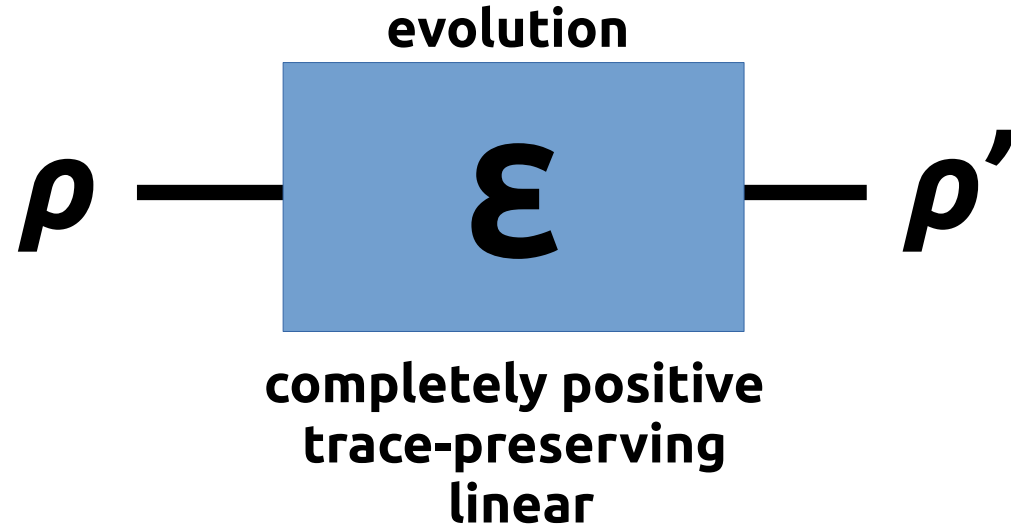


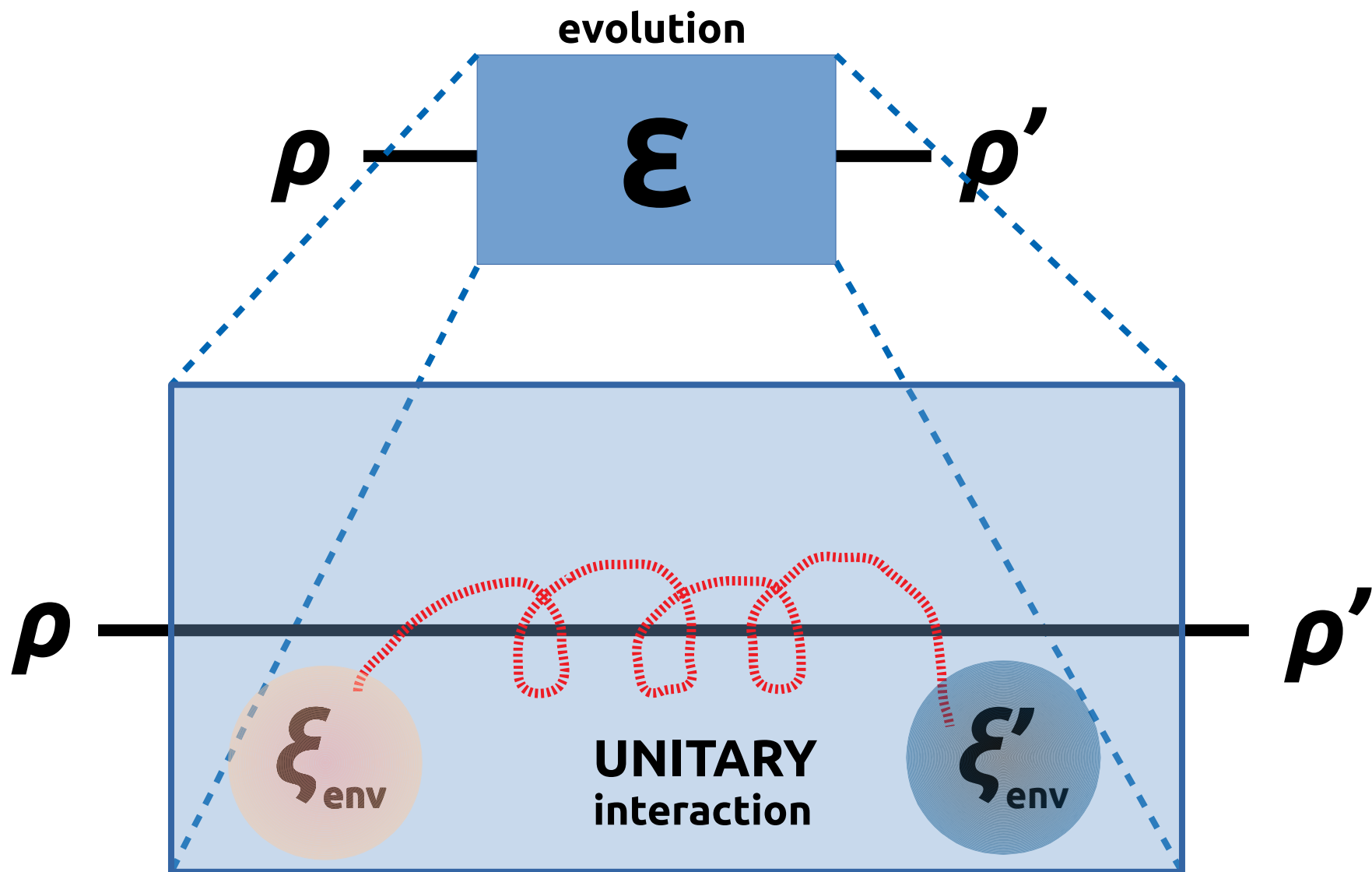
Probabilistic storing of quantum dynamics

