



STQ Workshop

Utilizing Timeslice KPIs For AI/ML-Based Voice Service Automation

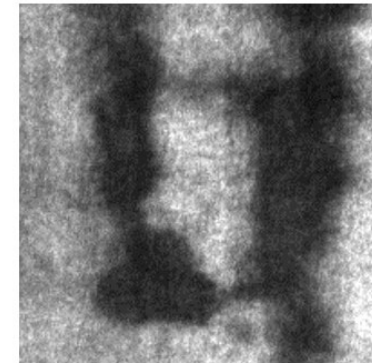
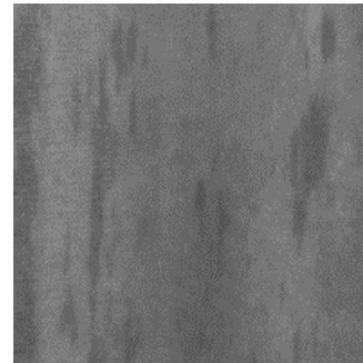
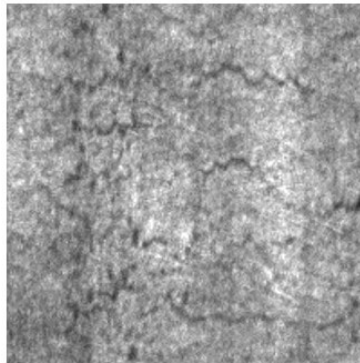
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22/11/2022



30 years ago...

- My first job offer as student helper: QA at Parsytec
- Parsytec created systems for optical surface inspection using pattern recognition
- High-speed detection of defects on metal and paper surfaces during production



Source: [ISRA Vision](#)

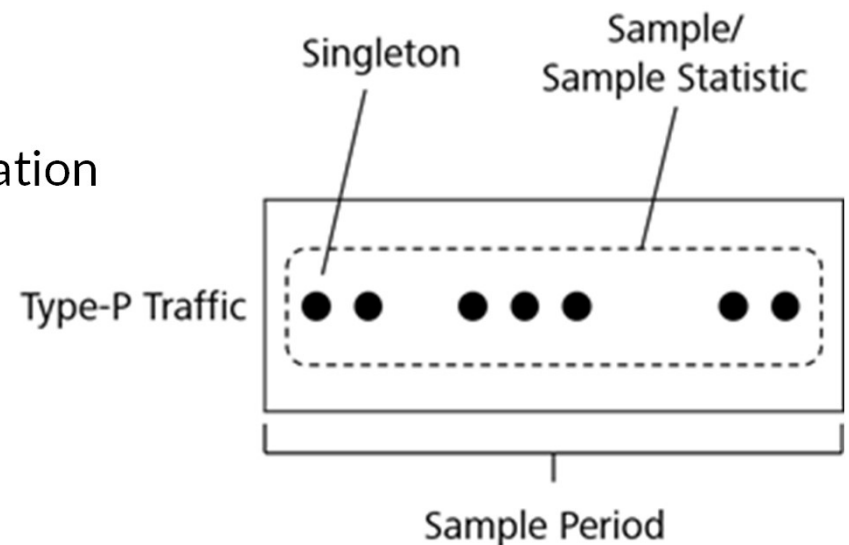
- We'll get back to this... 😊

Fast forward to 2021: ETSI STQ filling the media KPI gap

- Voice service performance monitoring traditionally focused on signaling
- IETF RFC 6076 is the de facto standard for VoIP signaling, defining e.g.
 - SEER – Session Establishment Effectiveness Ratio (comparable to NER)
 - SRD – Session Request Delay (comparable to Post-Dialing Delay, PDD)
- In 2018 ETSI STQ identified a gap in media KPIs and opened a new work item:
“Timeslicing KPIs for RTP based speech transmission”
- The resulting technical report TR 103 639 was published in March 2021

