2.networks.imdea.org/wng/madrid-lte-dataset>