

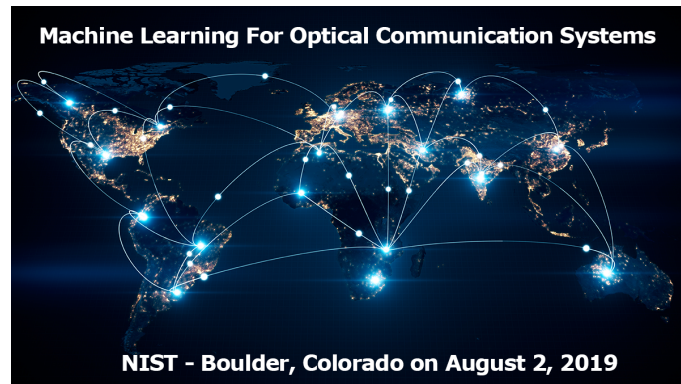
What data matters for optical communications

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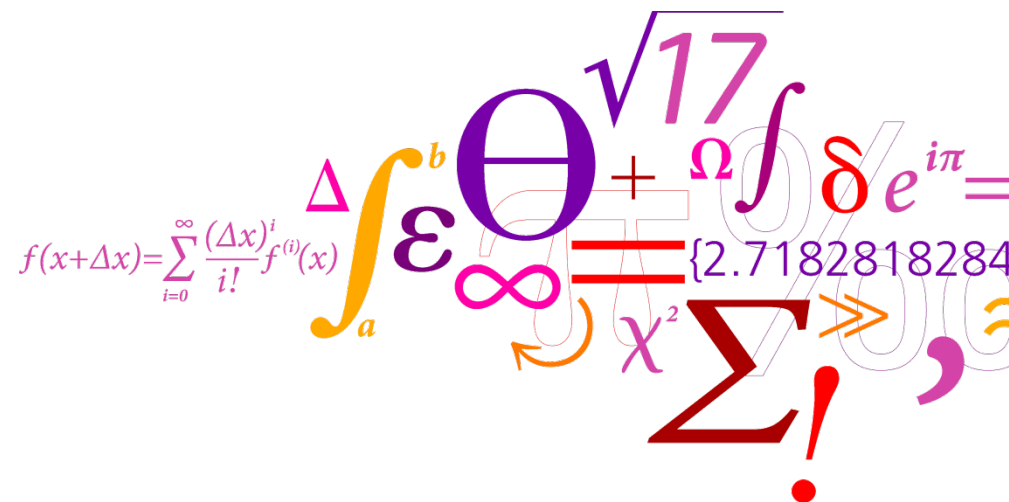
Machine Learning in Photonic Systems (M-LiPS) group

DTU Fotonik, Technical University of Denmark, DK-2800, Kgs. Lyngby



DTU Fotonik

Department of Photonics Engineering

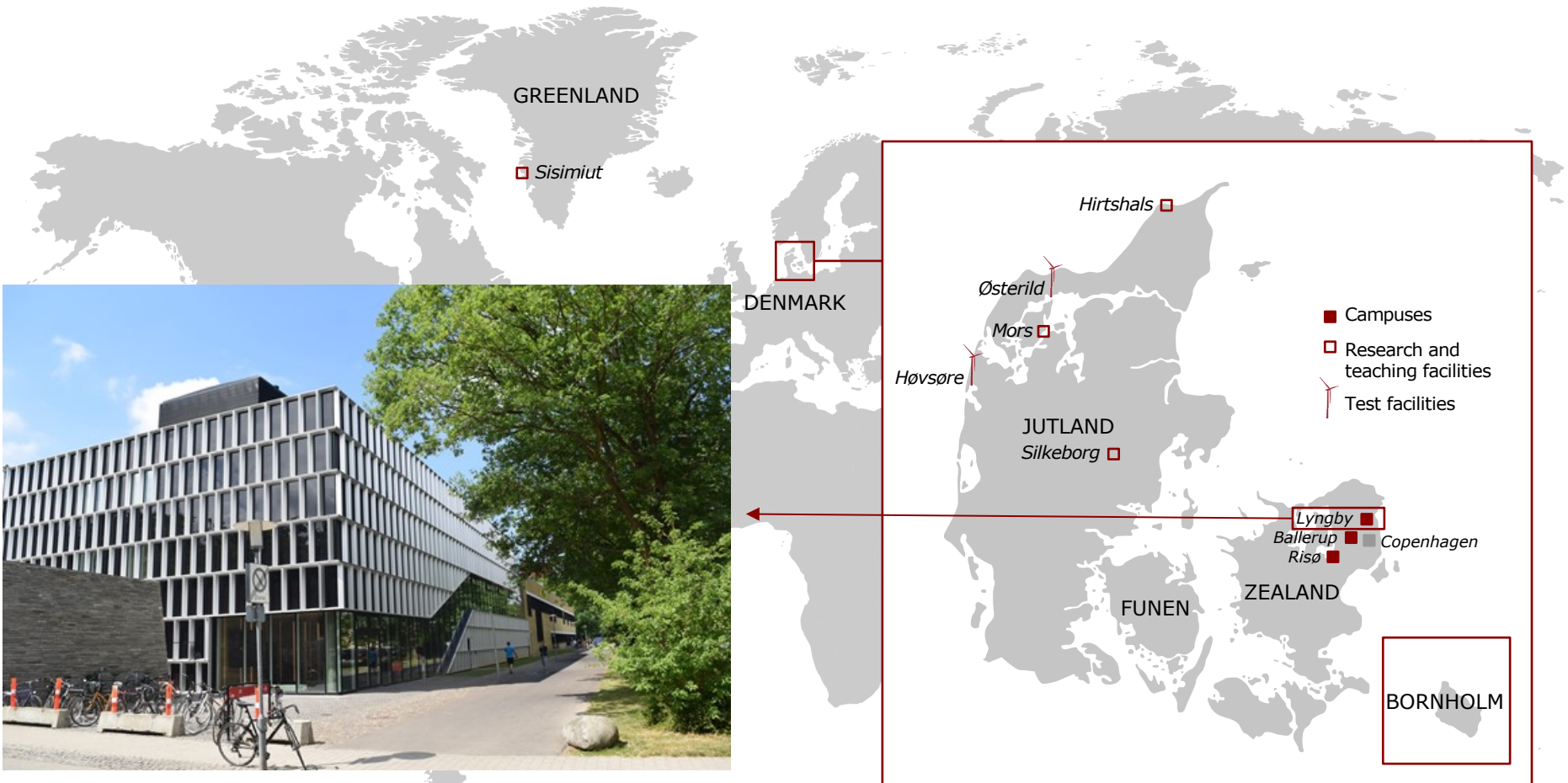


Outline



- Introduction to DTU and M-LiPS group
- Machine learning in optical communications
 - Relevant problems for machine learning
 - State-of-the-art
- **What data matters for optical communications**
- Summary

Technical University of Denmark (DTU)



Machine Learning in Photonic Systems (M-LiPS)



Group leader:

Darko Zibar

Senior Staff:

Francesco Da Ros
Simone Gaiarin
Hou-Man Chin
Uiara Celine Moura

PhD students:

Martin Djurhuus
Nicola De Renzis
Stenio Magalhaes Ranzini
Ognjen Jovanovic
Thyago Monteiro

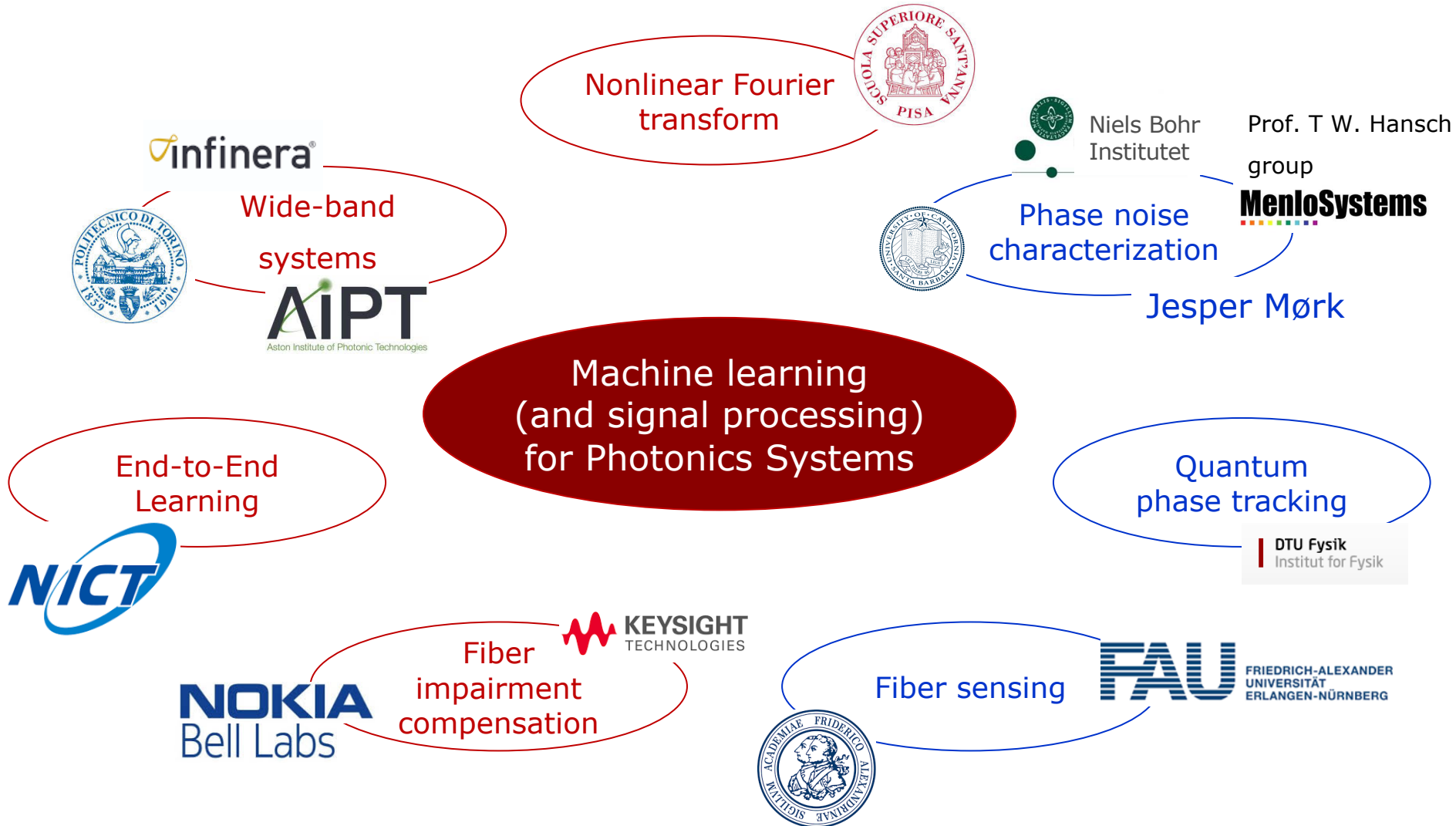
Research assistant:

Giovanni Brajato



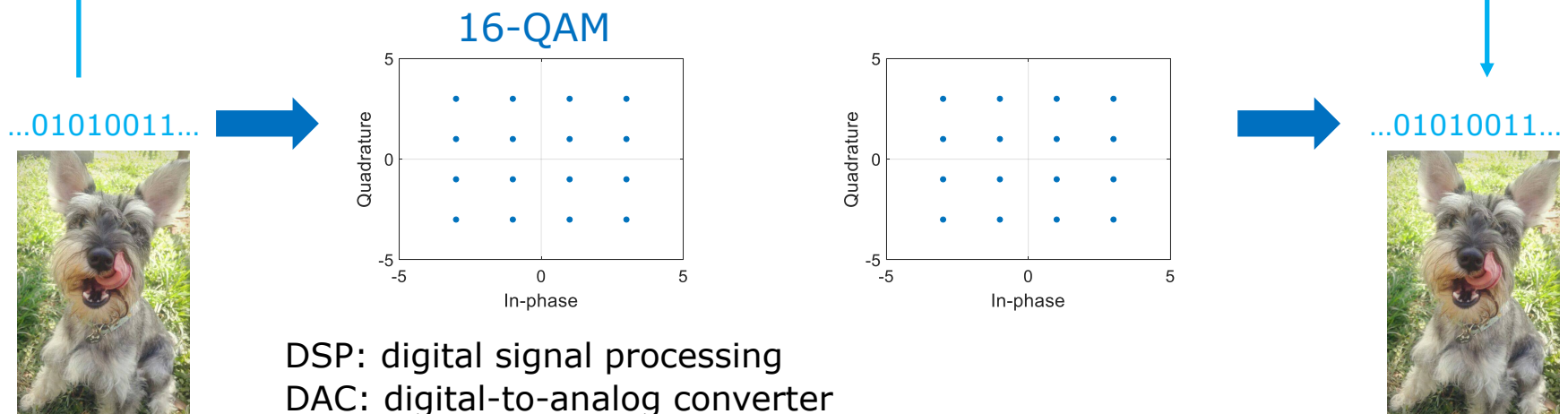
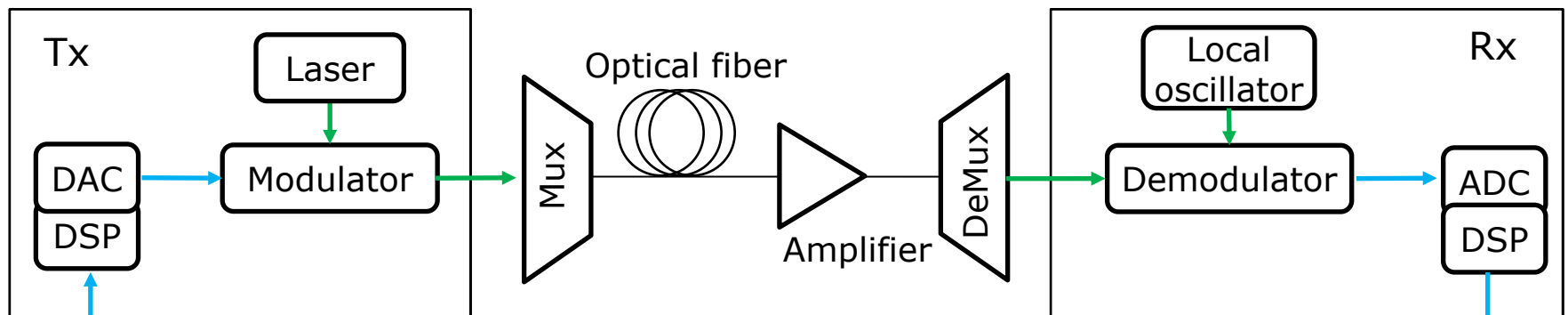
- In 2019, 11 group members
- 5 Ph.D. candidates
- 5 nationalities

