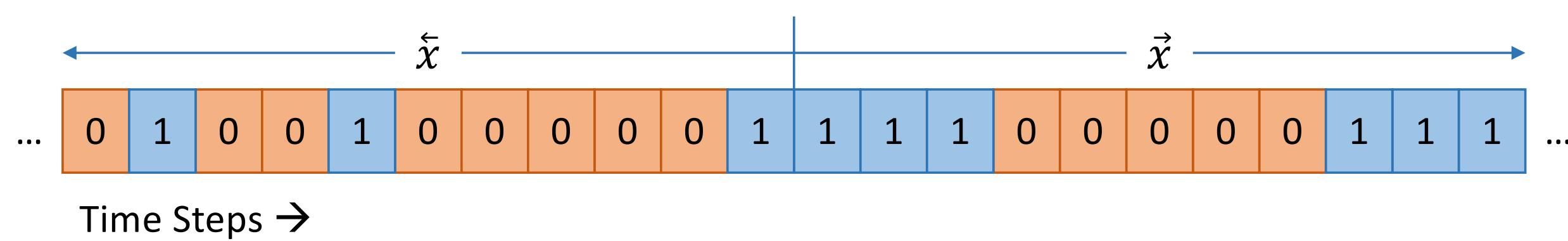


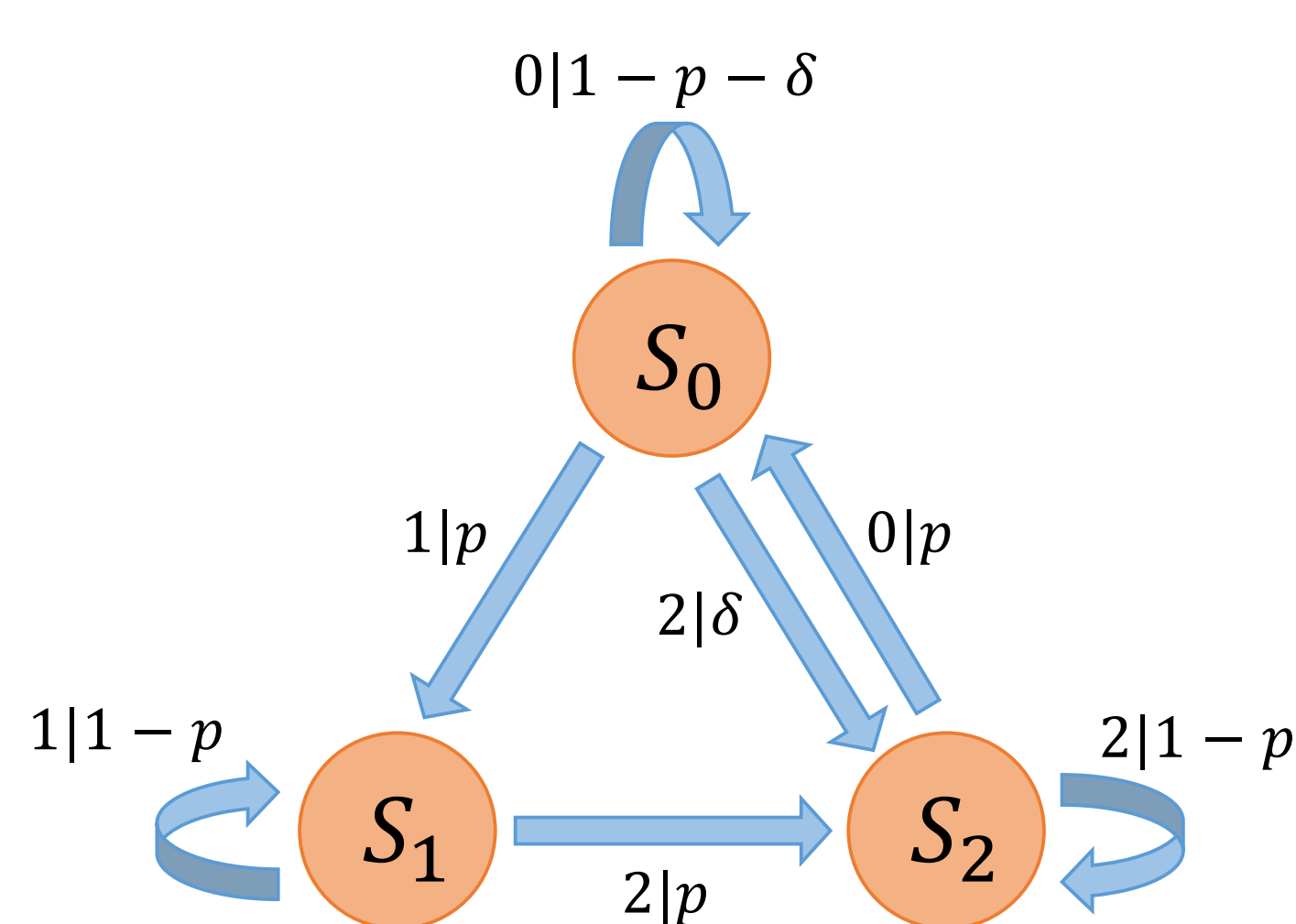
Stochastic Processes

We consider bi-infinite stationary discrete-time stochastic processes with

- a sequence of random variables $X_t \in \mathbb{N}$ with $t \in \mathbb{Z}$
- the *past* defined as $\vec{X} := \dots, X_{-2}, X_{-1}$
- the *future* defined as $\vec{X} := X_0, X_1, \dots$
- a governing joint probability distribution $P(\vec{X}, \vec{X})$
- a conditional distribution $P(\vec{X} | \vec{X})$, given a specific past instance \vec{X}



Example Process



