

From Near-Term to Fault-Tolerance: When Axis-Aligned QML Can Outperform Classical Counterparts in Earth Observation

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CLAIM UNIT

QML family



Preserved EO axis



Matched classical peer



Hardware horizon

**Not “does QML work?”
When, where, and under
which resource model?**

The useful question is narrower than “does QML work?”

Earth observation is a hard classical learning domain; the task itself has three defining characteristics.



instead of one field-level claim



state the competitive cell

model family + preserved axis + matched baseline + hardware/data-access horizon

Structural-Prior Alignment Organises Quantum Machine Learning for Earth Observation

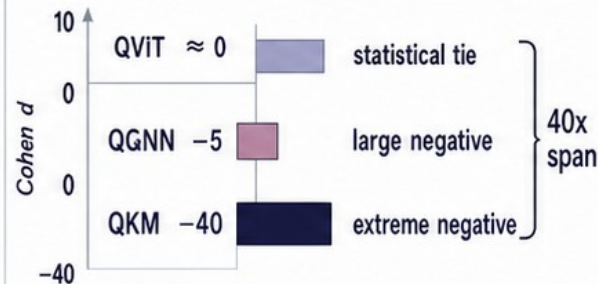
Overview of the position paper

I. Cross-family heterogeneity



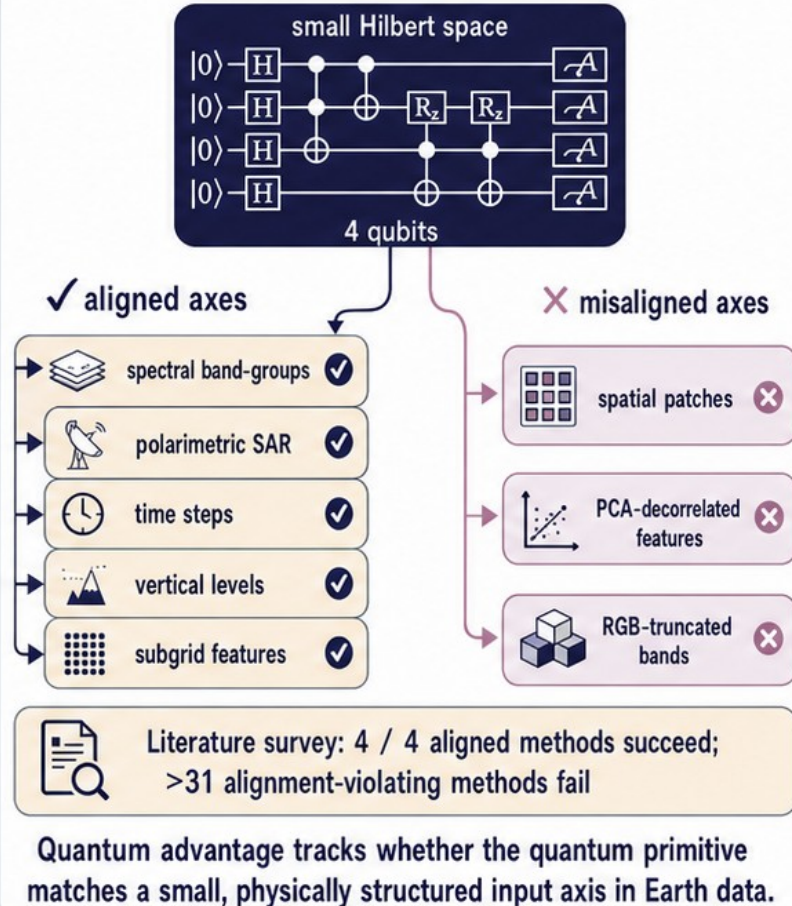
Shared dataset: m-EuroSAT-13b

Matched parameters

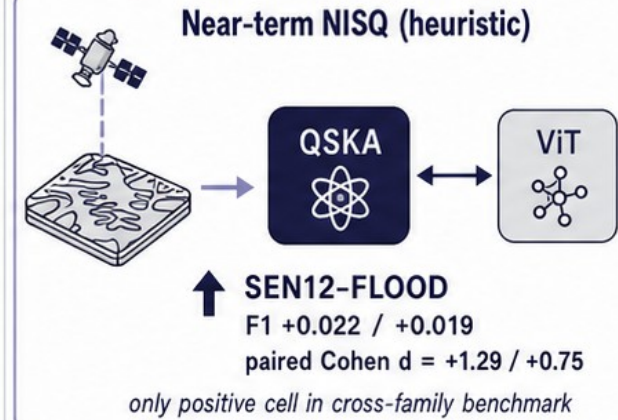


No single aggregate claim about 'QML for EO' can cover all three families.

II. Structural-prior alignment principle



III. Two clocks and the long-term anchor



Fault-tolerant (unconditional)



4.4–5.8 orders-of-magnitude memory advantage

Pavia Centre + EuroSAT-MS

Quantum Oracle Sketching (Zhao et al. 2026)

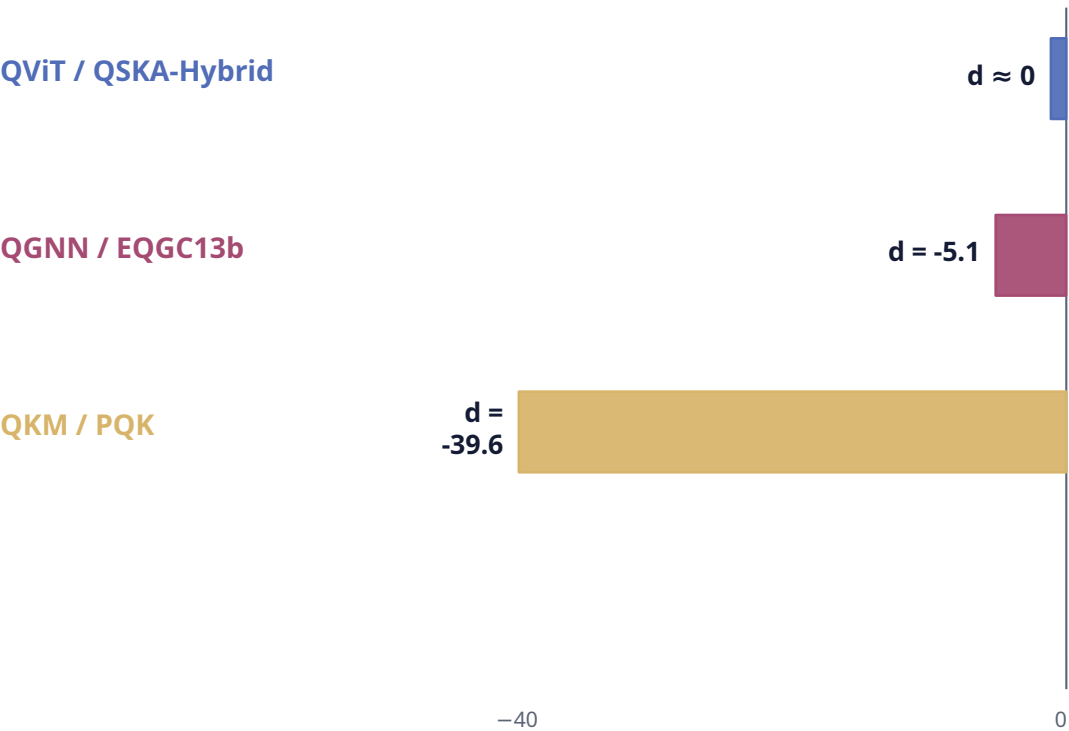
Retire aggregate 'QML for EO' framings

Replace with alignment-axis-specific claims at two clocks: NISQ heuristic, + FT unconditional

Matched-resource family results split immediately

The three representative families use the same EO benchmark but act on different axes through different access representations.