Mario Ziman

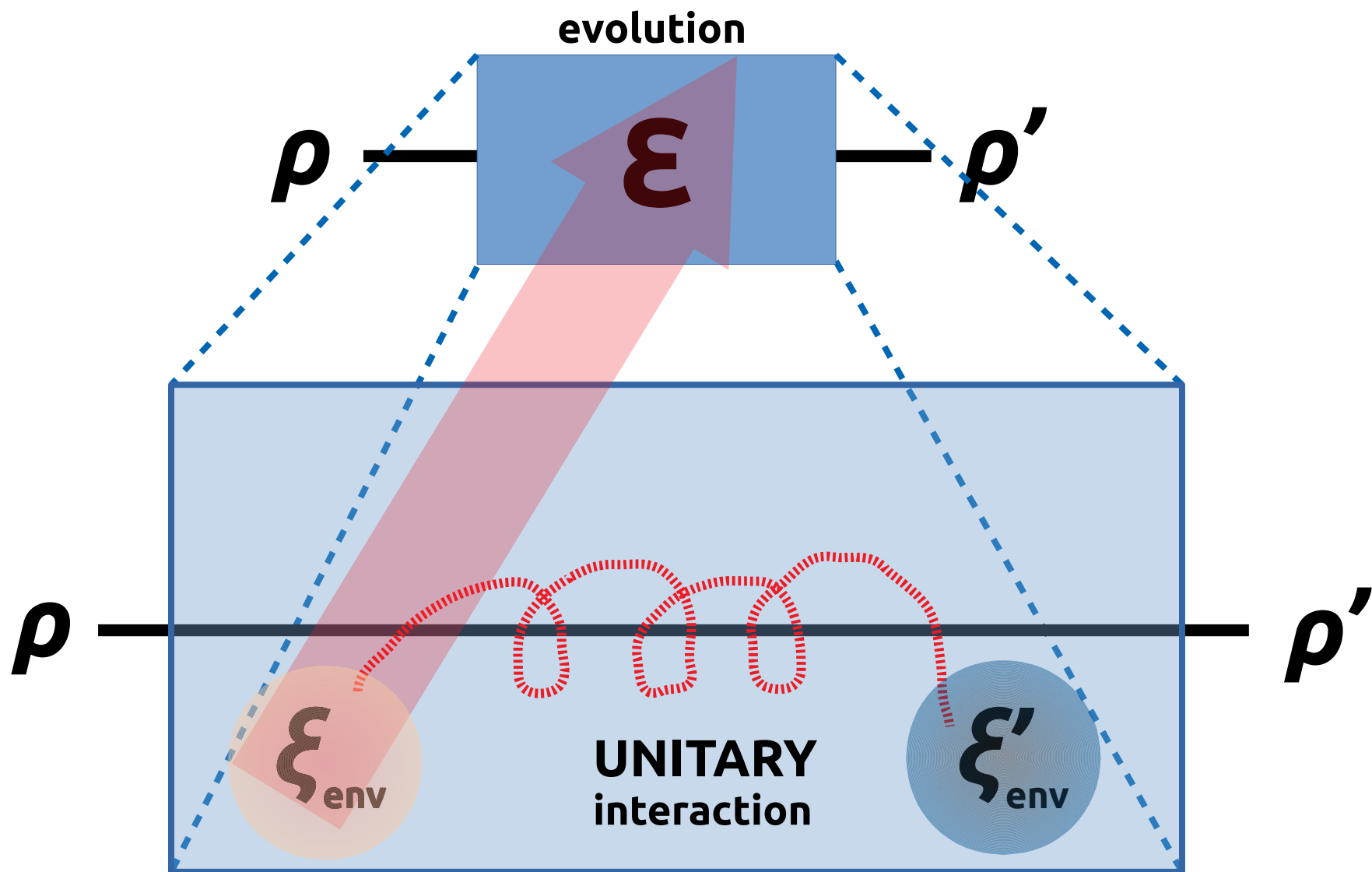
MESSAGE FROM STINESPRING



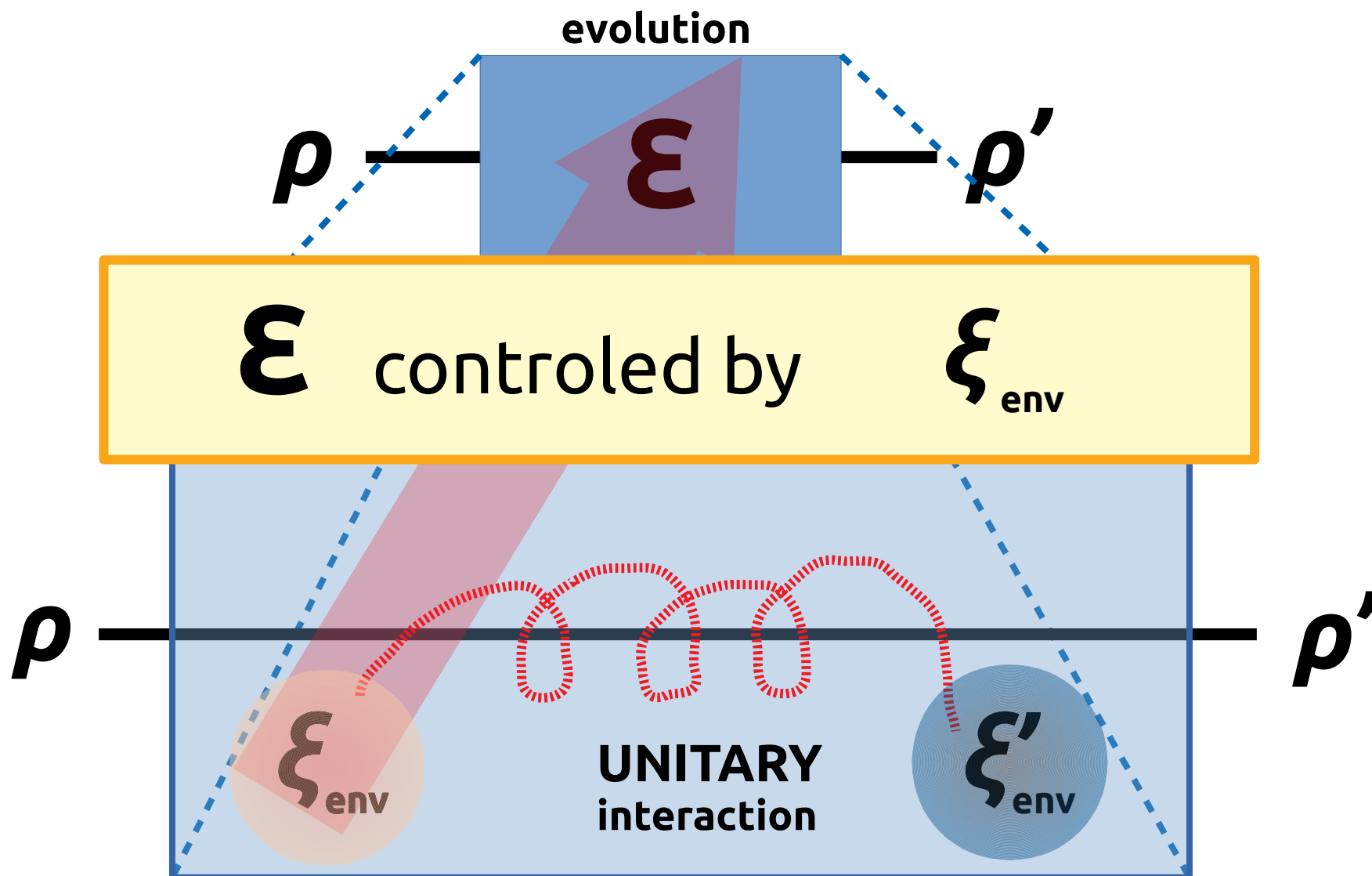