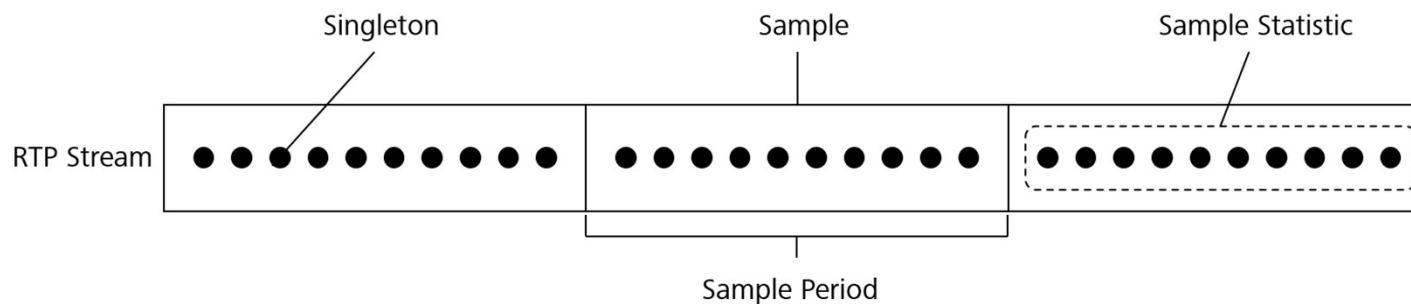
ETSI TR 103639 Basics

- TR 103639 provides a framework for defining timeslice media metrics and KPIs
- Uses concepts from the well-known IPPM (IP Performance Measurement) working group of the IETF
- IPPM Terminology
 - Type-P traffic: type of observed packets
 - Singleton metric: corresponds to single observation
 - Sample: collection of singleton measurements
 - Sample statistic: an aggregation of singleton measurements over a sample period

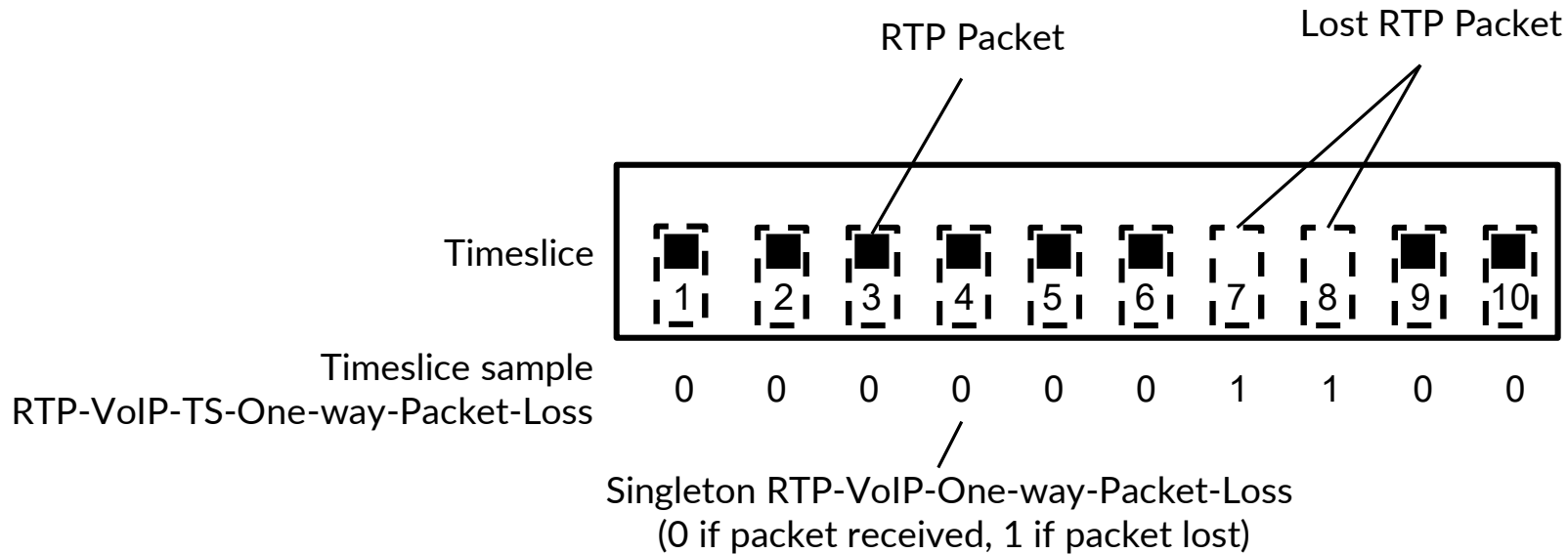


ETSI TR 103639 Terminology

- ETSI TR 103639 applies IPPM concepts with the following provisions:
 - Singleton metrics are of type "RTP-VoIP", i.e. they refer to packet streams transporting speech using RTP over UDP
 - Sample periods (timeslices) are of fixed duration and shorter than a typical call
 - All RTP packets in a sample period contribute to singleton metrics to fully characterize the timeslice
 - Samples are consecutive and continuous for the duration of each RTP stream