Cohen d: quantum cell vs matched classical peer



Takeaway: no single aggregate statement about “QML for EO” can cover all three families.

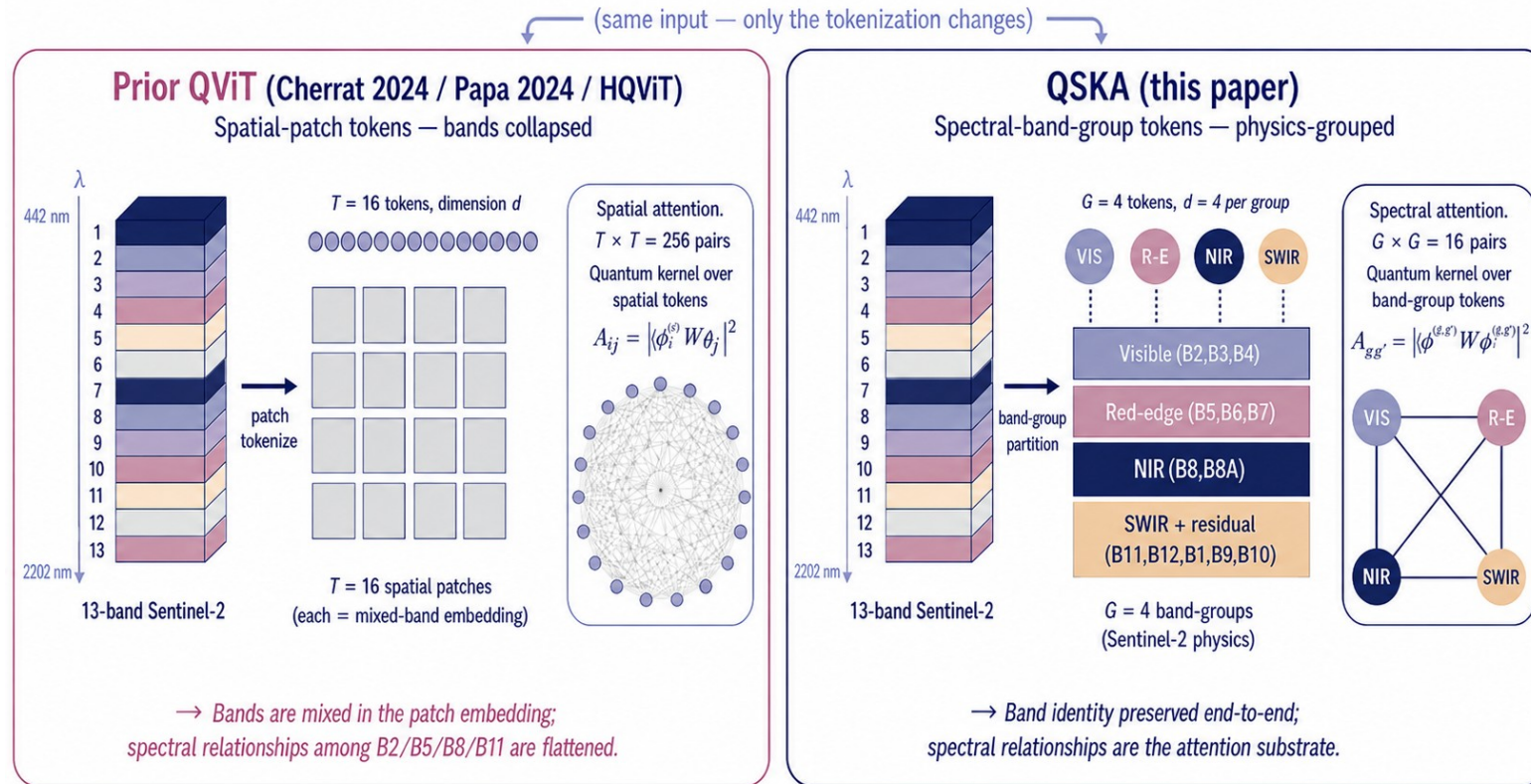
Axis by model family

MODEL	EARTH-DATA AXIS
QViT / QSKA-Hybrid	spectral channel axis
QGNN / EQGC13b	spatial graph axis
QKM / PQK	PCA feature axis

Quantum Spectral Kernel Attention Hybrid

Move the quantum attention block onto physically meaningful spectral band-group tokens.

