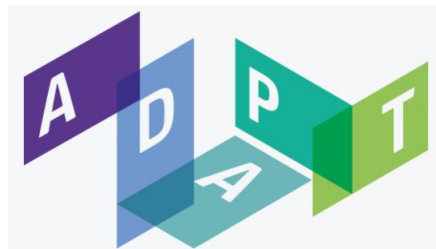


Machine Learning Algorithms for Optical Fiber Telecoms



Dr. Elias Giacomidis

26 March 2018



- **Bangor University, Wales, UK (PhD)**
 - Optical transmission for >40-Gb/s local and access networks
- **Athens Information Technology centre, Athens, Greece**
 - Passive optical networks (PONs)
- **Telecom Paris-Tech, France (collaboration with France Telecom-Orange Labs)**
 - Coherent optical communications for >100-Gb/s multi-channels
- **Aston University, UK**
 - Digital signal processing (DSP)-based fibre nonlinearity compensation
- **University of Sydney, Sydney, Australia**
 - Machine learning DSP for optical commun. and photonic-chip applications
- **Dublin City University (DCU), Ireland (visiting researcher at Xilinx-Ireland)**
 - Real-time machine learning DSP for optical communications



Machine learning for optical communications

Machine Learning

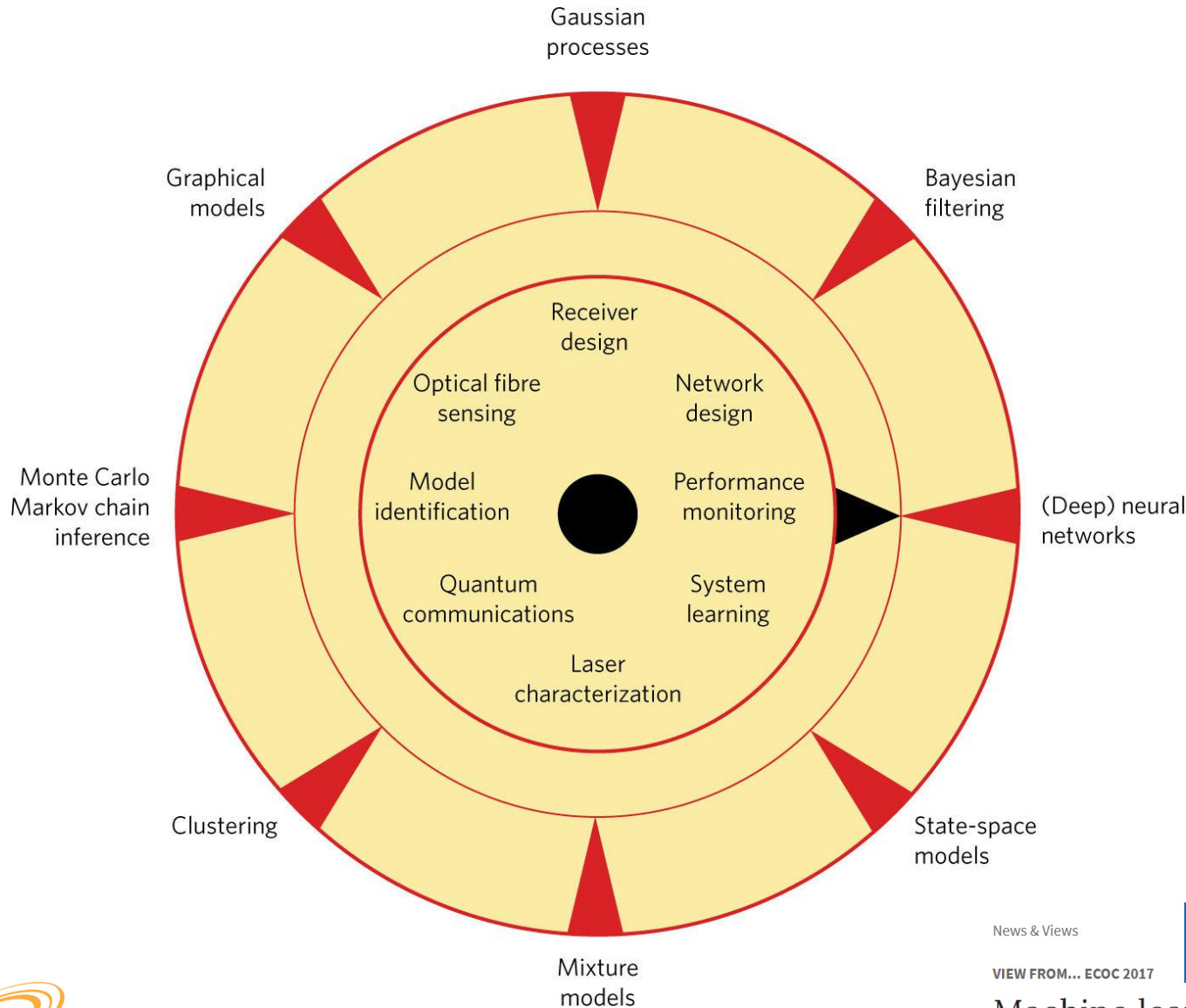
- DET, KDE, GMM
- PCA, ICA
- IS, MCMC
- HMM (EKF, UKF, PF)
- ANN (MLP, HNN, RBM, CNN, RNN)
- SVM (kernel: polynomial, RBF, sigmoid)
- Deep learning (DBN, etc.)

Apply

Optical Communications

- Linearity: CD, PMD
- Nonlinearity: SPM, XPM, XPolM, FWM
- Nonlinear equalization
- Polarization recovery
- Carrier phase recovery
- Nonlinear capacity analysis
- Coded-modulation design