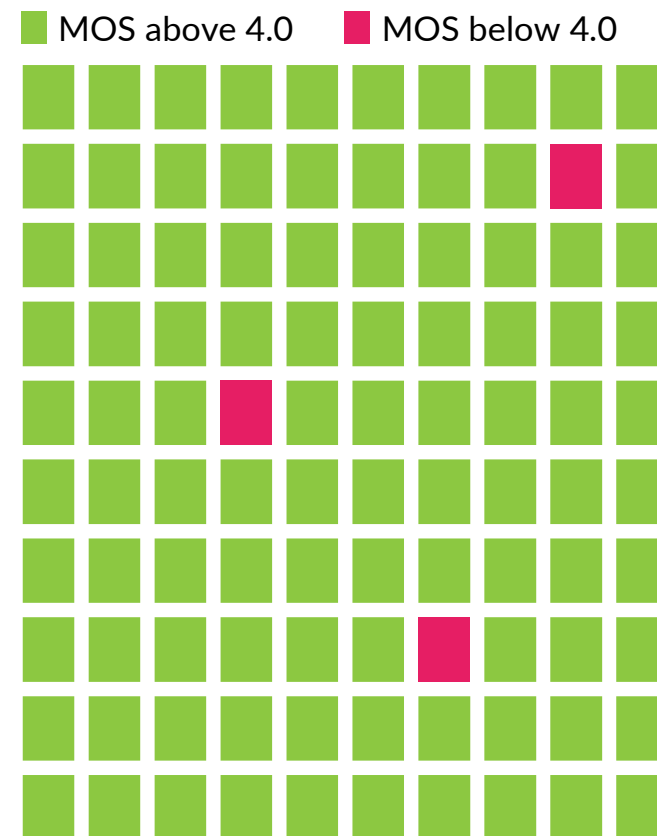
Example timeslice metric: one-way packet loss



- Example timeslice statistics based on this sample definition:
 - RTP-VoIP-One-way-Packet-Loss-Count
 - RTP-VoIP-One-way-Packet-Loss-Ratio
 - RTP-VoIP-One-way-Burst-Loss-Event-N

Annex A KPI example: Good Minute Ratio

- A 'good minute' is a set of timeslices representing one minute, where all time slices have MOS > 4.0.
- $GMR = \# \text{ good minutes} / \text{total number of minutes}$
- The set of time slices can be based on any group of media streams, e.g.
 - One direction of a call
 - All calls in the past 15 min to Bratislava
 - Incoming RTP on interconnection in the past year
- Example calculation: $GMR = 97/100 = 97\%$



Using timeslice KPIs for machine learning

- Prediction of time-series data is rather straight-forward
 - Timeslice KPIs can be provided periodically, e.g. for 5-, 10-, 15-minute intervals
 - Technically there is no difference to other time-series data
 - Prediction techniques are well-known
- Need to focus on standardisation/productization:
 - Define/select timeslice KPIs of interest, e.g. based on TR 103639 Annex A
 - Standardize data feed interfaces, e.g. as data source for NWDAF (3GPP TS 29.520) or comparable functions in other network segments
 - Develop respective ML-models making use of timeslice KPIs

Anomaly detection



Fed AWS Lookout for Metrics with signaling & timeslice media KPIs for a large set of SIP trunks



