# On the expressivity of embedding quantum kernels

#### Elies Gil-Fuster, FU Berlin

@eliesgf.





### 1. Quantum Machine Learning (QML)

Quantum model **Classical** data **Classical** training algorithm

### 2. Expressivity

Today: is there more to quantum kernels?

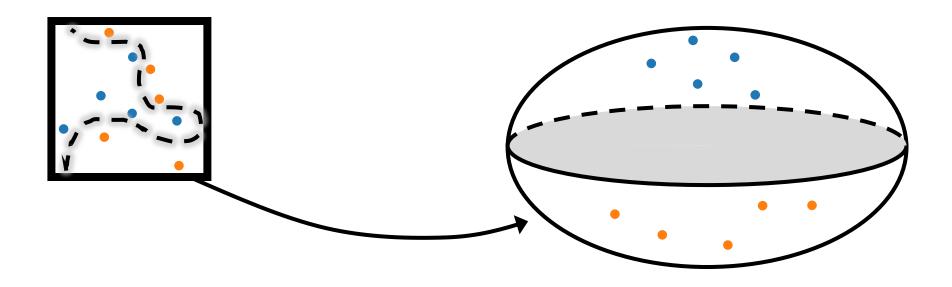
- Quantum kernels beyond the canonical approach?
- What is ultimately possible with quantum kernel methods?

Not today:

- How can I solve cool QML problems?
- How do kernel methods work?

### 3. Kernel methods

#### Linear optimization on a feature space:

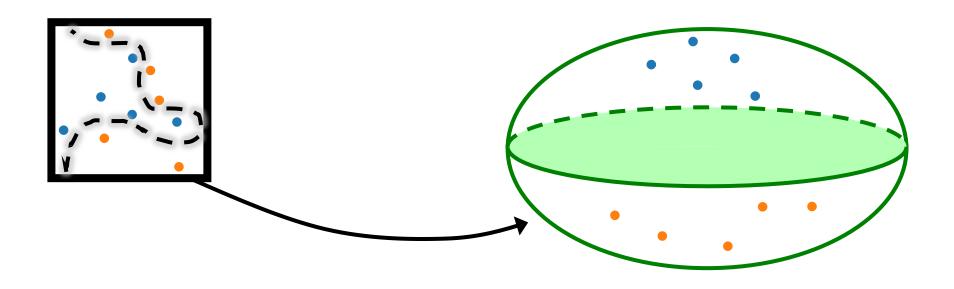


"A kernel is an inner product of feature vectors"

Hubregtsen et al, *PRA* **106** (2022) Schuld, *2101.11020* (2021)

## 4. Embedding Quantum Kernels (EQKs)

#### Linear optimization on quantum feature space:



"An Embedding Quantum Kernel is an inner product of quantum feature vectors"

Jerbi et al., *Nat. Comm.* **14** (2023) GF, Eisert, Dunjko, MLST. 5 (2024)

