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# Multi-Tech Sensing In-Field Demonstration on Terrestrial Optical Data Network

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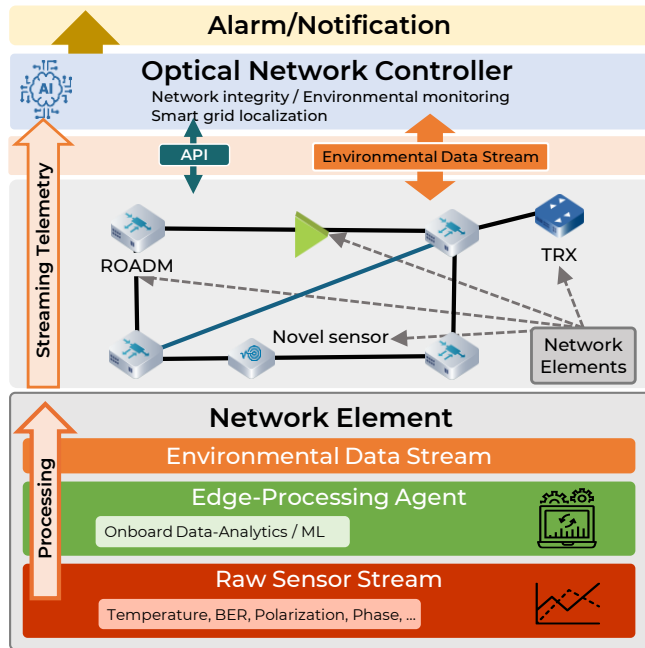
Transforming optical transport networks into a distributed sensing grid is a promising pathway to enhance infrastructure resilience and environmental awareness. To achieve widespread scalability without disrupting data traffic, operators can exploit the synergy between different sensing techniques naturally coexisting on the same infrastructure. In this paper, we present an in-field demonstration of a multi-technique sensing framework operating on live terrestrial optical networks, combining interferometric phase sensing and State of Polarization (SOP) monitoring. We first propose and validate an unsupervised anomaly detection pipeline based on autoencoders, applied to a 38 km regional link connecting Ascoli Piceno and Teramo (Italy) and show some examples of detected events. This framework proves capable of automatically identifying signal anomalies across different physical observables without relying on manual labeling. Subsequently, we move to a metropolitan urban scenario in the city of Turin and assess a multi-tech sensing analysis using coherent transceivers and state-of-polarization metrics obtained from polarimeters and cost-effective polarization beam splitters-based device. We highlight the correlation between specific sensing metrics and standard network telemetry data, such as Bit Error Rate (BER) and temperature, and validate the system capability to detect mechanical perturbations originated by human activity along the cable or by controlled experiments using a robotic arm. These results confirm the feasibility of aggregating heterogeneous data sources into a unified sensing plane to enable pervasive infrastructure supervision.

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## 1. INTRODUCTION

Optical fiber networks constitute the critical backbone of modern digital society, supporting global communication and data exchange. Due to the growth in bandwidth demand, these infrastructures have become pervasive, densely covering large geographical areas with thousands of kilometers of optical cables. In this context, the integrity and reliability of the physical layer are becoming a priority for operators and institutions [1]. However, the infrastructure's pervasiveness offers a unique opportunity: the transformation of the communication infrastructure into a vast, distributed environmental sensing grid [2]. This paradigm shift envisions the optical network not only as a data transport pipe, but as a multi-service, software-defined entity capable of

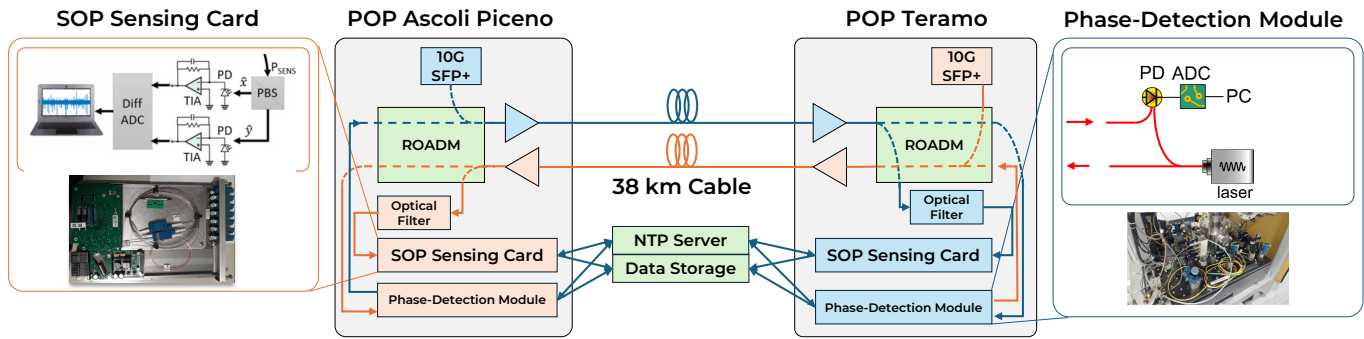
monitoring both its own health and the surrounding environment. While recent studies have been focused on field trials in the undersea environment [3–6], there is a huge interest in the investigation of sensing techniques in the terrestrial scenario [7–13]. In this case, typical meshed or ring topologies enable a smart-grid approach leveraging on large cross-correlation of data coming from diverse network sections. Moreover, the easier accessibility of the add-drop nodes to install additional sensors enables scalability to large geographical areas. As depicted in the architecture in Fig. 1, modern optical networks can exploit the streaming telemetry capabilities of deployed Network Elements (NEs) to retrieve environmental information [14–16]. Unlike dedicated sensing solutions that often require dark fibers, this



**Fig. 1.** Network sensing architecture: each network element produces telemetry data that can be processed in the edge and sent to a centralized controller for cross-correlation.

approach leverages the active components of the network—such as Reconfigurable Optical Add-Drop Multiplexers (ROADMs), amplifiers, and transceivers which are naturally equipped with embedded sensors [17]. These devices continuously monitor physical quantities like temperature, optical power, and Bit Error Rate (BER) for data network control and management purposes. Since these parameters are intrinsically sensitive to external stressors, they can be re-purposed to infer environmental conditions. Furthermore, to manage the vast amount of generated data, intelligent edge processing agents can be implemented directly at the NE level or within the node controller. Such agent may send to the controller only a portion of the acquired signal when an event is detected or even just a timestamped notification to the centralized Optical Network Controller (ONC). This allows for filtering significant data and transmitting only relevant features to the ONC, effectively enabling a scalable smart grid for wide-area monitoring. The ONC can take actions based on the sensing information received from network elements. For instance, if an earthquake is detected the ONC can send an early warning alarm to the local seismic monitoring agency. Alternatively, when events that threaten the network infrastructure integrity are detected (such as fiber shaking or bending), the ONC can perform proactive network restoration actions such as establishing a backup lightpath to avoid service disruption. To enhance the sensitivity of this architectural framework beyond standard telemetry, specific sensing techniques based on the analysis of light propagation are required. While Distributed Acoustic Sensing (DAS) offers high spatial resolution, its cost and need for dedicated fibers and spectral resources limit its deployment in active networks [18–20]. A more scalable alternative relies on monitoring the State of Polarization (SOP) and the optical phase of the data signals themselves. Optical fibers act as sensitive transducers; mechanical stresses, vibrations, and thermal variations induce birefringence changes and phase shifts that can be detected at the receiver [21]. Interferometric optical phase