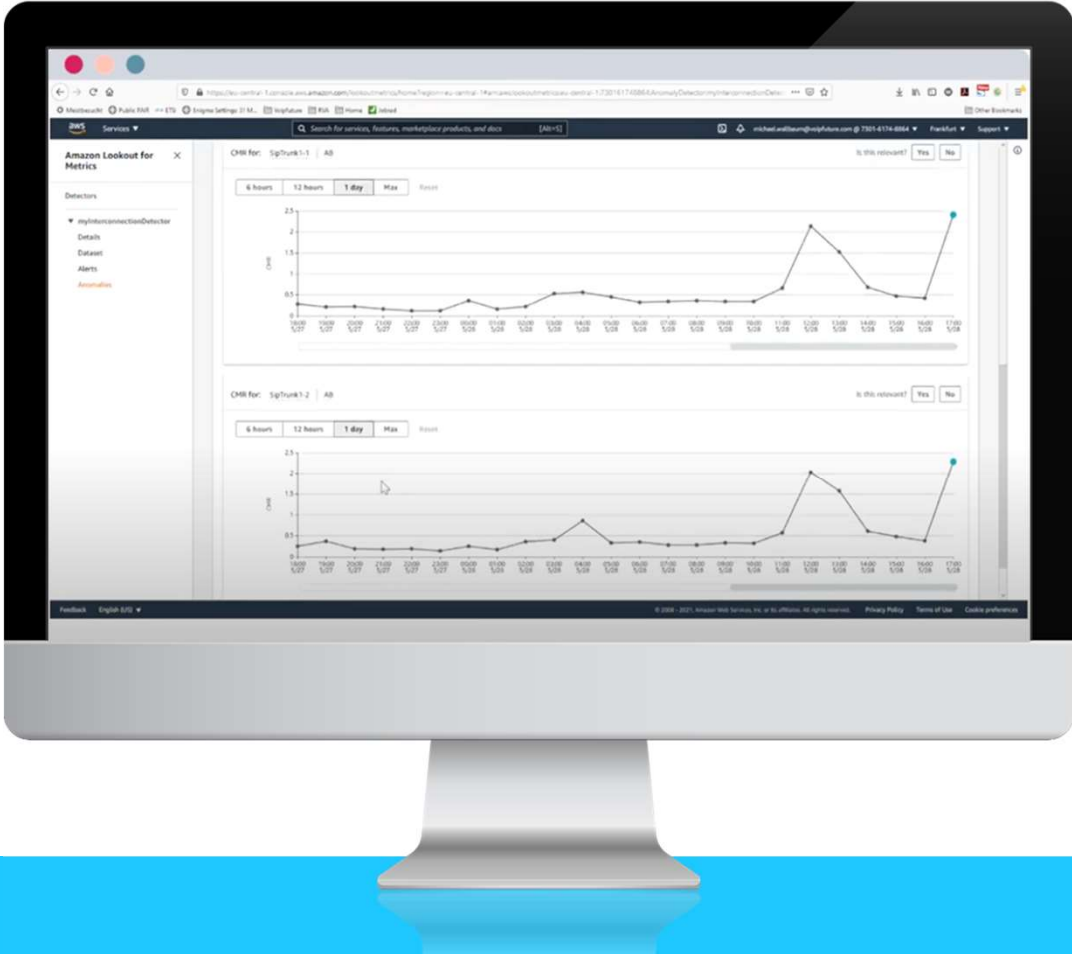
Lookout automatically detected anomalies in CMR (Annex A)



Lookout identified the affected trunks and media direction



Integrate to raise adaptive alarms



Adaptive alarming

- NOC dashboard created for a Voipfuture customer
- Selected KPIs used for anomaly detection
 - SER/NER
 - ACD
 - Proprietary signaling KPI
 - GMR (Annex A)



