

# Bringing the power of Quantum Computing to Earth Observation

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ESA EOP-S Φ-lab

01/06/2023



Time magazine  
February 2023 issue

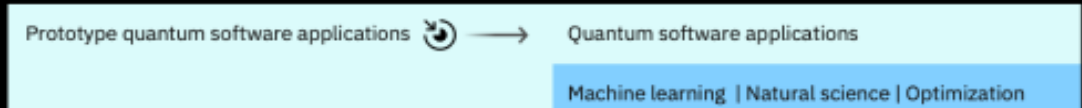


# Quantum Computing Trends in 2023

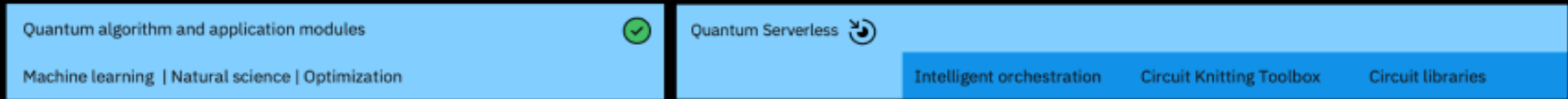


2019	2020	2021	2022	2023	2024	2025	2026+
Run quantum circuits on the IBM cloud	Demonstrate and prototype quantum algorithms and applications	Run quantum programs 100x faster with Qiskit Runtime	Bring dynamic circuits to Qiskit Runtime to unlock more computations	Enhancing applications with elastic computing and parallelization of Qiskit Runtime	Improve accuracy of Qiskit Runtime with scalable error mitigation	Scale quantum applications with circuit knitting toolbox controlling Qiskit Runtime	Increase accuracy and speed of quantum workflows with integration of error correction into Qiskit Runtime

Model Developers



Algorithm Developers



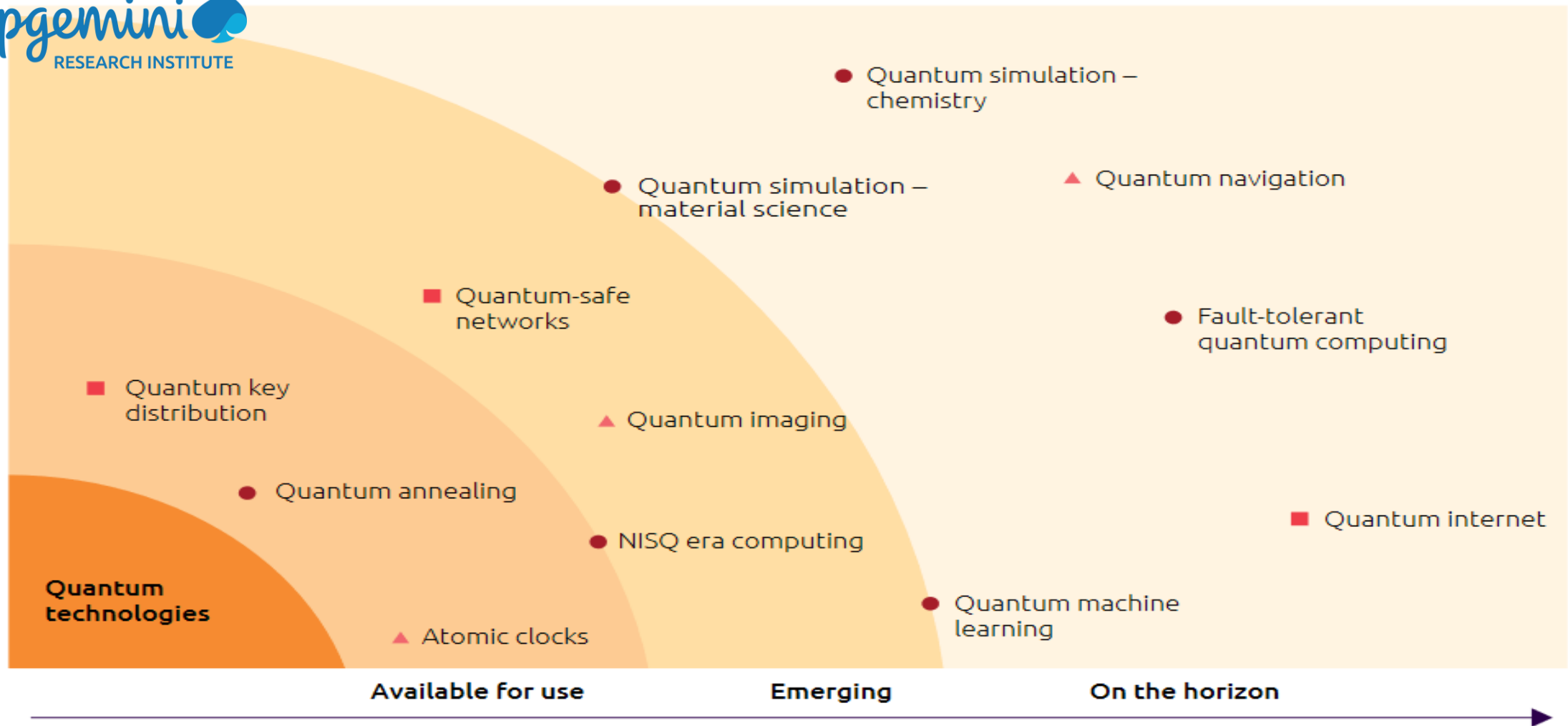
Kernel Developers



System Modularity



# Quantum Computing Trends in 2023



● Quantum computing

■ Quantum communications

▲ Quantum sensing

1. Roadmap definition  
(QC4EO studies)

2. QML and QC  
Exploratory activities



3. QC4EO Network

# Roadmap definition: QC4EO Studies

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2 projects during Q1 / Q2 2023 following ESA AO/1-11125/22/I-DT QUANTUM COMPUTING FOR EARTH OBSERVATION STUDY (QC4EO STUDY)



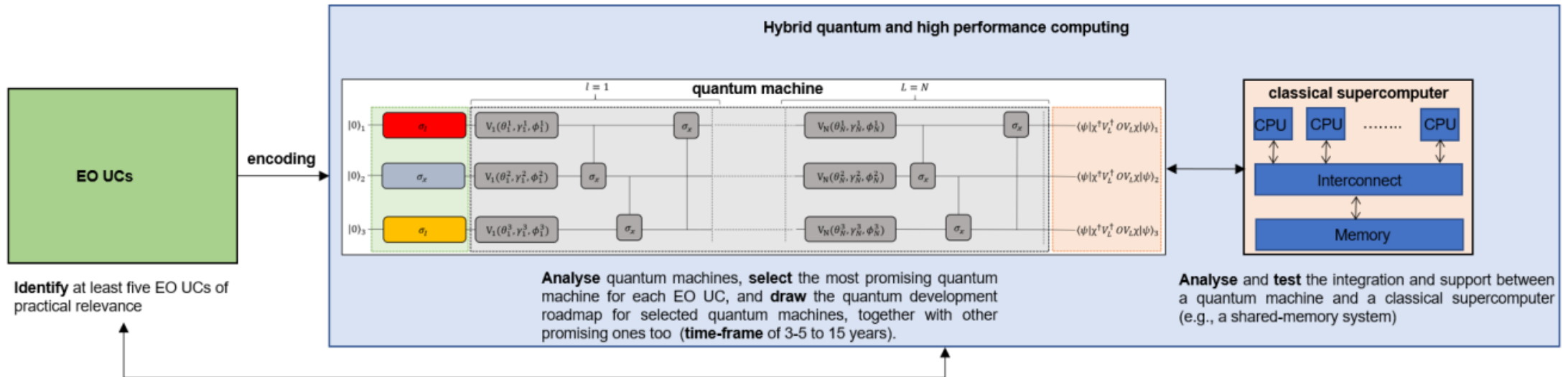
## Objectives:

- *Identify use cases relevant to the Earth Observation domain, for which QC is expected to dramatically enhance computational performances with respect to traditional methods.*
- *Provide options for QC or hybrid machine architectures required to solve the identified QC4EO use cases, with the relevant sizing, e.g. in term of Qubits.*
- *Perform a maturity and forecast assessment of the QC machine industry roadmaps; and*
- *Derive a credible QC4EO timeline of use cases that could take advantage of a QC approach*



# Quantum Advantage for EO (QA4EO) project overview

- Identify hard Earth observation use cases (EO UCs) for quantum computers (e.g., quantum machines) or a hybrid approach
- Analyse quantum machines according to their number of qubits, errors, and so on
- Draw the roadmap of quantum computers



Analyse the short- and long- term aim of utilizing a quantum machine and a classical supercomputer in the **time-frame** of 3-5 to 15 years to all identified EO UCs, and **draw timeline** for computing the identified EO UCs with respect to the quantum development roadmap.

