

# TESTING IOT / 5G TEST NETWORK

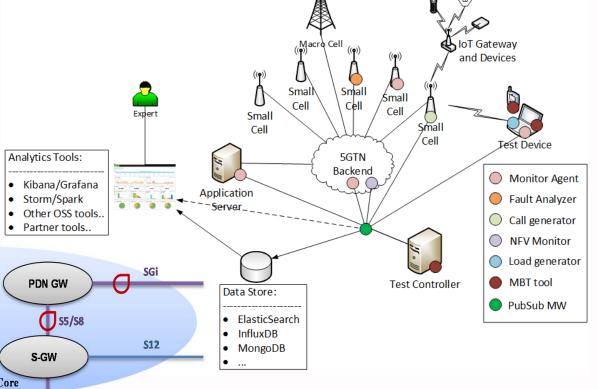
<u>Teemu Kanstrén</u>, Jussi Liikka, Jukka Mäkelä, Pekka Ruuska VTT Technical Research Center of Finland

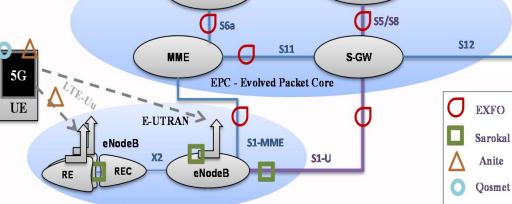






### **5GTN Overview**





HSS







### Internet of Things, Big Data, Machine Learning, ...

- Testing IoT?
  - Simple devices, complex combinations
- Using data analytics & machine learning for test analysis?
  - IoT, distributed systems, etc. produce "big data". How do we evaluate it?
- Using test generation to generate data for analysis?
  - Getting the required data to develop algorithms, tools, features can be a challenge















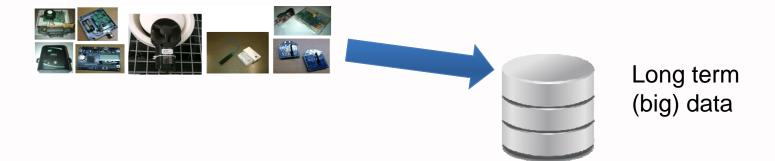










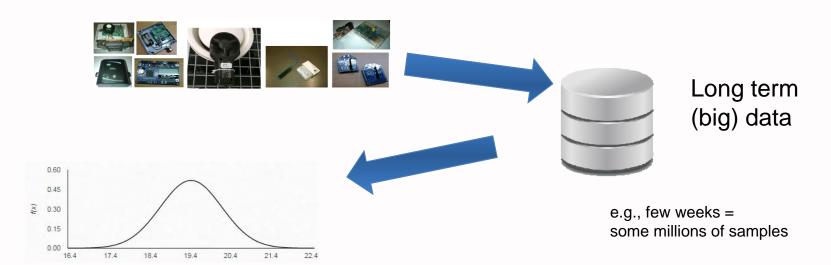


e.g., few weeks = some millions of samples







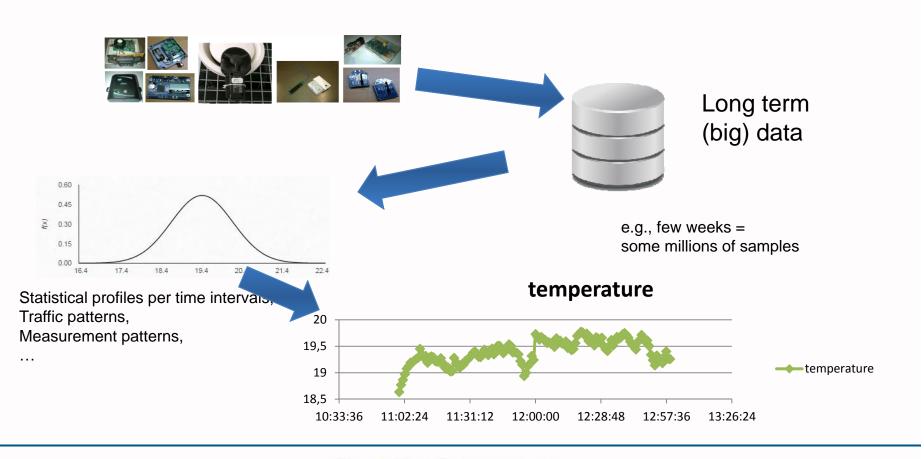


Statistical profiles per time intervals, Traffic patterns, Measurement patterns,

. . .







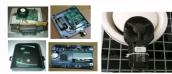








Similarly, e.g., network data profiles (packet sizes, frequencies, ...)