 $U_{\phi}^{\dagger}(\mathbf{x}')$ 

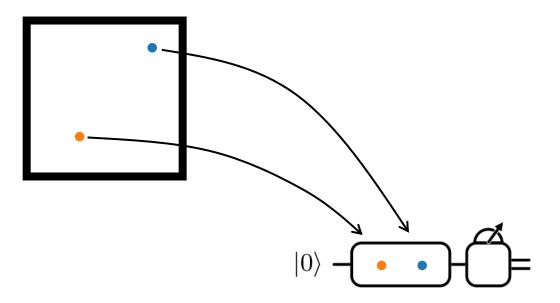
## 5. Quantum feature maps

#### The mathy picture: The quantum computer: $\rho(x) = U(x)|0\rangle\langle 0|U^{\dagger}(x)$ Data domain $\mathcal{X} \subseteq \mathbb{R}^d$ $\operatorname{tr}\left\{\rho(\boldsymbol{x})\rho(\boldsymbol{x'})\right\}$ x $|0\rangle$ HQuantum embedding $\rho \colon \mathcal{X} \to \mathcal{H}_Q$ $|0\rangle$ $\rho(x')$ Hilbert space of $|0\rangle$ density operators $\mathcal{H}_Q \subseteq \operatorname{Herm}(2^n)$ $\rho(x)$ |0) : $U_{\phi}(\mathbf{x})$ Hilbert-Schmidt inner product $|\dot{0}\rangle$ $\langle \cdot , \cdot \rangle_{\mathrm{HS}} \colon \mathcal{H}_Q \times \mathcal{H}_Q \to \mathbb{R}$ Quantum kernel $\operatorname{tr} \left\{ \rho(\boldsymbol{x}) \rho(\boldsymbol{x}') \right\} = \kappa_{\rho}(\boldsymbol{x}, \boldsymbol{x}')$

Embedding Quantum Kernel

### 6. Quantum functions

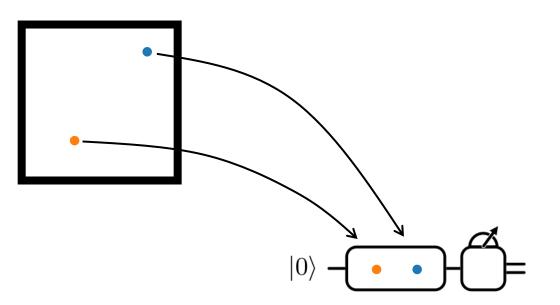
#### Prepare and measure:



GF, Eisert, Dunjko, MLST. 5 (2024)

### 6. Quantum functions

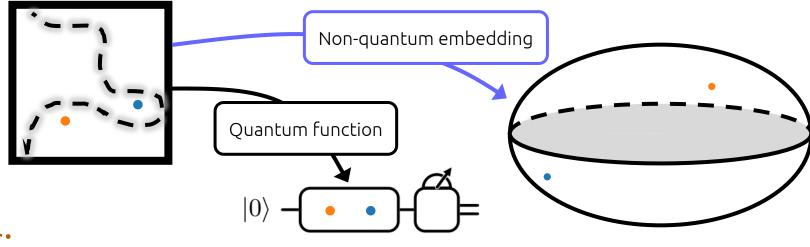
#### Prepare and measure:





### 7. non-Embedding Quantum Kernels?

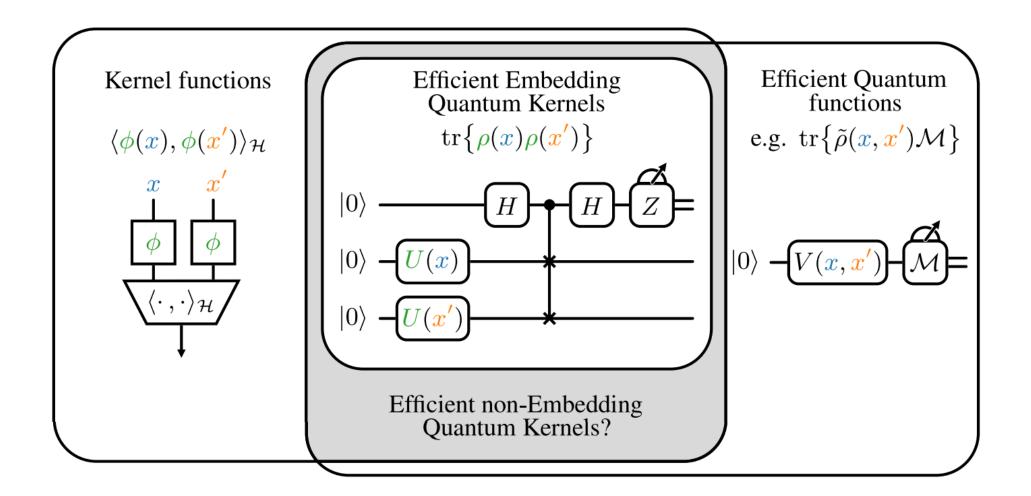
#### Question: what would a non-EQK look like?



#### Answer:

The feature map is a "non-quantum embedding".
The kernel function is still a quantum function.

#### 8. Our question



### 9. Our contribution

Hypothesis: Embedding Quantum Kernels (EQKs) are very expressive.