## Quantum Advantage for Earth Observation

Project duration: 4 months (February to May 2023)

Prime-contractor: DLR

Sub-contractors: VTT, CSC, SYDERAL, ETOS, Jagiellonian University

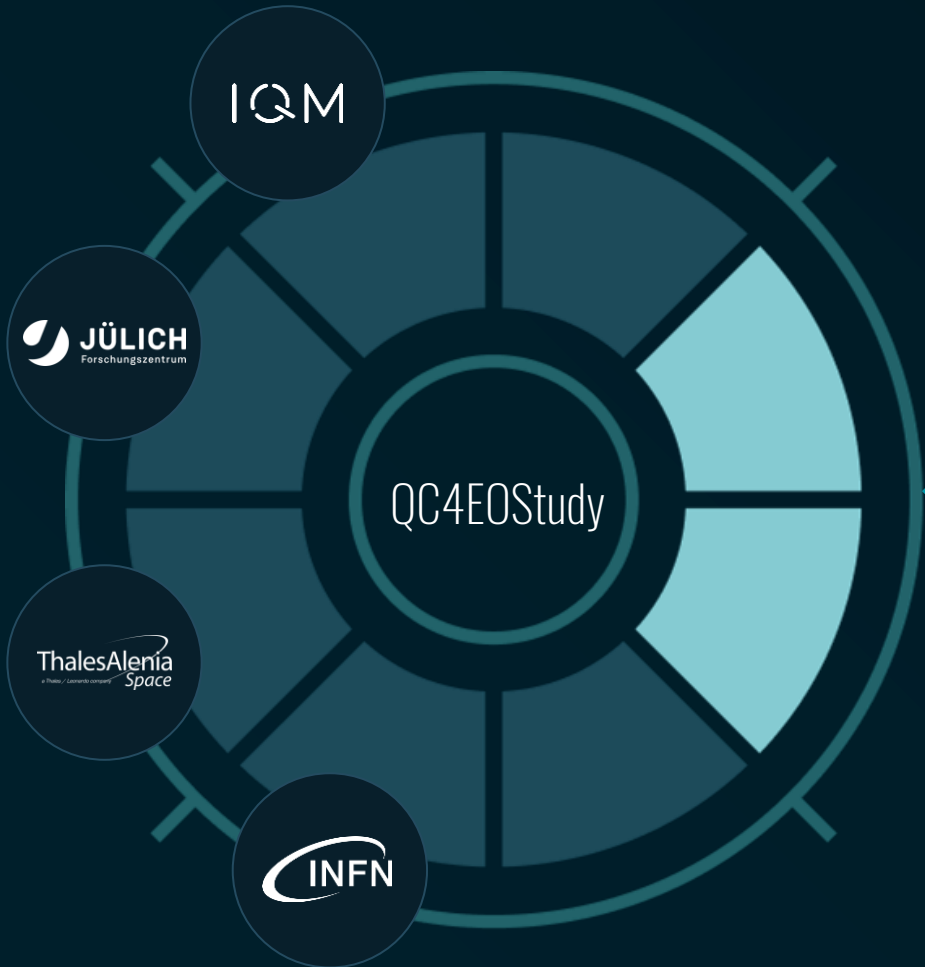
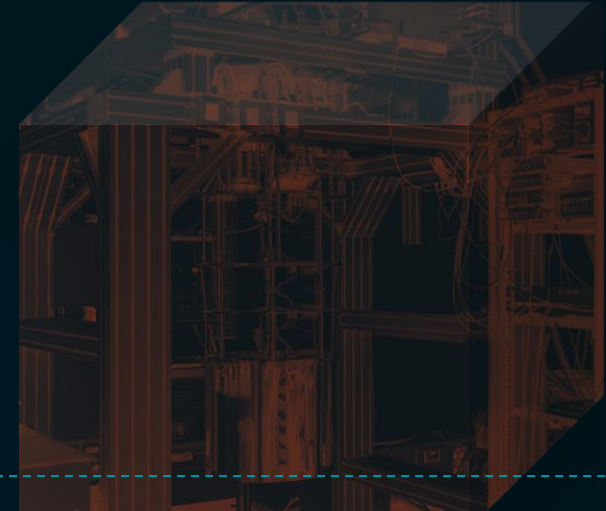
### Use-cases:

- I. Variational quantum algorithms for EO Image processing
- II. Climate adaptation digital twin hpc+qc workflow
- III. Earth land cover understanding
- IV. Feature selection for environmental monitoring hyperspectral imagery
- V. Uncertainty quantification for remotely-sensed datasets



# Quantum

## COMPUTING FOR EARTH OBSERVATION



Can QC offer advantages to EO applications within a medium to long timeframe (between the next 3-5 to 15 years)? What hardware developments are necessary to achieve this quantum advantage?

OPTIMAL SOLUTIONS

BETTER MODELLING

SPEED-UP

ENERGY EFFICIENCY

Potential quantum advantages

## Quantum Computing for Earth Observation



Project duration: 5 months (March to August 2023)

Prime-contractor: ForschungsZentrum Jülich (FZJ)