Photonics: machine learning under the spotlight



News & Views

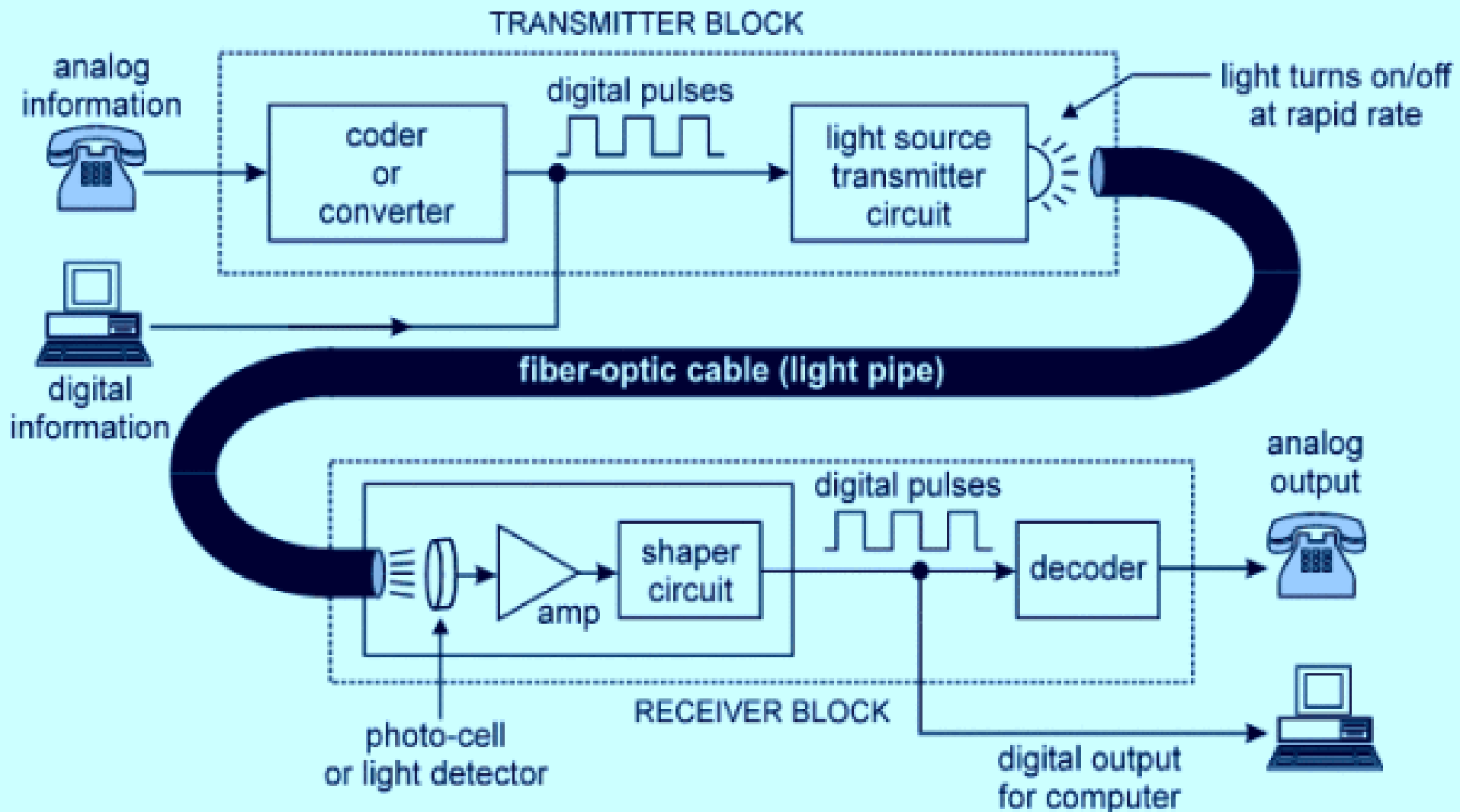
VIEW FROM... ECOC 2017

nature
photonics

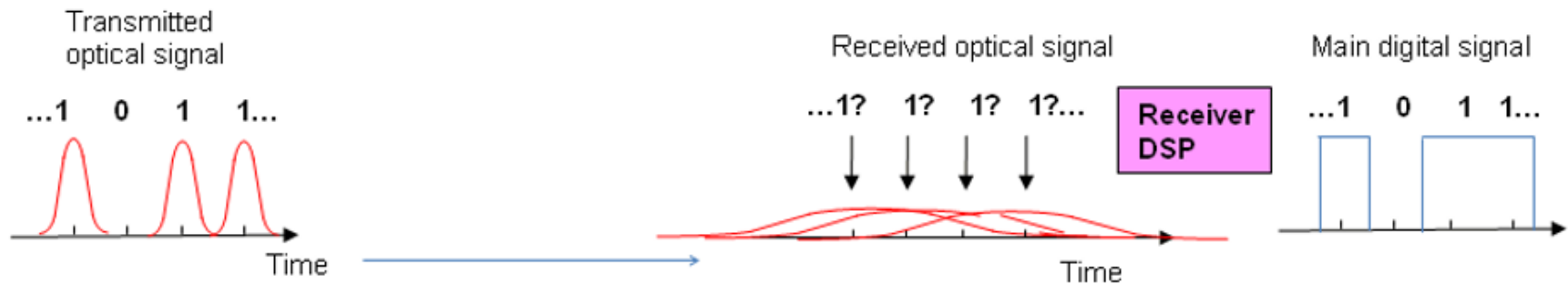
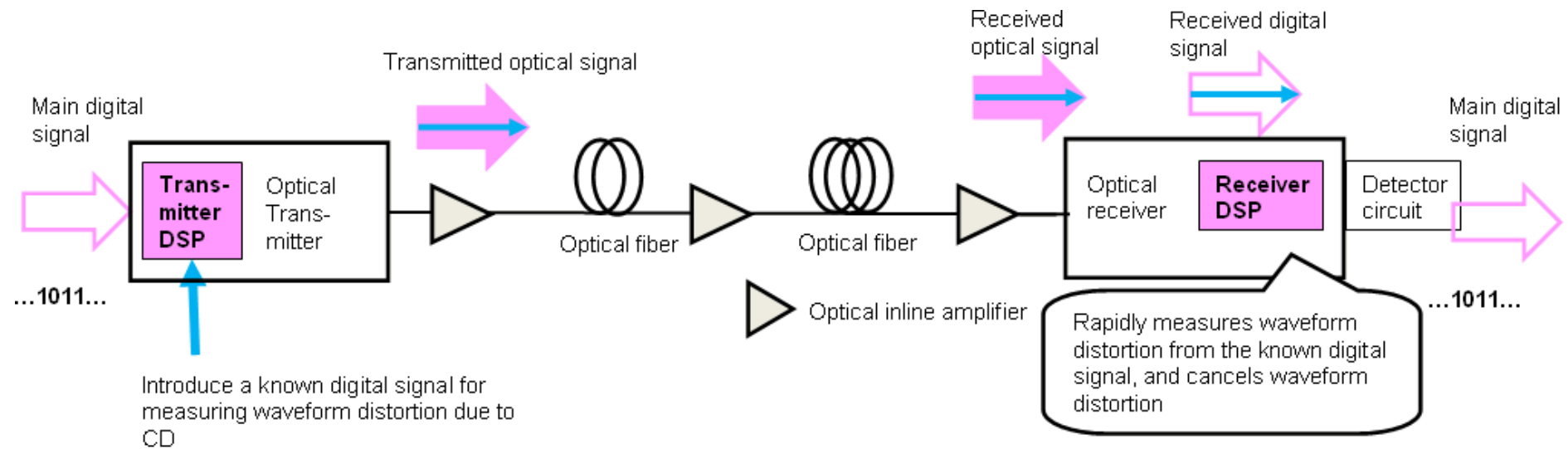
Machine learning under the spotlight

