

# Anomaly Detection in Optical Spectra via Joint Optimization

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# Introduction



- User information is enclosed into **spectra** and transmitted through fiber optics.
- A spectrum S contains a varying number of channels that follow an unknown trend  $\nu$ .
- The ASE provides a base power level in regions with no channels. ASE samples follow a trend  $\mu$ .
- The occurrence of faults cause **anomalies**
- Our goal is to automatically detect the anomalies in the spectrum, if any.

# Key Intuitions

- Anomalies are channels that do not conform to the channel trend.
- Fit the channel trend via robust fitting, such as RanSaC.  $\bullet$
- **Joint optimization** allows us to promote the similarity between channel and ASE trends.

# Joint Optimization

Our method accounts for the similarity of the channel and ASE trends,  $\nu$  and  $\mu$ , by optimizing the following loss:

similarity term data fidelity terms  $p \in \mathcal{N}$ coefficients residuals between residuals between samples in  $\mathcal{N}$  and  $\mu$ samples in  $\mathcal{C}$  and  $\nu$  $\nu$  and  $\mu$ balance the data fidelity and similarity terms

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### **Proposed Method**

**Algorithm 1** Joint Optimization

**Input:** Spectrum S, inlier threshold  $\varepsilon$ **Output:** Central frequencies  $c_k$  of the K anomalies

**Stage 1:** Initialization.

- 1: Fit a linear trend  $\ell \leftarrow \text{LO-RanSaC}(x_i, S(x_i), \varepsilon)$
- 2:  $h_r \leftarrow \text{Construct}$  the histogram of residuals.
- 3:  $\mathcal{C}, \mathcal{N} \leftarrow \text{Otsu}(h_r)$ .
- 4:  $\mathcal{C} \leftarrow \text{FindPeaks}(\mathcal{C})$ .
- 5:  $\nu \leftarrow$  Fit polynomial to the samples in  $\mathcal{C}$ .
- 6:  $\mu \leftarrow$  Fit polynomial to the samples in  $\mathcal{N}$ **Stage 2:** Joint optimization;
- 7:  $\nu^*, \mu^* \leftarrow$  Jointly estimate trends as in Eq. (1). **Stage 3:** Anomaly detection;
- 8:  $c_k \leftarrow \text{Recognize as anomalies } \{c_k \in \mathcal{C} : \operatorname{err}(p_k, \nu) > \varepsilon\}.$

#### Running example







### Experiments

#### Qualitative Experiments

#### Real Spectrum





#### **Quantitative Experiments**

	Accuracy		Precision		Recall		$F_1$ Score	
	Mean	Var.	Mean	Var.	Mean	Var.	Mean	Var.
Two-thresholds	0.662	0.054	0.329	0.119	0.697	0.141	0.391	0.112
Robust Fitting	0.951	0.010	0.800	0.117	0.810	0.110	0.789	0.106
Faster R-CNN	0.857	0.006	0.000	0.000	0.000	0.000	0.000	0.000
Joint Optimization	0.989	0.002	0.968	0.028	0.937	0.036	0.948	0.030

### Conclusions

- Our method overcomes currently employed solutions and provides us an estimate of the channel and ASE trends.
- The **joint optimization** procedure that allows us to fit the channel trend and extracting useful information encoded in the ASE.



#### Our joint optimization procedure enhances channel trend estimation, enabling the accurate detection of anomalies.