Sub-contractors: TASF, TASI, INFN, IQM



### High-level list of identified use-cases:

Scenario n° 1 – Phase Unwrapping problem for interferometric SAR applications

Scenario n° 2 – Quantum Fourier Transform for SAR raw data processing

Scenario n° 3 – Satellite Image Time Series Classification

Scenario n° 4 – Optical Agile Satellites Mission Planning

Scenario n° 5 – Multiple-view Geometry on optical images

Scenario n° 6 – Digital beamforming

Scenario n° 7 – Quantum algorithms for SAR raw data compression

Scenario n° 8 – Quantum algorithms for SAR image segmentation



# Exploratory activities

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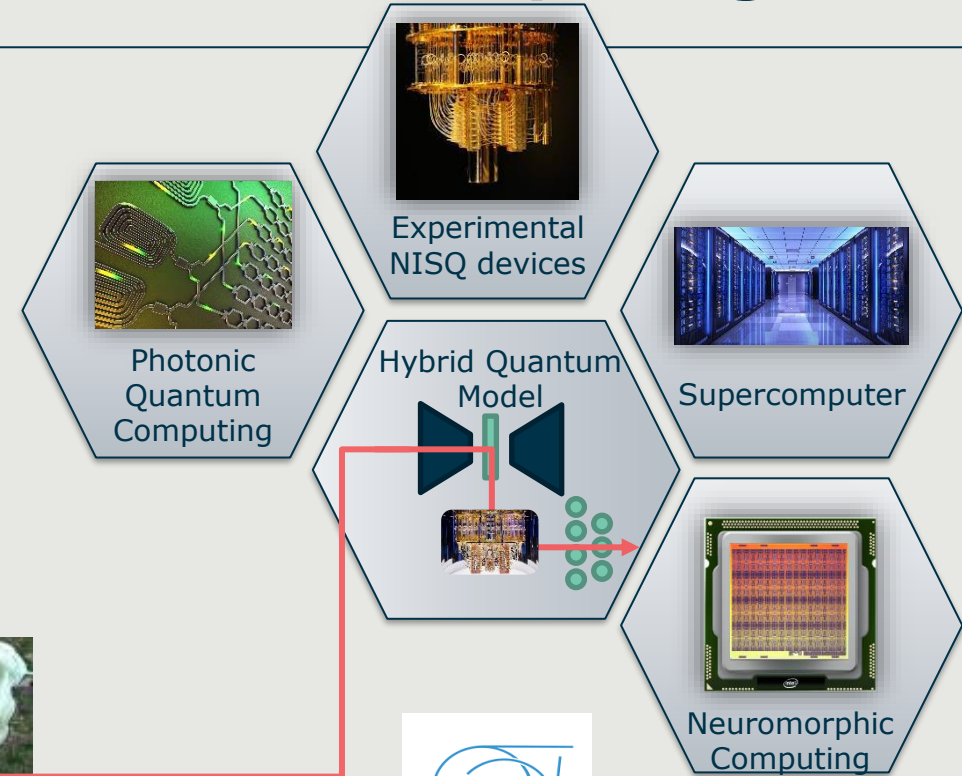


# Exploratory activities in QML and Quantum Computing

Explore the potential of Quantum Machine Learning for Earth Observation use cases

Devise hybrid quantum classical AI models in high performance computing environments

Build a strong community of experts in both Quantum Computing and Earth Observation



## ➤ Hybrid Classical Quantum Networks (Quantum Convnets, Quantum GANs, Recurrent nets)

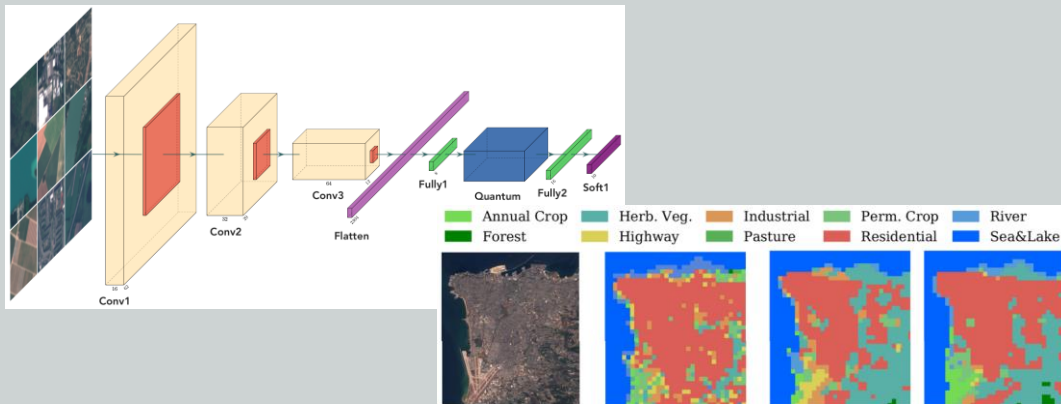
### Case Study 1: Hybrid QCNN for EO classification

**Use-case:** EO image classification for land-use and land-cover.

**Approach:** Hybrid Quantum Classical CNNs enrich standard conv nets with a quantum layer!

#### Findings:

- Successful Proof of concept, with slightly better performances than comparable CNNs thanks to entanglement.



Sebastianelli et al. "On Circuit-based Hybrid Quantum Neural Networks for Remote Sensing Imagery Classification", IEEE JSTARS (15) 2021



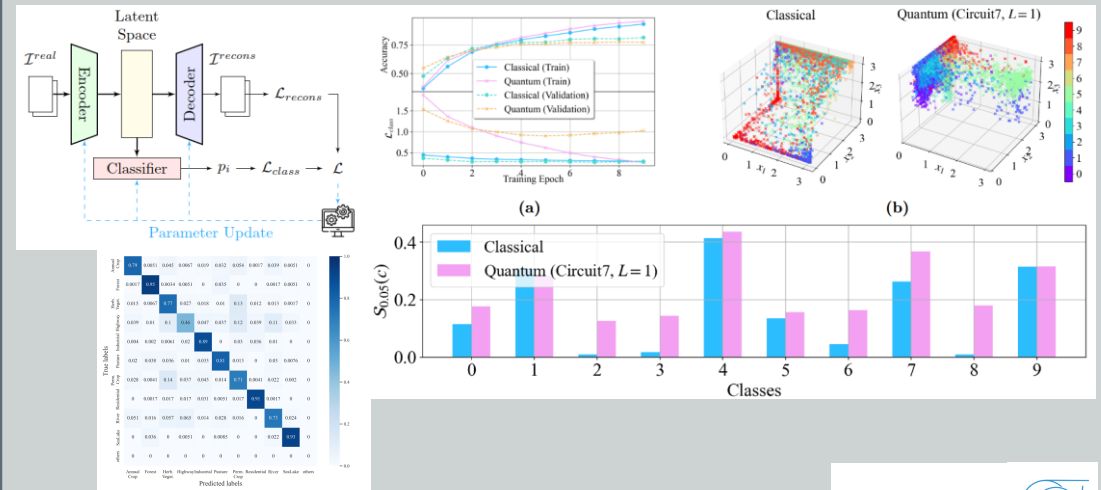
### Case Study 2: Hybrid QCNN Expressivity

**Use-case:** EO image classification

**Approach:** Hybrid models with latent space embedding

#### Findings:

- Investigation of Quantum Ansätze: better expressivity with circuits with two-qubit  $SU(4)$  state
- End-to-end Proof of Concept for EO image classification with SOTA performances



Chang et al., "Quantum Conv Circuits for EO image classif.", IGARSS 2022



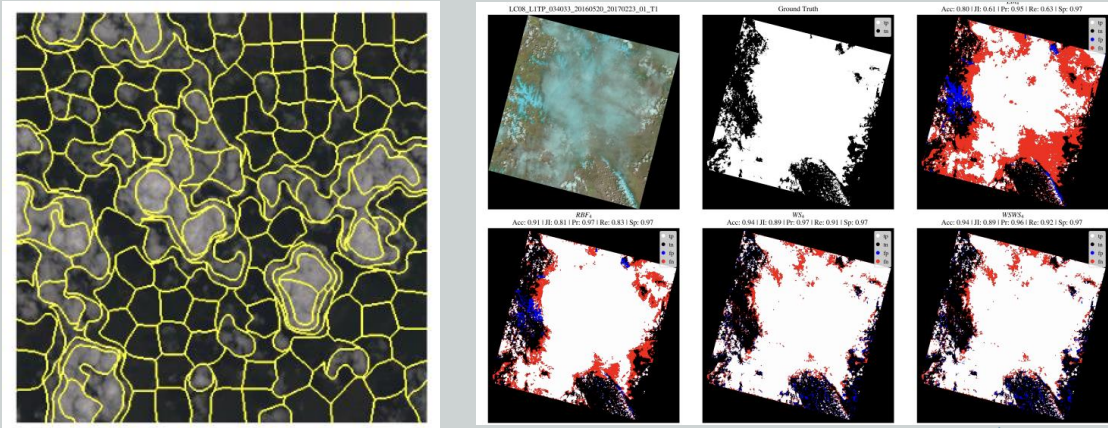
## ➤ Exploring Quantum Kernels (e.g. Projected Quantum Features, SVMs...)

### Case Study 3: Circuit-based Quantum SVM

**Use case:** cloud detection in multispectral EO images.  
**Approach:** Hybrid Support Vector Machines (SVMs) with **gate-based quantum kernels**.