MESSAGE FROM STINESPRING



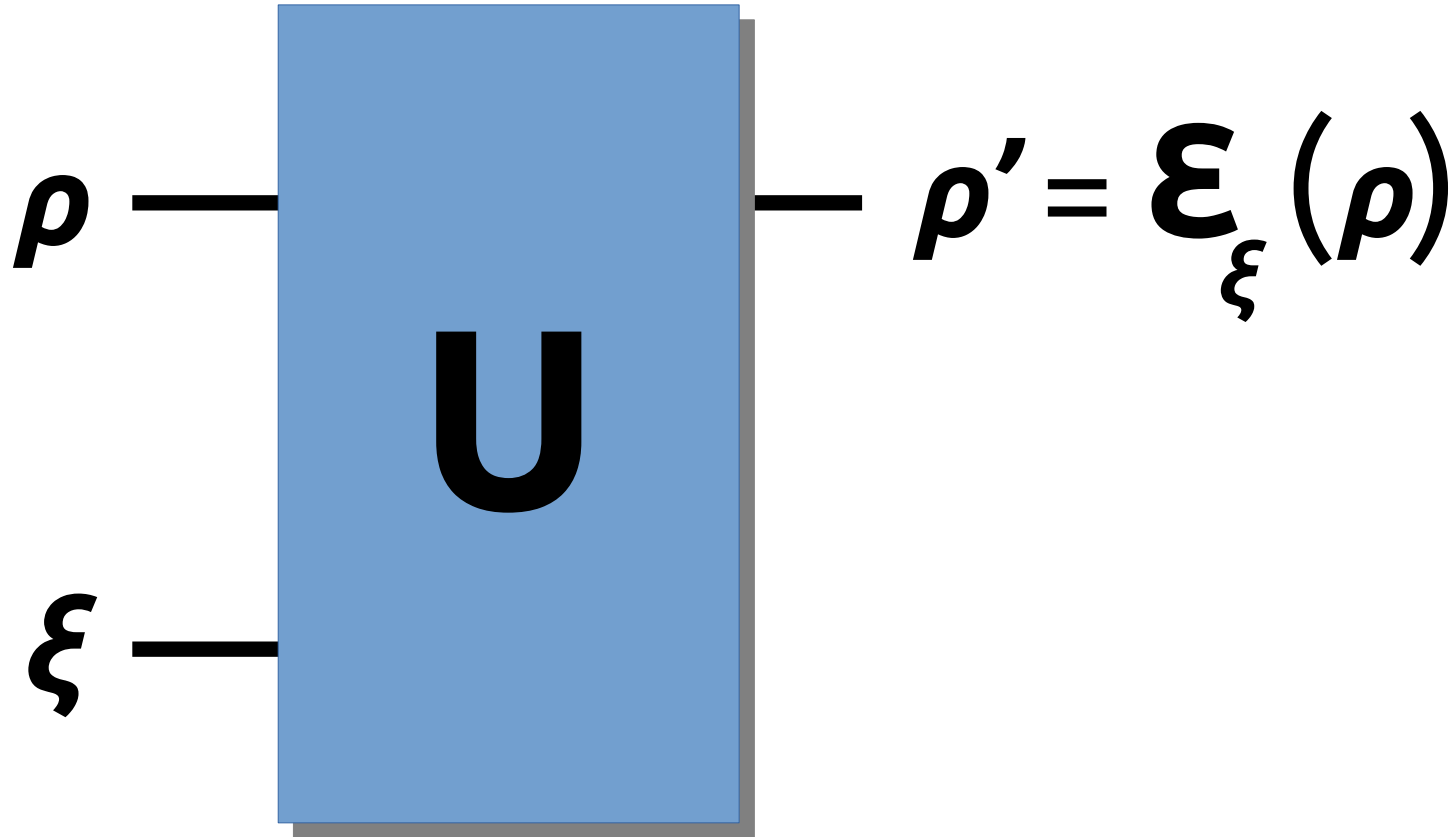
MESSAGE FROM STINESPRING



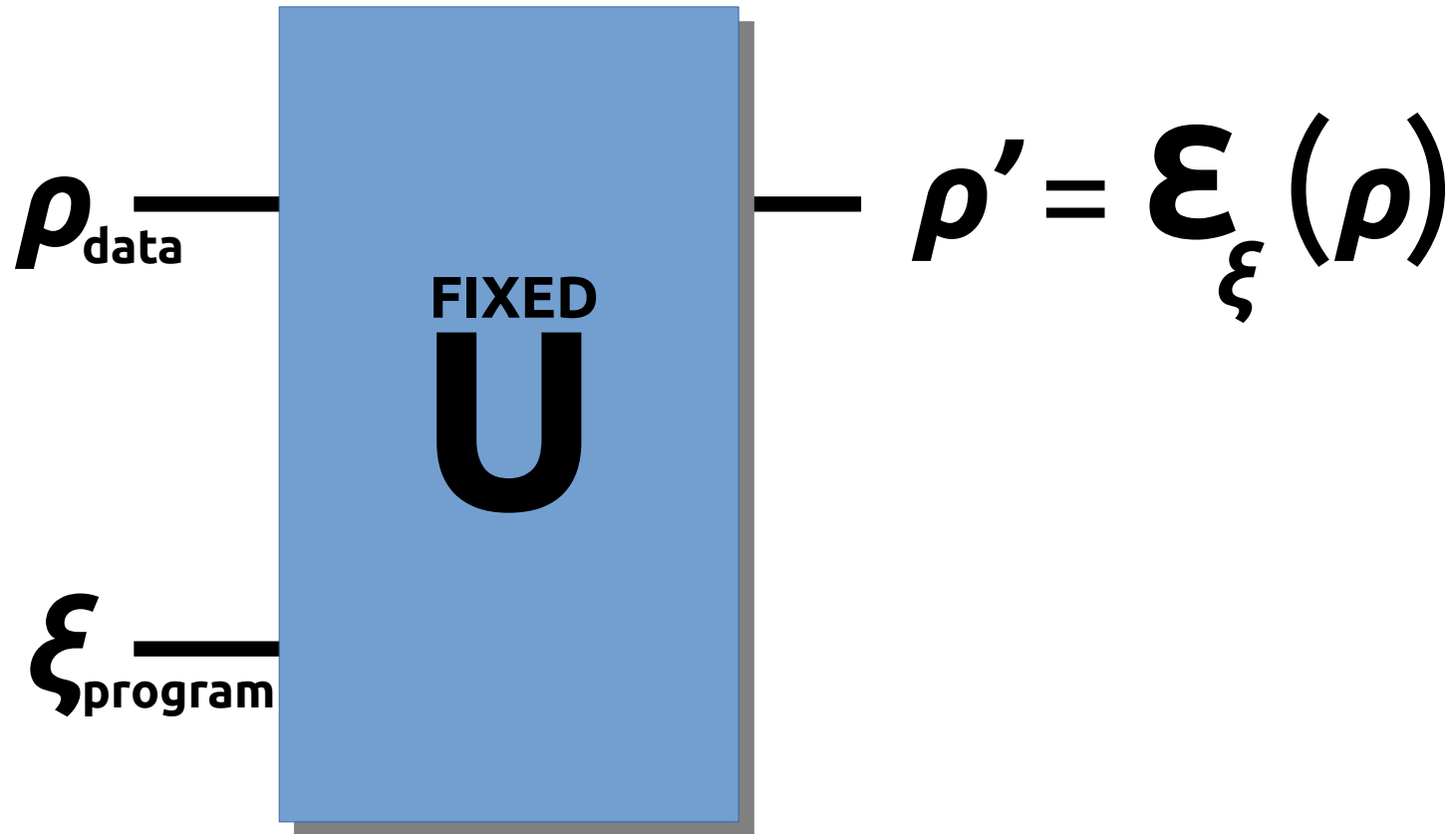
