

RICERCANDO:

Towards Mobile Broadband Measurement Mining

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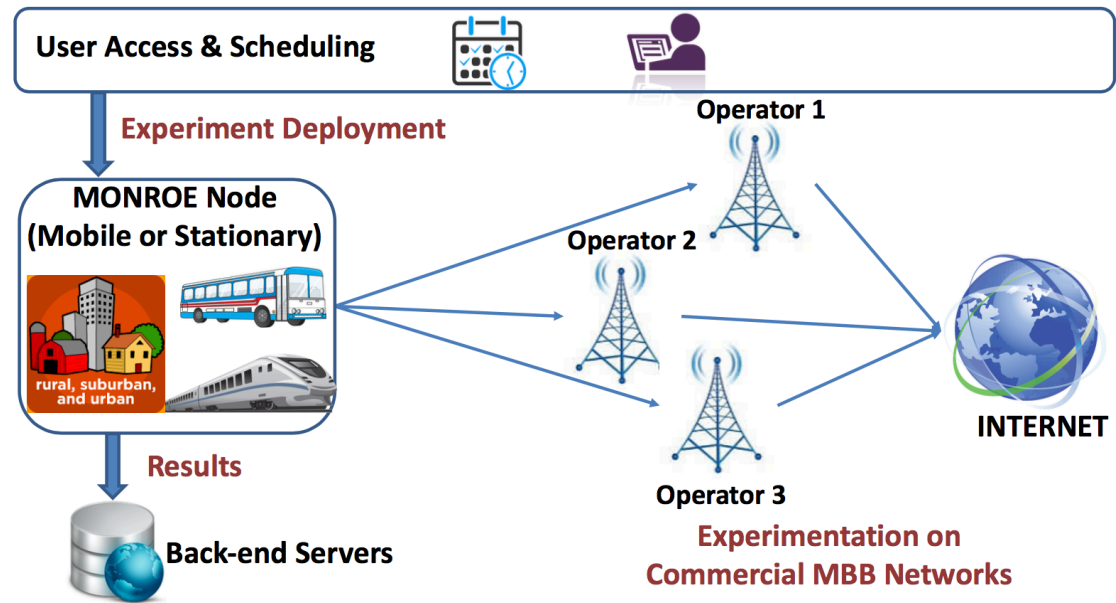
Cyber Alp Retreat,
Gstaad, Switzerland, January 2018

Fabio Ricciato sends his
greetings to everyone!