Research activities



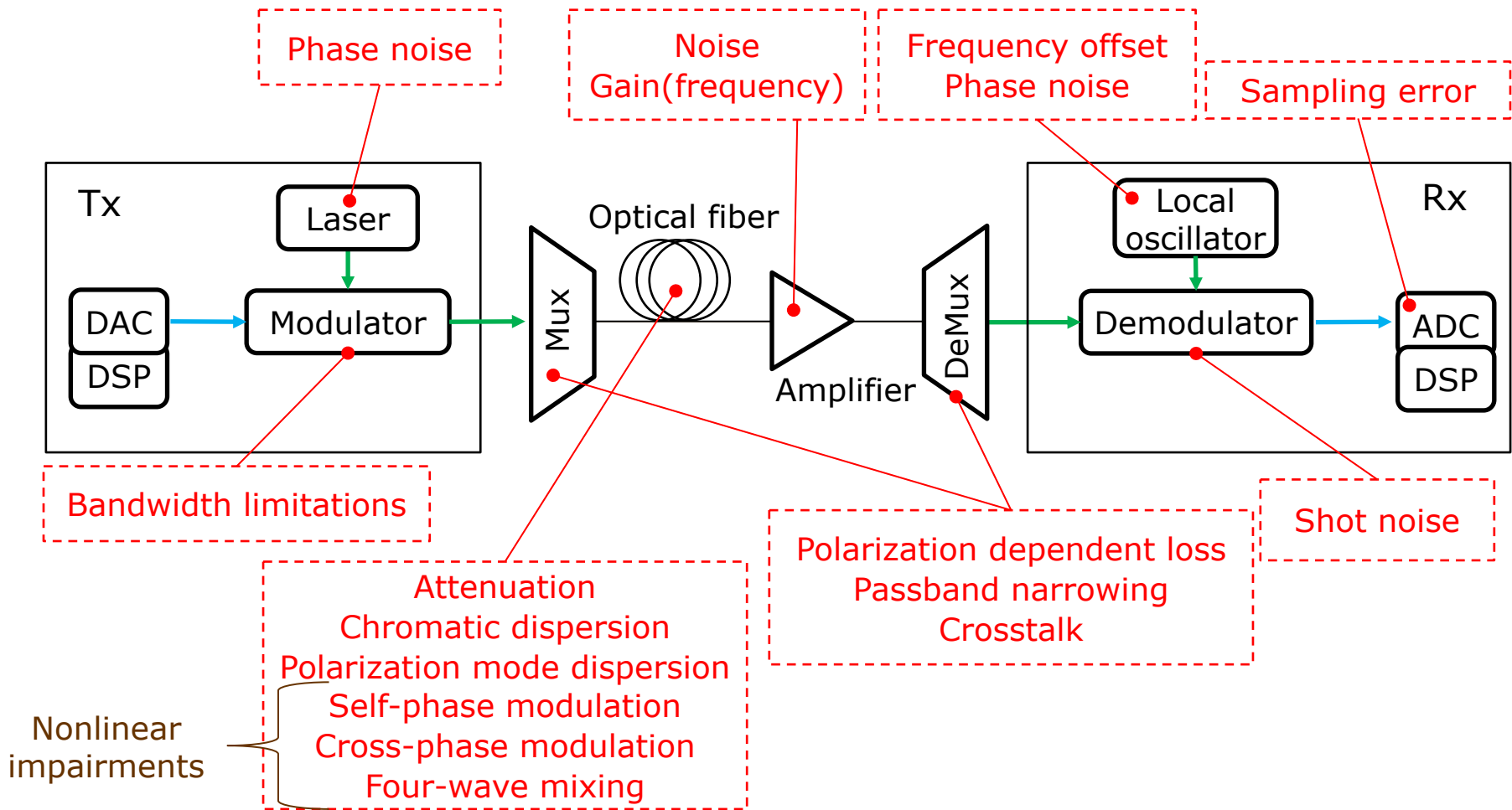
Optical communication systems

Optical coherent transmission illustration



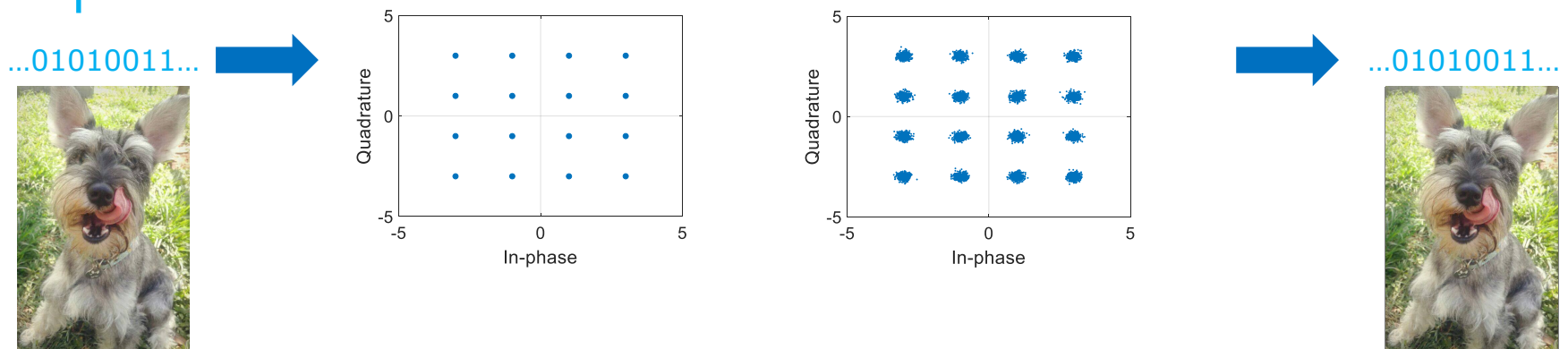
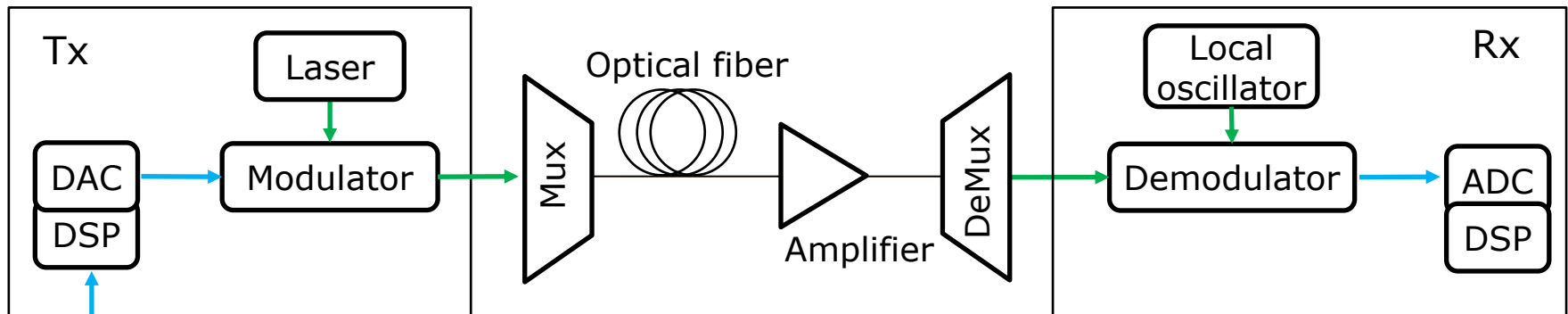
DSP: digital signal processing
DAC: digital-to-analog converter
ADC: analog-to-digital converter

Optical communication systems

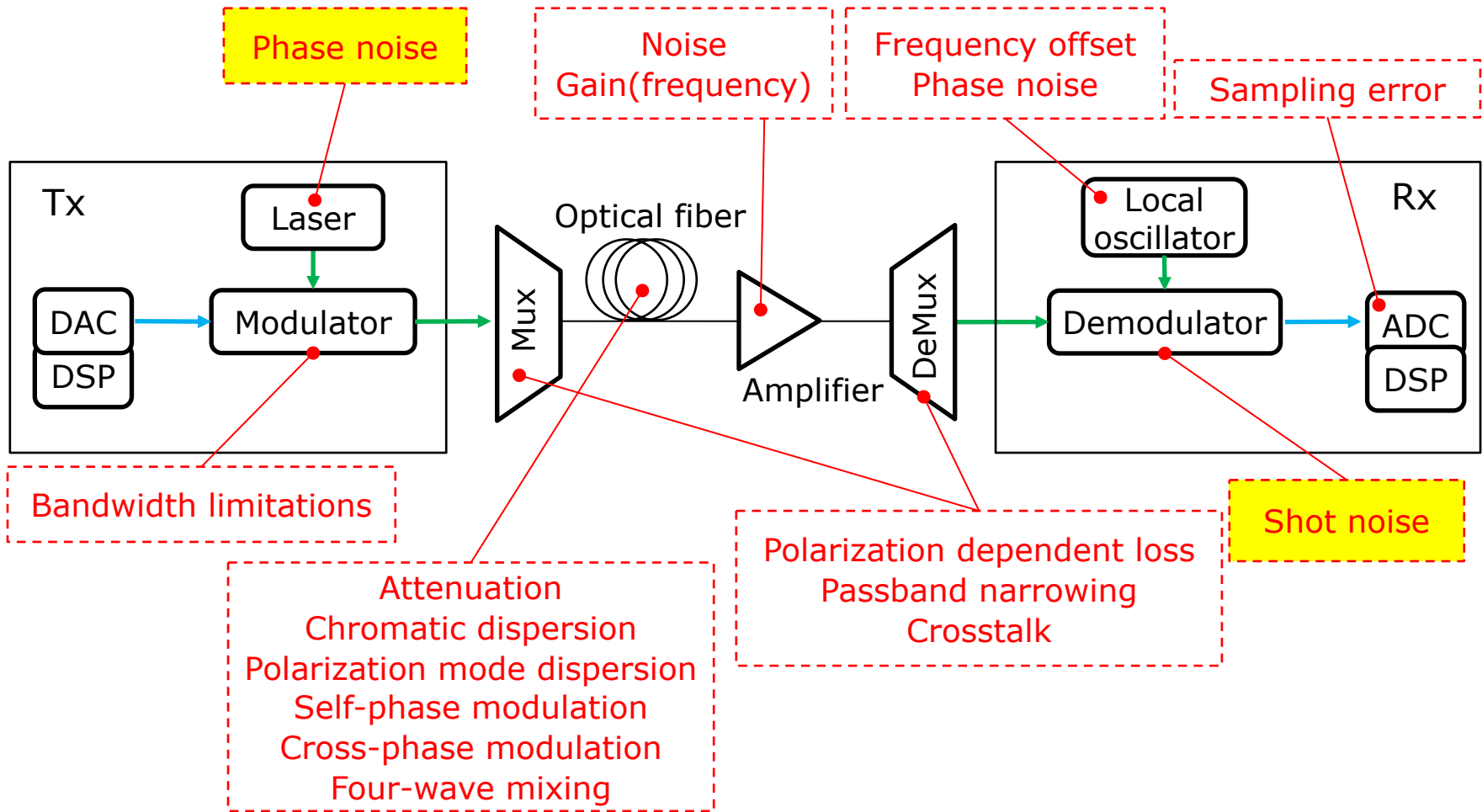


Optical communication systems

Optical coherent transmission illustration



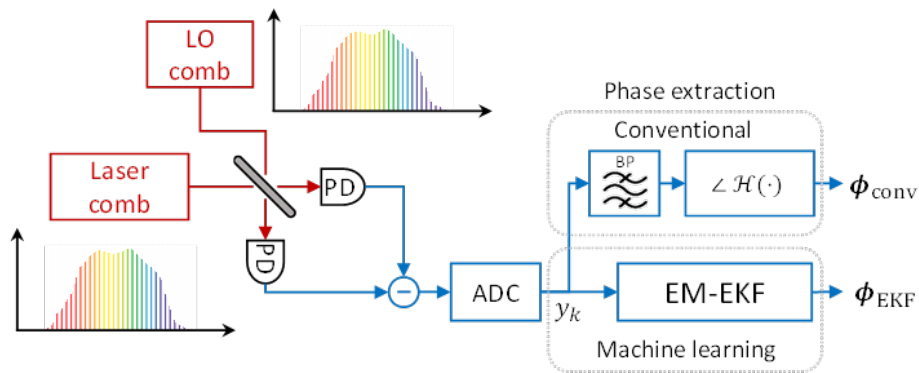
M-LiPS works



Frequency combs noise characterization

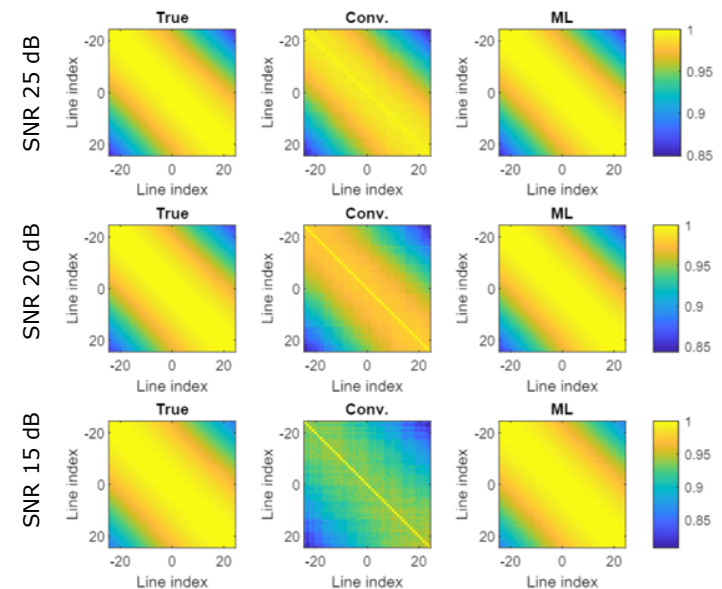
- *potential for becoming a reference tool*