MESSAGE FROM STINESPRING



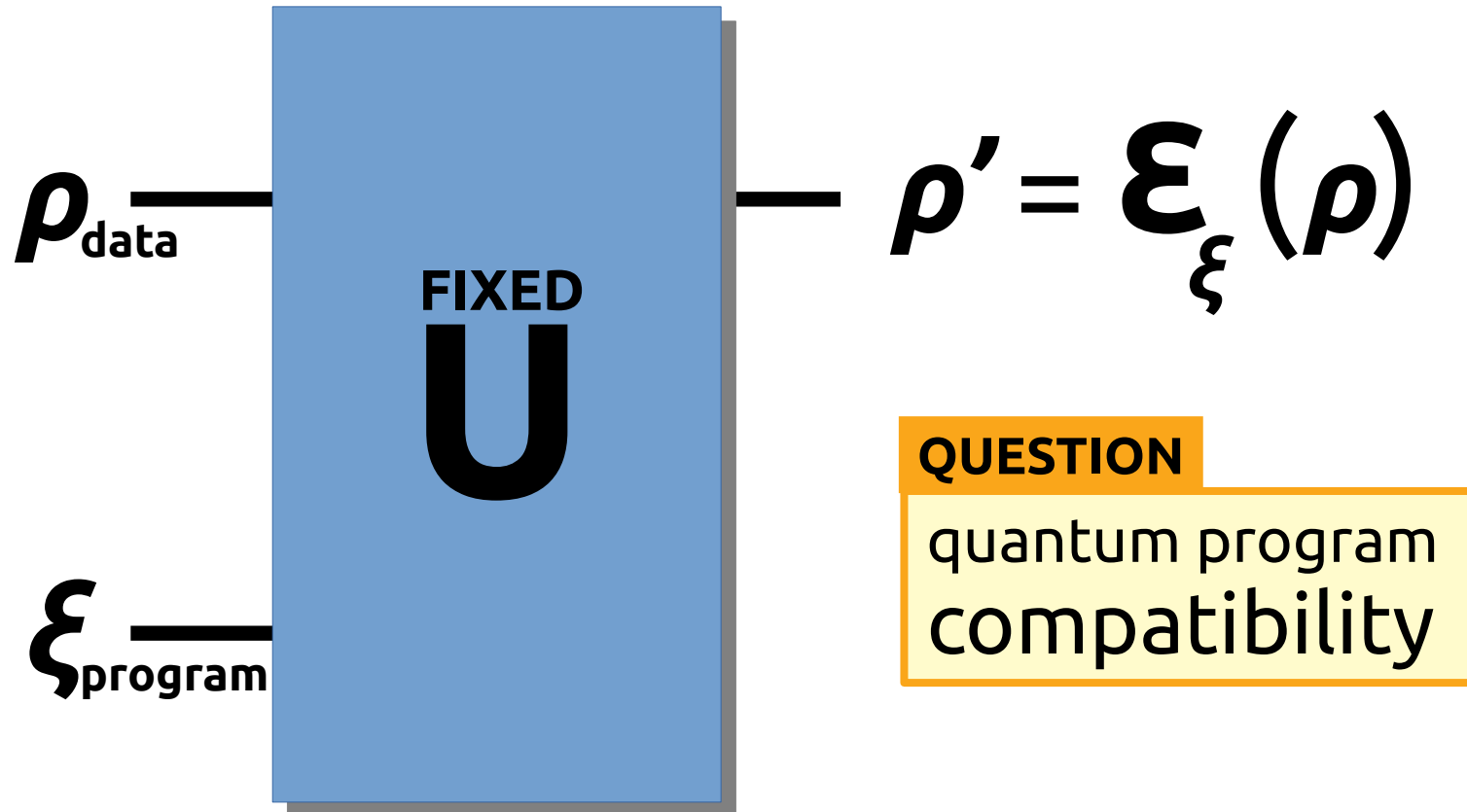
MESSAGE FROM STINESPRING