Darko Zibar , Henk Wymeersch  & Ilya Lyubomirsky 

Typical optical communication system

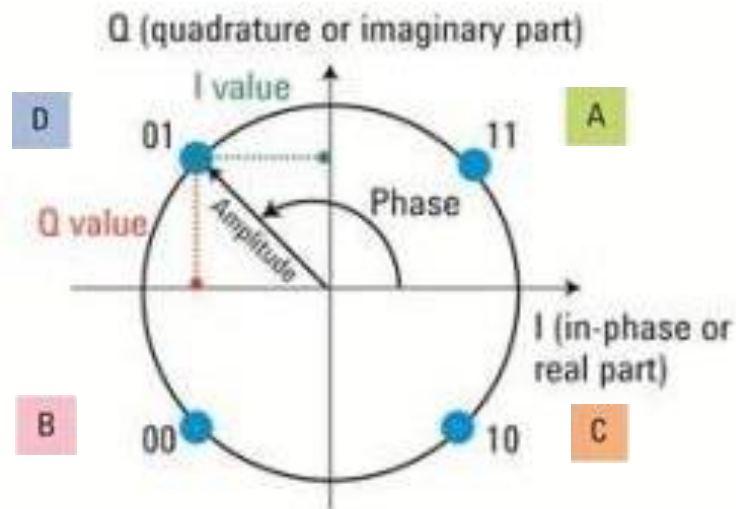


DSP importance in optical communications



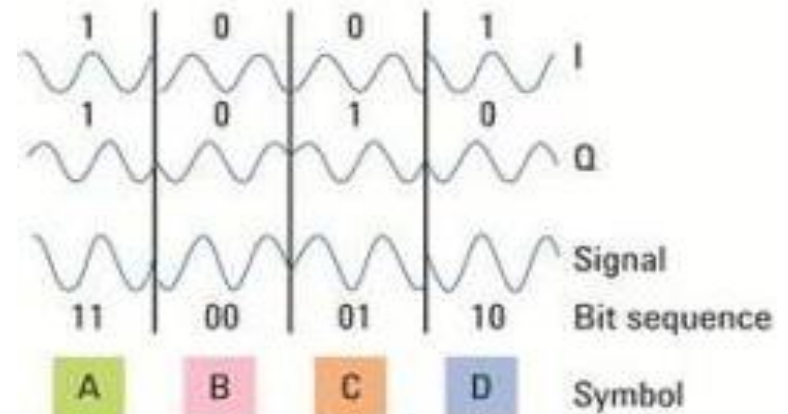
Constellation diagrams for modulation

Constellation diagram



- 1) Information is in amplitude
- 2) Information is in phase

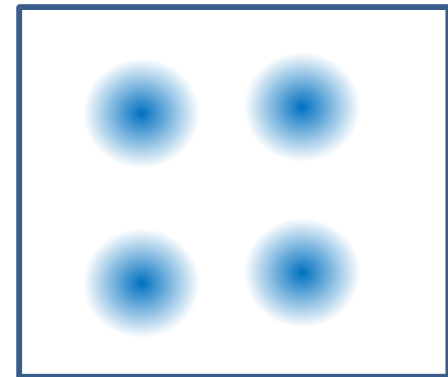
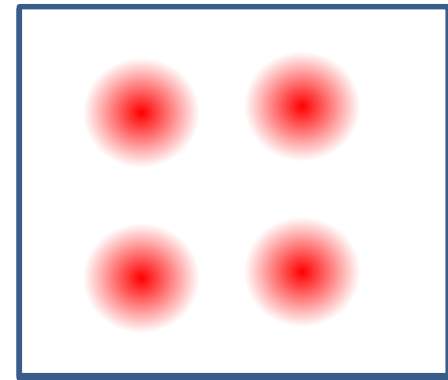
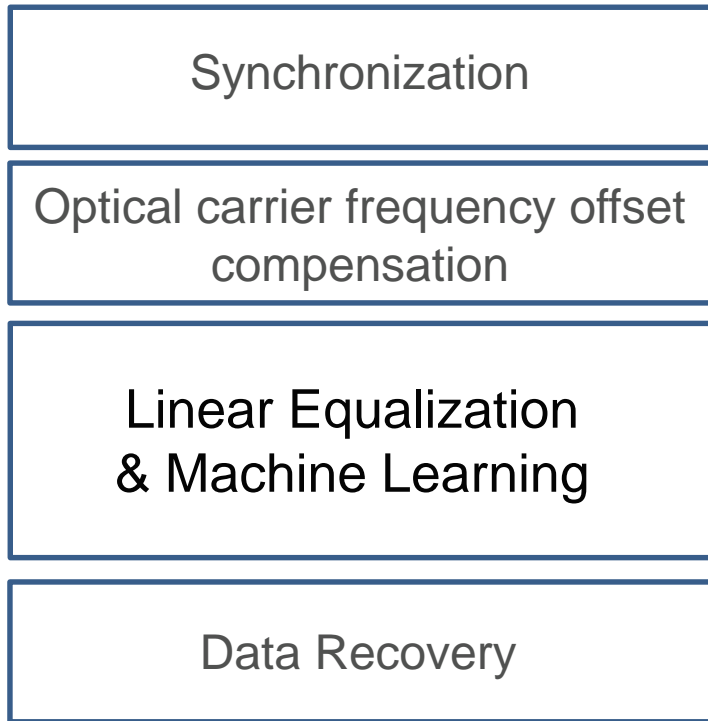
Time domain waveforms



We have constructed 4 vectors
→ One vector position in the complex plane codes 2 bits

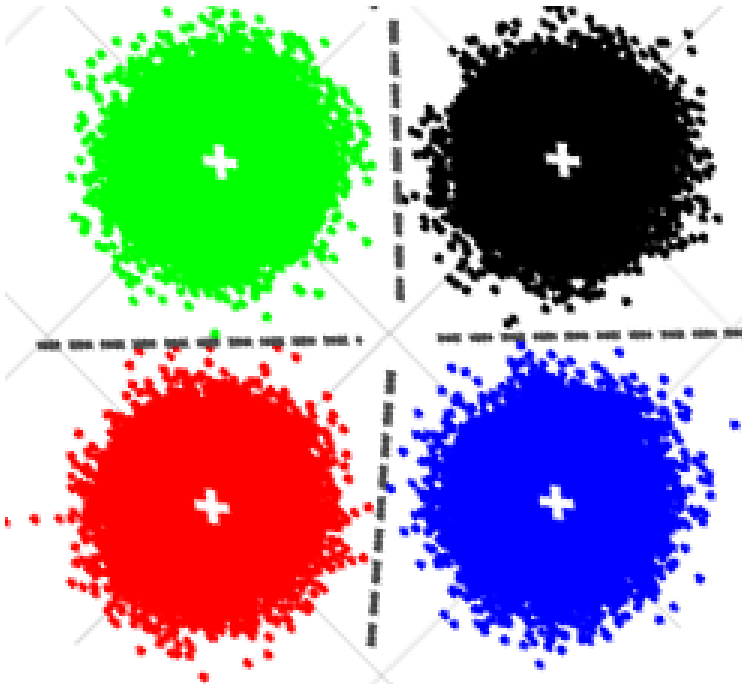
DSP receiver design with machine learning

■ DSP Receiver processing:

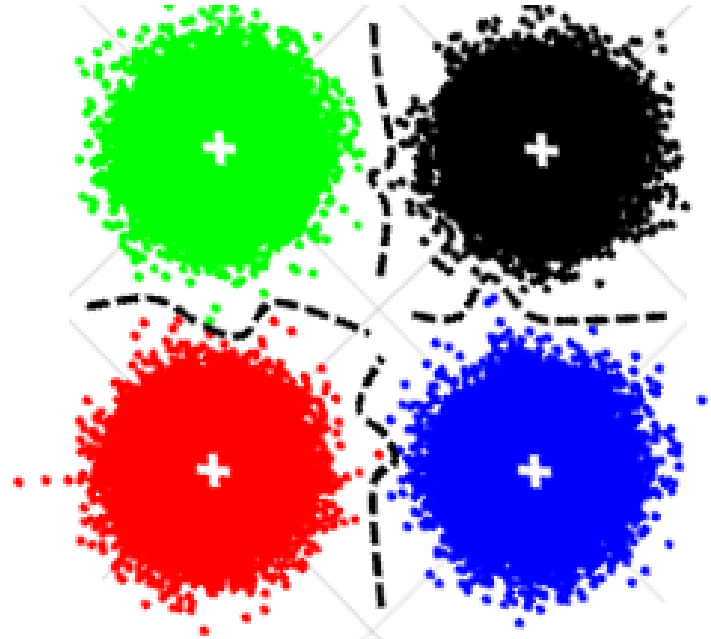


Clustering-based machine learning

K-means

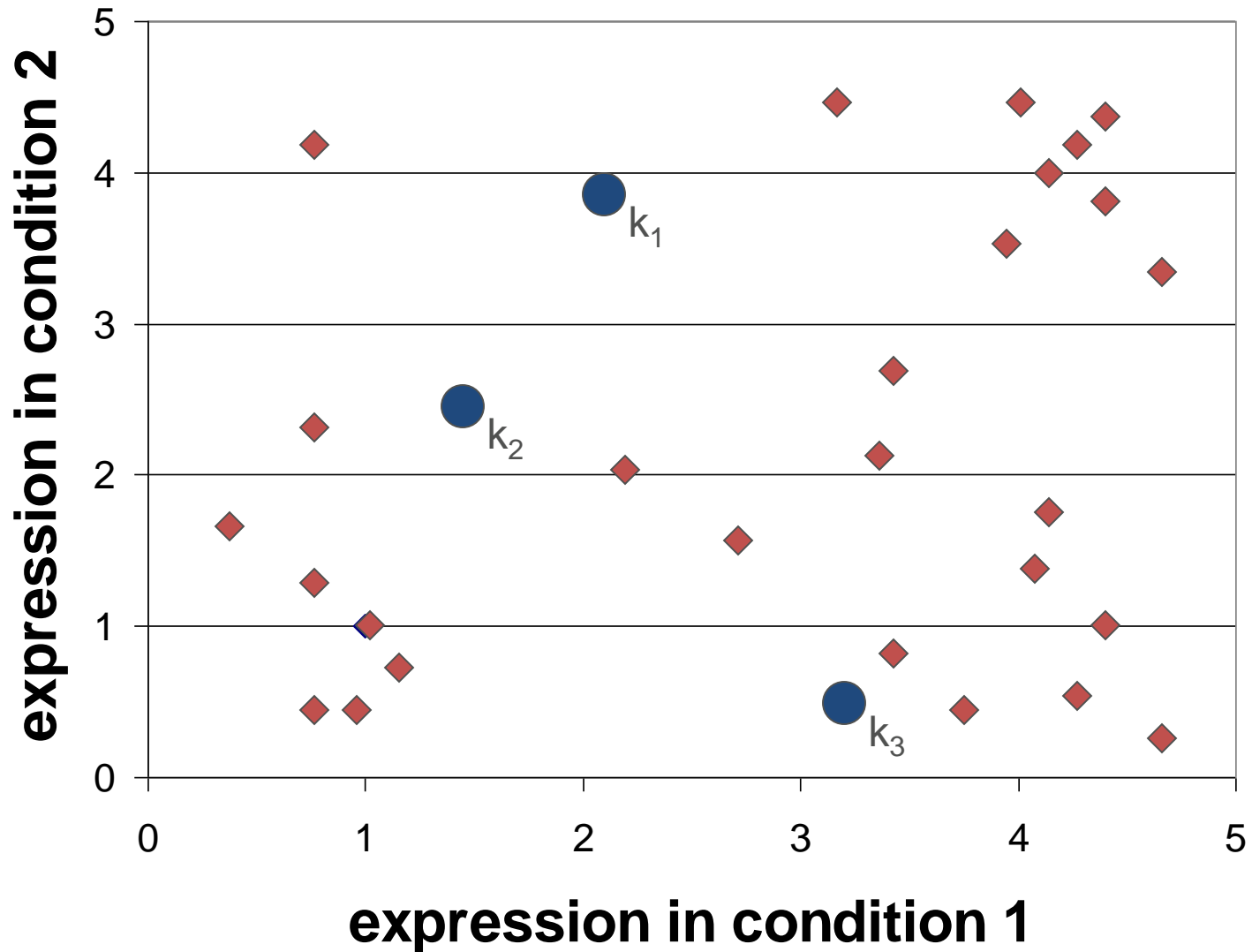


Fuzzy-logic c-means

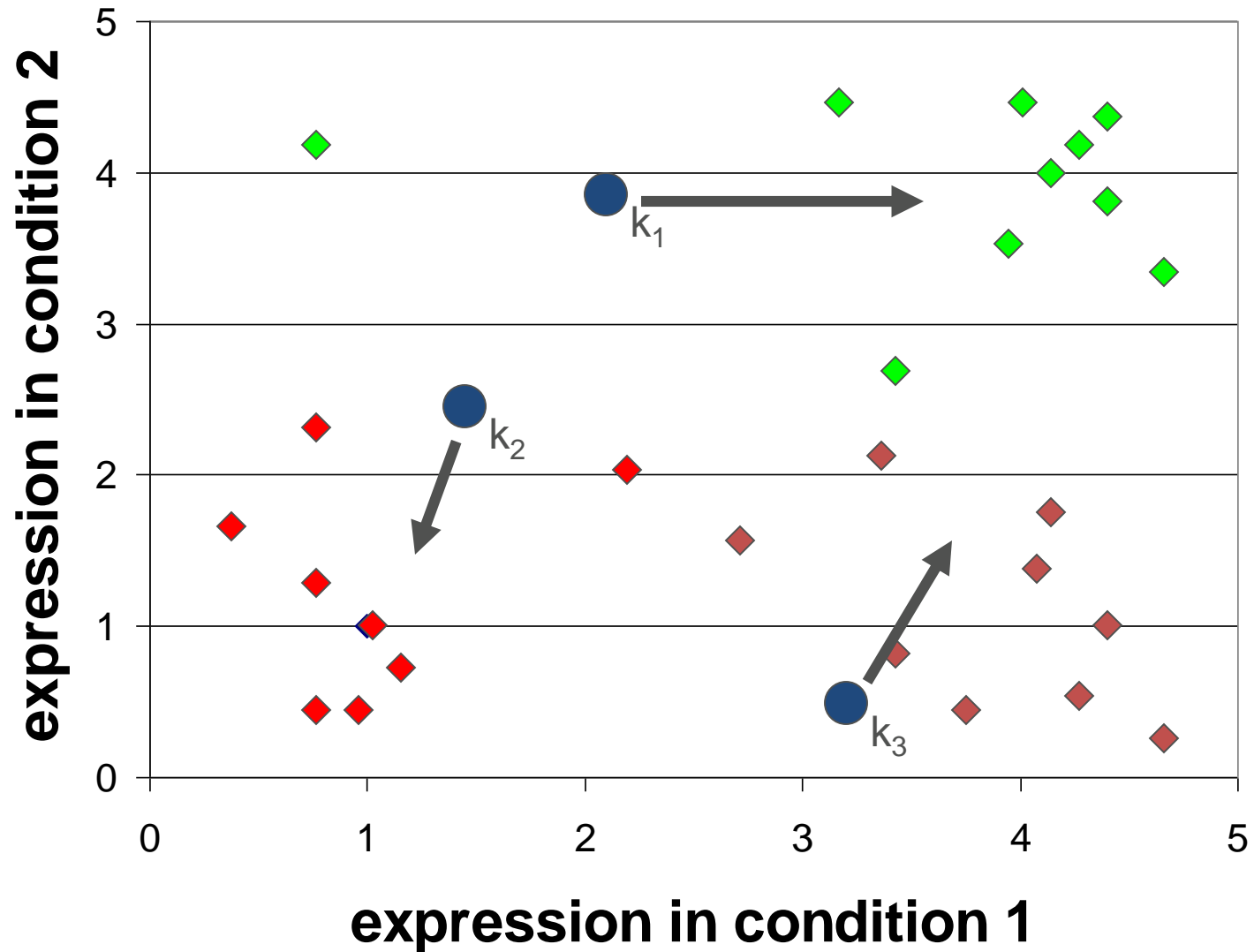


K-means: Step 1

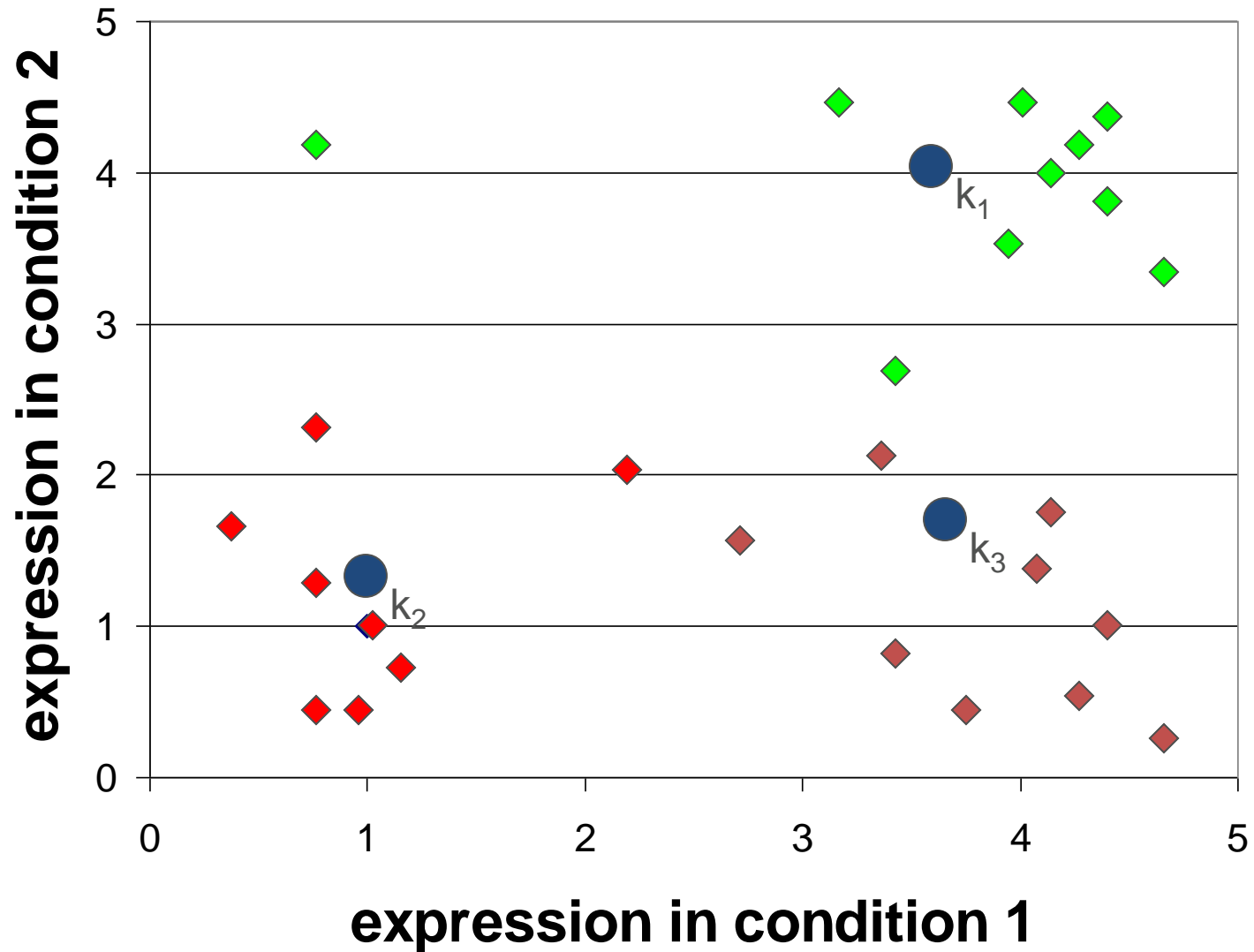
Algorithm: k-means, Distance Metric: Euclidean Distance