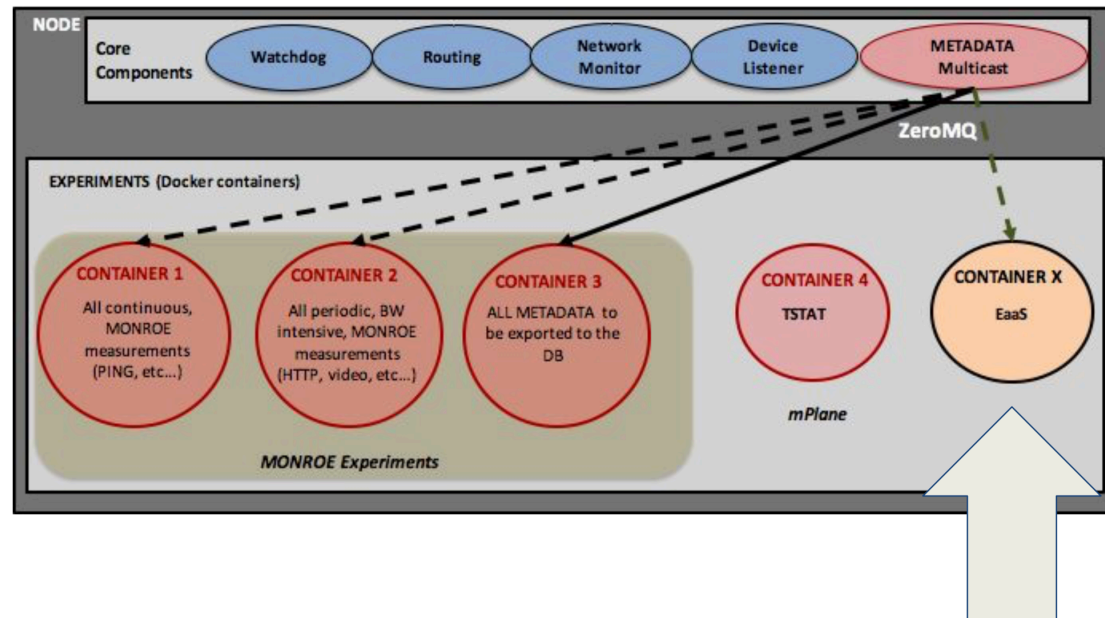
MONROE

*Design, build and operate an open, **European-scale**, and flexible hardware-based **platform** to run experiments on operational **3G/4G Mobile Broadband networks** with WiFi connectivity*



MONROE

- Experimentation:
 - Users create Docker containers with experiments that are uploaded to individual nodes and ran at scheduled times
 - Background ping throughput, traceroute, etc. measurement ran periodically



Network Measurement Data

- Performance indicators:
 - Packet delay, jitter, packet loss, open/closed ports, etc. whatever users design
- Metadata (in MONROE):
 - Ping time, upload/download speed
 - Network type (2G, 3G, LTE, WiFi)
 - Signal strength (RSSI, RSRP)
 - Node coordinates
 - Node CPU utilisation, temperature
 - Other experiments on a node, software versions, device config

With crowdsourced measurements also:

- Device model, OS
- Device artefacts (e.g. bent antenna)
- User's mobile plan and potential caps reached



RICERCANDO

- **Goal:** Facilitate network measurement data (and metadata) exploration, visualisation and interpretation
 - Ease problem discovery and troubleshooting
 - Separate **monitoring system glitches** from **network anomalies**
 - Support Root-Cause Analysis (RCA)



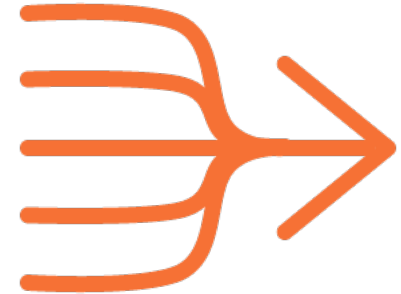
RICERCANDO Approach

- Develop a visual programming environment for data analysis
 - Jupyter-based rapid data exploration
 - Orange-based data analytics pipeline
 - “Data mining fruitful and fun” tool developed at FRI
 - Used in bioinformatics, statistics offices, etc.
 - <http://orange.biolab.si/>



Challenges

- Big time-series data
 - A few GB of data per day
 - Events with up to 10ms granularity
- Data merging
 - Different data comes with different granularity
 - Mismatched timestamps among different data sources
- Demand for visually-driven analysis
- Identify the most useful analytics tools