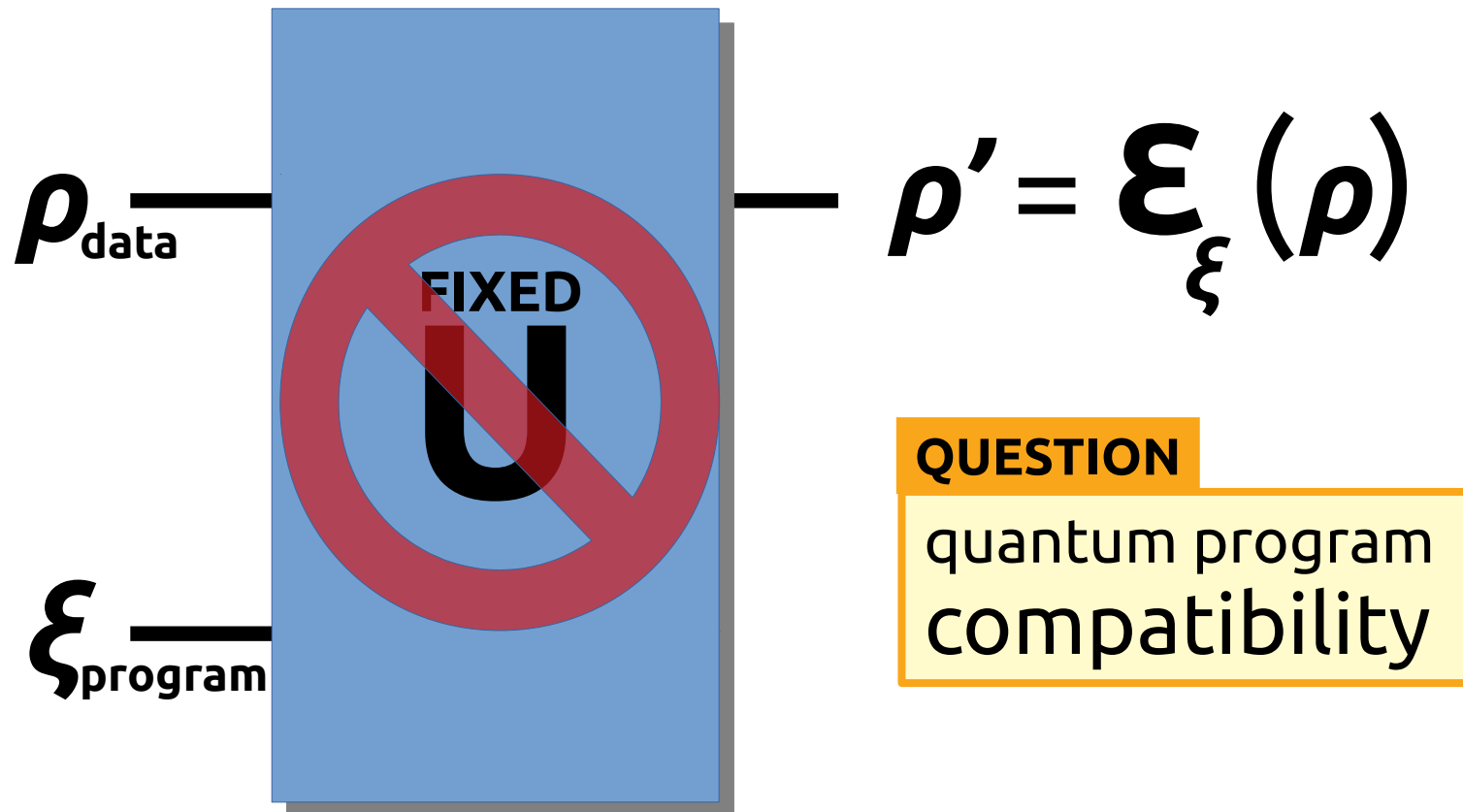
PARADIGM OF QUANTUM PROGRAMMING



PARADIGM OF QUANTUM PROGRAMMING

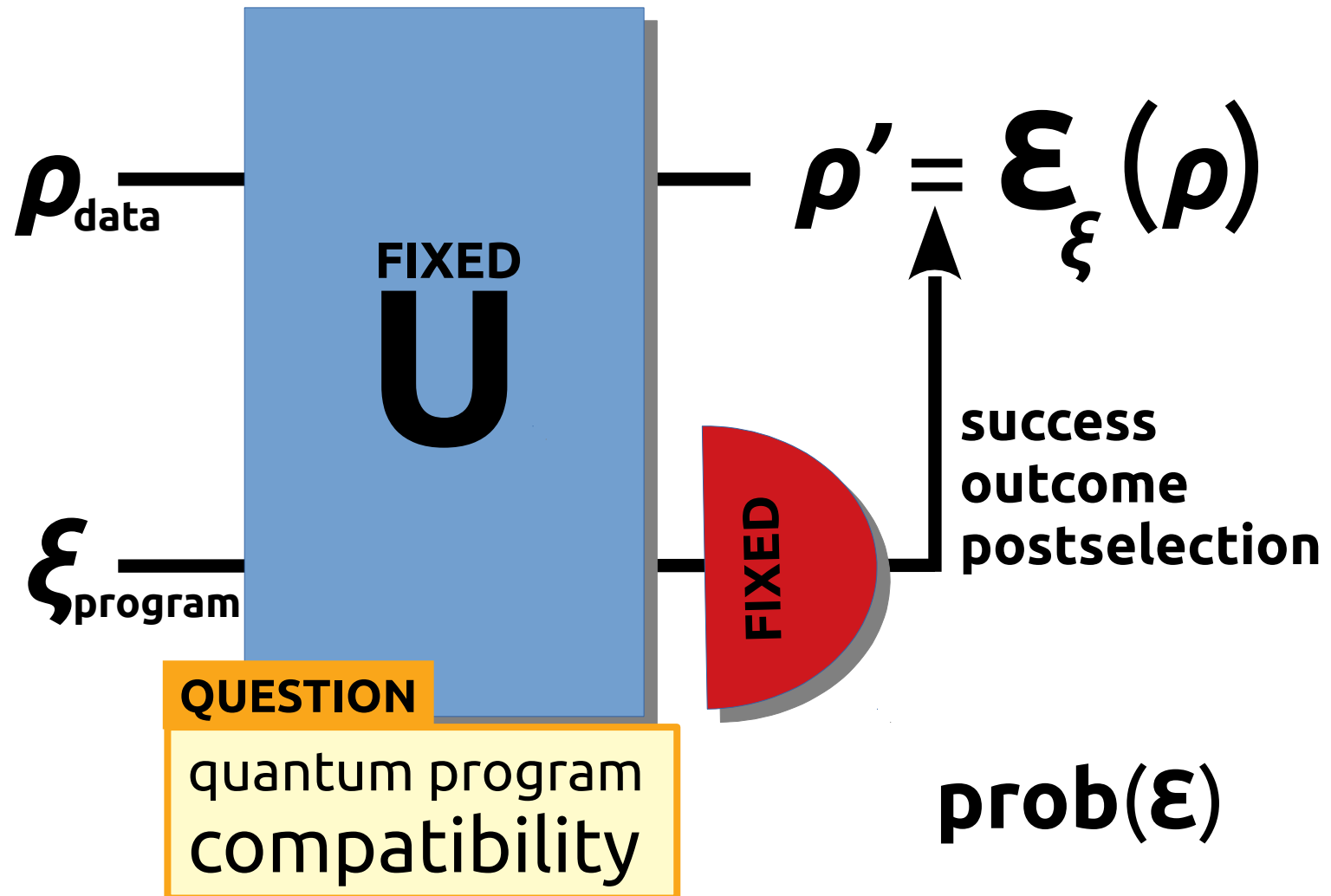