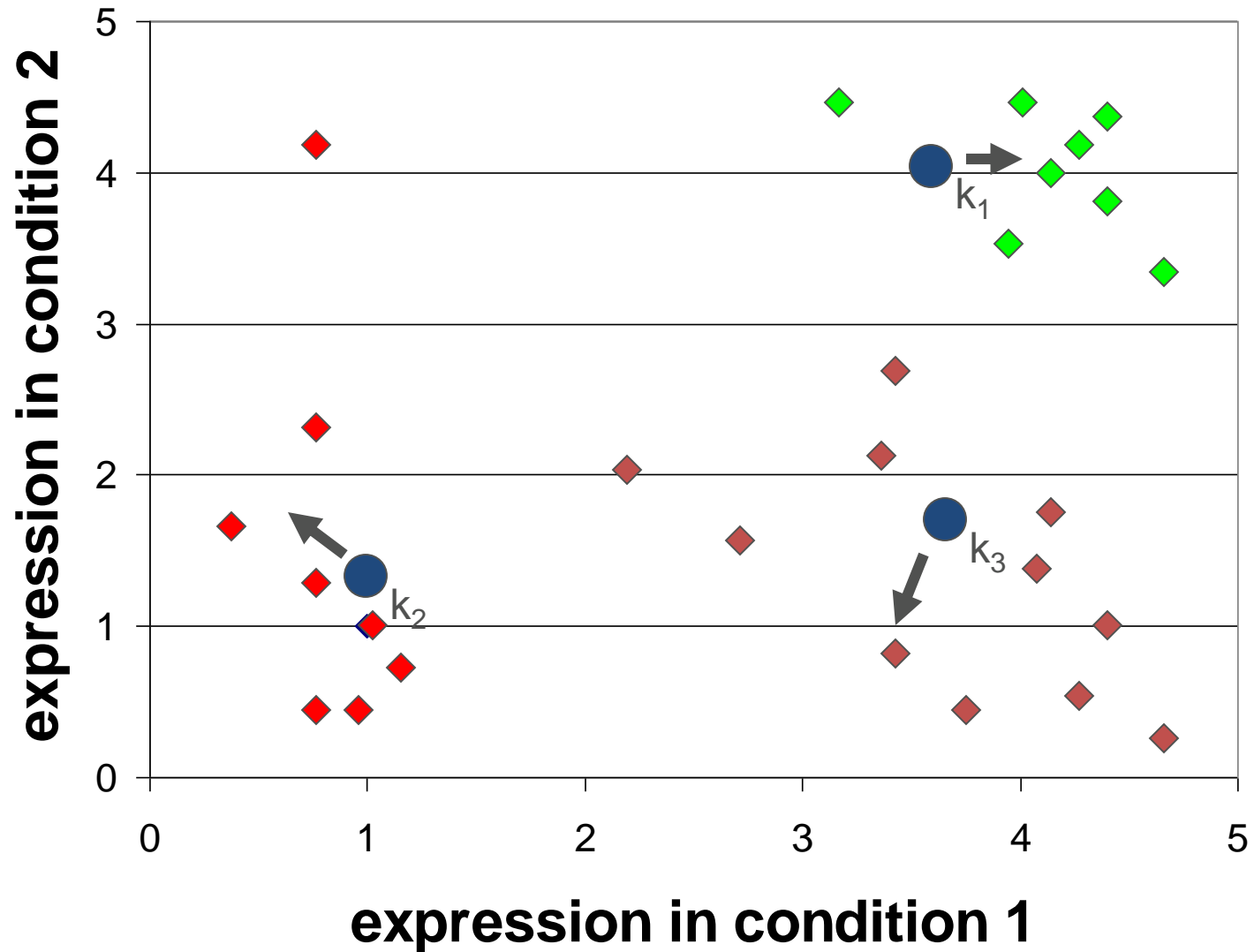
K-means: Step 2



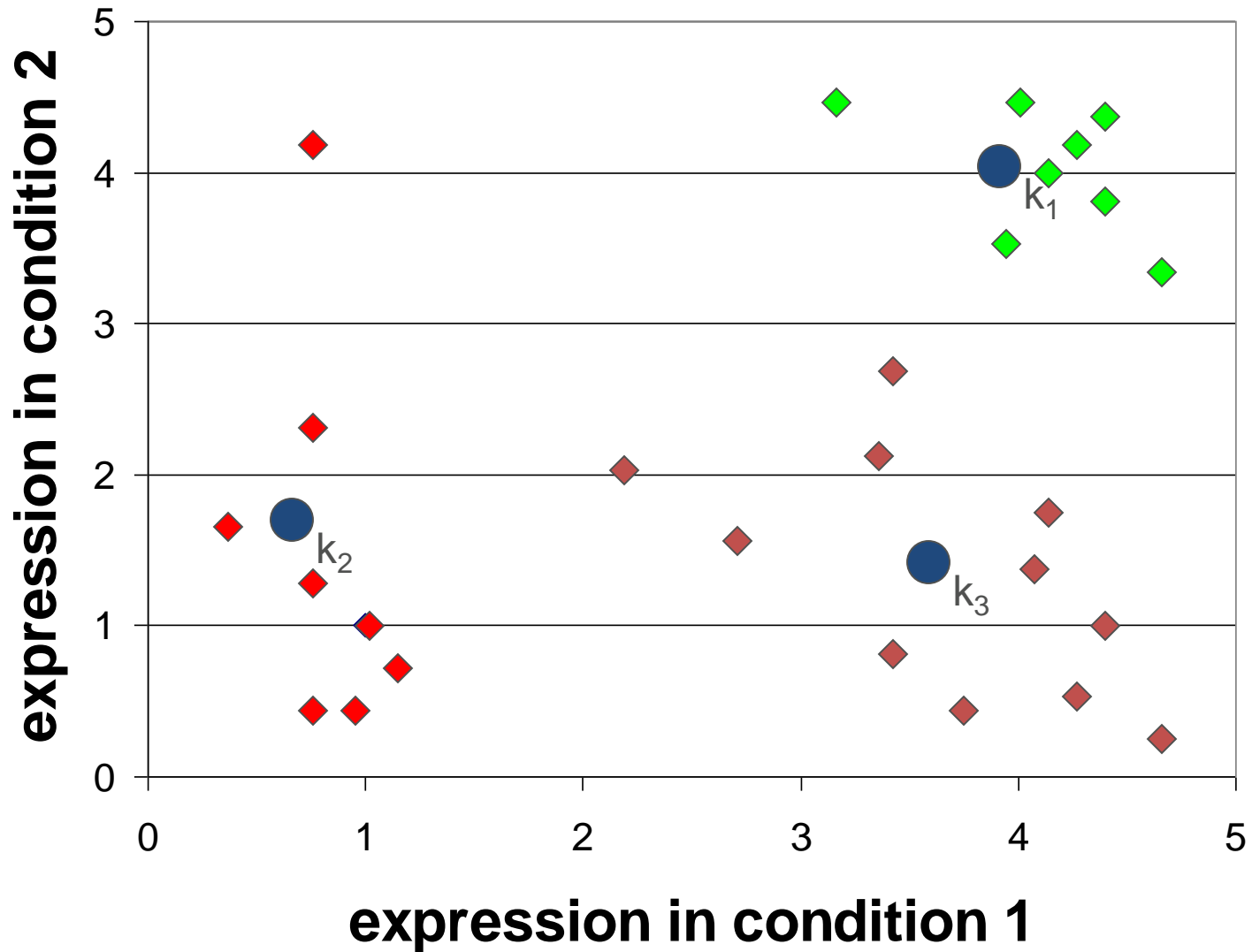
K-means: Step 3



K-means: Step 4



K-means: Step 5

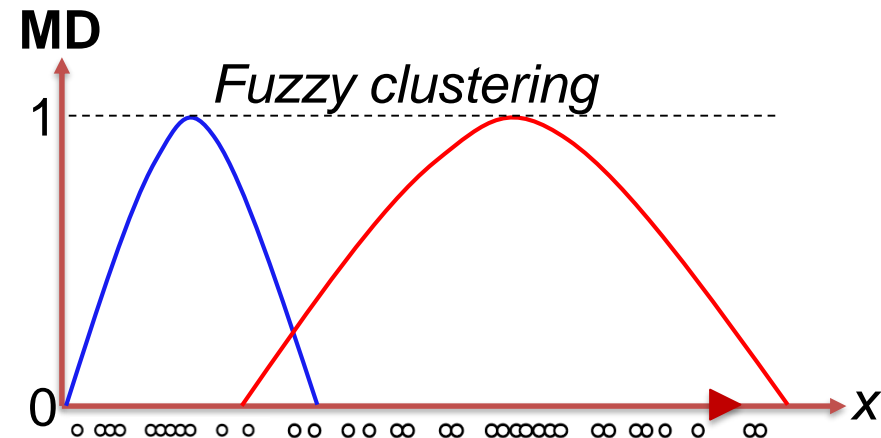
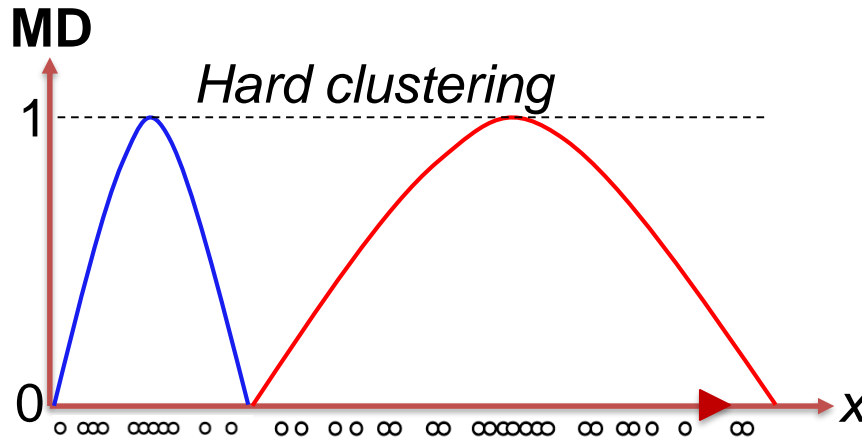