The right token for QML × multispectral Earth observation is the spectral-band-group, not the spatial patch



Prior Quantum Vision Transformer: bands collapse inside patch embeddings.

Quantum Spectral Kernel Attention: band identity is preserved end-to-end.

Structural-Prior Alignment Principle

Structure alignment asks whether the quantum primitive is placed on a preserved, physically meaningful Earth-data axis.

Three diagnostic questions

1

Does the quantum primitive operate on a small physical axis?

2

Is that axis preserved through preprocessing?

3

Does it perform the main axis-interaction step?

Feature examples

Recommended aligned features

- spectral band groups
- SAR polarizations
- time steps

Not aligned features

- RGB truncation
- PCA-decorrelated features
- generic CNN features

How the SEN12-FLOOD control experiment is designed

Each pair changes one design factor while the dataset, split, sample size, training protocol, and parameter scale stay fixed.

1 Synthetic aperture radar modality

without synthetic aperture
radar

vs.

with synthetic aperture radar

2 Quantum versus classical attention

classical linear
value / mixing map

vs.

quantum butterfly quantum
orthogonal layer

3 Spectral-attention path

without spectral path

vs.

with spectral path

4 Effect of token shape

spectral bands grouped
as physical tokens

vs.

separate optical + radar
modality tokens

Fixed protocol: SEN12-FLOOD dataset; event-group split; 50 / 100 / 200 samples per class; 3 seeds; matched parameter scale

Quantum versus classical attention under the same controls

This table isolates the “quantum versus classical attention” feature group: all other SEN12-FLOOD settings are held fixed.

Feature group being compared

Classical attention
linear value / mixing map

vs.

Quantum attention
butterfly quantum orthogonal layer

Compared at n = 50, 100, 200 samples per class; values are mean F1 over three seeds.

n / class	Classical attention F1	Quantum attention F1	Δ F1
50	0.746	0.756	+0.010
100	0.758	0.787	+0.029
200	0.776	0.797	+0.021

Result for this feature group: quantum attention is sign-consistently higher at all three sample sizes.

Interpretation: the pilot edge is more consistent with the quantum value/mixing parameterization than with token shape alone.

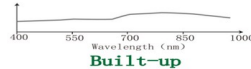
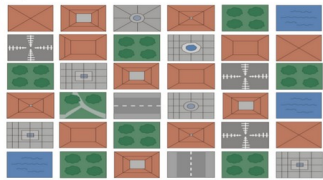
Controlled comparison: same dataset, event-group split, sample sizes, training protocol, spectral path, synthetic aperture radar modality, and matched parameter scale.

Quantum Oracle Sketching on Earth-observation data

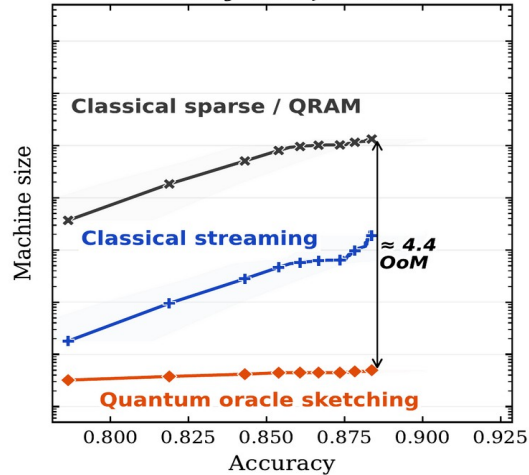
A fault-tolerant machine-size comparison under explicit data-access assumptions.