PARADIGM OF QUANTUM PROGRAMMING



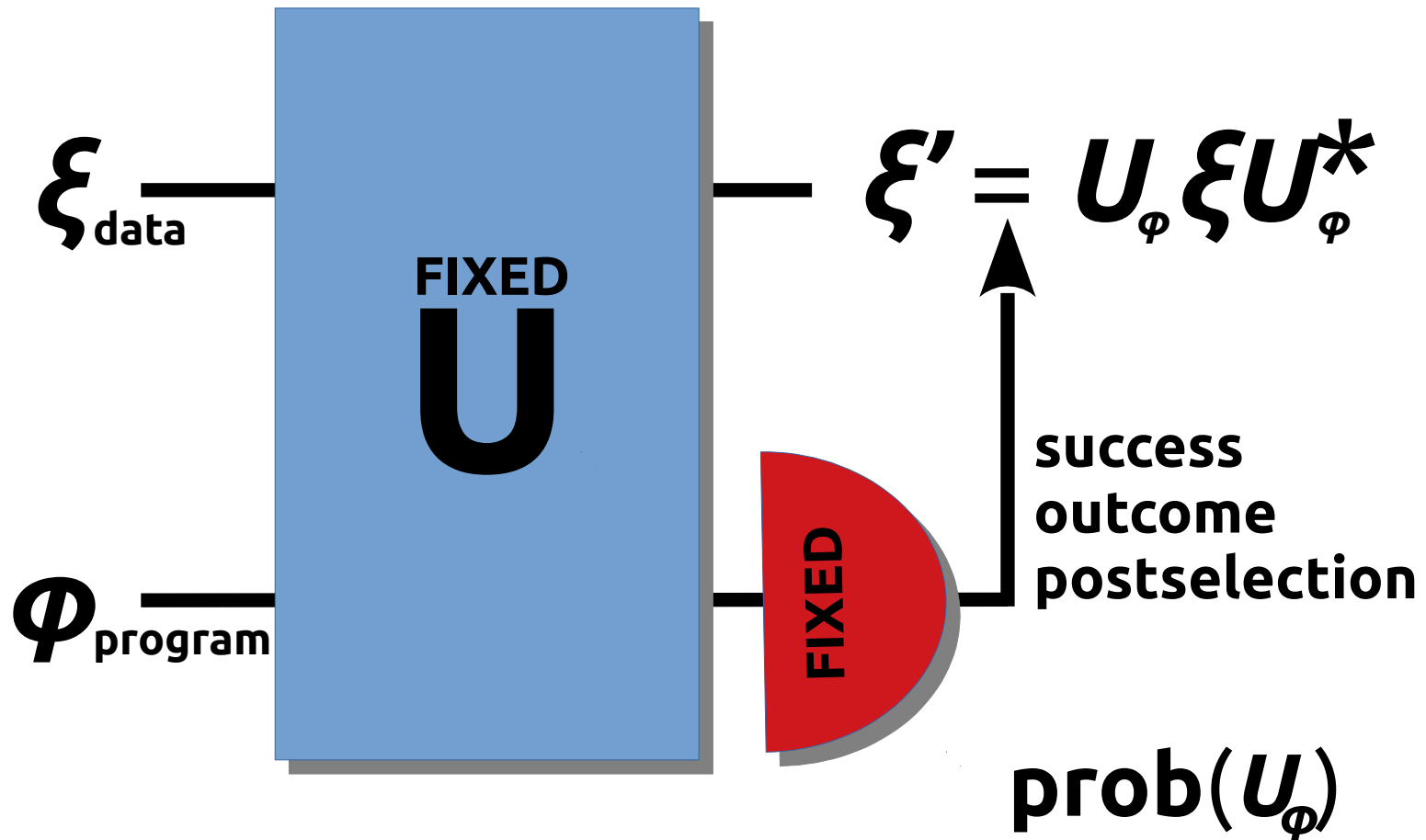
NO-PROGRAMMING THEOREM