sensing loops back an optical carrier to detect the phase shift caused by the small path variations due to fiber deformation [12, 22]. However, this typically requires expensive ultra-stable alien lasers [22] as absolute phase reference. SOP monitoring instead does not require absolute phase references as it measures the differential phase between two orthogonal polarizations. Hence, this technique is particularly promising for mass deployment as it can be performed on standard intensity-modulated direct-detected (IMDD) signals, still widespread in the deployed networks, and using cost-effective detection devices like Polarization Beam Splitters (PBS) [23–26]. Such enhanced scalability comes with a cost. Unlike DAS, both interferometric phase and SOP monitoring techniques are not inherently spatially resolved as they provide the *integrated* effect over the fiber length. In this scenario, localization of an event along the fiber length is possible when the sensing technique is duplicated in the opposite propagation direction. One can then cross-correlate the two acquisitions to estimate the time delay between a perturbation caused by the same event and translate it to a position along the cable [23, 27]. As dictated by the Cramer-Rao lower bound [28], the accuracy of this localization scheme scales down with the event’s signature signal-to-noise ratio (SNR), with its bandwidth and as the time synchronization between the two opposite directions is less precise. While this solution may enable precise event localization concentrated in a specific point along the fiber, events such as earthquakes may trigger detection on multiple monitored fiber segments, thus enabling a coarse earthquake origin localization by triangulation using a smart grid approach. In this framework, smart grid localization does not rely on intrinsic spatial resolution of a single sensing technique, but emerges from the network-wide aggregation of measurements. By leveraging the meshed topology of terrestrial optical networks and the availability of synchronized telemetry streams, external perturbations can be localized through correlation of their signatures across multiple links and nodes. This enables a form of distributed inference, where the network itself acts as a sensing grid and the localization accuracy depends on the spatial diversity and redundancy of the monitored paths. In this paper, which extends our work presented at ECOC 2025 [29], we take on the in-field implementation of this multi-technique sensing vision, with priority on the event detection. We shift the focus from single-parameter monitoring to the aggregated use of diverse physical quantities collected via a unified telemetry plane. The paper is organized into two main experimental sections. Section 2 describes a long-distance testbed connecting Ascoli Piceno and Teramo (Italy). Here, we analyze data from a live terrestrial network where interferometric phase sensing and PBS-based SOP monitoring coexist with data traffic. We propose an unsupervised anomaly detection framework capable to identify signal anomalies automatically and that can be potentially integrated in the ONC or in the edge processing agent of relevant network nodes. Section 3 focuses on an urban scenario in the city of Turin. Utilizing a metropolitan fiber ring owned by GARR carrying live traffic, we compare acquisitions using different sensing techniques, including coherent transceivers, polarimeters and a simplified PBS-based SOP monitoring device. In that section, we explicitly showcase the correlation between sensing-specific metrics (SOP) and standard network telemetry data (BER, Temperature), demonstrating the potential of the proposed sensing architecture. Finally, we report on controlled experiments using a robotic arm to simulate mechanical perturbations, validating the system’s ability to detect and characterize complex urban anomalies [30].



**Fig. 2.** Schematic overview of the experimental multi-technique sensing setup. The setup integrates the SOP sensing card (left) and the ultra-stable laser phase-detection module (right) on the same fiber cable (center).

## 2. ANOMALY DETECTION ON PHASE AND SOP SENSING TESTBED

In this section we focus on the experimental activities carried out in a section of the regional network between the cities of Ascoli Piceno and Teramo owned by Open Fiber. The proposed anomaly detection framework operates on data acquired from a multi-technique fiber sensing infrastructure deployed on a live terrestrial optical link. Two complementary sensing modalities are simultaneously employed along the same fiber cable: interferometric phase sensing and state-of-polarization (SOP) monitoring. While both techniques probe mechanical perturbations acting on the fiber, they rely on different physical observables and exhibit distinct sensitivity and scalability characteristics.

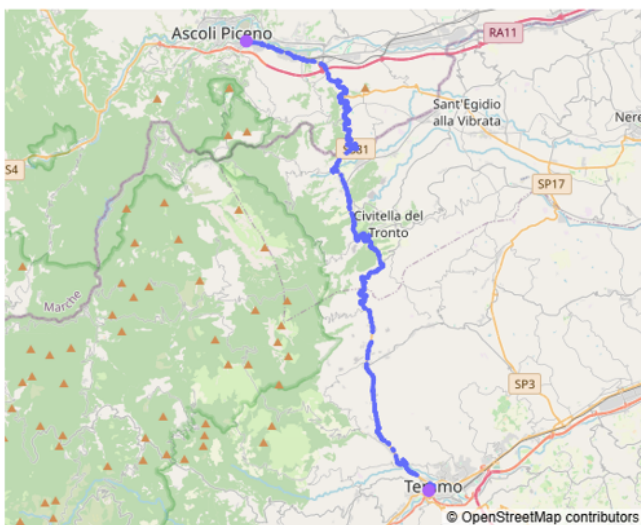
### A. Interferometric Coherent Phase Sensing

Interferometric phase sensing is based on the detection of optical phase fluctuations accumulated by a coherent optical carrier propagating along the fiber. External perturbations, such as seismic waves or anthropogenic vibrations, induce local strain and refractive index changes, resulting in measurable variations of the optical phase. By comparing the phase of the received signal with a stable local reference, it is possible to retrieve a

time-resolved measurement of the integrated deformation experienced by the fiber. In this work, the probe for interferometric sensing is an ultra-stable laser (shown in the right panel of Fig. 2), following the approach described in [12, 22]. The system employs coherent detection with sub-radian phase resolution and sensitivity to low-amplitude seismic signals over long distances. As previously mentioned, unlike conventional DAS, this technique is not able to pinpoint deformations along the cable. However, it features lower system complexity, data throughput and cost, while typically operating on alien wavelengths in coexistence with data traffic also on ultra-long-haul links and still providing good sensitivity and frequency response. Similar to polarization-based sensing, the phase observable represents the line integral of fiber strain along the optical path and is therefore intrinsically non-localized. Nonetheless, its high signal-to-noise ratio makes it a reliable reference for validating detections obtained through complementary sensing approaches. The ADC can sample the photodiode output up to a frequency of 10 KHz. In the following however we employ such signals downsampled at 100 Hz [22].

### B. PBS-based State Of Polarization Sensing

State of polarization sensing exploits the sensitivity of light polarization to mechanical perturbations acting on the fiber. External stress and bending induce birefringence variations, leading to measurable SOP fluctuations even in the absence of coherent detection. In the proposed setup (illustrated in the left panel of Fig. 2), SOP monitoring is realized using standard IMDD optical channels, such as optical supervisory channels, detected with a PBS and direct photodetection-based devices developed by the Italian optical transport solutions vendor SM-Optics [23, 31]. Each SOP sensing unit splits the received optical signal into two orthogonal polarization components using a PBS, which are independently detected and digitized using two photodetectors and DC coupled ADCs with 12 bit resolution and 100 Hz sampling frequency to retain slow variations. The digitized orthogonal components are then normalized with respect to the total received instantaneous power detected using a third photodiode and ADC in order to compensate possible small total power variations. The two normalized orthogonal components are then subtracted to cancel out the common mode noise and enhance the fluctuations caused by dynamic polarization changes induced by environmental and anthropogenic disturbances. To date, the SOP signal is acquired in 12 hours long sessions interrupted by a gap of about 2 minutes which are transferred daily to a remote data storage server. At 100 Hz sampling frequency



**Fig. 3.** Geographical layout of the Open Fiber terrestrial optical cable and location of the deployed sensing equipment.