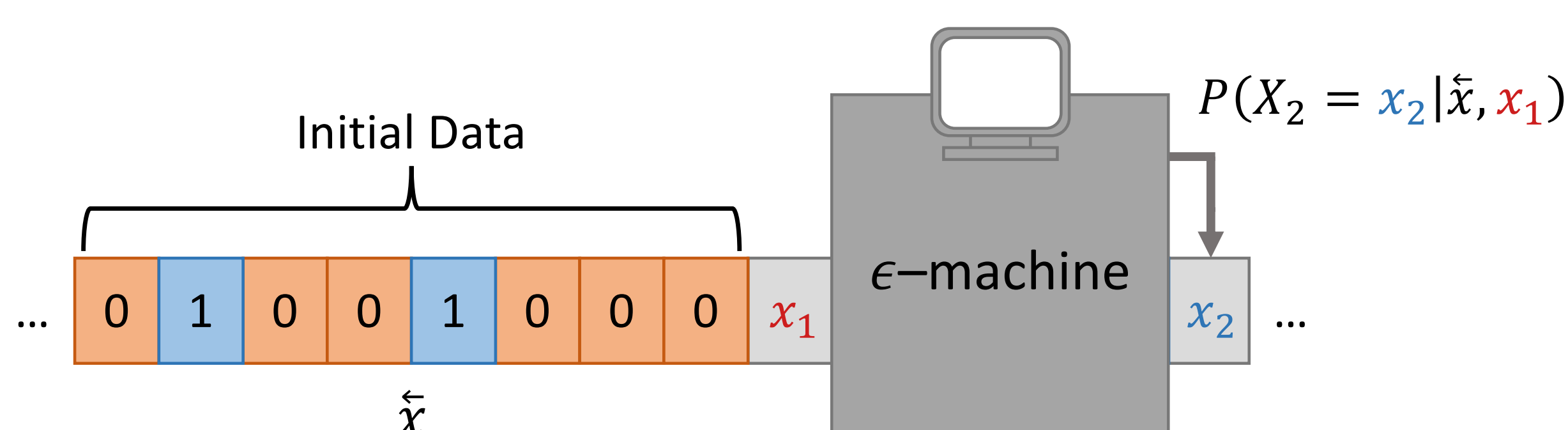
Graphical representation of the *quasi cycle process* with memory states s_i , emissions, and transition probabilities $x|P(x)$.

Classical Models

The provably optimal classical models are ϵ -machines, which are based on the equivalence relation

$$\vec{X} \sim \vec{X}' \iff P(\vec{X} | \vec{X}) = P(\vec{X}' | \vec{X}'). \quad (1)$$

For each class, one *memory state* $\mathcal{E}(\vec{X}) = s_i$ is allocated. The model initializes to a state defined by the input \vec{X} and outputs a single time step once at a time:

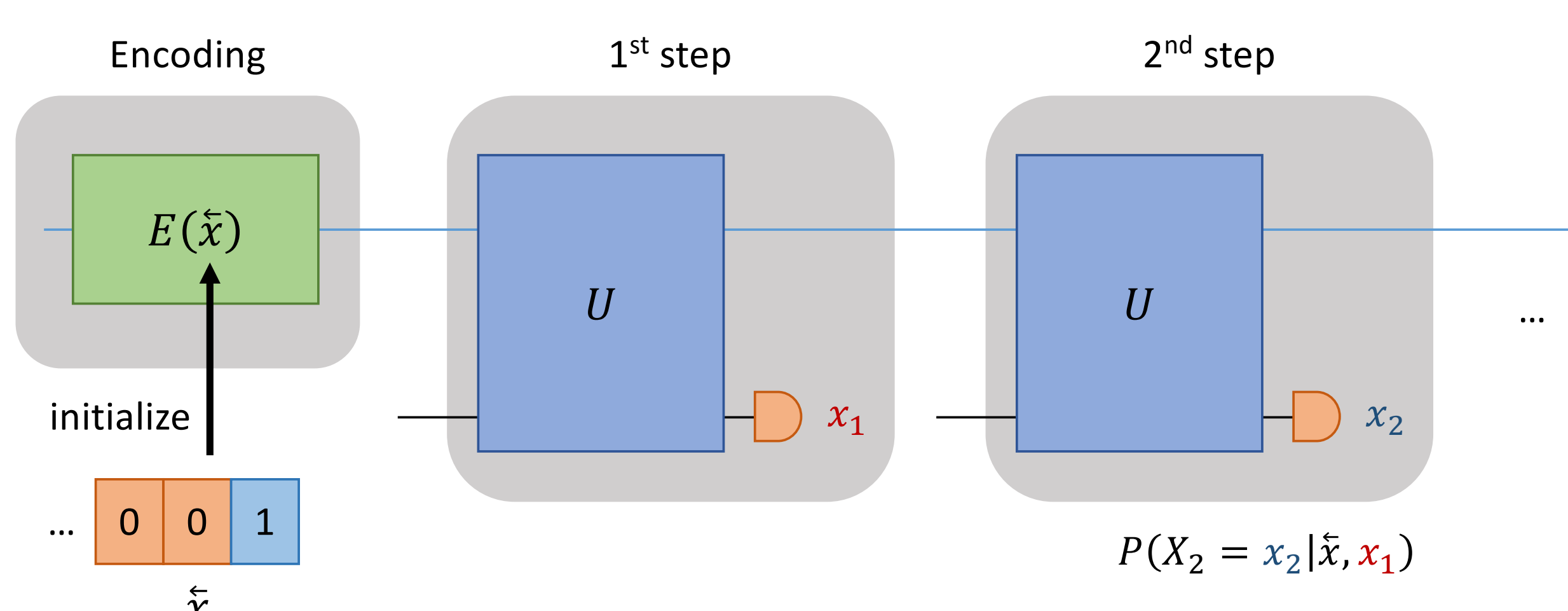


Quantum Models

The quantum analog is called q -simulator and operates onto quantum memory states $|s_i\rangle$ as

$$|1_i\rangle := U|s_i\rangle|0\rangle = \sum_x \sqrt{P(x|s_i)} |s_{\lambda(i,x)}\rangle |x\rangle, \quad (2)$$

where $\lambda(i, x)$ denotes the index to the next state. A measurement operation onto the second register of $|1_i\rangle$ thus outputs x with probability $P(x|s_i)$:



Low-rank Approximations

We aim for approximate models ($\hat{P} \approx P$) and start with the following

Theorem [1]: Given a stochastic process with P , λ , and $\{s_i\}_{i=1}^n$. A q -simulator exists iff

$$\langle s_i | s_j \rangle = \langle 1_i | 1_j \rangle \quad \forall i, j = 1, 2, \dots, n. \quad (3)$$

→ The main idea of this work is to perform a low-rank approximation of the overlap matrix $C_{ij} = \langle 1_i | 1_j \rangle$ and derive the quantum states $|s_i\rangle$ and unitary simulator U from it.