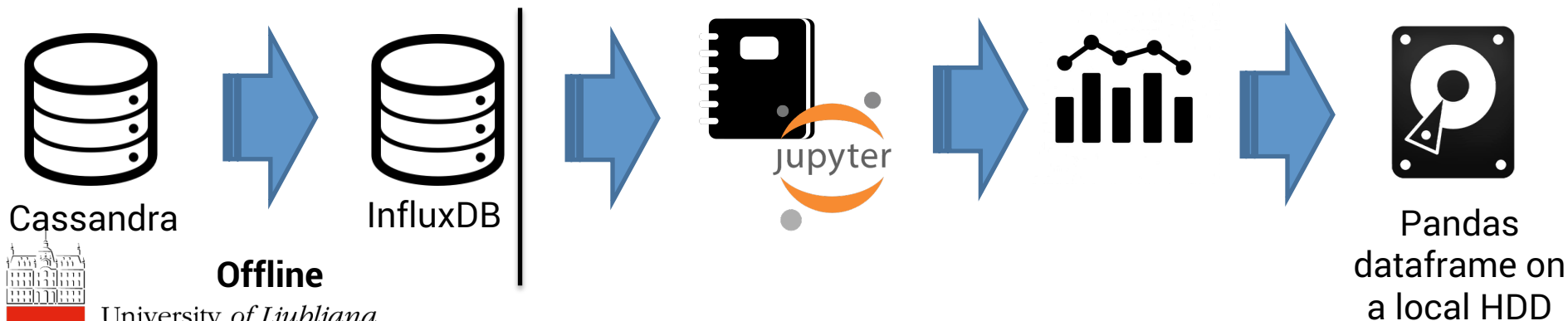
Temporal data handling and merging

- Standard databases are insufficient
 - From Cassandra to Influx DB
 - From noSQL to a DB optimised for time series data
 - “up to 168x faster response times for tested queries”
- Data merging:
 - Create 10ms granularity tables, but support user-defined granularity querying
 - Allow users to query the DB and get results interpolated in the desired way:
 - Without interpolation or with any of the pandas supported methods (e.g. linear, nearest, etc.)



Visual analysis

- **Jupyter** notebooks
 - Orange is very easy to use but the overhead prevents the analysis of massive data sets
 - Jupyter Notebooks allow highly customised analysis with substantial machine learning and visualisation support
 - Pipeline:



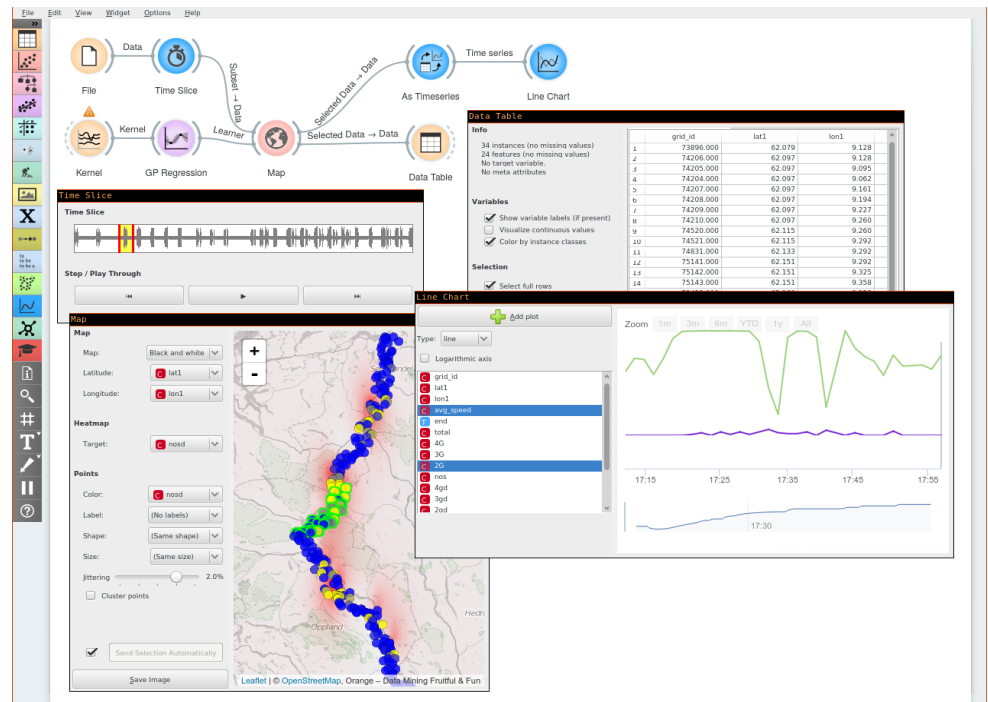
Visual analysis

- **Orange** analysis
 - Once we single out interesting data we send it for sophisticated analysis in Orange
 - Pipeline