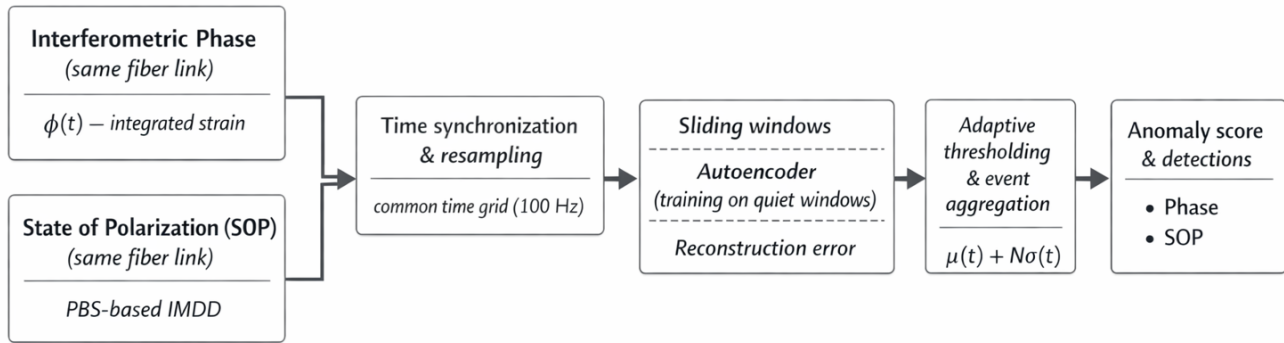


Fig. 4. Overview of the unsupervised multi-technique anomaly detection pipeline.

each session takes around 90 megabytes of storage. We envision the implementation of a real-time streaming service in the near future as step towards the sensing network architecture described in Fig.1. The resulting measurement thus provides partial polarization information with respect to full Stokes vectors acquired using a polarimeter and can be interpreted as the evolution of  $S_1$  in the receiver polarization frame. This approach may exhibit lower sensitivity compared to interferometric phase sensing as this depends on the average SOP at the receiver, which nonetheless slowly changes in time, for example, due to the thermal variations. In particular, the sensitivity worsens as the average receiver polarization state fades from linearly polarized towards circularly polarized states. However, it offers significant advantages in terms of scalability, cost, and ease of integration into existing network infrastructure [29], as it does not require signal loop-back in the backward direction. This is a clear advantage with respect to DAS which observes the Rayleigh back-scattered signal thus limiting the deployment in presence of an optical amplifier's optical isolators. With respect to interferometric approach, full forward sensing avoid wasting spectral slots in the return fiber. The SOP sensing device can operate continuously on live traffic fibers and can be deployed at multiple network points of presence. These characteristics makes it particularly attractive for large-scale, long-term sensing applications and motivate its use as a complementary observable within the proposed multi-tech anomaly detection framework.

### C. The Open Fiber Regional Cable Testbed

Fig. 3 illustrates the geographical layout of the monitored terrestrial optical link used in this study. The experimental setup is deployed on a production regional network operated by Open Fiber and consists of a bidirectional fiber cable connecting two network points of presence (POPs) located in central Italy. The two POPs are separated by approximately 39 km of optical fiber cable. Regarding optical phase technique, the laser source is injected in one fiber within the cable at the Teramo POP, it is then looped back on a different fiber of the same cable at the Ascoli POP and received back at Teramo. The oppositely propagated phase signal was not available for this study. Regarding the SOP technique, a 10 Gbps commercial transceiver is used as signal source launched in two different fibers of the same cable at each POP with -2 dBm power and received by the SM-Optics SOP sensing unit at the opposite POP. Both source signals are propagated within spectral slots made available by Open Fiber together with standard data signals partially loading the C-Band. Unfortunately, we do not have access to the details of the overall spectral loading in each fiber. However, the infrastructure

owner has not reported any significant degradation of the standard data signal transmission performance. A thorough study of the coexistence of such sensing signals and data channels is out of the scope of this paper. However, this scenario can be treated as the coexistence between coherent data channels and polarized continuous waves or IMDD data signals. Such problem is commonly addressed imposing proper guard-band between data and sensing signals [32]. The fiber route is predominantly buried inside underground conduits following road infrastructure, with the exception of a short aerial segment of approximately 100 m. This mixed deployment exposes the fiber to realistic environmental and anthropogenic perturbations, representative of typical terrestrial network conditions. As outlined in Fig.2, both the interferometry and SOP detectors store the acquisitions on a remote server provided by Open Fiber and their absolute time reference is obtained from the same Network Time Protocol (NTP) server.

### D. Anomaly Detection

Fiber sensing systems deployed for long-term monitoring are continuously exposed to a wide range of perturbations, originating from both environmental forcing and human activities. Beyond seismic signals, the acquired data routinely include transient signatures associated with vehicle traffic, maintenance operations or wind-induced oscillations of aerial cable spans or surrounding vegetation, among the others. The automatic detection of such events is a key requirement for infrastructural supervision and situational awareness, particularly in scenarios where continuous and quasi-real-time alerting is desired. Since the diversity and unpredictability of signal morphologies make supervised approaches impractical due to the need to collect large labeled training datasets, we investigate the use of unsupervised learning techniques that identify deviations from background behavior without relying on labeled datasets. Specifically, in this work we adopt and extend an autoencoder-based anomaly detection framework to a multi-technique distributed sensing scenario, jointly addressing phase and SOP observables.

#### D.1. Signal Pre-processing and Multi-Technique Alignment

Raw phase- and SOP-derived time series are first pre-processed to ensure consistency across sensing techniques. The two sensing systems are time-aligned using NTP which delivers about 10 ms of accuracy. This level of accuracy is sufficient to compare the anomaly detection outcomes and assess that a perturbation detected using each method corresponds to the same event in time. Stricter synchronization may be instead required to localize the events across the fiber with a bidirectional, time delay estimation-

based sensing setup [28], which is out of the scope of this paper. Measurements are then interpolated onto a common temporal grid and downsampled to a sampling frequency of 100 Hz. This value represents a compromise between computational efficiency, data size and the ability to resolve fast transients, such as impulsive anthropogenic events or short-lived environmental vibrations. This harmonization step enables the use of a unified machine learning architecture and detection logic for both sensing techniques, facilitating a direct comparison of their anomaly statistics and temporal response to the same physical perturbations. The simultaneous availability of interferometric phase and SOP measurements provides a redundant and synergistic view of fiber perturbations, increasing detection robustness and supporting cross-validation of automatically detected events.

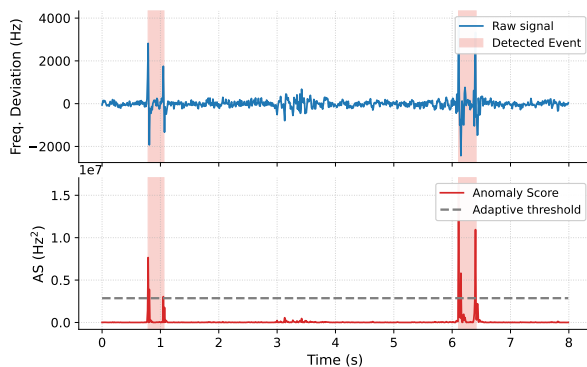
### D.2. Autoencoder-based Anomaly Detection Pipeline

To detect anomalies within the continuous data streams without relying on manual labeling, we employ an unsupervised methodology centered on a dense autoencoder. A high-level schematic of the workflow is presented in Fig. 4. Following pre-processing, the time series are segmented via a sliding window with length  $w$  and unit step. A crucial aspect of our approach is the definition of the training set: to ensure the model exclusively learns the stationary background noise rather than transient disturbances, we implement a data-driven filtering strategy. All signal windows are sorted based on their Root-Mean-Square (RMS) amplitude, and only the lowest 30%—representing the quietest periods—are utilized for the training phase. This enforces a robust definition of "normal" behavior. The neural network features a symmetric, fully connected architecture constrained by a low-dimensional bottleneck. Training is performed by minimizing the Mean Squared Error (MSE) between the input window and its reconstruction. Once the model weights are frozen, the autoencoder is applied in inference mode across the entire dataset to identify deviations from the learned background.