Joint phase noise tracking



EM-EKF = expectation maximization algorithm with extended Kalman filter

Extracted correlation matrix for different SNR scenarios

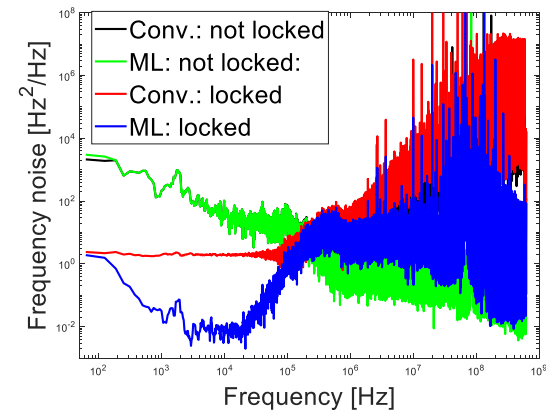
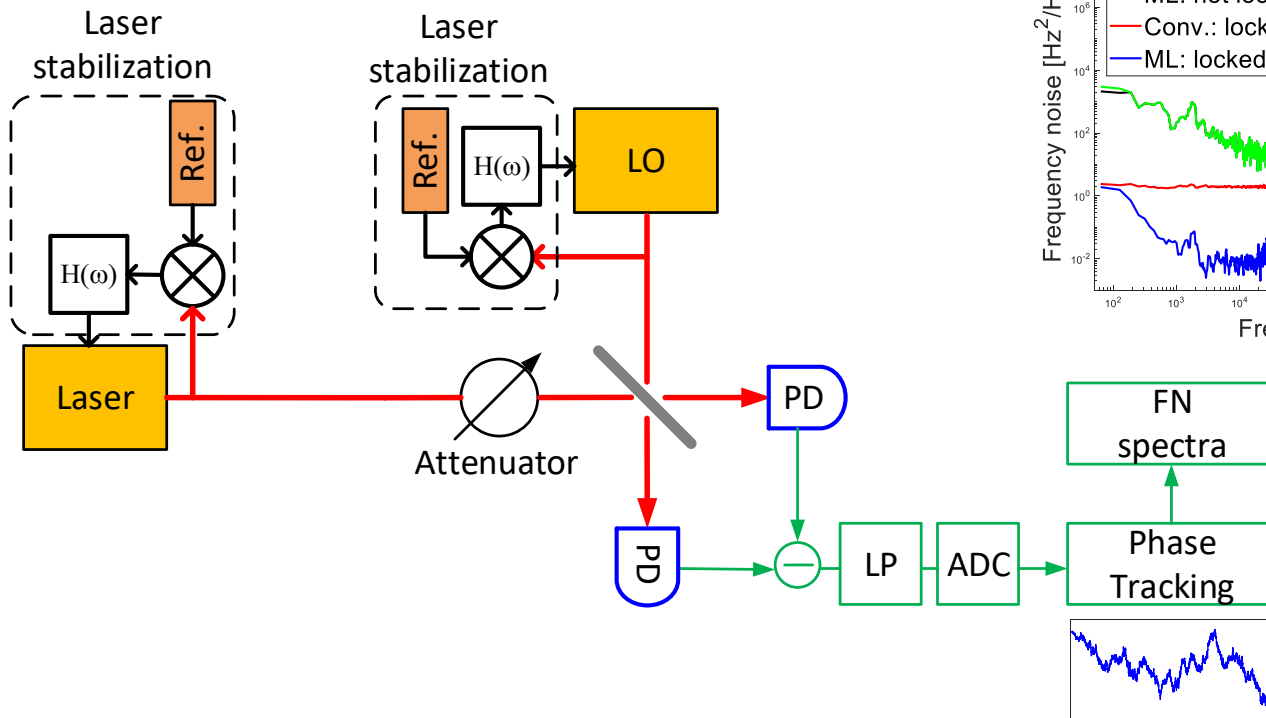


[1] B. Giovanni et al., Optical Frequency Comb Noise Characterization Using Machine Learning, accepted ECOC 2019.

Laser noise characterization

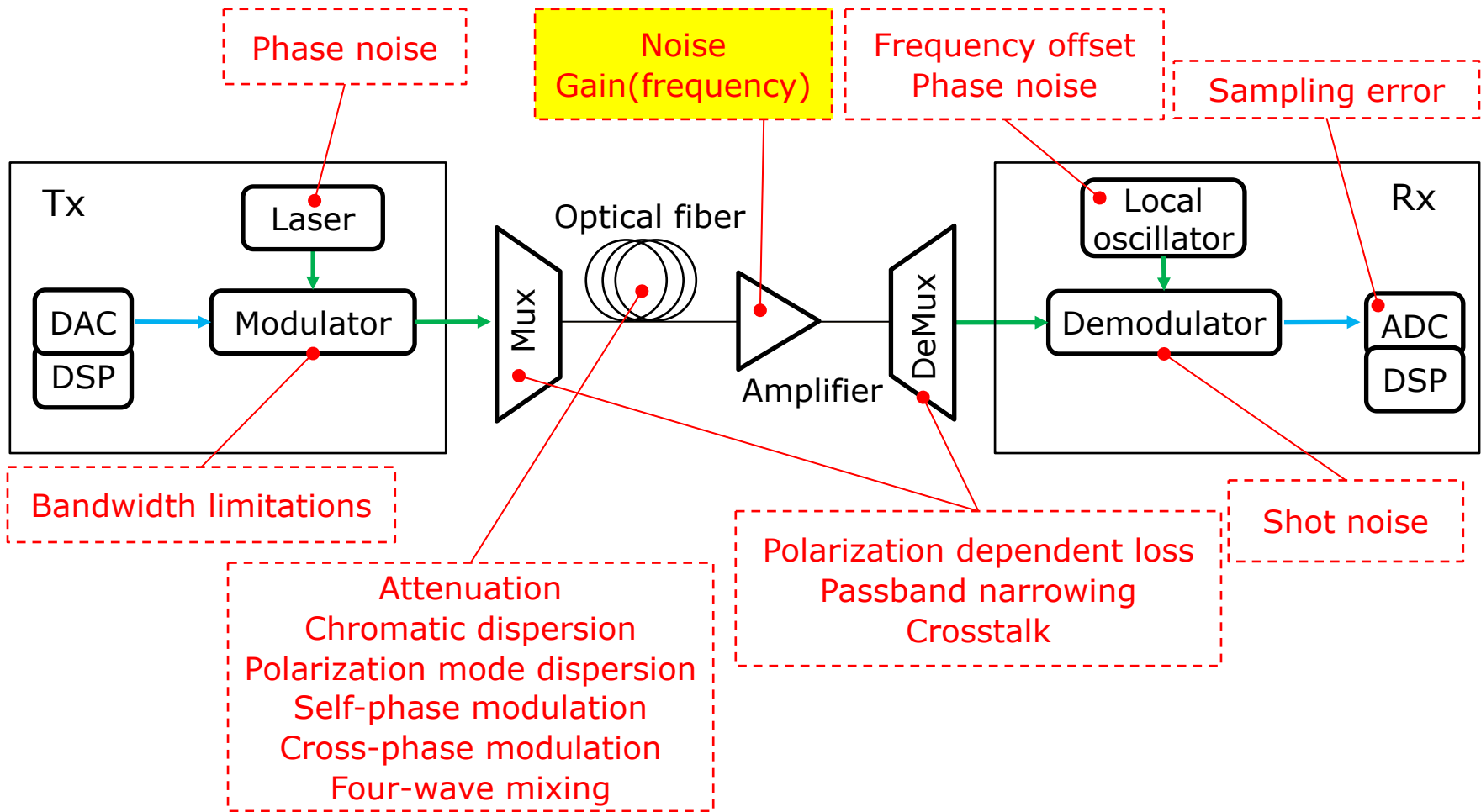
- *potential for becoming a reference tool*

Use Bayesian inference to remove measurement noise (shot noise)



[1] H. Chin, D. Zibar, N. Jain, T. Gehring, and U. L. Andersen, Phase Compensation for Continuous Variable Quantum Key Distribution, in *CLEO 2019*.

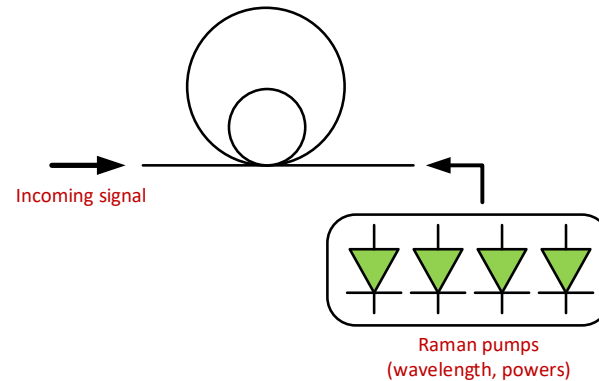
M-LiPS works



Raman amplifier for optical communication

Name	O-Band	E-Band	S-Band	C-band	L-Band
Wavelength range (nm)	1260-1360	1360-1460	1460-1530	1530-1565	1565-1625
C-band system				35 nm	
C+L-band system				95 nm	
WON	365 nm				