Identifying the Most Useful Tools

- Geo-referenced visualisation:
 - Signal coverage maps
 - Statistical comparison of measurements in different regions
 - Location-dependent anomaly detection



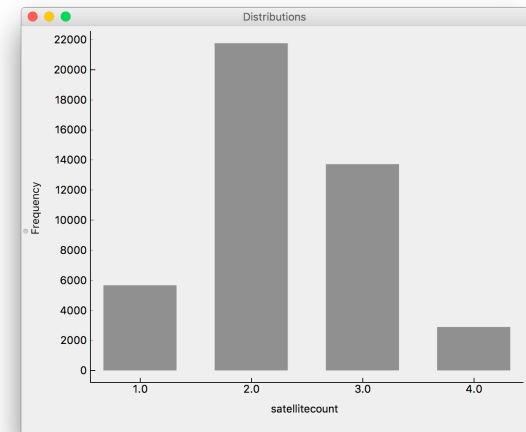
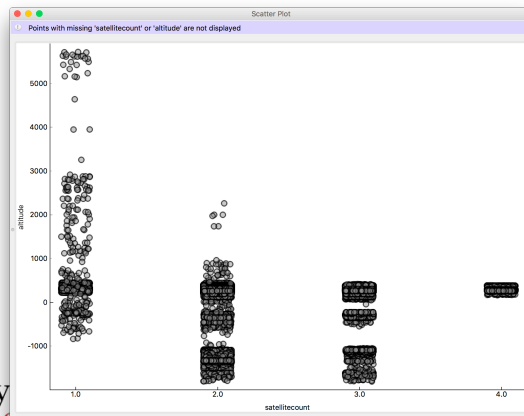
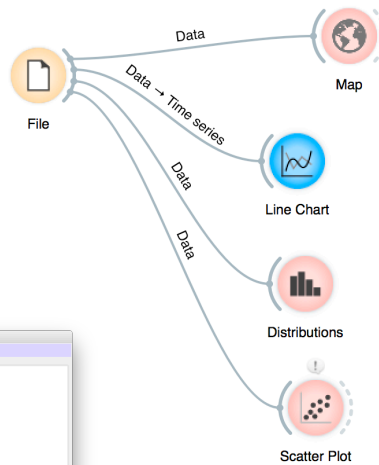
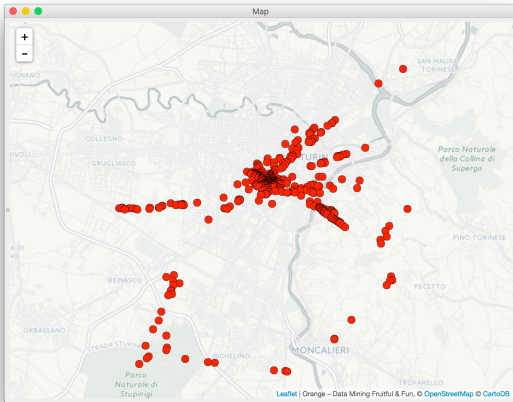
Identifying the Most Useful Tools

- Significant factor analysis
 - Which (combinations of) attributes impact(!) measurements results and how much?
 - For different types of data we implement different tests:
 - Hypergeometric test example
 - N – population size (say, all port 25 blocking tests)
 - K – success count
 - n – number of draws (say, all tests with iPhone on LTE)
 - k – number of observed successes
 - *Enrichment* = $(k/n)/(K/N)$