### D.3. Adaptive Thresholding and Event Post-Processing

During the inference stage, the reconstruction error is evaluated for every window. Because the sliding window approach results in overlaps where a single time sample is included in multiple windows, we derive a continuous, sample-wise Anomaly Score (AS). This is calculated by averaging the reconstruction errors across all  $k$  windows that cover a specific time instant  $t$ :

$$AS(t) = \frac{1}{k} \sum_{i=1}^k (x_i(t) - \hat{x}_i(t))^2. \quad (1)$$



**Fig. 5.** Example of automatically detected passage of two vehicles in the phase trace.

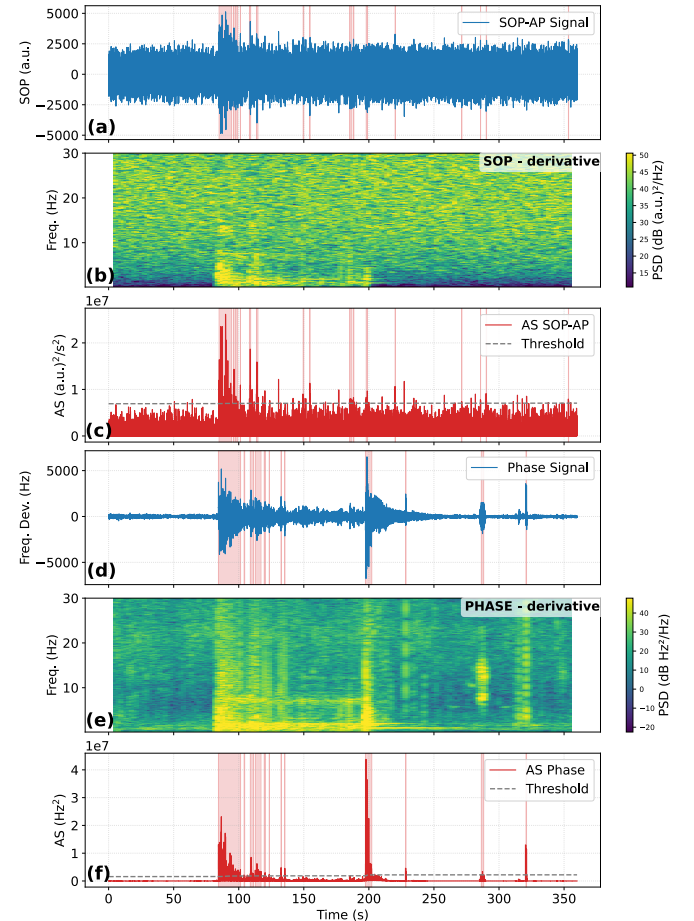
To distinguish genuine anomalies from the noise floor while accounting for slow environmental fluctuations, we apply a dynamic thresholding logic. We compute the moving average  $\mu(t)$  and standard deviation  $\sigma(t)$  of the AS over a 30-minute horizon. A sample is flagged as anomalous if it exceeds the statistical bound defined by:

$$AS(t) > \mu(t) + N\sigma(t) \quad (2)$$

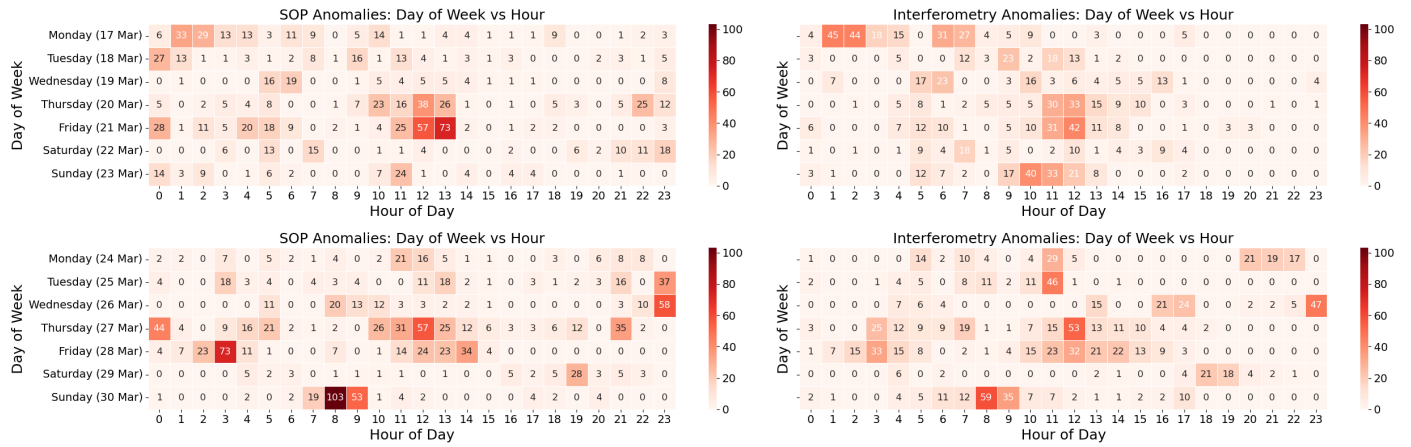
For this field trial, we set  $N = 5$  to balance sensitivity to small events against the rejection of false positives. Finally, a cleanup procedure transforms the raw per-sample triggers into a catalog of distinct physical events. This step involves filtering out negligible, extremely short-duration spikes and merging consecutive or closely spaced anomalous intervals into single logical entities.

### E. Examples of Detected Events

Fig. 5, 6 show representative examples of anomalies detected by the proposed framework running on phase-based detection data. Both are recurring events that have been previously verified by in-site human inspection during the years of deployment of the sensing testbed. Fig. 5 corresponds to the passage of two vehicles at around 1 and 6 seconds marks. Each vehicle passage



**Fig. 6.** Observation of the wind-induced oscillations of an aerial fiber span interacting with nearby vegetation. (a,b,c): the SOP time trace, spectrogram, anomaly score, respectively, recorded at Ascoli Piceno POP. (d,e,f): the interferometric phase frequency deviation time trace, spectrogram, anomaly score, respectively.



**Fig. 7.** Heatmap of detected anomaly counts of SOP-based measurements (left column) and interferometric phase (right column) over the same observation period: week of 2025 March 17-23 (top row), week of 2025 March 24-30 (bottom row).

342 shows up as two sharp peaks caused by the rear and front wheel  
 343 axles passing over a section of damaged asphalt exposing the  
 344 optical fiber cable. Peaks are clearly visible as distinct localized  
 345 fast transients in both the raw signal and the derived anomaly  
 346 score. The automatic detection accurately identifies the duration of  
 347 both events without prior knowledge of their signatures. A  
 348 second example, shown in Fig. 6, highlights wind-induced oscil-  
 349 lations of an aerial fiber span interacting with nearby vegetation.  
 350 The figure compares SOP (top three panels) and interferomet-  
 351 ric phase (bottom three panels) across three domains: time series,  
 352 time-frequency spectrograms and the resulting anomaly scores.  
 353 The perturbation, lasting approximately 120 seconds  
 354 (from  $t \approx 80$  s to  $t \approx 200$  s), exhibits a quasi-periodic structure  
 355 clearly visible in the spectrograms. While the interferometric  
 356 phase spectrogram reveals sharp, well-defined harmonic lines  
 357 extending up to 20 Hz, the SOP spectrogram displays a more dif-  
 358 fuse spectral energy concentrated in the lower frequency range  
 359 ( $< 10$  Hz). Identification of the source is based on these har-  
 360 monic patterns, which correspond to the natural frequencies  
 361 of a suspended cable. Such differences arise from the different  
 362 nature of the compared sensing technologies. Interferometric  
 363 phase is mostly sensitive to longitudinal fiber strain, while SOP is  
 364 impacted also by the cable transversal compression which mod-  
 365 ulates birefringence intensity. Moreover, even different fibers  
 366 inside the same cable may exhibit different response to the same  
 367 event as that is strongly dependent on the cable coupling to the  
 368 ground. Hence, a more detailed and quantitative investigation  
 369 deserves dedicated observation that must be carried out in a  
 370 controlled laboratory environment as the response is influenced  
 371 by too many uncontrollable factors in a production testbed of  
 372 several kilometers. Despite the different spectral signatures and  
 373 Signal-to-Noise Ratios (SNR), the derived anomaly scores (third  
 374 and sixth panels) demonstrate that the unsupervised pipeline  
 375 consistently detects the event, crossing the adaptive threshold  
 376 synchronously in both modalities. These examples demonstrate  
 377 the ability of the autoencoder-based detector to identify events  
 378 of diverse physical origin without the need for manual labeling  
 379 or explicit modeling assumptions.