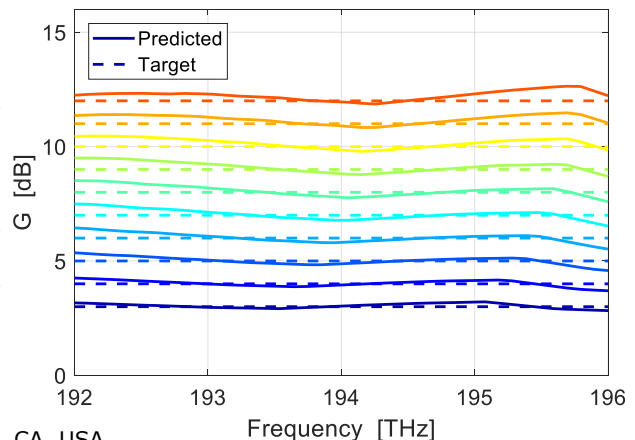
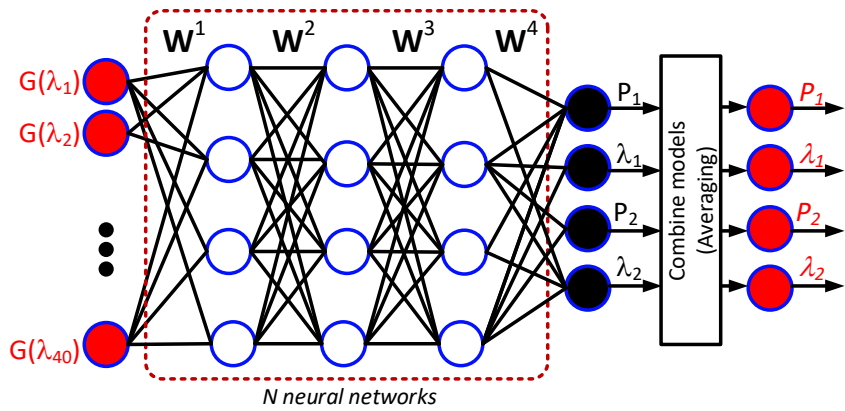
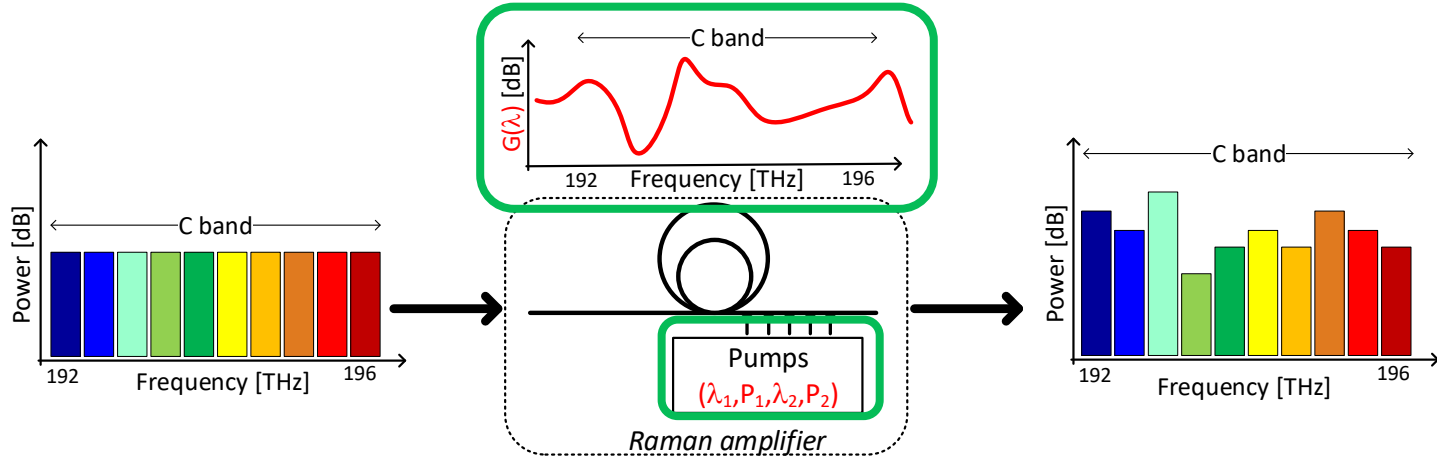
Fig. 1: Optical wavelength bands in the low-loss window of single-mode fibres. Wideband optical networks (WON) offer more than 10x increased optical bandwidth compared to C-band systems.



Employing O, E, S and L band requires rethinking optical amplification design

Raman amplifier inverse design

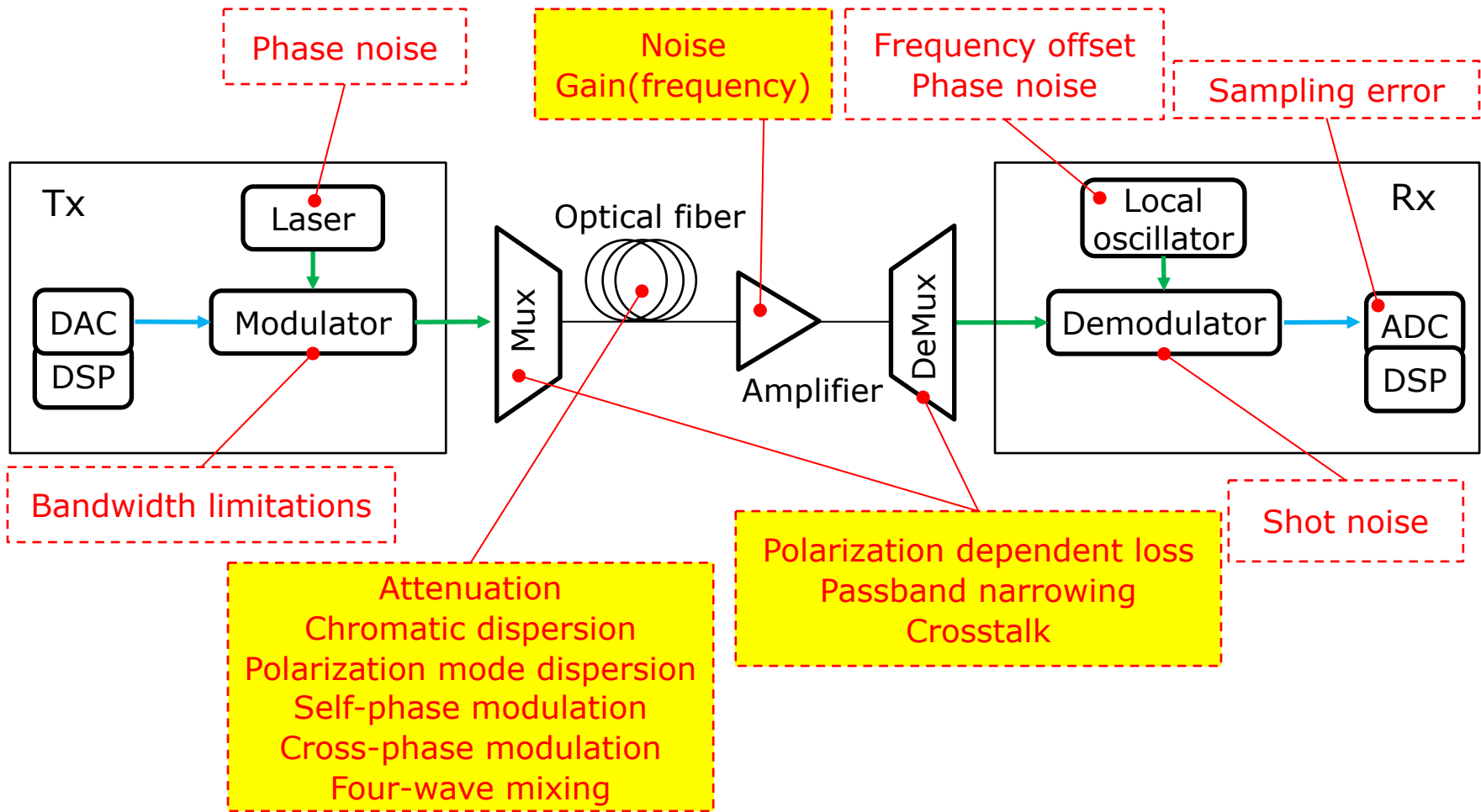
Objective: given a Raman gain profile determine pump powers and wavelengths



[1] D. Zibar et. al., Machine Learning-Based Raman Amplifier Design, OFC 2019, San Diego, CA, USA.

[2] D. Zibar et. al., Inverse System Design using Machine Learning: the Raman Amplifier Case, submitted JLT.

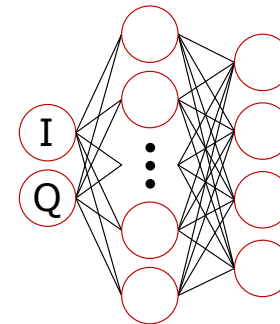
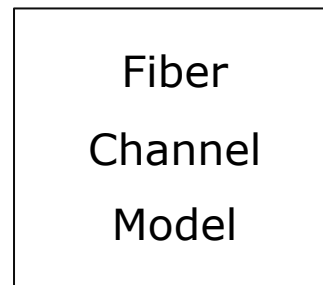
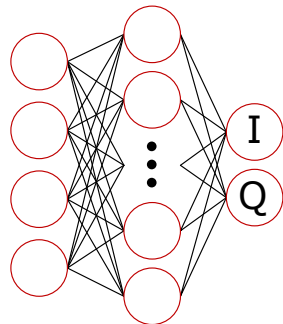
M-LiPS works



Learning to communicate using auto-encoders

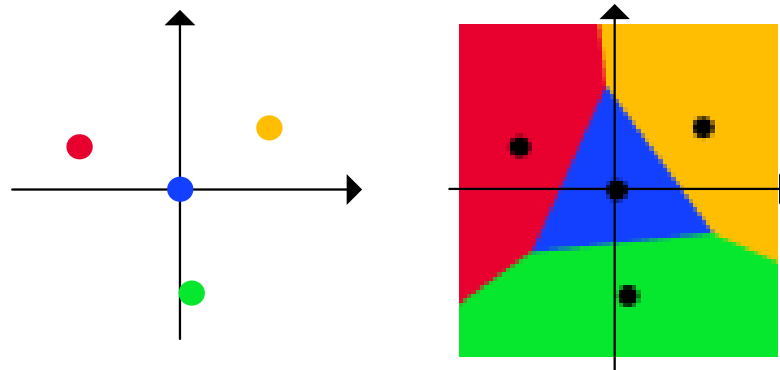
Input Space:

1	0	0	0
0	1	0	0
0	0	1	0
0	0	0	1



Output Space:

.8	.1	.1	.0
.1	.8	.0	.1
.1	.0	.8	.1
.0	.1	.1	.8

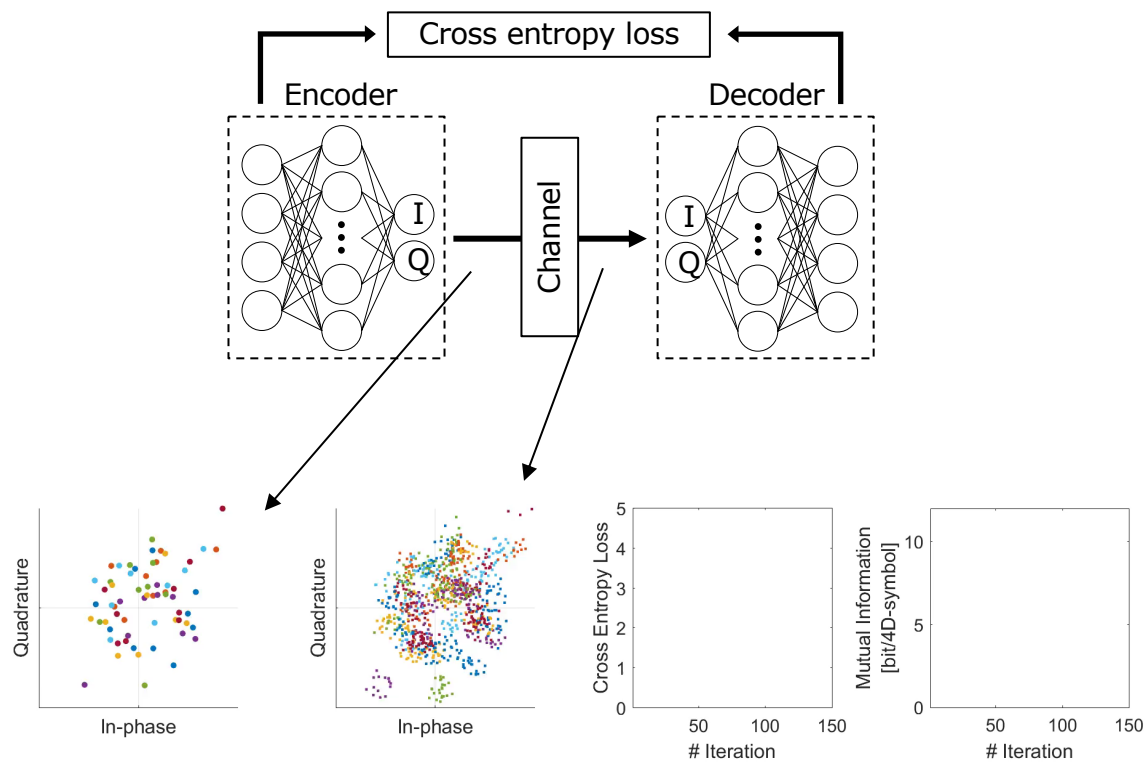


Objective: increase the transmitted information over a nonlinear channel

- [1] R. Jones et al., Deep Learning of Geometric Constellation Shaping including Fiber Nonlinearities, in Proceedings of ECOC 2018.
- [2] R. Jones et al., Geometric Constellation Shaping for Fiber Optic Communication Systems via End-to-end Learning, submitted to JLT
- [3] R. Jones et al., End-to-end Learning for GMI Optimized Geometric Constellation Shape, accepted ECOC 2019.

Learning to communicate using auto-encoders

Auto-encoder learning constellation robust to channel impairments



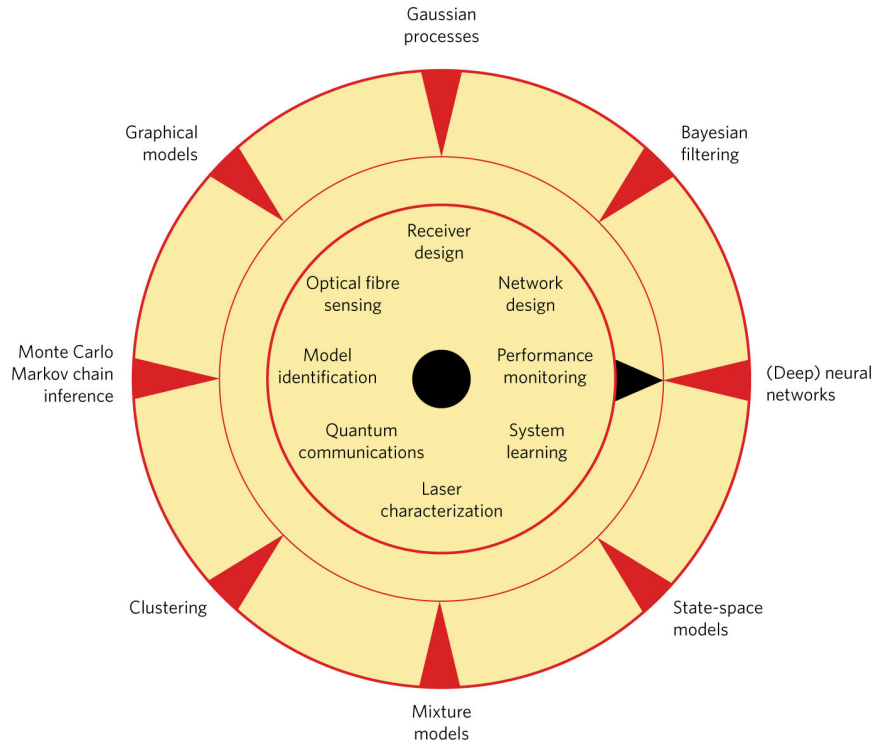
- [1] R. Jones et al., Deep Learning of Geometric Constellation Shaping including Fiber Nonlinearities, in Proceedings of ECOC 2018.
- [2] R. Jones et al., Geometric Constellation Shaping for Fiber Optic Communication Systems via End-to-end Learning, submitted to JLT

Group mission statement: research with impact



- Unifying framework for noise characterization of lasers, frequency combs and mode-locked lasers - move all functionalities into digital domain
- Nonlinear distortion-free communication over the nonlinear optical fibre channel
- Orders of magnitude improvements in accuracy of optical fibre sensor and quantum measurements using machine learning

Machine learning in optical communication



[1] D. Zibar et al., Machine learning under the spotlight, Nature Photonics, (11) 749-751, 2017

New topics 2019:

- **Photonic neural network**
- **Optical Amplifier design**
- **End-to-end learning**
- **Back-propagation learning**

[2] F. Musumeci, C. Rottondi, A. Nag, I. Macaluso, D. Zibar, M. Ruffini, M. Tornatore, "An Overview on Application of Machine Learning Techniques in Optical Networks," in IEEE Communications Surveys & Tutorials, vol. 21, no. 2, pp. 1383-1408, 2019.

Problems that will benefit from ML



- Noise characterization of lasers and frequency combs:
 - Optical phase tracking at the quantum limit
 - Noise correlation matrix of frequency combs lines
- Design of optical components (inverse system design):
 - Given laser linewidth and noise find the physical parameters
 - Given modulator bandwidth find the physical parameters
 - Instead of running time-consuming simulation build fast ML based models
- Optical amplifiers for multiband-wavelength and SDM systems:
 - Complex relation between pumps and gain
 - Pump power and wavelength allocation for specific gain profile challenging
 - Minimization of mode dependent loss
- Communication over the nonlinear fiber-optic channel:
 - Channel highly complex
 - Capacity **unknown?**
 - Optimum receiver architecture **unknown**
 - Optimum modulation and pulse-shapes **unknown**

SDM: spatial division multiplexing

Problems that will not benefit from ML



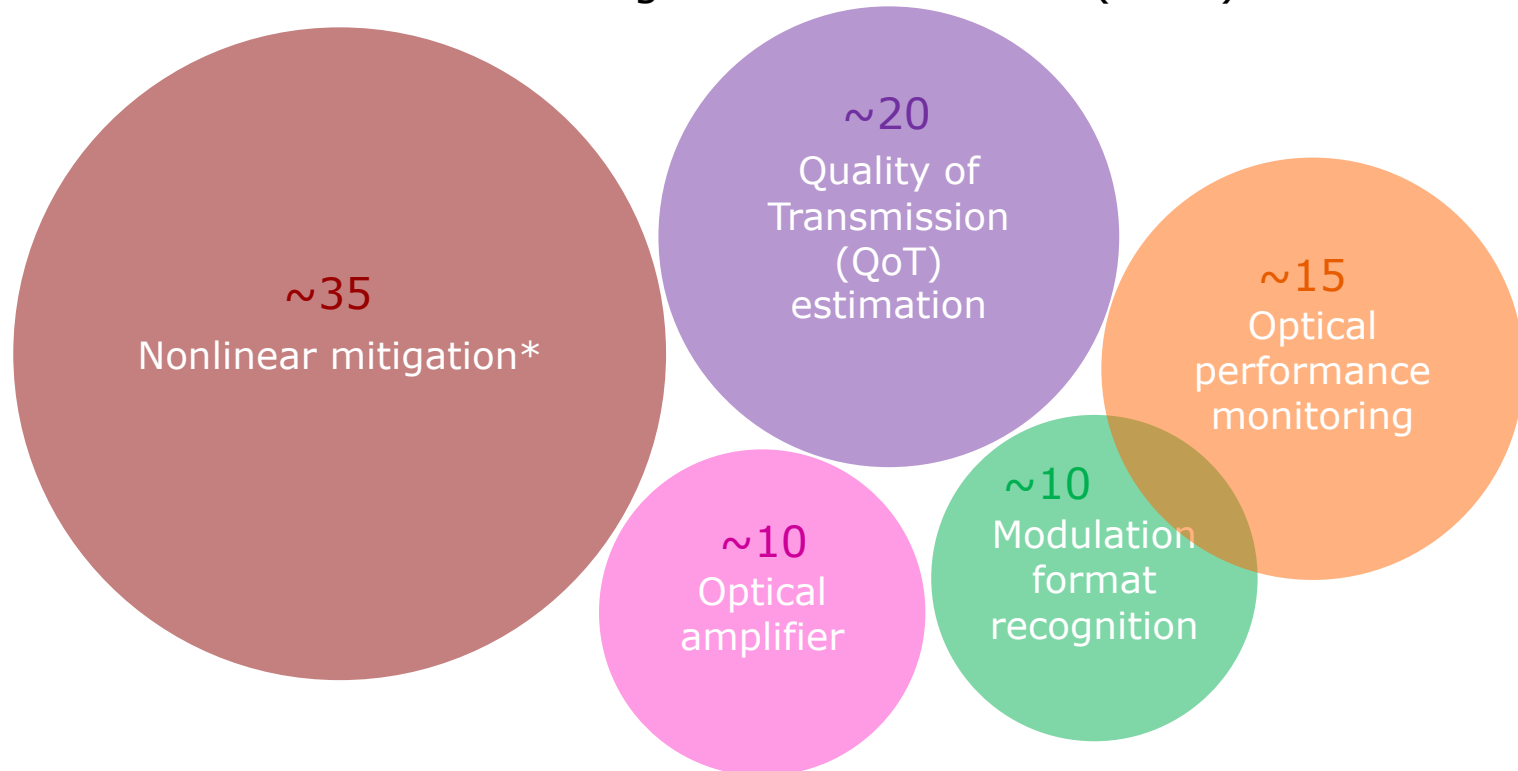
- Problems that we have a good knowledge, (low cost) models and analytical solutions such as:
 - Linear impairment compensation (chromatic dispersion) in coherent systems
 - EDFA design*