Test:

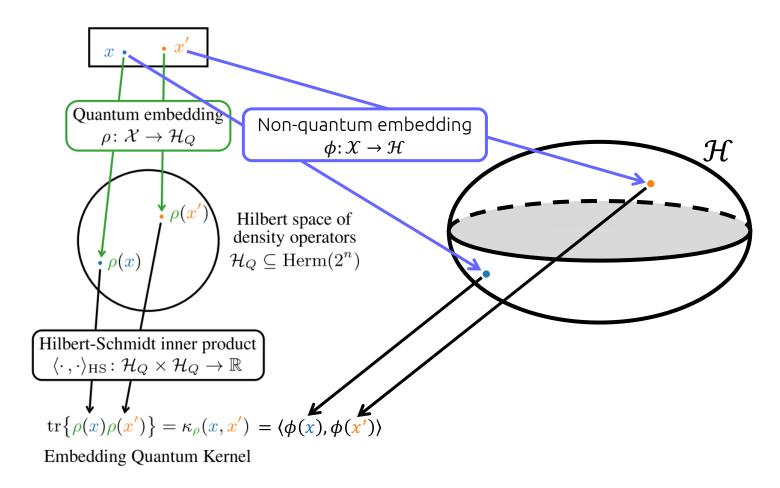
"Take a kernel, can it be expressed as an EQK?"

Results:

"EQKs are <u>universal</u> for important families of kernel functions".

### 10. non-EQKs can also be EQKs

Things aren't always what they seem!



**Results:** *"EQKs are <u>universal</u> for important families of kernel functions".* How? Explicit construction

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2. Construct a feature map

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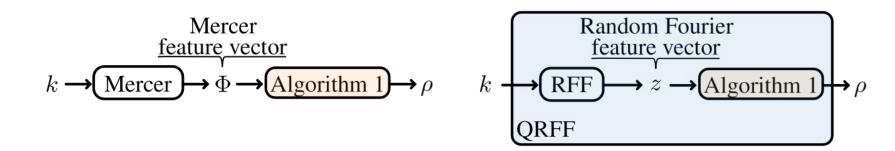
1. Take kernel

2. Construct a feature map

3. Make it a quantum embedding

**Results:** *"EQKs are <u>universal</u> for important families of kernel functions".* How? Explicit construction

1. Take kernel 2. Construct a feature map 3. Make it a quantum embedding



### 12. Take-home message

- 1. Although non-EQKs can exist *in principle,* we are not aware of any.
- 2. If you use EQKs, you're fine... for now.

<u>Invitation:</u> Help us search! We give many possible avenues in the paper.

### References

"QML smells like kernels":

Schuld, Killoran, *PRL* **122 4** (2019) Havlícek *et al., Nature* **567** (2019)

#### "QML is just kernels":

Schuld, *arXiv:2101.11020* (2021)

"QML is not just kernels":

Jerbi *et al., Nat. Comm.* **14** (2023)

Our work: "Are we considering all possible kernels?": <u>GF</u>, Eisert, Dunjko, *MLST*. 5 (2024)





## Thank you for your attention!

On the expressivity of Embedding Quantum Kernels Mach. Learn.: Sci. Technol. 5 025003 (arXiv:2309.14419)

### The kernel trick

Polynomial kernel  $(x, x' \in \mathbb{R}^d)$  Gaussian kernel  $k(x, x') = \langle x, x' \rangle^r$   $k(x, x') = \exp(-\|x - x'\|^2)$ 

#### <u>Option 1</u>: 2d + r - 2 ops.

- $\langle x, x' \rangle$ : d products and d 1 sums
- $(...)^r: r 1$  products

#### <u>Option 1</u>: 3d - 1 ops + sth.

- $||x x'||^2$ : 3d 1 ops.
- exp(...): dunno..., not much?

 $\frac{\text{Option 2: }\infty.}{k(x, x') = \langle \phi_d(x), \phi_d(x') \rangle}$ 

 $\phi_d(x)$  all monomials.

There are infinitely many of them. Construct two ∞-length vectors Total ∞.