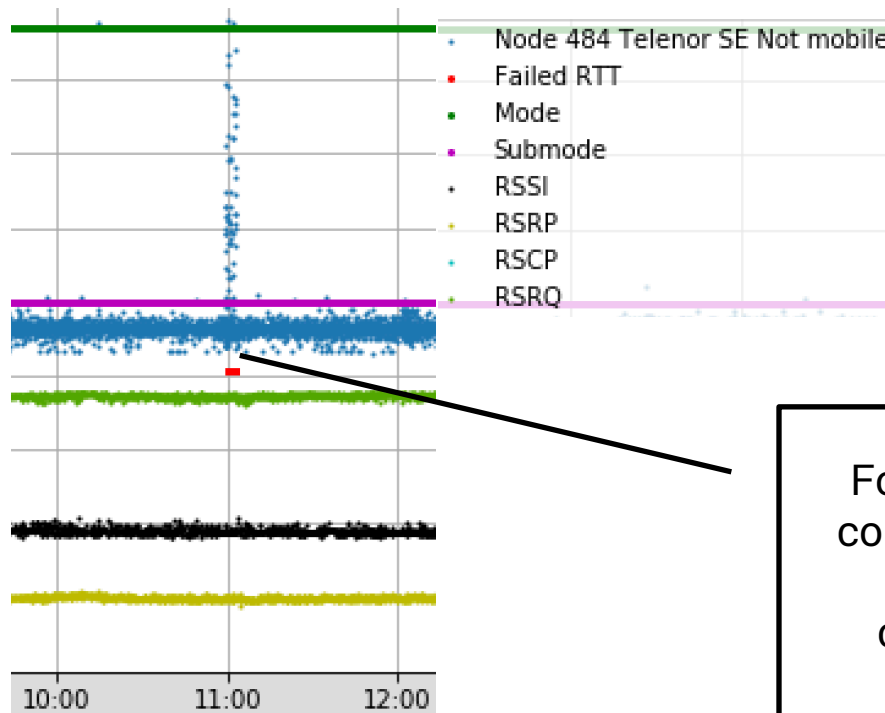
RICERCANDO in Action

- Anomaly examination



RICERCANDO in Action

- Experiment interplay – visual analysis



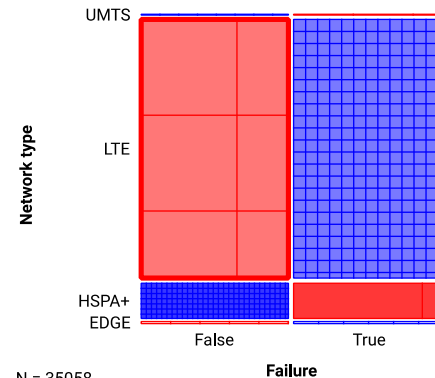
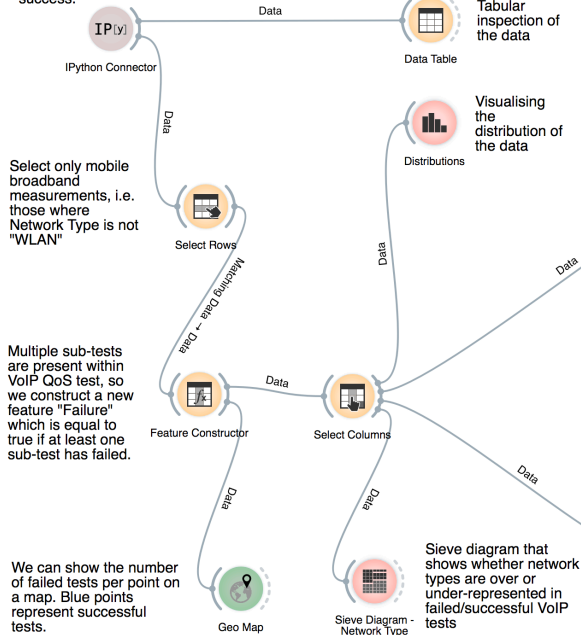
Found that RTT spikes correlate with increased CPU activity which correlates with user experiment activity



RICERCANDO in Action

- Analysing VoIP test performance for Slovenian mobile broadband operators

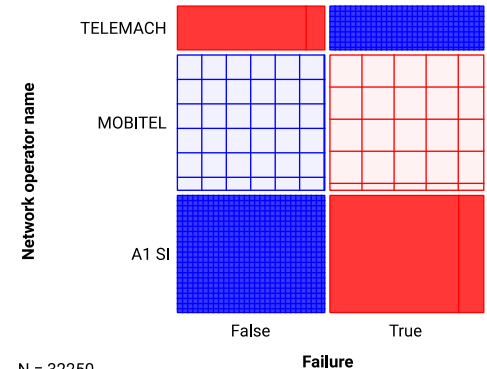
IPython Connector widget is used to load locally stored data (from IPython Notebook) on VoIP test success.