Fuzzy-logic c-means

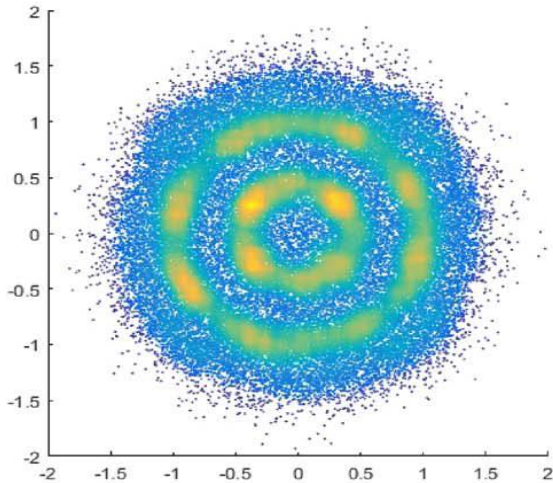
Single-dimensional data



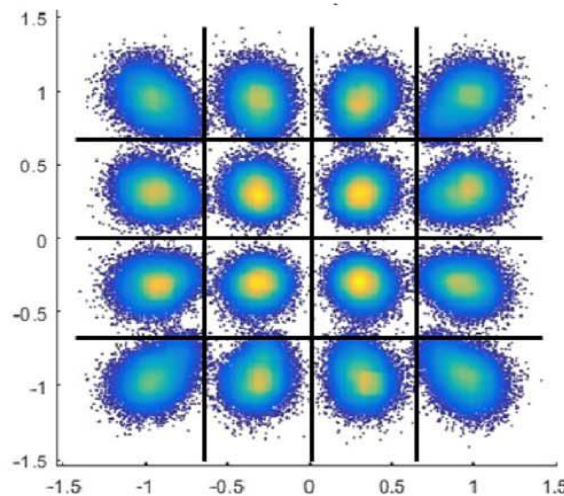
MD: Membership Degree



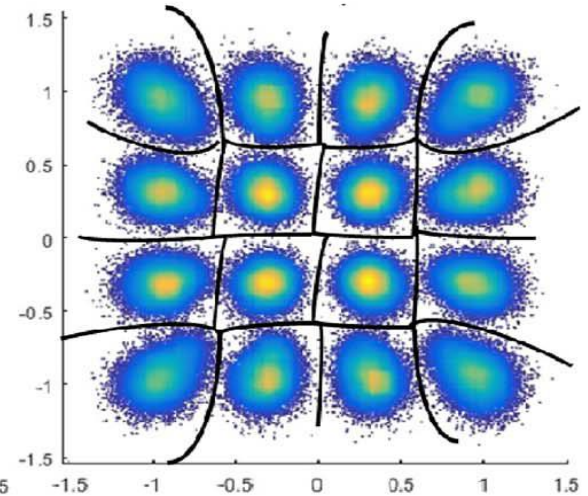
Received constellation diagrams for 16-QAM