#### Findings:

- End-to-end pipeline to embed and process EO data with small NISQ circuits.
- Successful Proof of Concept, with results on par with standard SVM thanks to Quantum Kernel Target Alignment.



Miroszewski, A., Mielczarek, J., Czelusta, G., Szczepanek, F., Grabowski, B., Le Saux, B., & Nalepa, J. (2023). Detecting Clouds in Multispectral Satellite Images Using Quantum-Kernel Support Vector Machines. arXiv preprint arXiv:2302.08270.

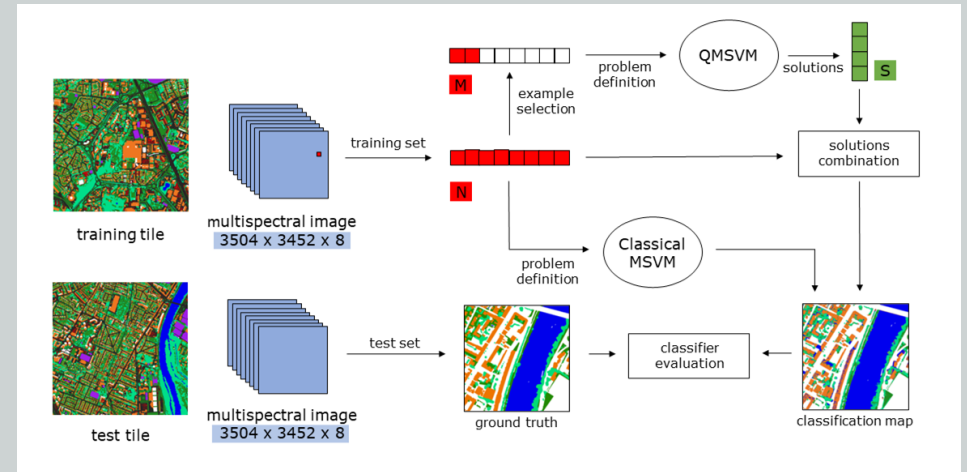


### Case Study 4: Annealing-based Quantum SVM

**Use case:** Classification of multispectral EO data.  
**Approach:** Hybrid Support Vector Machines (SVMs) with **Julich SC Quantum Annealer**.

#### Findings:

- Advantage Annealer operates only a limited number of samples for Q optimization...
- .... But execution times increase linearly!



Delilbasic, A., Le Saux, B., Riedel, M., Michielsen, K., & Cavallaro, G. (2023). A Single-Step Multiclass SVM based on Quantum Annealing for Remote Sensing Data Classification. arXiv preprint arXiv:2303.11705.





## ➤ Hybrid Classical Quantum Networks (Quantum Convnets, Quantum GANs, Recurrent nets)

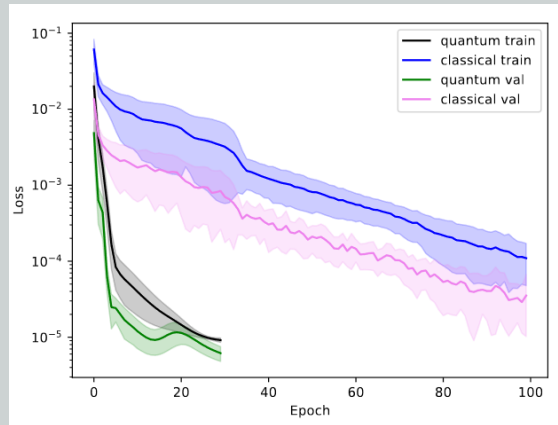
### Case Study 5: Continuous-variable QC RNNs

**Use case:** Earth Systems observation and prediction.

**Approach:** Recurrent Neural Networks for time-series on Continuous-Variable QC.

#### Findings:

- Promises of faster training convergence
- For a small number of trainable parameters, it can achieve lower losses than its classical counterpart.



Siemaszko, McDermott, Buracsewski, Le Saux & Stobinska  
 "Rapid training or recurrent quantum neural networks", **QTML**  
 2022 / Qu Mach. Intell. 2023



### Case Study 6: Quantum Generative AI

**Use case:** Generative modelling and synthesis of images.

**Approach:** Quantum Generative Adversarial Networks (QGANS) :  
 Quantum Generator + Classical Discriminator.

#### Findings:

- Trick #1: Latent space embedding by pretrained autoencoder
- Trick #2: Continuous, Style-based quantum GAN
- Successful image generation for varied image types
- Faster and better performances (in terms of distribution mapping) with less parameters



Chang et al., to appear



# Hybrid Quantum-Classical Neural Networks

Su Yeon Chang [CERN, EPFL]

23/05/2022



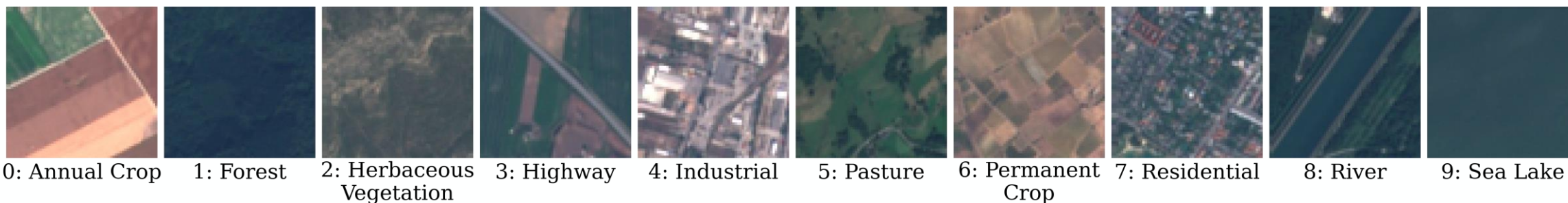
QUANTUM  
TECHNOLOGY  
INITIATIVE



# Motivation and objectives

- **Quantum Machine Learning (QML)**
- Intersection between Machine Learning (ML) and Quantum Computing (QC)
- Potential to improve the existing ML techniques
- Can be efficiently simulated on the real quantum hardware
- Increasing studies on application of QML on Earth Observation (EO) images