N = 35058
 $\chi^2=636.90$, $p=0.000$

Significant Groups

Perform hypergeometric test with (VoIP test) Failure as a dependent and network_type and network_operator_name as independent variables. The result tells us which combinations of variables are more/less represented in a set of failed tests than expected, if a uniform distribution is assumed.



N = 32250
 $\chi^2=6698.78$, $p=0.000$

RICERCANDO

- A toolbox for supporting discovery in mobile broadband data
- Orange tools are publicly available:
 - Main environment <http://orange.biolab.si/>
 - Add-on enabling local InfluxDB mining and stored Jupyter data access with additional tests <https://github.com/biolab/monroe-anal/>
- **Mine your data**, not Bitcoins!



...since it's my first time at Cyber Retreat

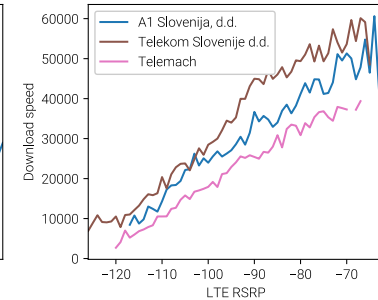
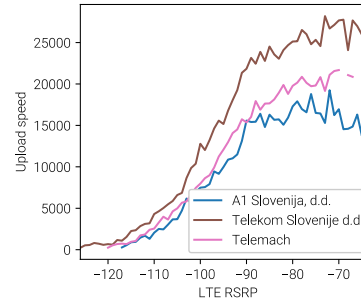
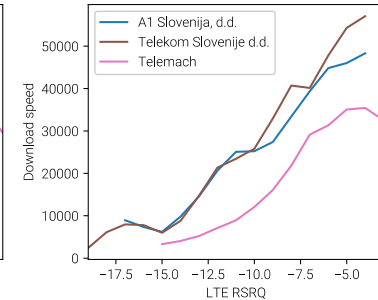
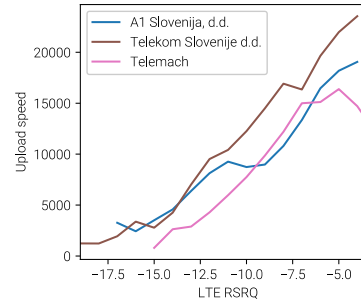


Mobile and Wireless Research at FRI

Mobile broadband
measurement data
analysis

Collaborative wireless
decoding

**Mobile sensing for
behaviour modelling**



Mobile Sensing for Behaviour Modelling Projects

Big Data

- Telecom call data records
- Modelling and examining disease spreading and containment
 - Where do people move – How does the disease spread
 - Who do people call – How does the information spread?



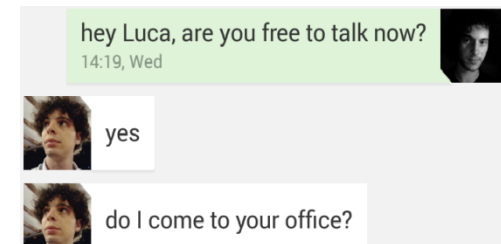
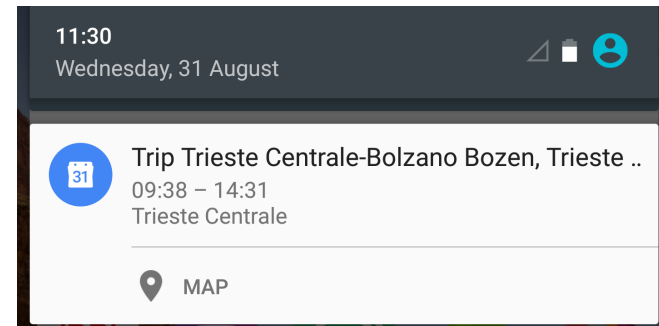
Smaller Data

- Smartphones are personalised devices carried at all times
- Smartphone sensors readings can reveal (semantic) location, physical activity, communication patterns, emotion, etc.