No equalization



**Linear equalization -
hard decision
boundaries**

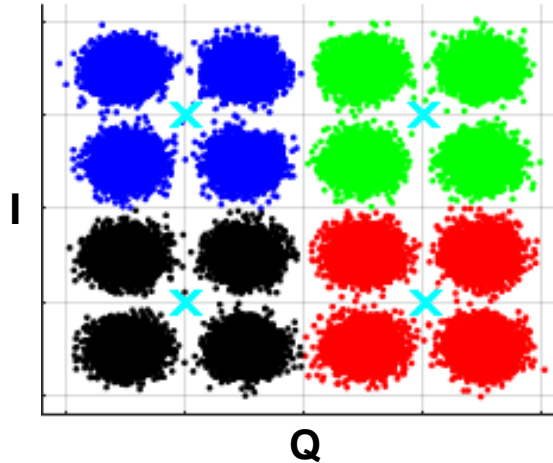


**Machine learning -
soft decision/nonlinear
boundaries**

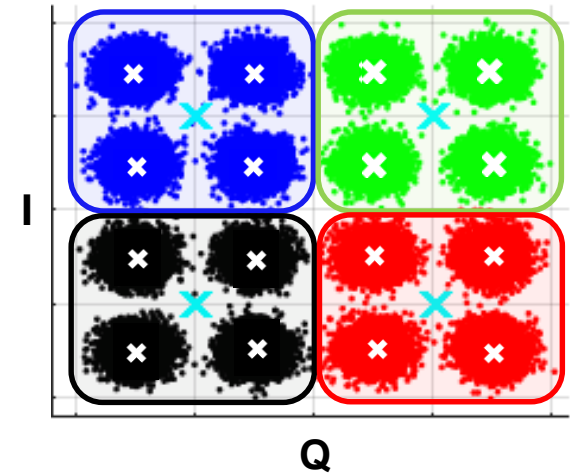
Alternative design for 16 clusters

CASE-1

Step 1: large group of 4 clusters

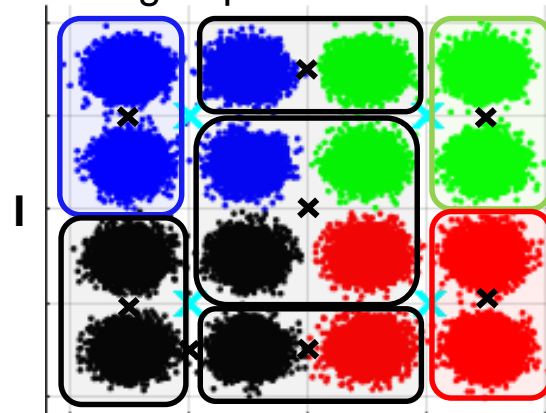


Step 2: 4 groups of 4 clusters

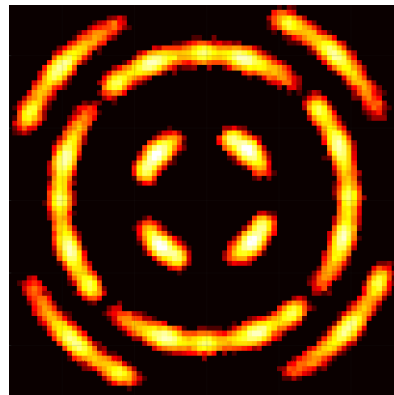
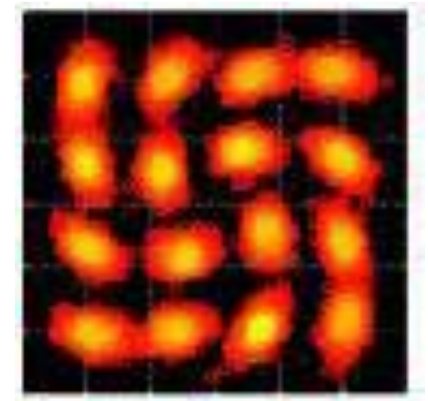
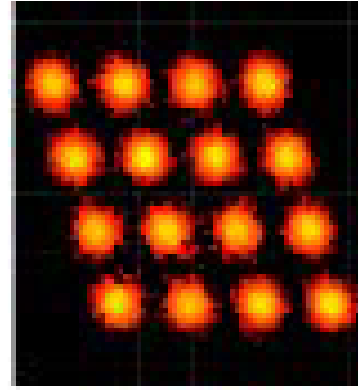
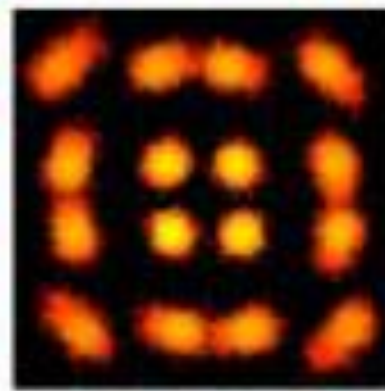
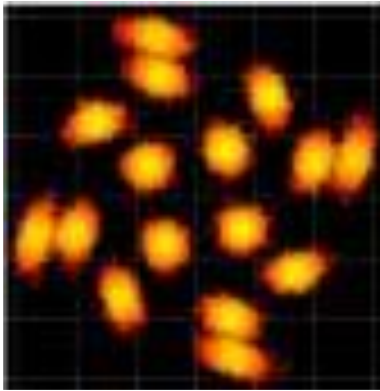


Single-step: 1 group of 4 clusters &
6 groups of 2 clusters

CASE-2

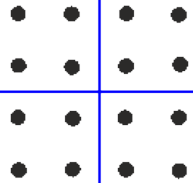


Shapes of constellation diagrams



Transceiver setup

16QAM



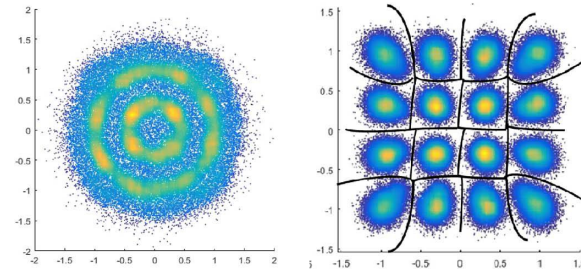
MATLAB

Electrical Transmitter



MATLAB

Electrical Receiver



DSP
transmitter

Digital-to-
Analogue
Conversion

Optical
modulation

Optical
channel

Photo-
Detection

Analogue-
to-Digital
Conversion

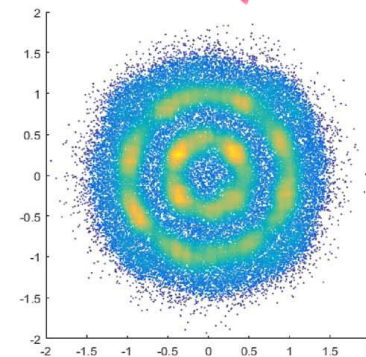
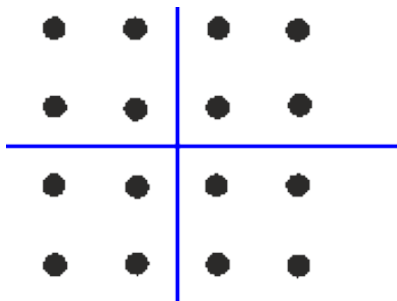
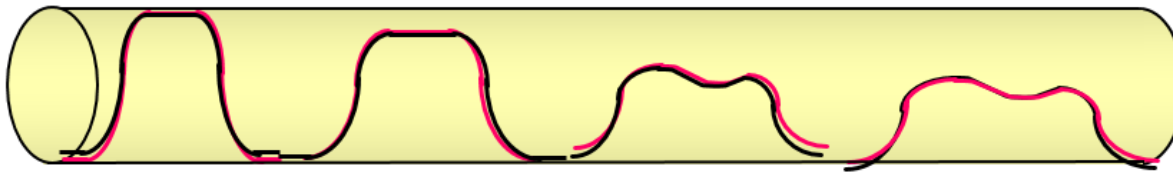
DSP receiver with
machine learning



Nonlinear distortion

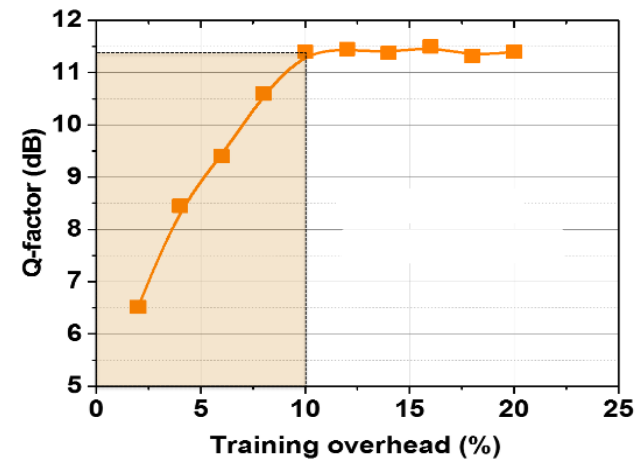
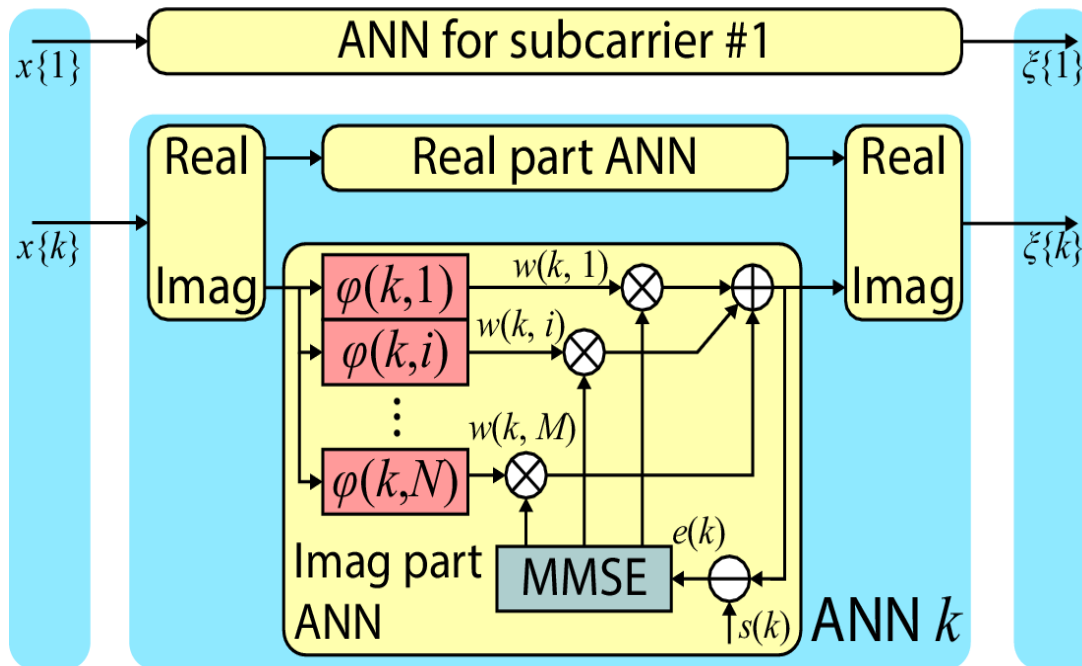
High power intensity → Refractive index change → Frequency chirp → Spectrum broadening → Waveform distortion due to chromatic dispersion