**PhD Objective** : Implement quantum generative models to reproduce EO images

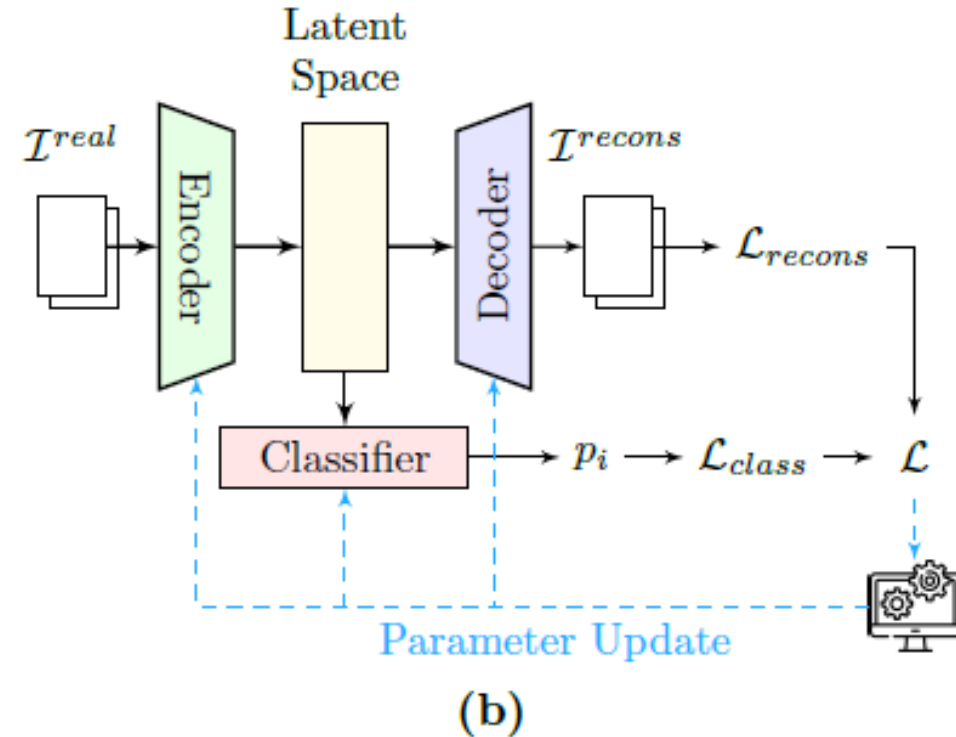
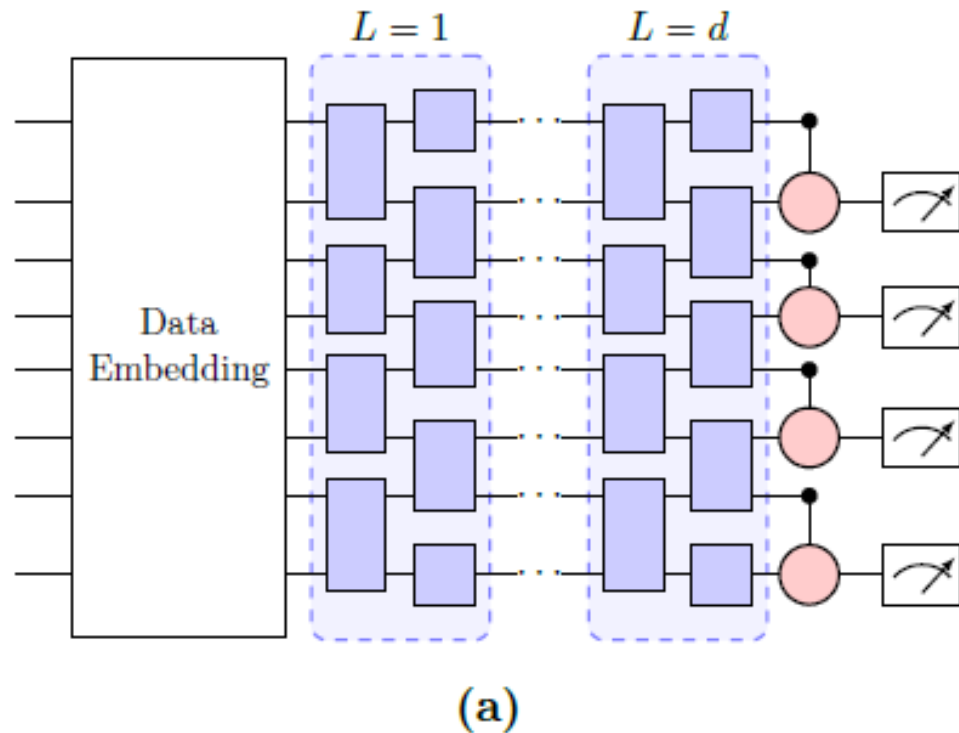


# Challenges

1. **Scale EO data** for current quantum simulators → Dimensionality reduction techniques
2. Transfer classical data to quantum & quantum data to classical  
→ How to **interpret the quantum circuit output** into classical data?
3. Find the **quantum circuit architecture** which is the most trainable, efficient and accurate  
→ Hardware Efficient Ansatz (HEA), Quantum Convolutional Neural Networks (QCNN), Alternating Layered Ansatz (ALT), etc...
3. Possibility of **quantum advantage**  
→ Quantum advantages expected in case the training is classically intractable  
→ Can quantum generative model learn hidden image properties that classical ML cannot?
5. Trainability on **real quantum hardware**  
→ Aim to run quantum circuit on different quantum devices, e.g. superconducting chips (IBMQ), Ion trap (IONQ)

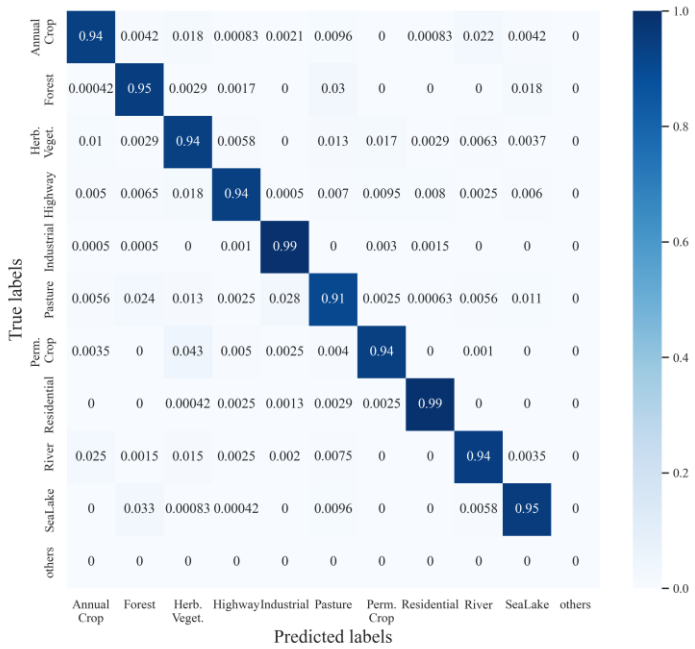
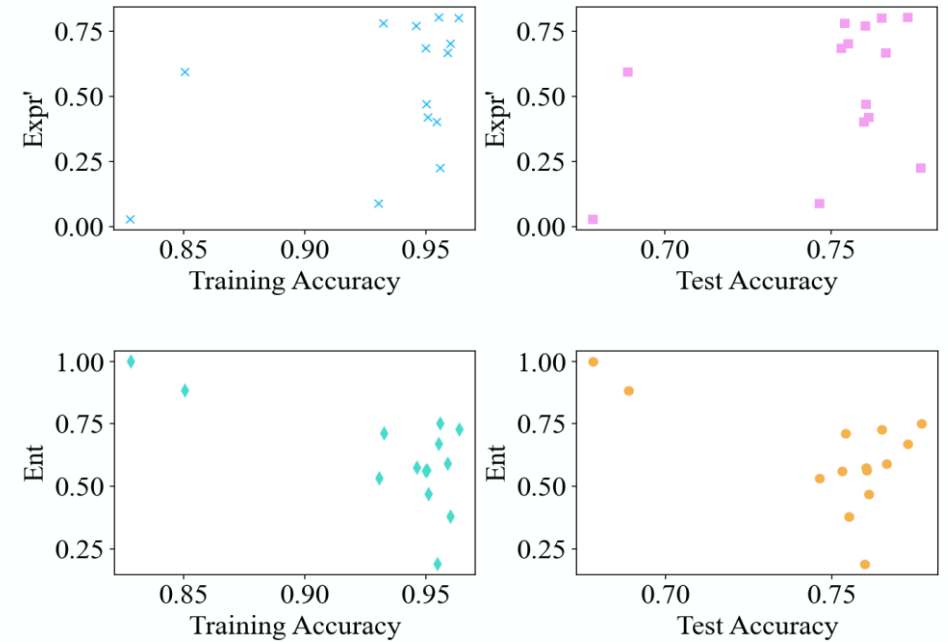
# Multi-task hybrid model for image classification

- Work presented in [Quantum Information Processing 2023 conference](#)
- Hybrid classical-quantum model** for image reconstruction and classification
- Classical autoencoder for dimensionality reduction & image reconstruction
- Quantum classifier for feature classification