**E.1. Comparative Analysis Between SOP and Interferometric Data**

380 The availability of both SOP- and interferometry-based observ-  
 381 ables enables a direct comparison of their anomaly detection  
 382 performance. Figure 7 shows a heatmap of anomaly counts per  
 383

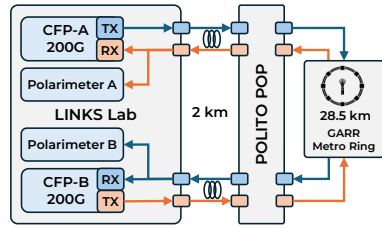
384 hour of day of both techniques over the same observation period  
 385 of 2 weeks, from 2025 March 17th to March 30th. The two sens-  
 386 ing modalities exhibit similar distributions of detected events,  
 387 indicating that the proposed pipeline captures mostly genuine  
 388 physical perturbations rather than technique-specific artifacts.  
 389 At the same time, subtle differences in detection density sug-  
 390 gest complementary sensitivities, which could be exploited for  
 391 cross-validation or enhanced event characterization in future de-  
 392 velopments. For example, on Thursday 27th, 21:00, SOP exhibits  
 393 35 anomalies and no anomalies on phase. Conversely, phase  
 394 counts 46 anomalies on Tuesday 25th, 11:00 and zero on SOP.  
 395 As previously mentioned, these differences may arise from the  
 396 different underlying mechanism of the two techniques and their  
 397 sensitivity to the fiber to the ground coupling, showing how  
 398 both can be used complementarily to enrich the set of detected  
 399 anomalies.

**3. MULTI-TECH SENSING ON TURIN URBAN TESTBED**

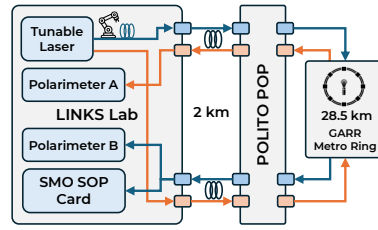
400 In the previous section we have shown a deployment of a multi-  
 401 tech sensing testbed using coherent interferometry and the SM-  
 402 Optics (SMO) PBS-based polarization sensing unit in a inter-  
 403 urban scenario involving two cities. In this section instead, we  
 404 focus on the observation of the measurements obtained in differ-  
 405 ent stages during the second half of 2025 on a second, urban-only,  
 406 testbed deployed in the city of Turin. Here we use a metropolitan  
 407 fiber ring owned by the GARR consortium [33] and involving  
 408 transceiver Q-Factor and full polarimeter-based sensing tech-  
 409 niques to showcase their potential in an urban environment.

**A. Metropolitan Ring Experimental Setup**

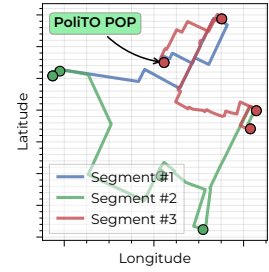
410 The GARR metropolitan fiber cable is a 28.5 km long fiber ring  
 411 deployed underground around the city center of Turin. Fig. 8c  
 412 shows the cable map, made up of three segments (displayed  
 413 with different colors) and the available point of presence (POP).  
 414 The actual geographical locations in the city are confidential.  
 415 Cable access is possible from the POP placed in Politecnico di  
 416 Torino (PoliTO). The PoliTO POP is connected to the LINKS  
 417 Foundation laboratory (Fig. 8) with a 2 km long fiber. Within  
 418 the activities of several research projects, GARR made available  
 419 a 100 GHz WDM slot in both propagation directions to carry  
 420 out experimental activities. The rest of the spectrum is partially  
 421 loaded with standard coherent data signals managed by GARR.  
 422  
 423



(a) Experimental setup 1 using coherent data signals source and 200G transceiver BER telemetry.



(b) Experimental setup 2 using tunable laser source, SMO PBS-based SOP monitoring unit and robotic arm on the AB link.



(c) Map of the GARR 28.5 km metropolitan ring.

**Fig. 8.** Block diagrams of experimental setup 1 (a) and setup 2 (b) at LINKS laboratory. (c) Map of the GARR 28.5 km metropolitan ring.

Unfortunately we do not have access to the detailed spectral load description. However, even in this scenario, GARR has not reported significant transmission penalties. Having access to the metro cable directly from the LINKS laboratories enabled much more flexibility and allowed us to perform acquisitions using different devices and techniques. In particular, here we consider the two setups reported in Fig. 8. In setup 1 we propagate two 200G signals in opposite directions through the fiber ring using the available 100 GHz slot. In the A-to-B (AB) direction, the signal travels in order through the segments 1, 2 and 3 coming back to PoliTO POP; the opposite in B-to-A (BA) direction. Data signals are generated by two commercial CFP-2 coherent transceivers hosted within a whitebox Galileo Phoenix transponder by NEC using DP-16QAM at 32 GBaud. After ring propagation, each signal is received back by the CFP-2 module at LINKS laboratory and the BER is stored and timestamped using NTP protocol on our streaming telemetry platform with an average sampling frequency of 0.06 Hz (one BER sample every 16 seconds). Before coherent detection, a portion of the each received optical signal is tapped and fed to two Novoptel PM1000 polarimeters to retrieve and store the full Stokes vector time evolution. The polarimeters' sampling rate is set to 95 Hz, which is a good compromise between the capability to support long time measurements and to detect anomalies generated by anthropic activity. We thus obtain two BER and Stokes vector streams measured from two coherent signals propagating in the same cable in opposite directions AB and BA. These measurements have been carried out during August 2025. Setup 2 instead has been employed during July and October 2025. In this case the Stokes detection configuration remains the same as setup 1, while the source is changed. A tunable laser generates an unmodulated carrier at 1565.496 nm which feeds both the propagation directions. After ring propagation, while the BA signal is fed to the polarimeter A only, the AB signal is split and received by both the polarimeter B and the SMO SOP device. We thus collect two polarimeter streams in opposite directions and one SMO SOP stream both at 100 Hz. Due to SMO device malfunctioning however, we have been able to collect only polarimeters traces during July acquisitions, while SMO device was operative in October. We had also employed a variation of setup 2 during the measurements of October 2025. In this scenario, at the tunable laser output in the AB direction only, we employed an Arduino-based robotic arm to simulate fiber shaking in a controlled environment. This variant was employed within another research activity focusing on the identification of mechanical disturbances which can impair telecom infrastructure integrity through SOP monitoring [30]. The

robotic arm was configured to interact with the fiber through a vertical "shaking" motion, maintaining a 90° angle relative to the fiber's longitudinal axis to maximize stress induction. We defined two specific mechanical regimes: "Slow Shaking" (1 to 5 Hz), mimicking manual cable handling and tapping, and "Fast Shaking" (5 to 10 Hz), which replicates the high-frequency transients associated with critical vibrations which could possibly lead to fiber cut. Although both setups feature two oppositely propagated polarimeter acquisitions, it has not been possible to exploit them to perform localization because the polarimeters time base was not accurately synchronized relying only on a coarse NTP synchronization.