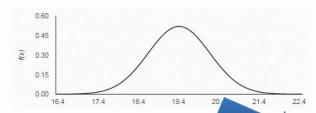






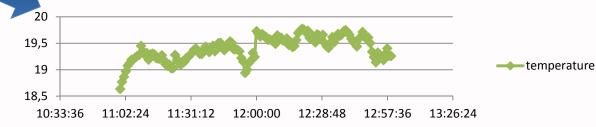
temperature

Long term (big) data



e.g., few weeks = some millions of samples

Statistical profiles per time intervals, Traffic patterns, Measurement patterns,



User Conference on Advanced Automated Testing

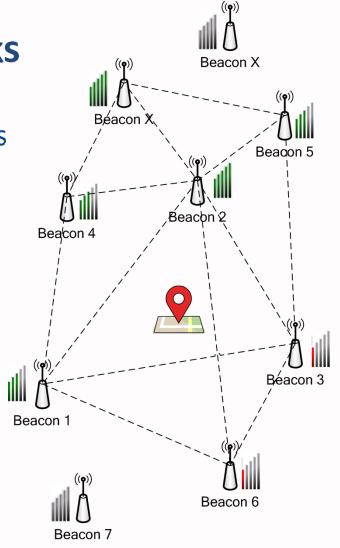






## **More Complex Sensor Networks**

- For example, indoor location via beacons
  - Several beacons evaluated at any time
  - Location = value function of sensors
- Need to consider sensor relations
  - Signal strength vs others
  - Environmental factors
- Or relation to sensor itself
  - Smoothing curve vs hop
  - Moving: allowed, expected, service functions, ...
  - ..











## **Analysing the Data (Test Oracle?)**

 With thousands of parameters and up to millions or billions of data values from the tests...

- What happened?
- What matters?

86,136,185,3246,141,11,244,27,218,86,2,20,0,0,3,638,0,1381,4000,91,2738,82,24,0,214,545,0,331,462,699 30.1136.185.3273.1146.11.244.27.194.101.2.19.0.0.15.694.0.1284.3871.64.2789.57.17.0.198.576.0.306.316 19.1107.185,3176,1117,11,244,27,314,117,2,19,0,0,1,696,0,1221,3701,91,2695,69,24,0,142,563,0,426,502,4 47,999,185,3138,1005,11,244,27,591,207,2,20,0,0,1,682,0,1172,3668,55,2668,68,12,0,163,653,0,453,451,65 70.69.185.3096.69.11.244.27.521.189.2.19.0.0.7.687.0.1393.4097.54.2820.106.26.0.338.807.0.755.1600.345 53.622.185.3118.637.11.244.27.665.178.2.20.0.0.2.723.0.1333.4169.52.2834.114.59.0.224.640.0.612.932.22 35,508,185,3218,513,11,244,27,258,101,2,20,0,0,2,612,0,1267,3934,98,2793,81,18,0,118,621,0,299,412,889 30, 1238, 185, 3027, 1253, 11, 244, 27, 237, 89, 2, 19, 0, 0, 5, 604, 0, 1336, 3969, 95, 2737, 111, 37, 0, 166, 580, 0, 333, 368, 28.2883.185,3098,2903.11,244.27,238.100.2,20.0.0,3,719.0,1209,3514.40,2654.92.41.0,350.821.0,890.1010 94,375,185,3206,384,11,244,27,213,78,2,20,0,0,2,684,0,1335,3898,50,2739,103,42,0,353,637,0,352,804,176 52,1056,185,3329,1052,11,244,27,252,101,2,19,0,0,4,710,0,1192,3783,28,2674,101,22,0,222,705,0,600,108 20,357,185,3213,357,11,244,27,590,196,2,19,0,0,1,698,0,1214,3792,63,2722,69,8,0,131,621,0,394,759,1356 39,58,185,3305,58,11,244,27,691,251,2,20,0,0,3,681,0,1398,3969,111,2939,142,82,0,775,1898,0,2509,4054 50.60.185.3398.65.11.244.27.458.169.2.20.0.0.1.579.0.1317.3902.105.2808.161.85.0.1028.2106.0.2779.386 92,70,185,3450,70,11,244,27,641,270,2,19,0,0,1,588,0,1259,3876,126,2886,261,168,0,1154,1972,0,2655,39 37,26,185,3161,34,11,244,27,504,231,2,20,0,0,2,718,0,1423,3983,70,2942,222,143,0,556,1122,0,1335,2901 70.114.185.3270.124.11.244.27.401.134.2.20.0.0.7.657.0.1364.4019.52.2780.103.6.0.132.651.0.387.815.11\$ 90,101,185,3068,111,11,244,27,256,111,2,19,0,0,1,601,0,1363,4190,161,2818,150,25,0,132,714,0,554,1031





- Automated analysis of large test data sets
- E.g.,
  - performance spikes, drops,
  - Important data elements,
  - causes...