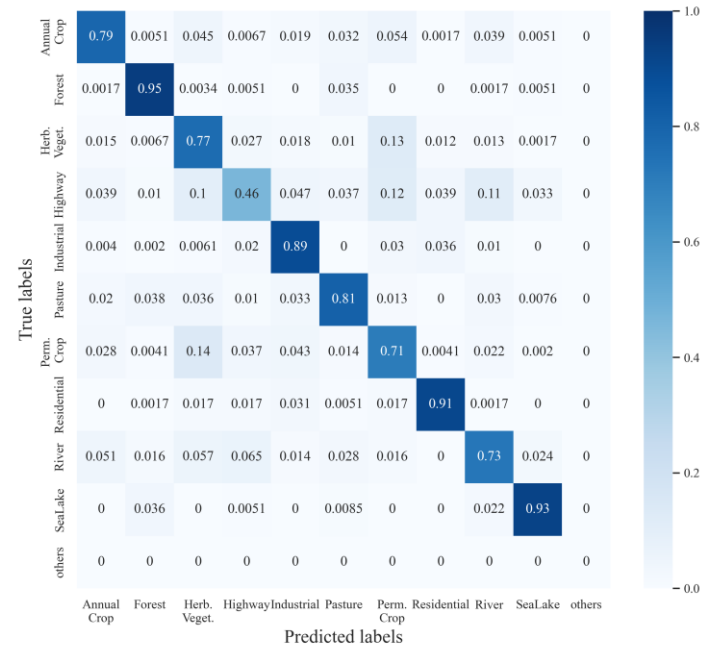


# Multi-task hybrid model for image classification

- Successful classification for EuroSAT dataset (10 classes)
  - Challenge : Cannot observe quantum advantage with current model
- Investigate latent feature arrangement to understand the limitation



Train set



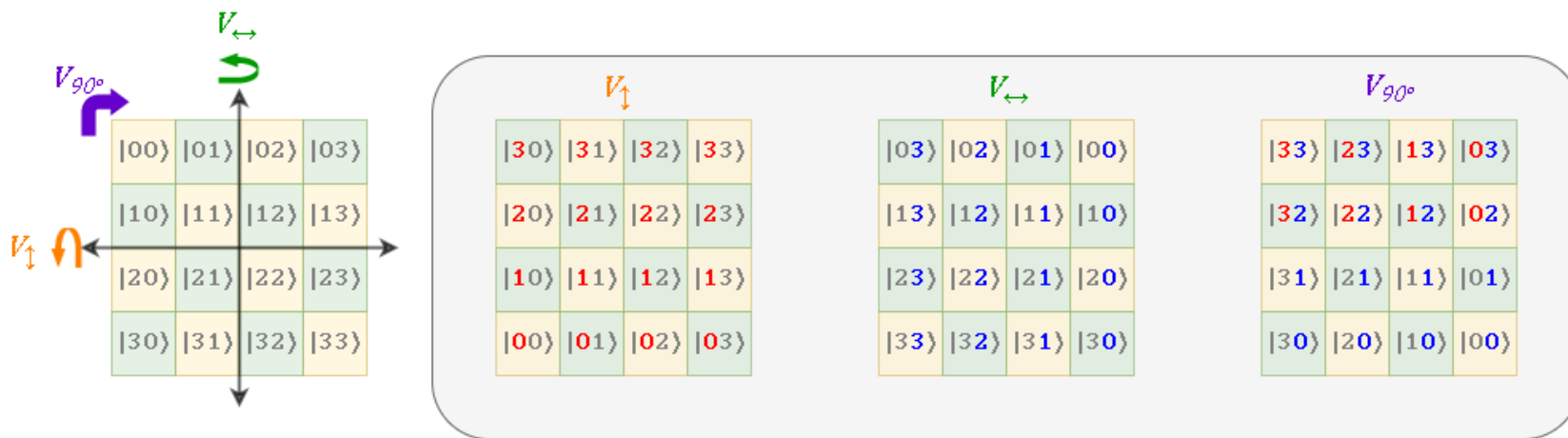
Test set

**Study correlation between PQCircuit architecture & classification accuracy**

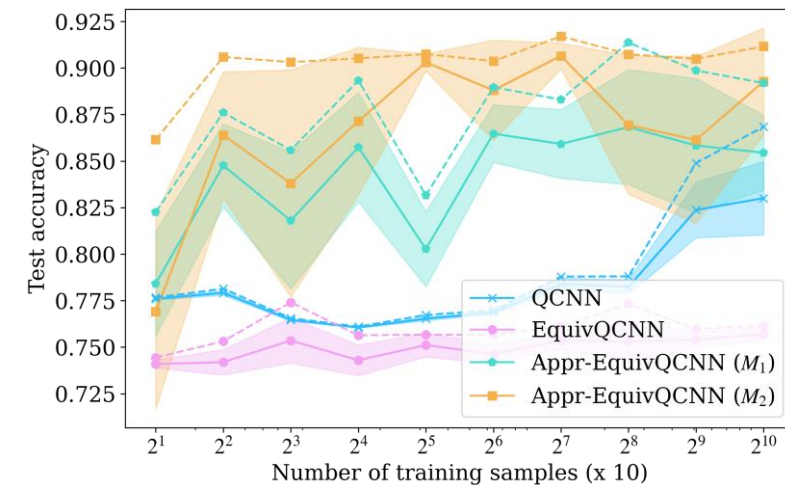
- Pearson Correlation Coefficient** calculated for Expressibility /Entanglement capability v.s. Accuracy
- Higher Expressivity → Higher Accuracy
- Higher Entanglement → Lower Accuracy

# Equivariant Quantum CNN

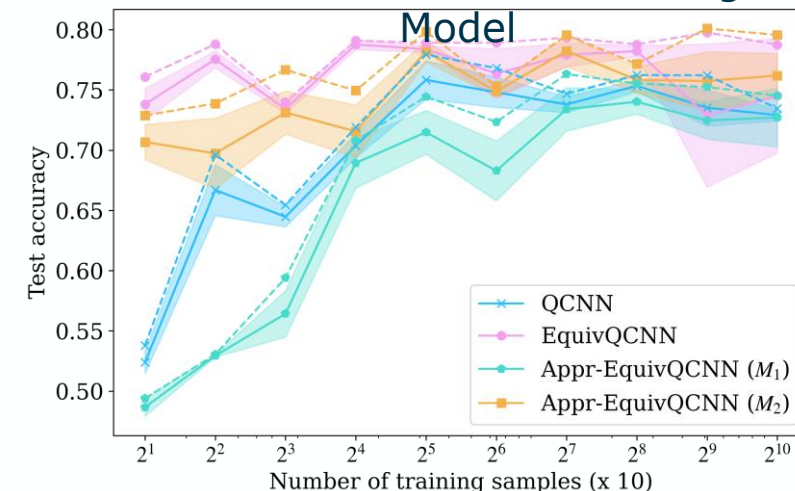
- Work submitted to [IEEE Quantum Week 2023](#)
- Construct an equivariant quantum CNN for image classification under rotation & reflection symmetry
- Better generalization power compared to the non-equivariant model
- Challenge** : Extend the application to larger RGB datasets



MNIST Image classification (digits 4,5)