## B. Experimental Data Analysis

We first focus on the observation of the experimental measurements using setup 1. Fig. 9a shows the time series acquired from August 1st, 16:30 to August 8th 4:30 (Europe/Rome timezone). Rather than showing the raw Stokes vector, we have preferred to rework the polarimeter acquisitions in the Jones domain. This allows us to observe the proper phase difference between orthogonal polarization components. The received electric field Jones vector  $\mathbf{J}$  can be expressed as:

$$\mathbf{J} = \begin{bmatrix} E_x \\ E_y e^{i\delta} \end{bmatrix} \quad (3)$$

$$\delta = \phi_y - \phi_x$$

$$\alpha = \frac{E_x}{E_y}$$

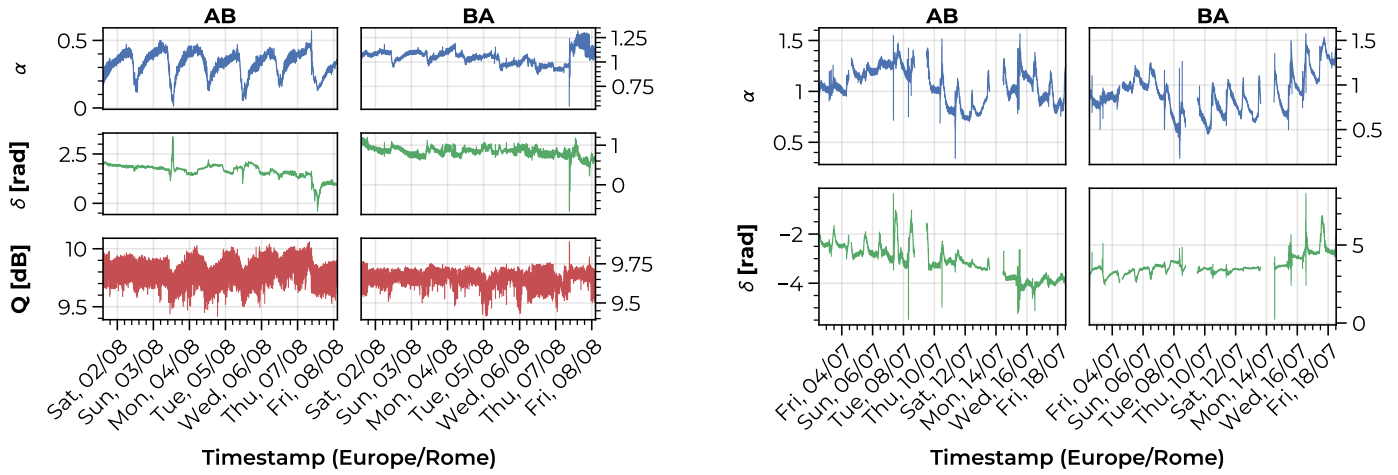
where  $E_x, E_y$  are the orthogonal electric field components.  $\phi_x, \phi_y$  are their absolute phases. Thus  $\delta$  is the phase difference between the orthogonal components and  $\alpha$  the amplitudes' ratio. These quantities are obtained from the Stokes vector using the following relations:

$$\alpha = \arctan\left(\frac{1 - S_1}{1 + S_1}\right) \quad (4)$$

$$\delta = \arctan\left(\frac{S_3}{S_2}\right)$$

Together with the Jones parameters, Fig. 9a shows also the Q-Factor in dB obtained from the simultaneously acquired transceivers' BER:

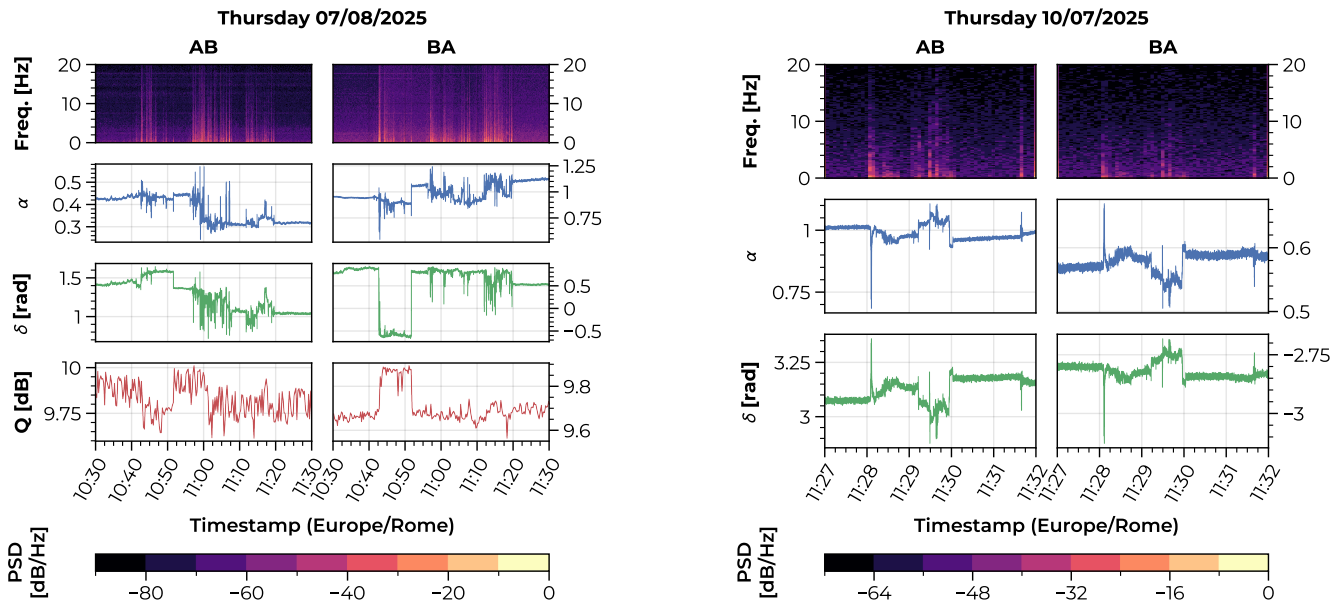
$$Q = \sqrt{2} \operatorname{erfc}^{-1}(2 \cdot \text{BER}) \quad (5)$$



(a) August 1st 2025 16:30 - August 8th 2025 4:30

(b) July 2nd 2025 14:30 - July 18th 2025 14:30

**Fig. 9.** Experimental measurements taken in (a) August 2025 using setup 1 and (b) July 2025 using setup 2 without SMO card. Left columns: AB direction, Right columns: BA direction. 1st/2nd rows: Polarimeter Jones  $\alpha$  and  $\delta$  traces. 3rd (a) row: Q-Factor (dB) from transceiver BER. Time axis is shared and each labeled tick corresponds to the day's midnight.



(a) Event of Thursday 2025 August 7th, 10:30. Left column: AB direction, Right column: BA direction. 1st row: time spectrogram. 2nd/3rd rows: Polarimeter Jones  $\alpha$  and  $\delta$  traces. 4th row: Q-Factor (dB) from transceiver BER.

(b) Event of Thursday 2025 July 10th, 11:27. Left column: AB direction, Right column: BA direction. 1st row: time spectrogram. 2nd/3rd rows: Polarimeter Jones  $\alpha$  and  $\delta$  traces.

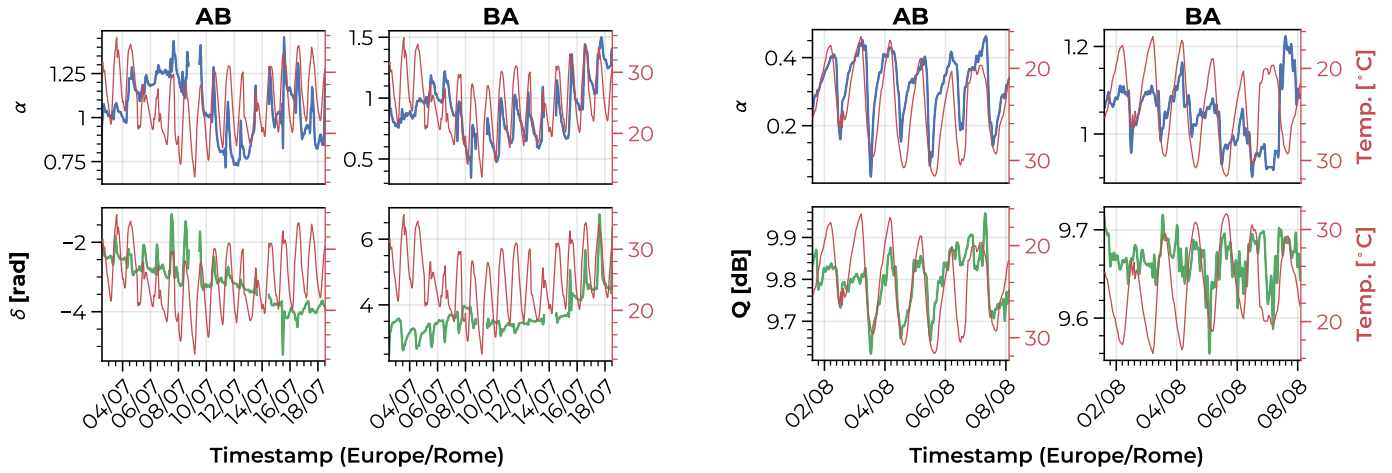
**Fig. 10.** Example of detected anomalies during the August and July 2025 acquisitions.