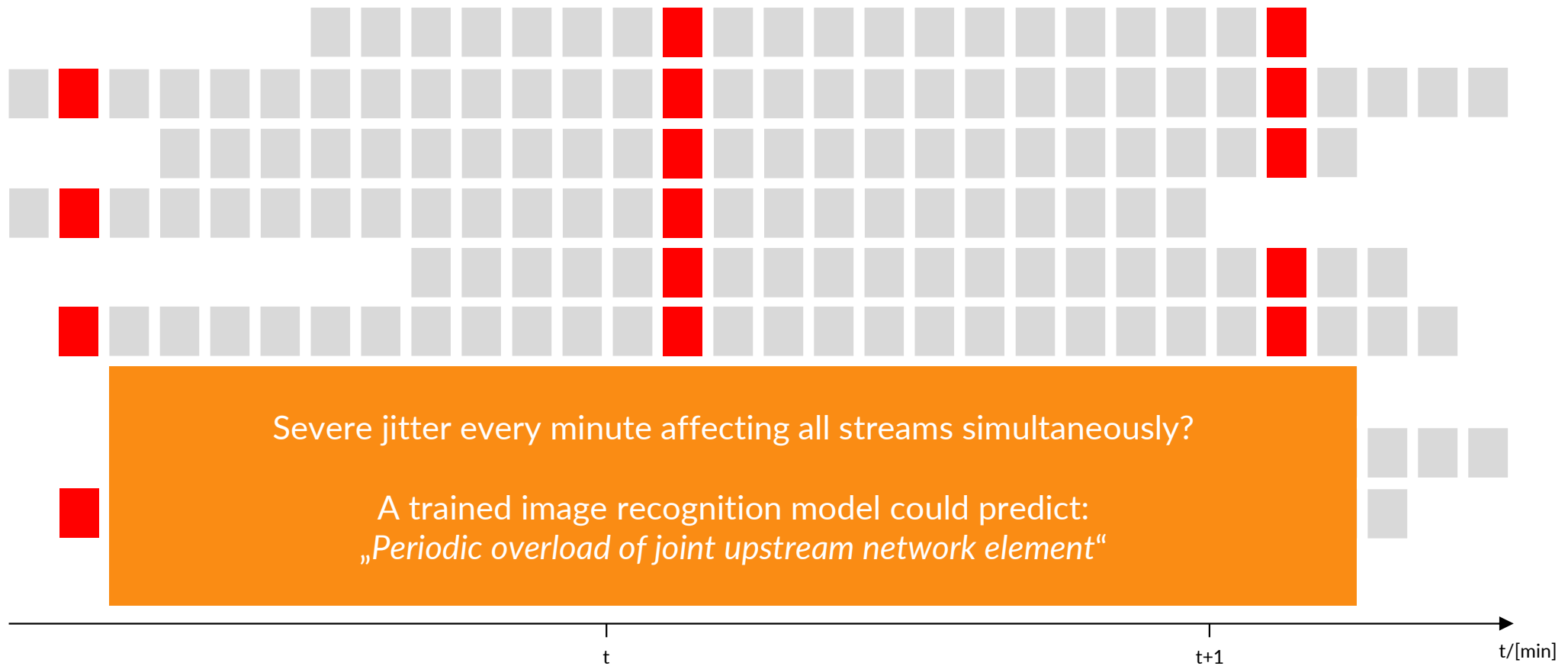
Coming back to detection of surface defects

- Passively monitoring live VoIP traffic in an operator's network generates millions and billions of timeslices
 - Now interpret each timeslice as the pixel of an image
 - Now assume timeslice metrics define the „color space“
- Get the picture? Idea is to apply image classification techniques from machine learning to detect impairment patterns affecting the RTP traffic
- Next examples assume passive monitoring system with 5s timeslice duration