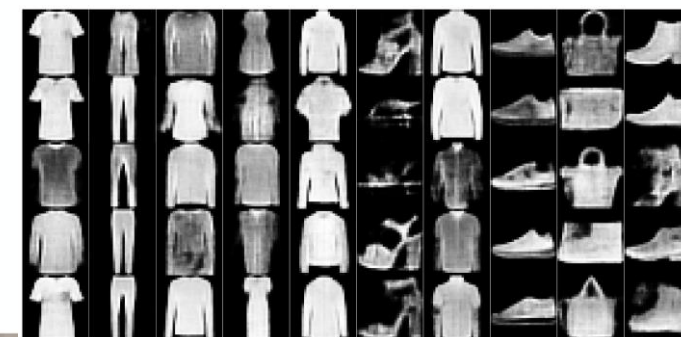
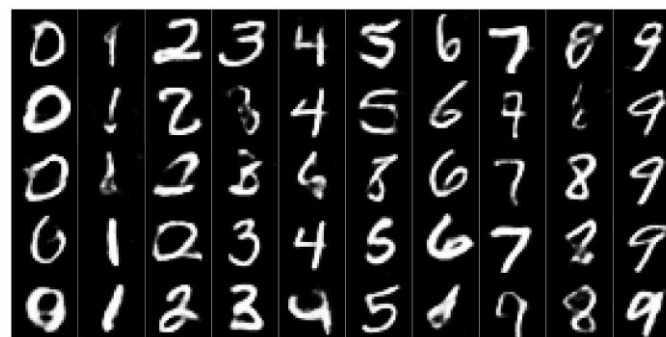
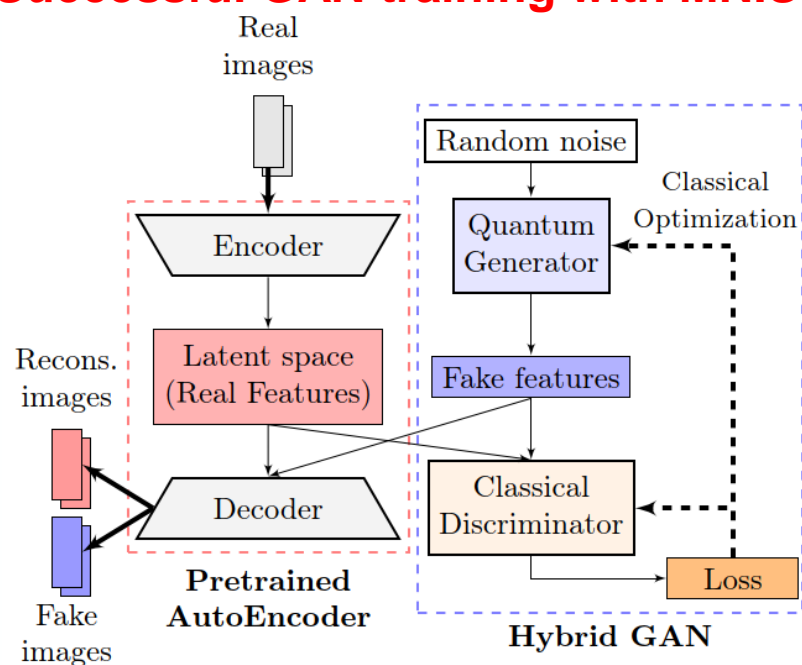


Phase classification for Ising Model



# Style-based quantum GAN for image generation

- Quantum Generative Adversarial Networks : **Quantum Generator + Classical Discriminator**
- Hybrid approach allows to scale up the model for realistic use cases
- Features extracted from images via a **pretrained autoencoder** used as GAN training set
- Generated features passed back to the autoencoder to reconstruct images
- Successful GAN training with MNIST, FashionMNIST and SAT4 dataset**



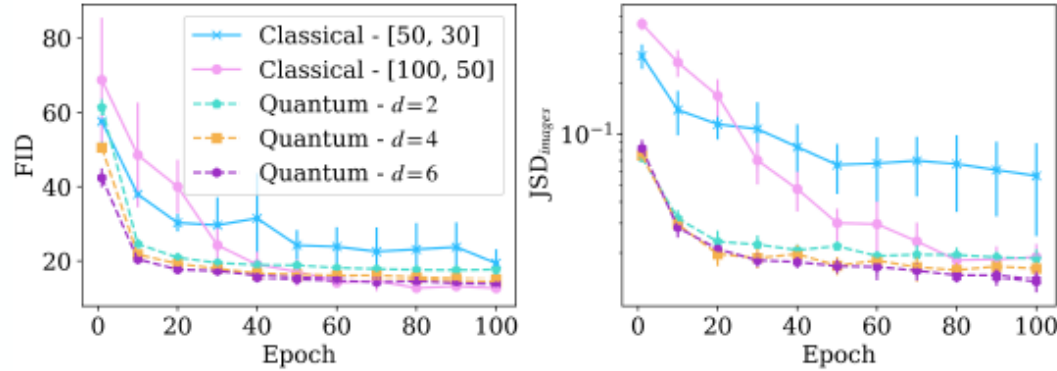
Training schema of hybrid GAN model with pretrained autoencoder

Samples generated with quantum GAN

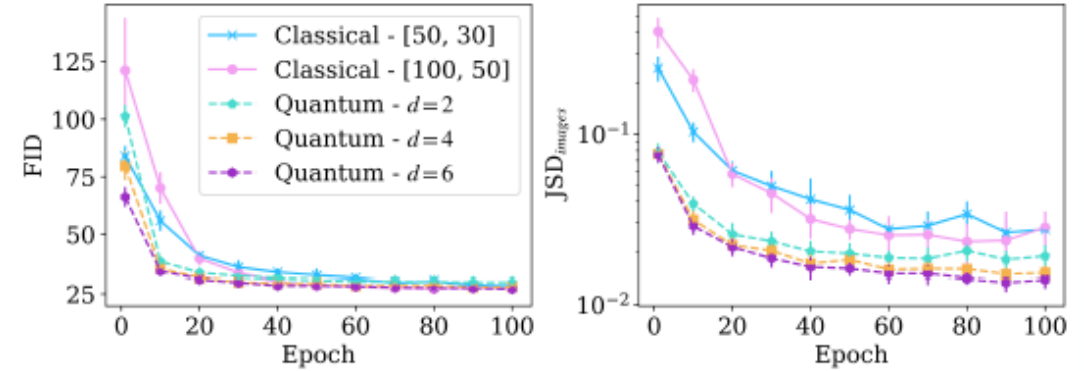


# Style-based quantum GAN for image generation

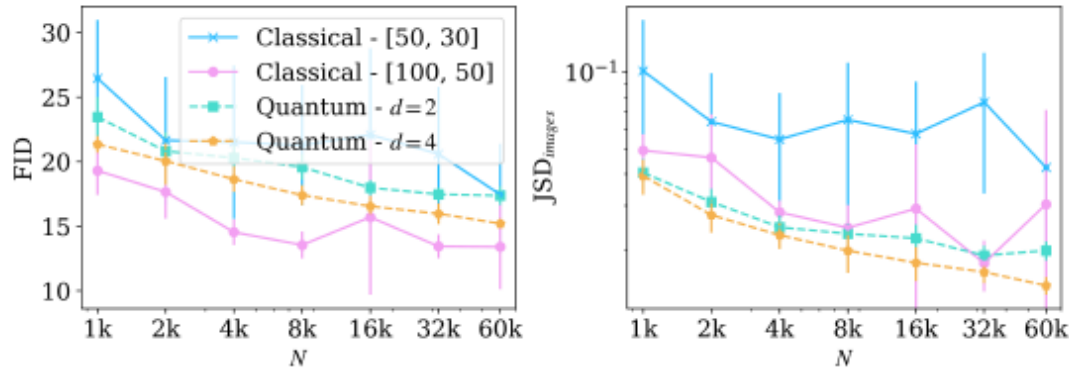
- **Faster convergence** using similar number of parameters
- **Better generalization** power with less number of training samples