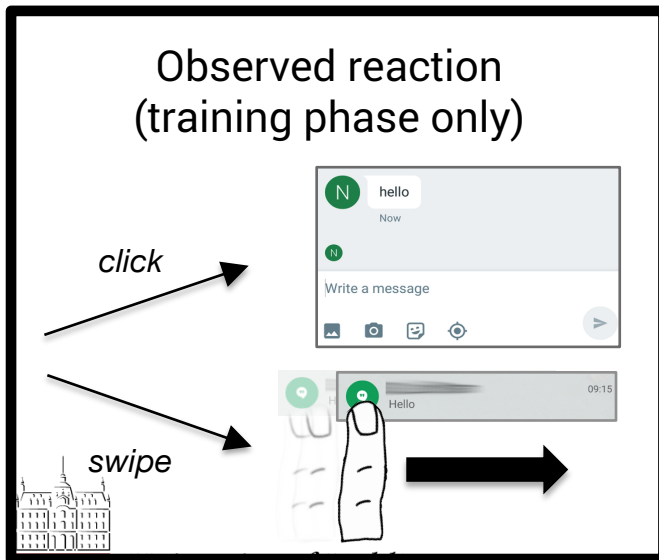
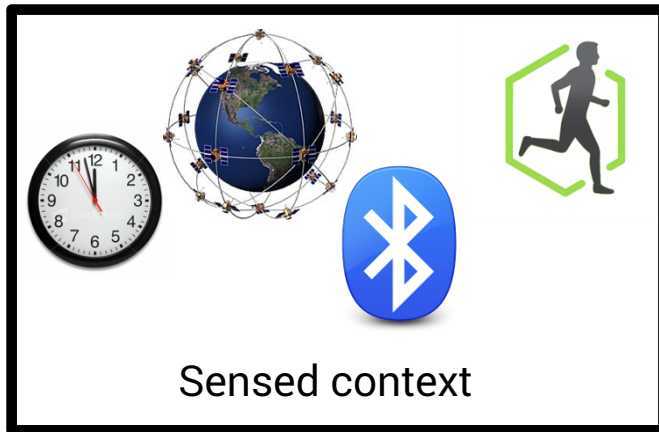


Mobile Sensing for Behaviour Modelling Project Example

- **Managing attention** of a mobile user
 - When/how to efficiently deliver information to a user?



Mobile Sensing for Behaviour Modelling Project Example



Classifiers:

- Batch, online
- Naïve Bayesian, Tree-based, Boosting

Notification response presence, time, sentiment

InterruptMe Android Library for notification management

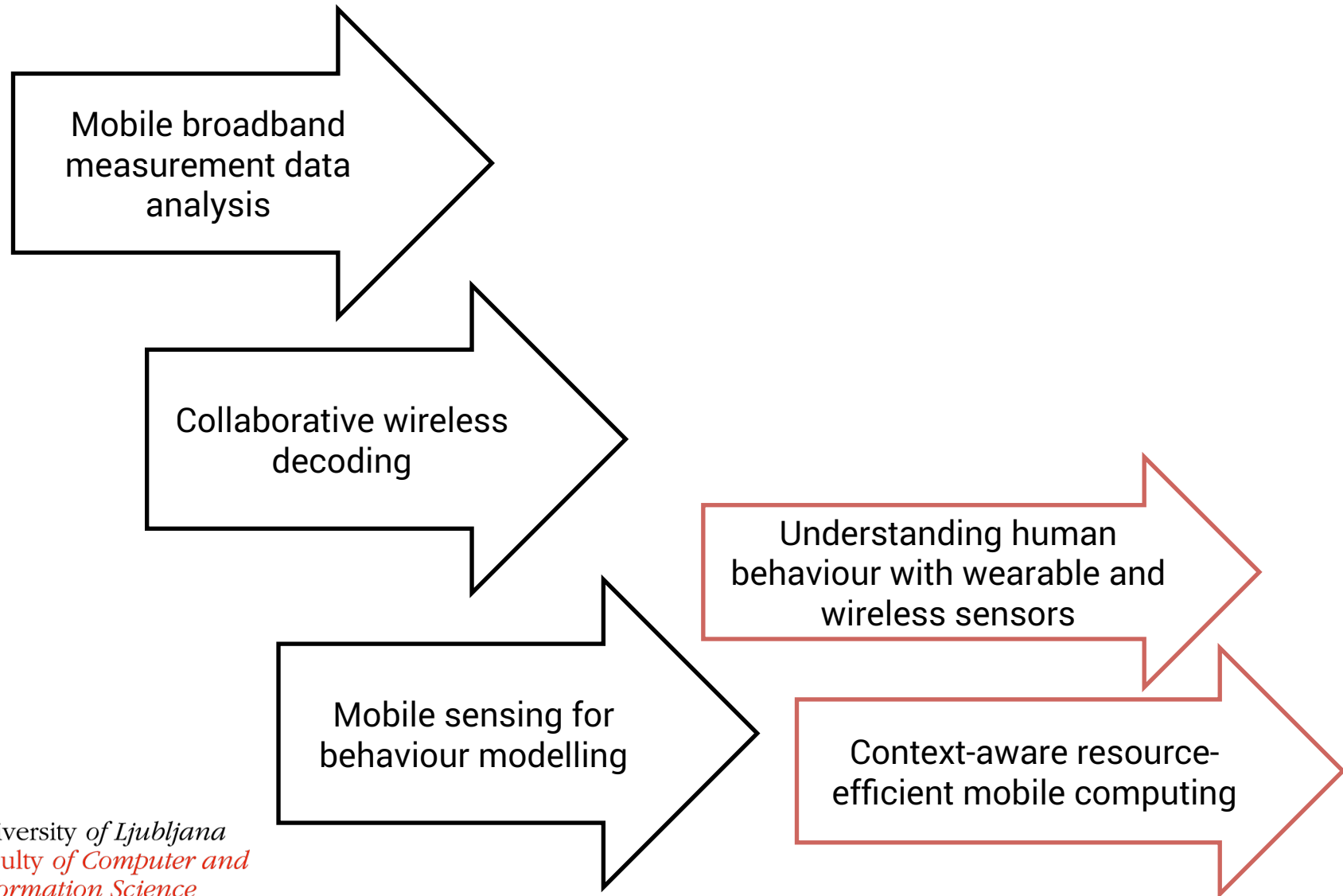
bitbucket.org/veljkop/intelligenttrigger
github.com/vpejovic/MachineLearningToolkit

Mobile Sensing for Behaviour Modelling Project Example

- **Managing attention** of a mobile user
 - When/how to efficiently deliver information to a user?
- Our attempts:
 - Using sensor data and experience sampling, on-device machine learning
 - Monitoring real-world notification usage
 - Taking into account application type, contact type (e.g. family, friend, work-related)
 - Sensing cognitive engagement with a phone



Mobile and Wireless Research at FRI



Collaborators

- **FRI:**
 - Prof Blaž Zupan
 - Blaž Repas
 - Ivan Majhen
 - Dr Miha Janež
- **Outside:**
 - Prof Mirco Musolesi, Dr Abhinav Mehrotra (UCL)
 - Christoph Anderson (Kassel University)
- **In between:**
 - Prof Fabio Ricciato



Thank you!

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@veljkoveljko

**We have an opening
for a professor in
cybersecurity closing
on 20/2!**



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