First example image

RTP traffic incoming to A-SBC

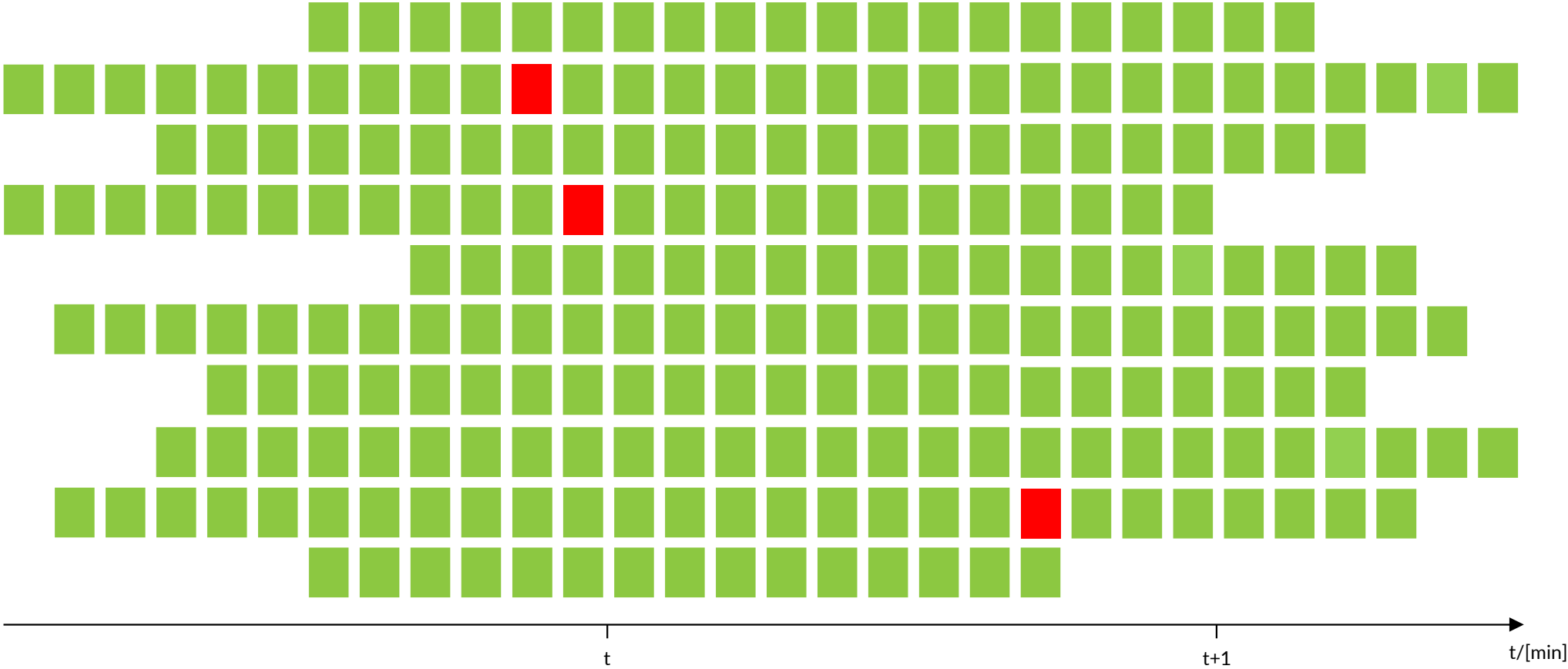
■ IPDV ≤ 40ms ■ IPDV > 40ms



Second example: overall „image“ in terms of user experience

RTP traffic between two I-SBCs

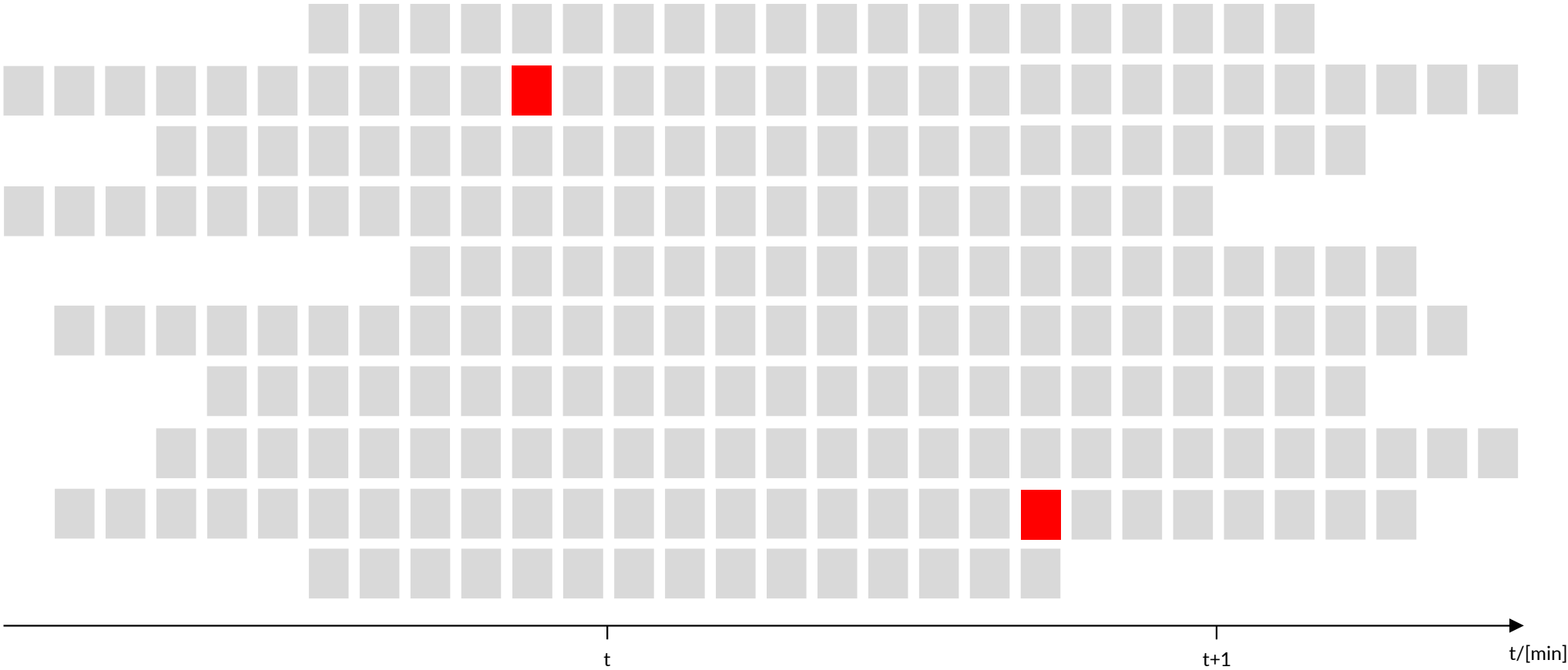
■ MOS above 4.0 ■ MOS below 4.0



Jitter „color layer“

RTP traffic between two I-SBCs

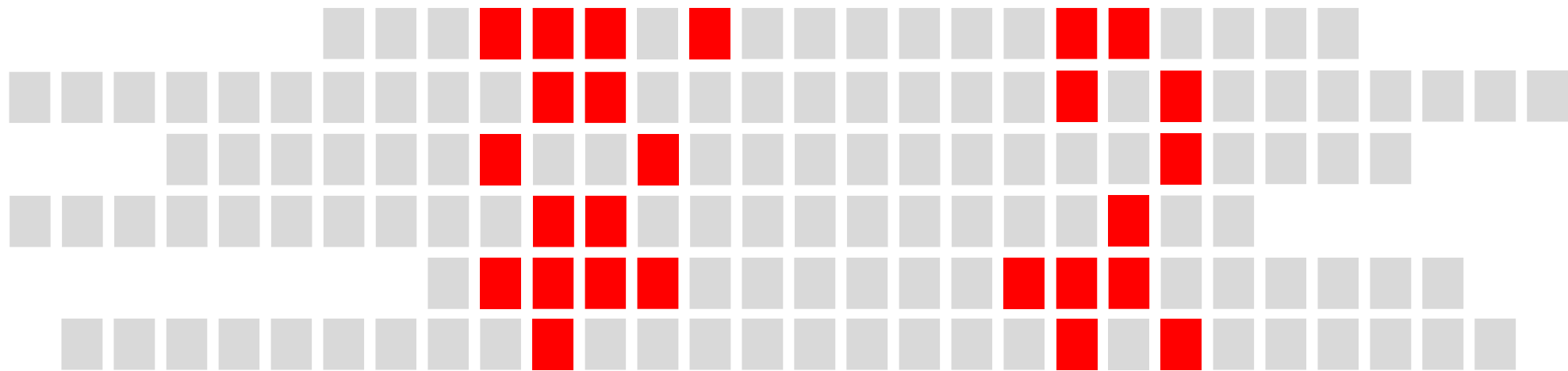
■ IPDV ≤ 40ms ■ IPDV > 40ms



Burst loss „color layer“

RTP traffic between two I-SBCs

■ No burst loss ■ Burst loss



No jitter, but lots of semi-synchronous burst loss affecting all streams on a link?

A trained image recognition model could predict:
„Hardware issue affecting communication link“

t

t+1

t/[min]

Summary and outlook

- ETSI TR 103639 timeslice KPIs provide a data foundation for AI/ML-based prediction of media traffic
- Specifically Annex A KPIs are actively being used in the industry, incl. for AI/ML-based anomaly detection and adaptive alarming
- Application of image analysis techniques to timeslices
 - More complex than use of timeslice KPIs, however approach retains temporal detail
 - Potential for faster and more accurate/detailed predictions regarding root causes
 - More research needed...

Thank you!