499

500 Coherent data signals are inherently depolarized, thus not  
 501 having a fixed polarization state with equally distributed power  
 502 between the orthogonal components and degree of polarization  
 503 (DOP) near zero. Hence, one may argue that observing their  
 504 polarization state with a polarimeter should not deliver any  
 505 meaningful information about external disturbances. However,  
 506 the acquisitions show an oscillation on  $\alpha$  with one day period  
 507 and several peaks that could be caused by human or urban  
 508 activities on the fiber cable. Upon further inspection on shorter  
 509 timespans, here not reported for space constraints, the polarization  
 510 measurements also show oscillations with a period ranging

511

512 from 30 to 60 minutes and lasting for several hours. We may  
 513 argue that variations with such periods are caused by cooling  
 514 or some form maintenance in GARR PoPs. However, we cannot  
 515 be sure on the validity of this hypothesis as currently we do  
 516 not have access to these locations or to the building internal  
 517 temperature data. Moreover, Q-Factor variations seem strongly  
 518 correlated with SOP, at least on longer periods. Hence, this  
 519 suggests that observing a coherent channel SOP using a simple  
 520 polarimeter, rather than accessing the transceiver estimated  
 521 Jones matrix, can still provide precious data on environmental  
 522 events. This is a significant advantage as the access to transceiver  
 Jones matrices is not easy due to closed nature of DSPs and it



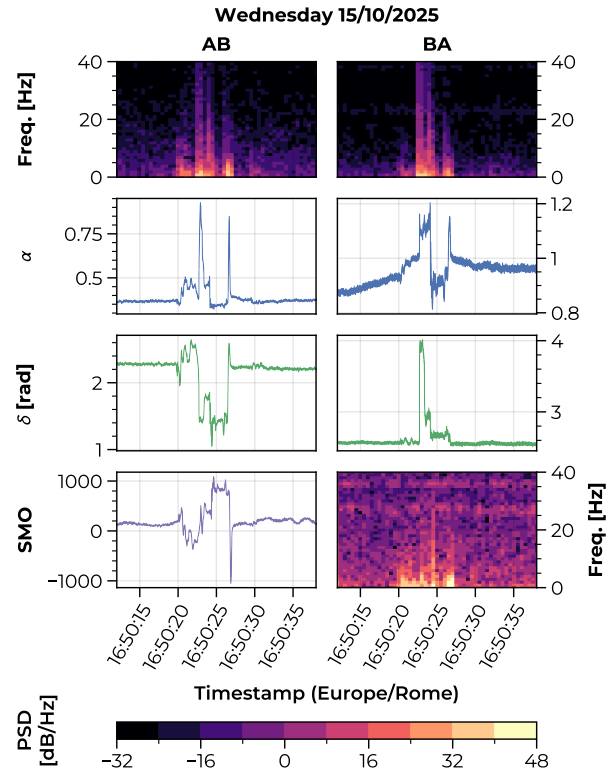
(a) July 2nd 2025 14:30 - July 18th 2025 14:30

(b) August 1st 2025 16:30 - August 8th 2025 4:30

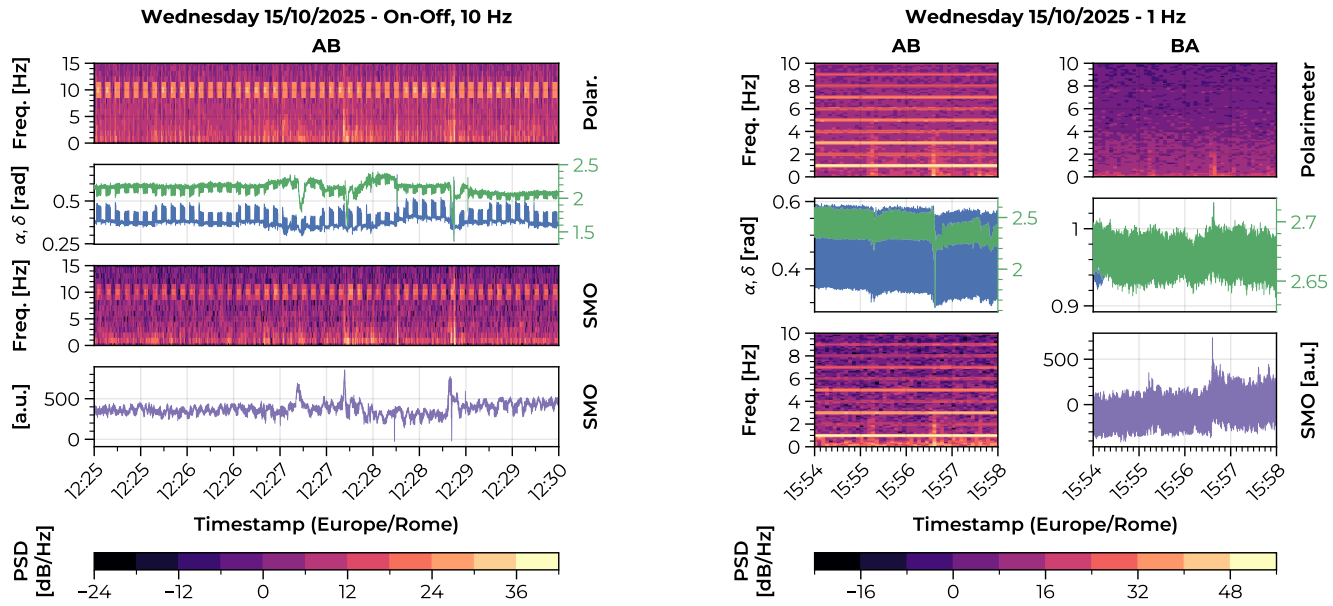
**Fig. 11.** SOP/Q-Factor vs Temperature taken in (a) July 2025 using setup 2 without SMO card and (b) August 2025 using setup 1 filtered with a 4th order Butterworth filter with 14 mHz bandwidth. Left columns: AB direction, Right columns: BA direction. 1st rows: Polarimeter Jones  $\alpha$  trace. 2nd (a) row: Polarimeter Jones  $\delta$  trace. 2nd (b) row: Q-factor in dB from transceiver BER. Left y-axes show the recorded temperature (Celsius) of nearby weather station. Time axis is shared and each labeled tick corresponds to the day's midnight.

enables SOP sensing avoiding waste of spectral slots to transmit a dedicated laser source. This feature can be justified with the polarization dependent loss (PDL) introduced by optical amplifiers and add-drop nodes along the lightpath increasing the residual DOP [34]. However, this mechanism may also cause sensitivity variation along the route which is difficult to isolate in a non-controlled environment as a production cable. However, a back-to-back SOP measurement at the transponder output may help to isolate this effect as in absence of PDL one would expect minimal SOP variation. A deeper understanding of this effect thus deserves dedicated laboratory experiments which will be targeted in future works. As an example, Fig.10a reports a zoom of the largest event in Fig.9a on the August 7th, although more lower magnitude anomalies can be found upon zooming the time series. The temporal spectrogram plots show the sum ( $S_{xx,1} + S_{xx,2} + S_{xx,3}$ ) of the three single Stokes vector components  $S_i$  spectrograms  $S_{xx,i}$ . In each spectrogram shown throughout this paper lower energy components are clipped out to underline the significant spectral content. Upon inspection, the peak is actually a roughly 1 hour long sequence of disturbances whose shape is similar to human-induced stressors in [7] and whose spectral content is mostly limited to about 10 Hz. We underline that this event should be likely have happened somewhere along the cable since it is visible in both propagation directions and the LINKS laboratory was closed during the August measurements, while the university campus was mostly uncrowded due to the summer break. Moreover, the Q-Factor show performance changes synchronized to the SOP variations, thus confirming detectable correlations between SOP variations caused by external disturbances and typical telemetry metrics available from optical network elements. It is then interesting to compare the coherent channel-based acquisitions to those obtained using setup 2 employing a tunable laser as signal source to confirm the validity of the coherent-sourced SOP detection. In this session, only polarimeters acquisitions were available and their time series are shown in Fig.9b, spanning from July 2nd 2025 14:30 to July 18th 14:30, thus providing

more than two weeks of measurements. Data presents three gaps which have been necessary to restart the acquisitions after polarimeter buffer saturation: from 14:20 to 18:10 on July



**Fig. 12.** Event of Wednesday 2025 October 15th, 16:50. Left column: AB direction, Right column: BA direction. 1st row: time spectrogram. 2nd/3rd rows: Polarimeter Jones  $\alpha$  and  $\delta$  traces. 4th row: SMO device time trace (left) and time spectrogram (right).