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- A simple case: statistical profiling and automated outlier detection









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### **Profiling via test scenarios**

- Labeled datasets
  - IoT = L1
  - Video = L2
  - Idle = L3

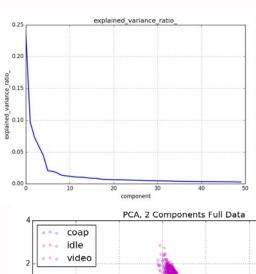
loT (L1)	Video (L2)	Idle (L3)
55659	30025	64
55985	28009	1
55604	28021	5
55303	35789	0
55744	28950	0
		variable data
Single variable data from thousands		

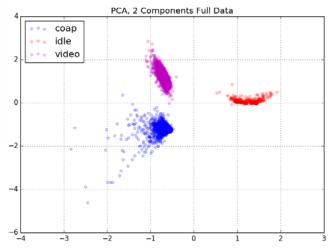
- What are the impacts?
- Inputs to learning algorithms,
- To find labeled targets (categories)

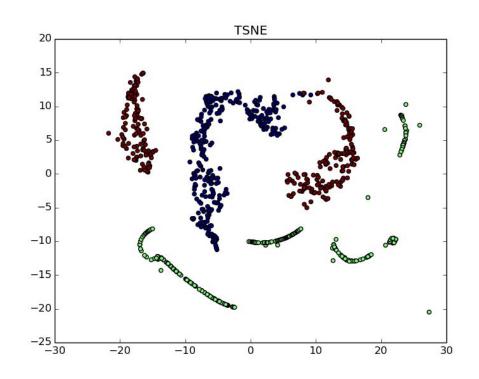




## **Example: Visualizations with PCA, TSNE, ...**





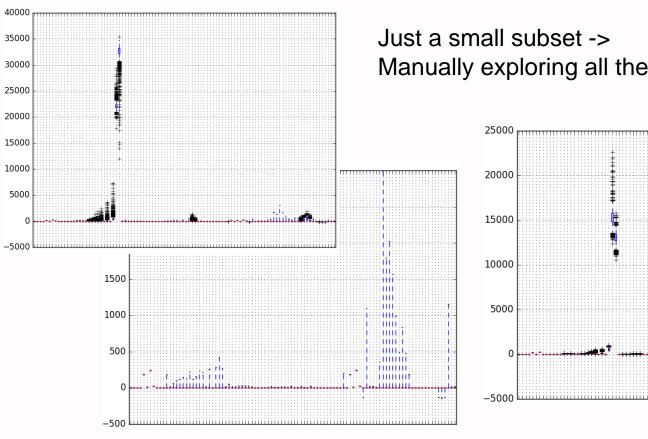




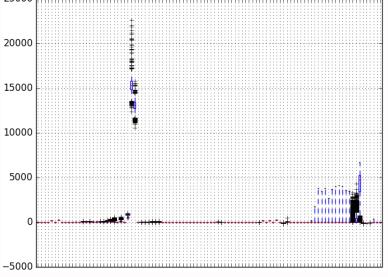




## Finding the relevant features



Manually exploring all the data is hard

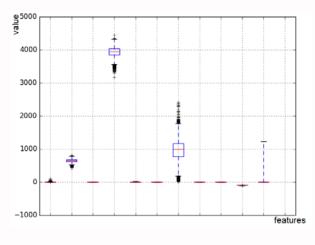


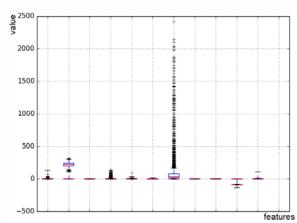
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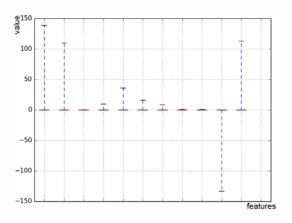


## Finding important features: Automation





e.g., Recursive feature elimination (RFE) algorithm







#### What for?

- Test devices, networks, services against new types of input profiles
- Provide new inputs and datasets for development
- Analyze and evaluate results, identify interesting properties and characteristics
  - Profiles for users, devices, sensors, environmental factors, ...
  - Performance statistics





#### **Conclusions**

- IoT properties & challenges
  - Simple sensors form complex combinations
  - Combining with big data, data analytics & machine learning
- Testing here
  - Exploring the impact of parameters, profiles, configurations, performance results, functionality
  - Providing inputs to features, algorithms, optimizations, ...

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