Optical fiber



Artificial Neural Network design

■ ANN: Artificial Neural Network



$\varphi_{k,i}(x)$ = nonlinear transformations of subcarrier k

N = level of constellation mapping

w = weights

e = error

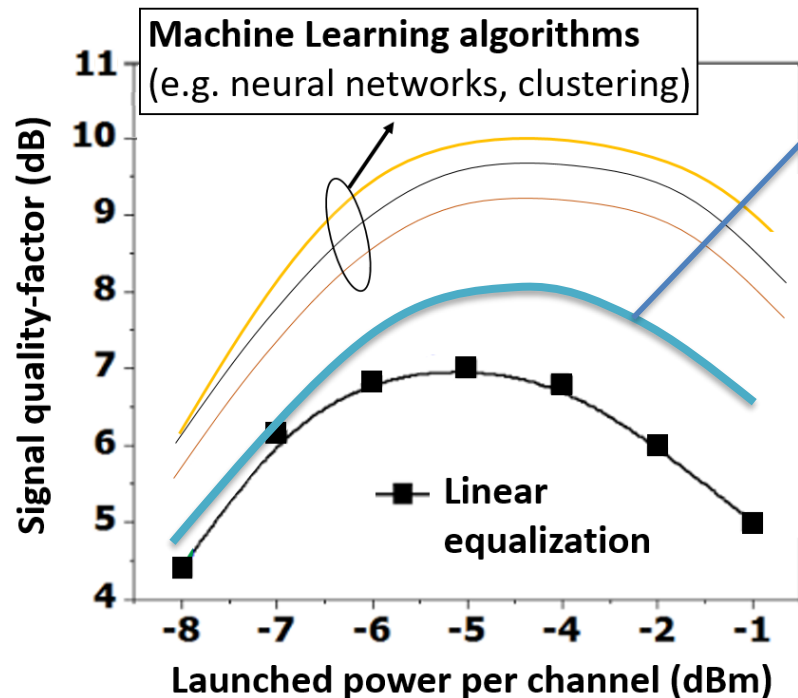
s = signal

MMSE = minimum-mean square-error

$$e(k) = s(k) - \hat{s}(k)$$

$$\hat{s}(k) = \sum_{i=1}^N w_{k,i} \varphi_{k,i}(s(k))$$

Why machine learning is good for us?



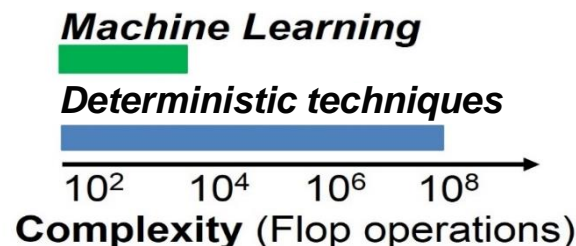
Deterministic techniques

- Machine Learning tackles stochastic noises in optical networks without knowledge of the fibre link parameters (*versatile learning*).
- It has benefit over wireless systems because optical link has stable parameters.

[1] E. Giacoumidis et al, OSA Opt. Lett. 12, 123 (2016)

[2] E. Giacoumidis et al, IEEE JLT 10, 234 (2017)

Complexity comparison (Number of operations)



Comparison with benchmark technologies

Nonlinear compensation techniques:

- Digital back propagation (DBP)
- Optical phase conjugation (OPC)
- Phase conjugated twin wave (PC-TW)

nature
photonics

ARTICLES

PUBLISHED ONLINE: 26 MAY 2013 | DOI: 10.1038/NPHOTON.2013.109

Phase-conjugated twin waves for communication beyond the Kerr nonlinearity limit

Xiang Liu*, A. R. Chraplyvy, P. J. Winzer, R. W. Tkach and S. Chandrasekhar

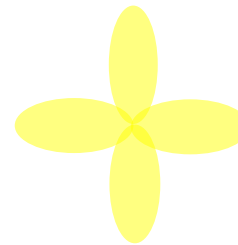
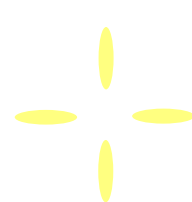
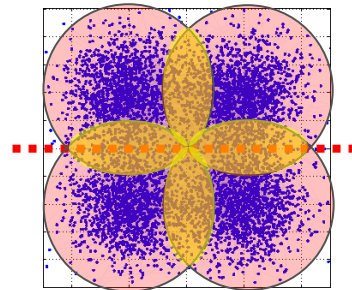
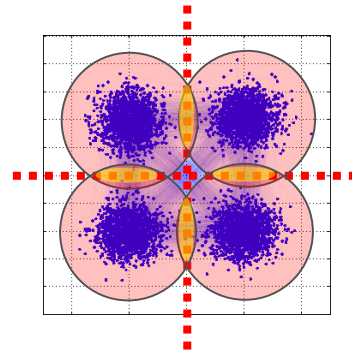
	DBP	OPC	PC-TW	Proposed
Performance	✓	✓	✓	✓
Complexity	☹	☹	✓	✓
WDM links	☹	✓	✓	✓
Links with ROADM	☹	☹	☹	✓
Overhead	✓	✓	☹	✓
Flexibility	✓	☹	✓	✓

✓ Advantage ☹ Limited benefit ☹ Challenge

Crucial points



Real-time signal processing on FPGA

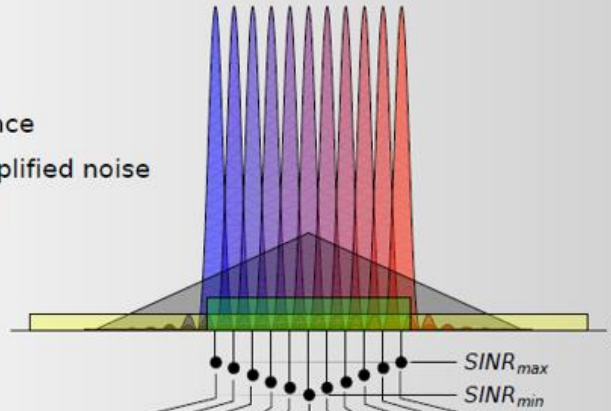


areas where errors
are most likely

3D deep learning?

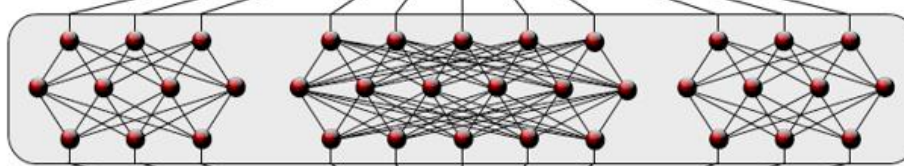
Before equalization

- FWM interference
- Parametric amplified noise
- Additive noise



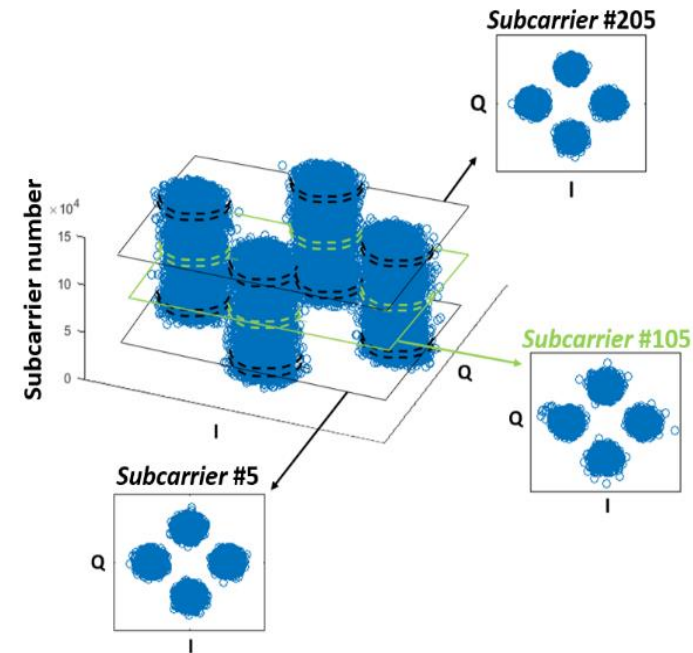
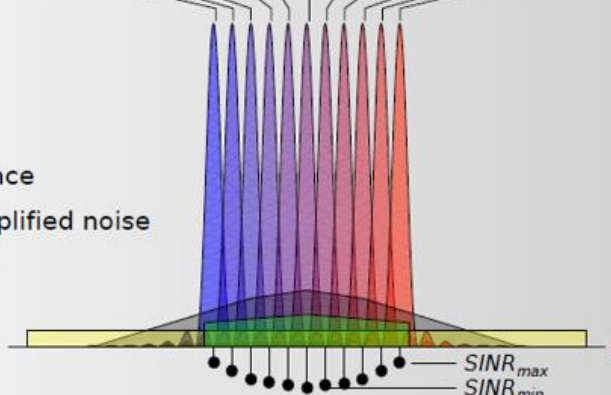
Adaptive
3D deep learning
equalization

● Neuron



After equalization

- FWM interference
- Parametric amplified noise
- Additive noise



Thank you for your attention !!!