(a) Event of Wednesday 2025 October 15th, 12:25 (AB direction only). 1st: Polarimeter time spectrogram. 2nd: Polarimeter Jones  $\alpha$  (blue) and  $\delta$  (green) traces. 3rd: SMO device time spectrogram. 4th: SMO device time trace.

(b) Event of October 15 2025, 15:54. Left column: AB direction, Right column: BA direction. 1st row: Polarimeter spectrograms. 2nd row: Polarimeter Jones  $\alpha$  (blue) and  $\delta$  (green) traces. 3rd row: SMO device time trace in the AB direction (right) and spectrograms (left).

**Fig. 13.** Events detected on 2025 October 15th using polarimeter and SMO PBS-based SOP monitoring device with AB fiber shaken by robotic arm. (a) Intermittent shaking every 3 seconds at 10 Hz, (b) Continuous shaking at 1 Hz.

4th, from July 8th 19:50 to July 9th 14:30 and from July 13rd 16:15 to July 14th 13:00. In  $\delta$  calculation, phase has been cumulatively unwrapped assuming that no major phase variations happened during the gaps to preserve the visible trend in the long run. The overall July trend is comparable to the August acquisitions, thus validating the data acquired with a coherent source. Figures however show some different patterns between opposite propagation directions, which are more notable in the August polarimeter acquisitions, although the main trend and stronger peaks are in agreement. Inspection of the raw Stokes components, not shown here due to space constraints, show that this major difference is caused by the average SOP being in different states between AB and BA in August, while the two directions in July remained steady around similar polarization states. With respect to the August acquisitions, however, the July set of measurements shows more anomalies as the large peaks July 6th, 7th, 10th, while much more with different intensities can be observed upon further inspection. In Fig. 10b we report as an example the event of July 10th 2025 at 10:30. Even in this case, the peak is a sequence of perturbations of the parameters  $\alpha, \delta$  lasting around 5 minutes, which may suggest maintenance along the fiber path. The time evolution is compatible to the event of Fig. 10a with a similar spectral content still limited to around 10 Hz. Regarding the slow variations with evident daily periodicity we observe that while during July this trend fades also through the phase difference  $\delta$ , in the August dataset is strongly visible only through  $\alpha$  because during this period the SOP varied along a circularly polarized state. In Fig. 11, we show that they are caused by the night and day temperature variations, reporting  $\alpha$  and Q-Factor for the August dataset and both  $\alpha$  and  $\delta$  for the July dataset. All the signals have been filtered using a 4th order Butterworth low-pass filter with a cut frequency of 14 mHz in order to retain variations within a 2 hours period. On top of them we reported the temperature evolution obtained using the

Open-Meteo API [35] from a weather station located in the city center of Turin and sampled every hour. The results show a very good match between the temperature curve and both the SOP and Q-Factor features, thus showing also the dependence of transmission performance on daily temperature excursion. It is worth noting the opposite polarity of August and July SOP with respect to the temperature and of the AB Q-Factor with respect to BA Q-Factor. This can be explained with the transition of the signal between opposite polarization states, from left circular to right circular polarization, for example.

We finally tackle on the acquisitions carried out during October 2025 using setup 2. In this case, no long term continuous measurements have been carried out. We thus show a subset of the observed events as the availability of the SMO PBS-based sensing unit enables testing of a multi-tech sensing scenario and the comparison to the reference polarimeter measurement. The SMO time traces show the difference between the two raw ADCs counts as depicted in the PBS-based device block diagram in Fig. 2. Fig. 12 shows a 20 seconds perturbation detected on October 15th 2025. Robotic arm was deactivated during this acquisition, thus we are able to test the SMO device capability to detect the event alone. Even in this case the polarimeter trace is similar to the events previously seen in Fig. 10. The SMO device (receiving the AB signal) is able to detect the event and shows a time trace compatible with the polarimeter phase difference  $\delta$ , although with opposite polarity. Both polarimeter and SMO spectrograms are compatible with the event spectral content below 40 Hz. In Fig. 13 we show instead two time windows where an anomalous external perturbation stroke while the robotic arm was shaking the fiber to emulate perturbation caused by human activities in construction sites which may be harmful to telecom infrastructure integrity. Fig. 13a shows the polarimeter and SMO device acquisitions between 12:25 and 12:30 of October 15. In this case, the BA polarimeter acquisition

was not available. The robotic arm shakes the fiber at 10 Hz intermittently every 3 seconds. While SMO device time trace seems noisy, frequency analysis reveals the intermittent pattern of 10 Hz shaking, thus enabling detection upon proper filtering. Moreover, three anomalies happening from 12:27 and 12:29 are clearly detected by both sensing methods. Finally, Fig.13b shows both directions polarimeters' acquisition together with the SMO device trace between 15:54 and 15:58 of October 15th 2025. As per Fig.8b, the arm excites only the AB fiber so that the BA polarimeter acquisition has not been impaired. Here the robotic arm continuously shakes the fiber with a frequency of 1 Hz. Both the polarimeter and SMO time spectrograms show the 1 Hz fundamental tone carrying most of the energy and several harmonics up to 10 Hz and further, although we have limited the frequency axis to 10 Hz for cleaner visualization. However, also upon time traces inspection, two anomalies at around 15:55 and 15:57 struck while the robotic arm was shaking the fiber. These are still clearly visible in the spectrogram. Moreover, they appear on the BA polarimeter trace which is not impaired by the robotic arm's shaking. This enables the possibility to selectively detect, validate and possibly localize anomalies by cross correlation of different signals measured at different points along the network and impaired by diverse phenomena.

#### 4. CONCLUSIONS

In this work, we presented an in-field demonstration of a multi-technique sensing framework operating on live terrestrial optical networks. By exploiting the synergy between interferometric phase sensing and SOP monitoring, we shifted the focus from single-parameter observation to an aggregated sensing approach that leverages the pervasive nature of the optical infrastructure. We validated an unsupervised anomaly detection pipeline based on autoencoders using data from the long-distance Ascoli-Teramo testbed. This framework proved capable of automatically identifying anomalies across different physical observables—phase and polarization—without relying on large, manually labeled datasets, ensuring robustness against environmental variability. Furthermore, the experimental activities conducted in the Turin metropolitan area highlighted the potential for correlation of specialized sensing metrics with standard network telemetry data, such as transceiver Q-Factor. Moreover, results have shown that measuring the polarization evolution of a coherent data signal hooking up a polarimeter can provide alternative and useful SOP data source, despite the depolarized nature of the coherent signals. The cross-validation performed using coherent transceivers, polarimeters, and cost-effective PBS-based devices confirmed the reliability of the proposed scalable solutions, while controlled experiments with a robotic arm demonstrated the capability to detect and characterize complex events. All demonstrated techniques operated in coexistence with data traffic, confirming the feasibility of transforming the optical transport infrastructure into a pervasive, distributed sensing grid. The move towards a unified telemetry plane that aggregates heterogeneous data sources — from high-precision phase sensors to standard network elements — paves the way for the implementation of cognitive, dual-purpose networks capable of simultaneous data transmission and environmental supervision.

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**Data availability.** Data underlying the results presented in this paper are not publicly available at this time but may be obtained from the authors upon reasonable request.

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