* Possible in an SDM scenario

State-of-the-art (physical layer)



- Number of solutions according to the use case^[1-3] (2019)



* Including receiver/transmitter design and end-to-end learning

[1] J. Mata, I. de Miguel, R. J. Durán, N. Merayo, S. K. Singh, A. Jukan, M. Chamania, Artificial intelligence (AI) methods in optical networks: A comprehensive survey, *Optical Switching and Networking*, 2018.

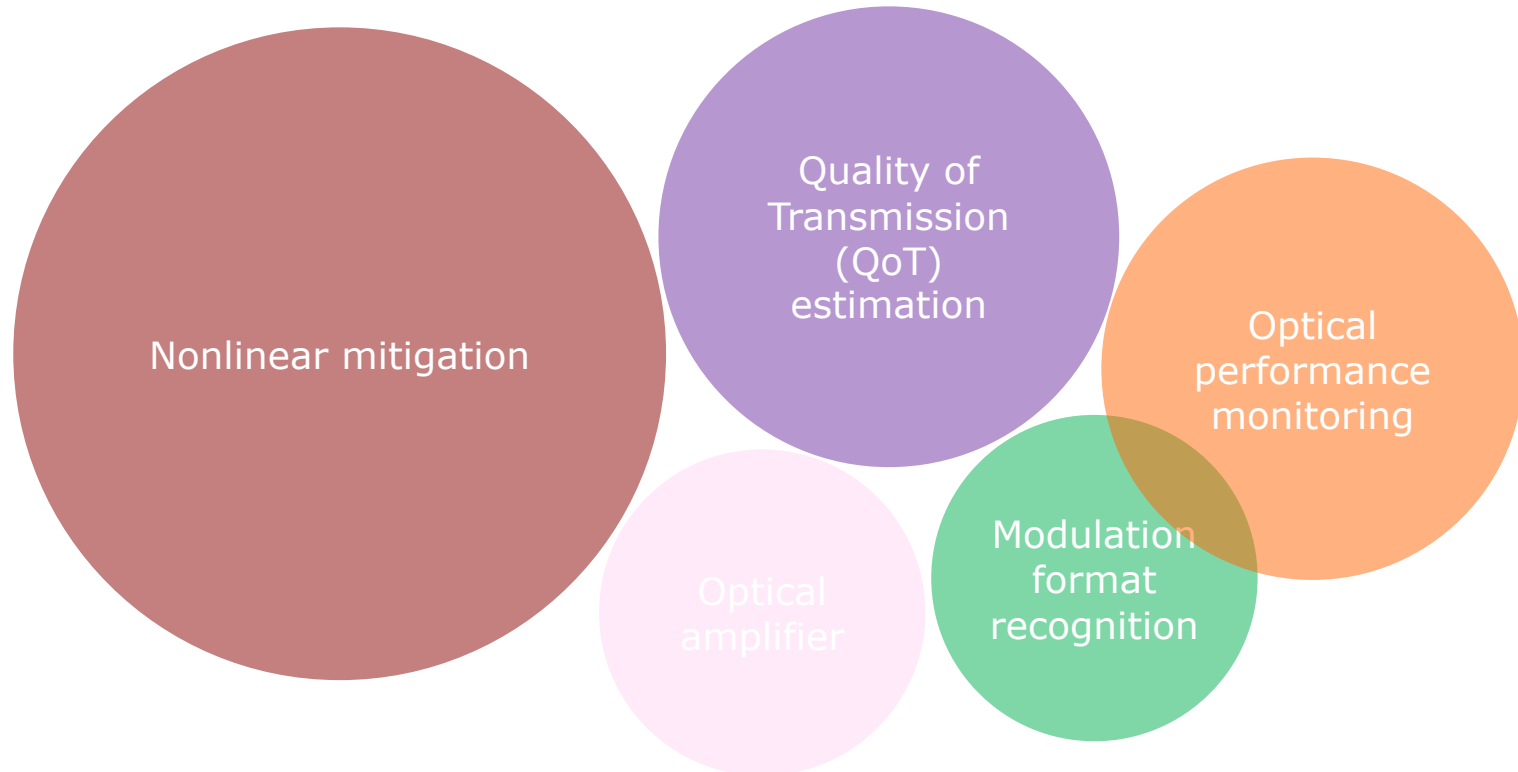
[2] F. Musumeci, C. Rottondi, A. Nag, I. Macaluso, D. Zibar, M. Ruffini, M. Tornatore, "An Overview on Application of Machine Learning Techniques in Optical Networks," in *IEEE Communications Surveys & Tutorials*, vol. 21, no. 2, pp. 1383-1408, 2019.

[3] F. N. Khan, Q. Fan, C. Lu and A. P. T. Lau, "An Optical Communication's Perspective on Machine Learning and Its Applications," in *JLT*, vol. 37, no. 2, pp. 493-516, 2019.

What data matters (M-LiPS view)

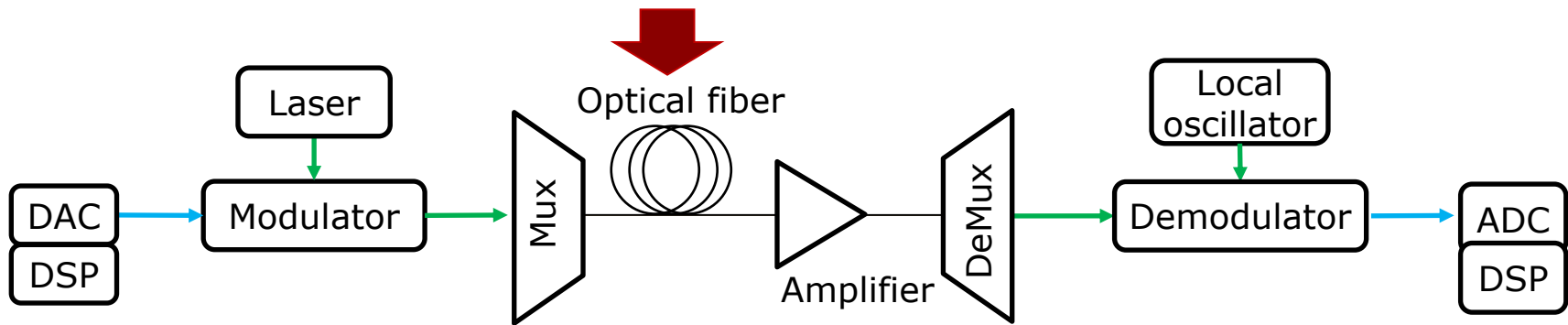


Does it depend on the use case?



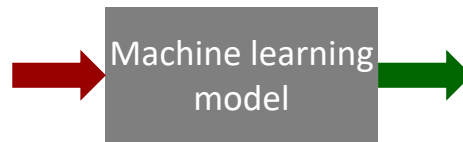
What data matters for nonlinear mitigation

Self-phase modulation (SPM)
Cross-phase modulation (XPM)
Four-wave mixing (FWM)



Aid/replace: DSP

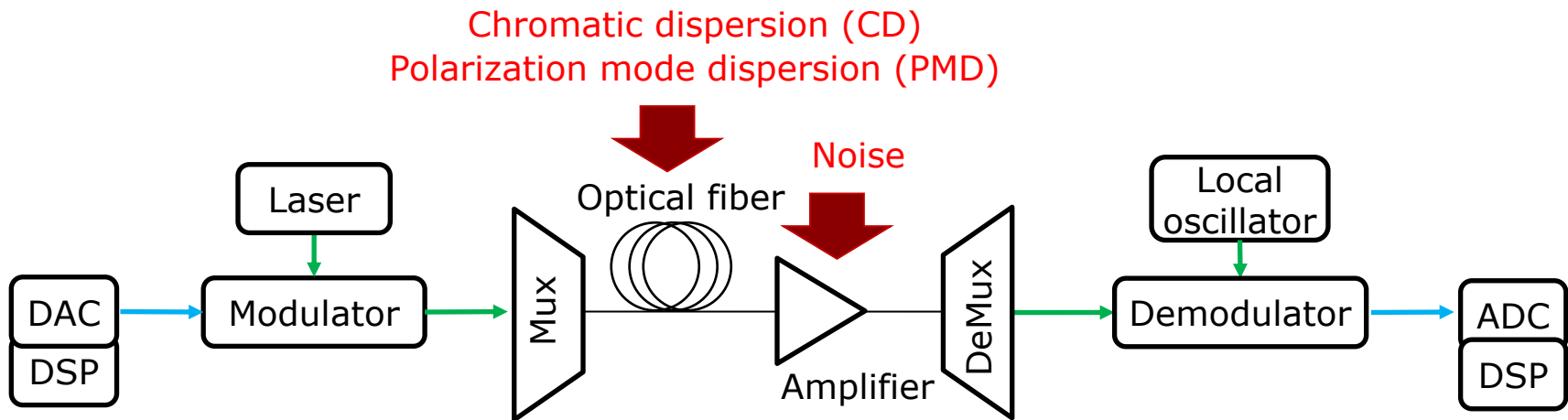
- Received constellation
- Received symbols