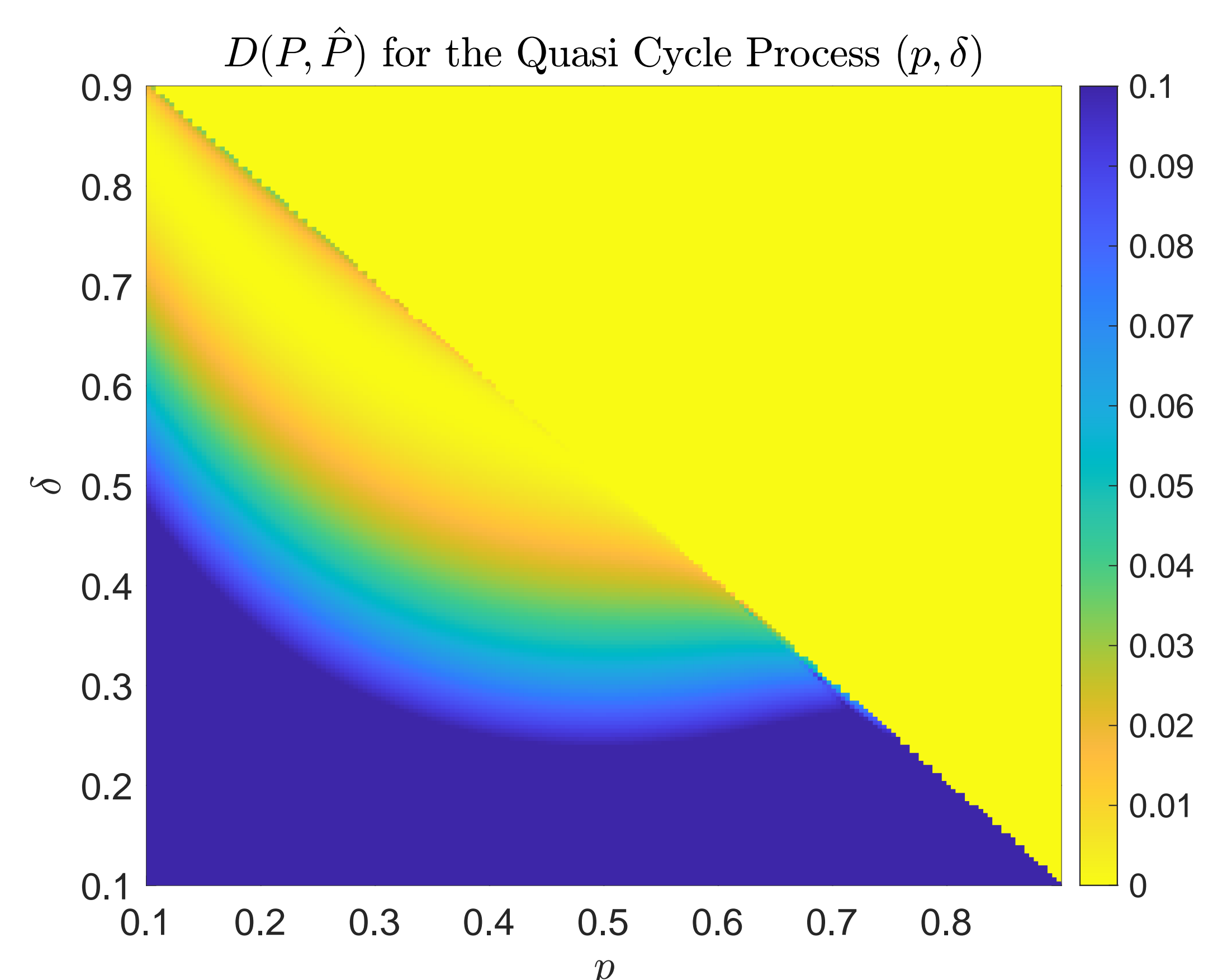
Sketch of the algorithm

- 1 Construct the overlap matrix $C \leftarrow P, \lambda, \{s_i\}_{i=1}^n$
- 2 Perform a SVD of $C = V\Sigma W^\dagger = V\Sigma V^\dagger = V\sqrt{\Sigma}\sqrt{\Sigma}V^\dagger$
- 3 Shrink to a low-rank approximation $C \rightarrow C^{(d)} = V\sqrt{\Sigma^{(d)}}\sqrt{\Sigma^{(d)}}V^\dagger$
- 4 Identify the quantum states $|\hat{s}_i\rangle$ as columns of $\hat{S} = \sqrt{\Sigma^{(d)}}V^\dagger$
- 5 Construct the approximate one-step matrix \hat{F} with columns $\hat{F}_i = |\hat{1}_i\rangle$
- 6 Approximate the unitary simulator as

$$\arg \min_{\hat{U}} \|\hat{U}\hat{S} \odot |0\rangle - \hat{F}\|_F^2 \quad \text{s.t.} \quad \hat{U}^\dagger \hat{U} = I \quad (4)$$

Interpretation: The approximate states $|\hat{s}_i\rangle$ are the quantum states to a slightly different stochastic process \hat{P} simulated by \hat{U} following (2).

Results



→ Via comparison with the best classical models [2], these approximate quantum models are superior with respect to the KL divergence

$$D(P, \hat{P}) = \sum_{\vec{X}, \vec{X}'} P(\vec{X}) P(\vec{X}' | \vec{X}) \log_2 \frac{P(\vec{X}' | \vec{X})}{\hat{P}(\vec{X}' | \vec{X})}. \quad (5)$$

Future Directions

- Derive upper bounds onto the KL divergence
- Include complex phases for the overlap matrix C
- Derive QML ansätze to learn approximations based only on data

References

- [1] Felix C. Binder, Jayne Thompson, and Mile Gu. Practical unitary simulator for non-markovian complex processes. *Phys. Rev. Lett.*, 120:240502, Jun 2018.
- [2] Chengran Yang, Andrew J. P. Garner, Feiyang Liu, Nora Tischler, Jayne Thompson, Man-Hong Yung, Mile Gu, and Oscar Dahlsten. Provably superior accuracy in quantum stochastic modeling. *Phys. Rev. A*, 108:022411, Aug 2023.