Pavia Centre

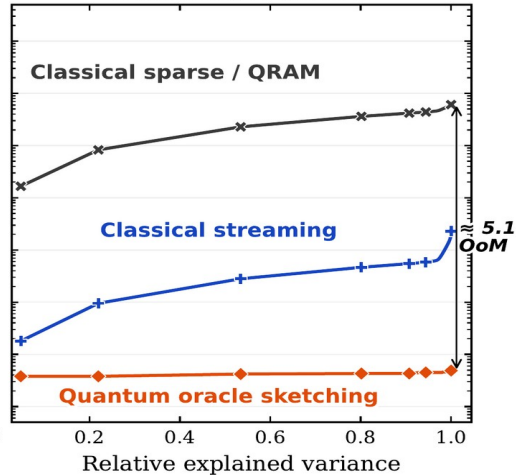
(a) Pavia Centre



Binary classification

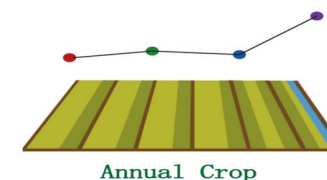
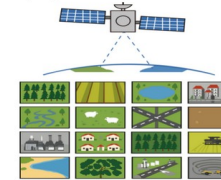


Dimension reduction

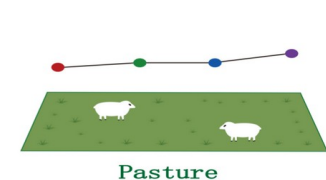


EuroSAT-MS

(b) EuroSAT-MS

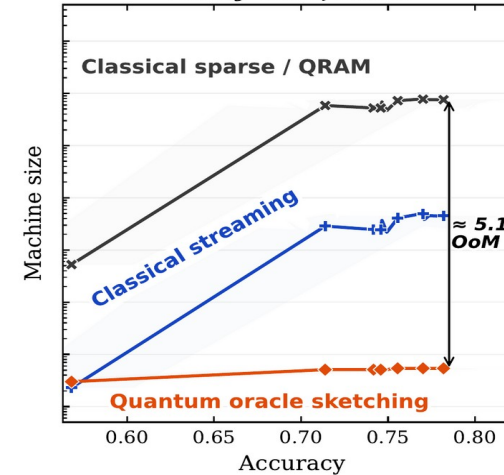


Annual Crop

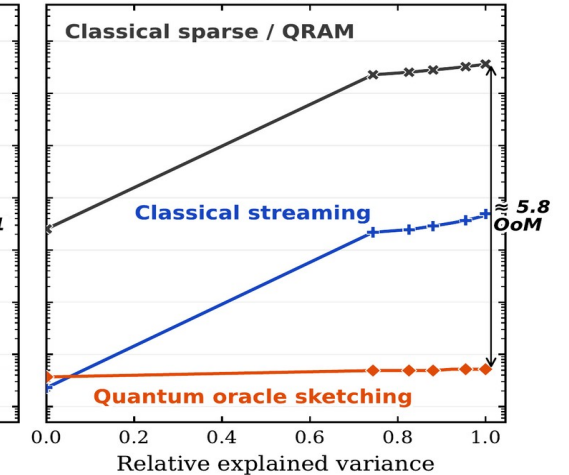


Pasture

Binary classification



Dimension reduction



Reported memory gap: 4.4–5.8 orders of magnitude; quantum machine size $\approx$ 50–54 logical qubits.

Three claims of the paper

Replace broad “QML for Earth observation” promises with axis-specific claims at two hardware clocks.

1

Evaluation unit

Report a concrete quantum family, architecture, preserved Earth-data axis, matched classical peer, and resource setting.

2

Near-term cell

Quantum Spectral Kernel Attention Hybrid is a bounded parameter-efficient positive cell on SEN12-FLOOD, not a universal Quantum Vision Transformer claim.

3

Fault-tolerant path

Quantum Oracle Sketching gives a separate memory-scaling anchor when streaming, sparse, high-dimensional access assumptions hold.

Final message: when, where, and under which resource model?