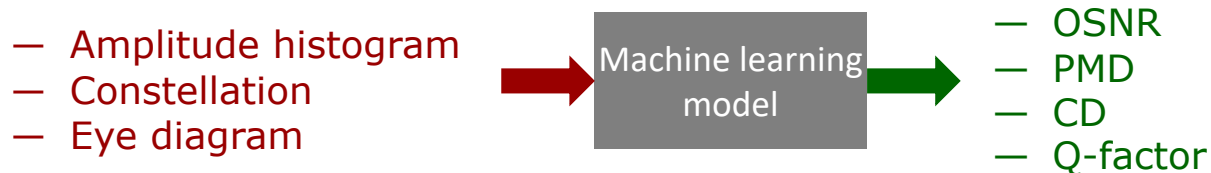
- Decoded symbols with impairment estimated or mitigated
- Nonlinearity mitigated constellation points
- Symbol decision boundaries

Data generation: random sequence of bits

What data matters for optical performance monitoring

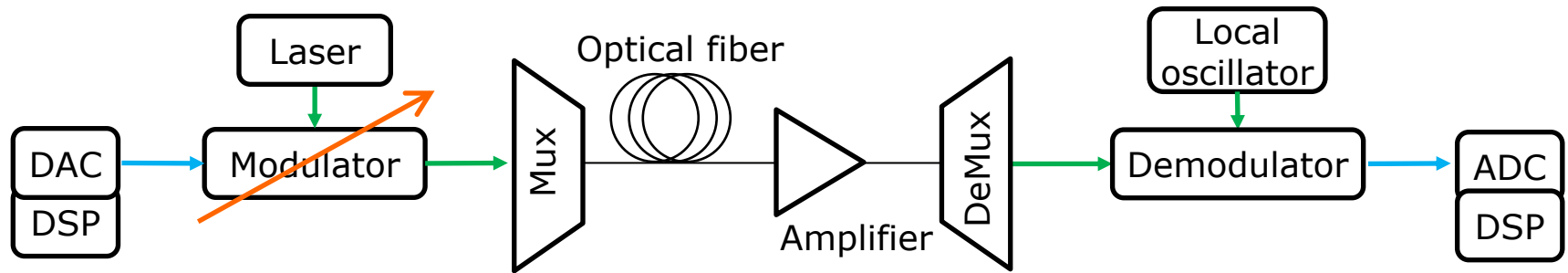


It will trigger actions depending on the current performance

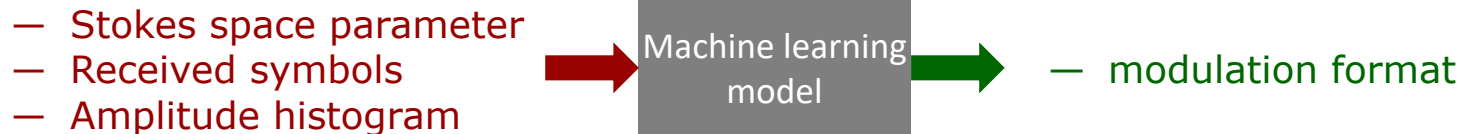


Data generation: vary the interested parameter (CD, PMD, noise) or a combination + random sequence of bits

What data matters for modulation format recognition

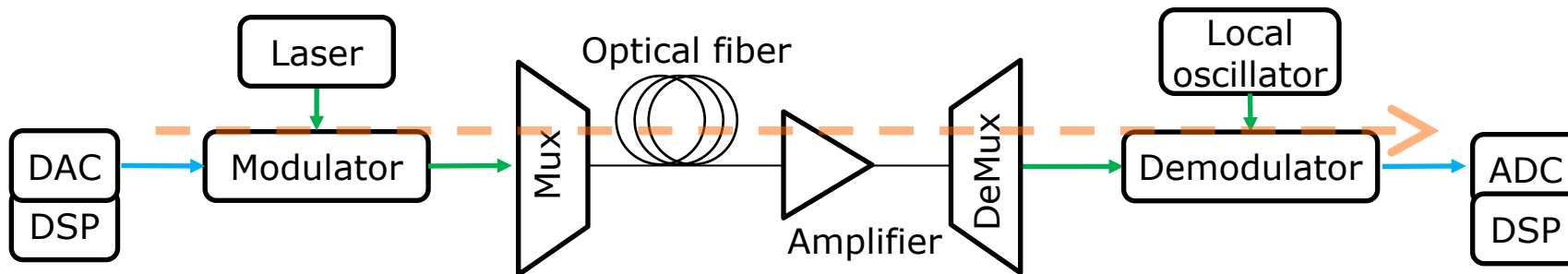


Flexible transmitters/receivers



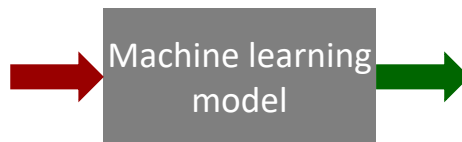
Data generation: vary the modulation format + random sequence of bits

What data matters for quality of transmission estimation



Estimation before connection deployment

- Historical OSNR
- Historical SNR
- Lightpath features*



- BER
- OSNR
- SNR

Data generation: historical data collected from monitors or generate different connections with different lightpath features with the real QoT measured.

*Lightpath features: number of links, links' lengths, number of amplifiers along the link, modulation format, baud rate, channel launch power, ...

What data matters (M-LiPS view)



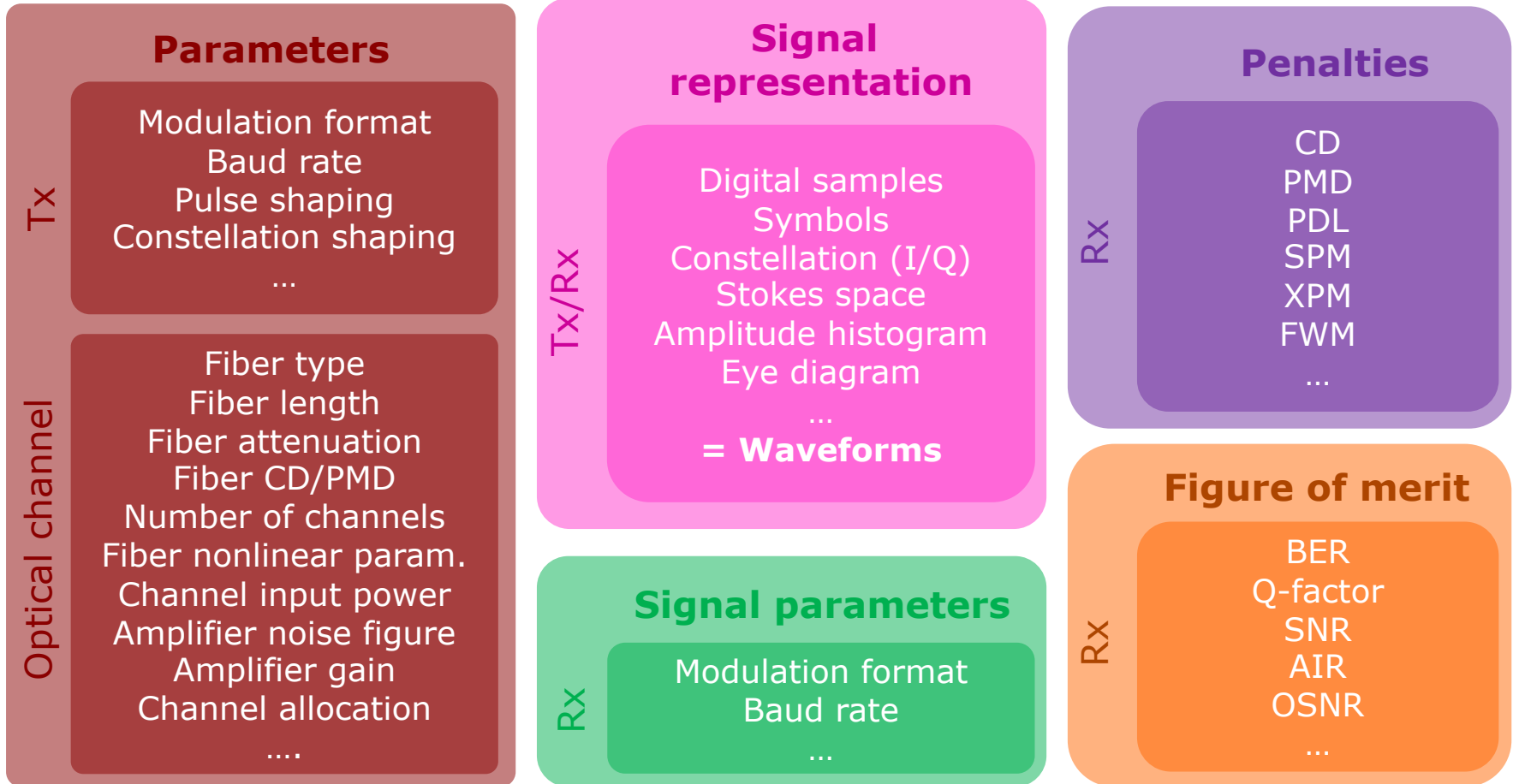
Does it depend on the use case?

Yes

But for most of the cases on the literature, the **input data** can be represented by the received waveforms (after ADC)

And the **output data** will depend on the application

Summarizing



Thank you for your attention.

Acknowledgements



CoG FRECOM (grant agreement no. 771878)



European Union's Horizon 2020 research and innovation programme under the Marie Skłodowska-Curie grant agreement No 754462.