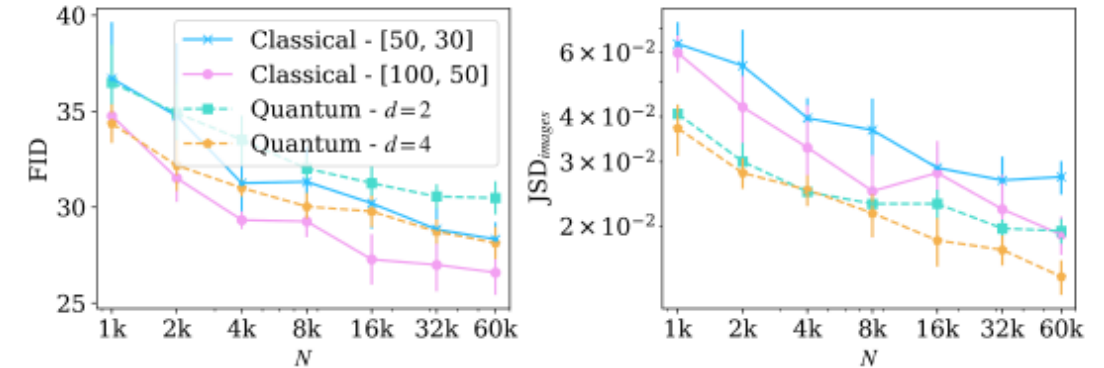
(a) MNIST



(b) FashionMNIST



(a) MNIST



(b) FashionMNIST

# QC4EO Network

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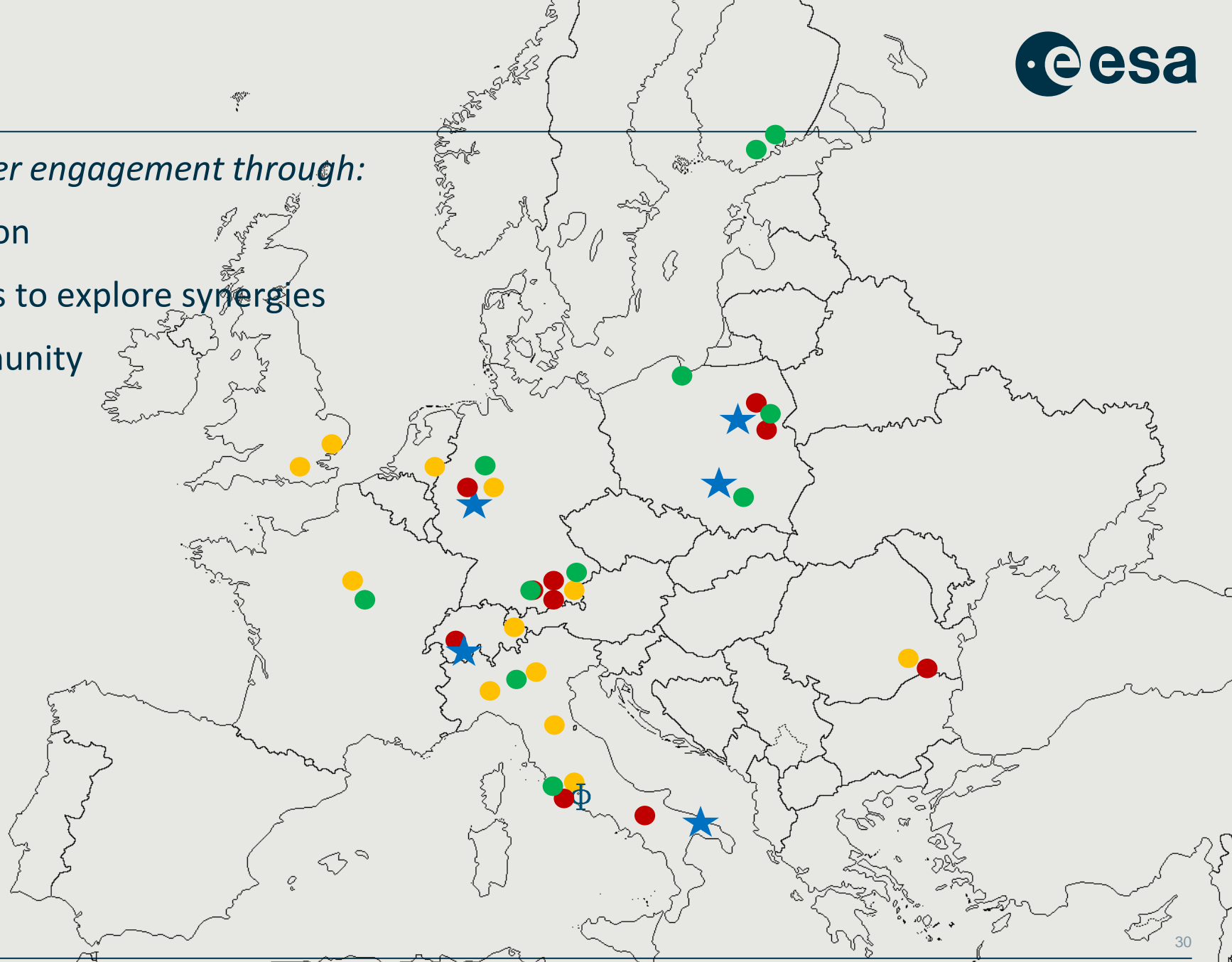


# QC4EO Network

Community building and stakeholder engagement through:

- Workshop and event organisation
- Consult QC and EO communities to explore synergies
- Support emerging QCxEO community

- QC4EO Study
- ★ Co-funded research
- Partners / visitors
- Community / events







# QC4EO Network: visiting researchers



## Senior visiting researchers:



Mihai Datcu  
(Politehnica Uni of Bucharest)



Gabriele Cavallaro  
(Forschungszentrum Jülich)



Piotr Gawron  
(CAMK / Polish Acad of Sciences)

## Early-career researchers:



Michal Siemaszko  
(PhD, Univ. Warsaw)



Alice Barthe  
(PhD, CERN / Leiden Uni)



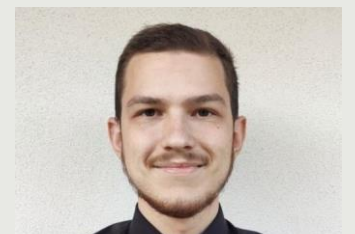
Andrea Ceschini  
(PhD, La Sapienza)



Francesca de Falco  
(MSc., La Sapienza)



Francesco Mauro  
(PhD, Uni. Sannio)



Amer Delilbasic  
(PhD, Uni of Iceland / FZ Jülich)

- We are welcoming visiting researchers from academia and industry!
- Spend short stays or residencies at the  $\Phi$ -lab to mingle with EO, AI, and QC experts!
- Let's get in touch!



# Conclusions

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# ESA Φ-lab's Initiative on Quantum Computing for Earth Observation (QC4EO)

General perspectives:

- *Increase the mutual awareness of the needs and capabilities of the **Quantum Computing and Earth Observation** communities*
- *Create new **synergies**, building on shared experience in **AI**, optimisation, and **high-performance computing***
- *Prepare the ground for the opportunities that will be presented when the quantum community will be able to produce hardware and software for applied problems*



# ESA $\Phi$ -lab's Initiative on Quantum Computing for Earth Observation (QC4EO)

Practical perspectives:

- *Look for practical applications and use-cases, enabled by increased quantum volume*
- *Understand the **advantages** (faster, better, etc.?) brought by QC with **exploratory activities***
- *Design hybrid computing frameworks including traditional CPU, GPU, HPC and new paradigms such as quantum and neuromorphic computing for optimal problem solving*

# Bringing the power of Quantum Computing to Earth Observation

- Follow us: <https://www.esa.int/> / <https://philab.esa.int/>
- Join ESA  $\Phi$ -lab @ESRIN: <https://jobs.esa.int/>
- Contact: [alessandro.sebastianelli@esa.int](mailto:alessandro.sebastianelli@esa.int),  
[bertrand.le.saux@esa.int](mailto:bertrand.le.saux@esa.int)

*SAVE THE DATE!*

*6th ESA Quantum Technology Conference | 19 – 21 September 2023 | Matera, Italy*