HERALDED UNIVERSALITY



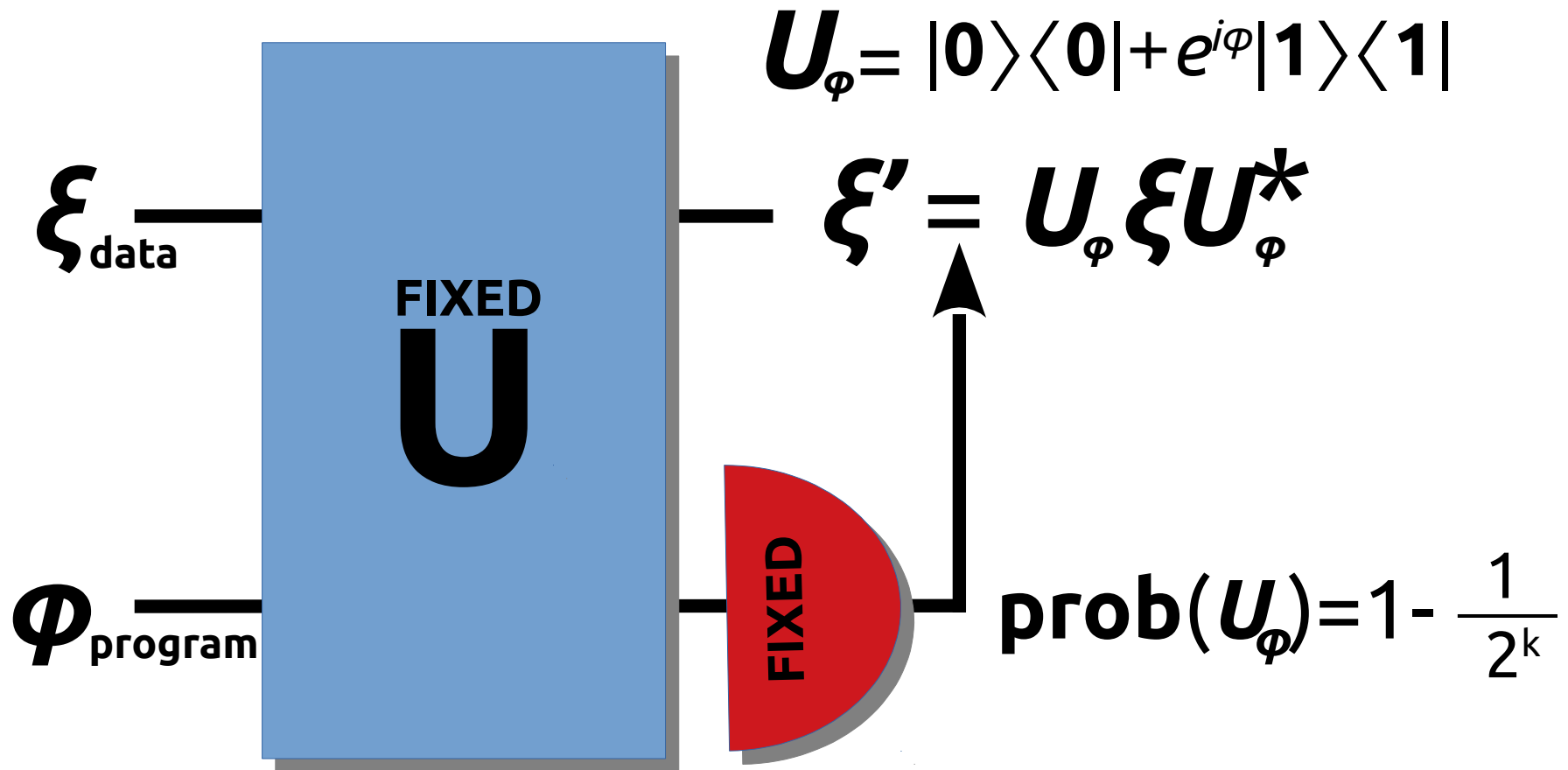
PROGRAMMING THE PHASE

$$U_\varphi = |0\rangle\langle 0| + e^{i\varphi}|1\rangle\langle 1|$$



optimize $\dim(H_{\text{program}})$ and success probability

PROGRAMMING THE PHASE



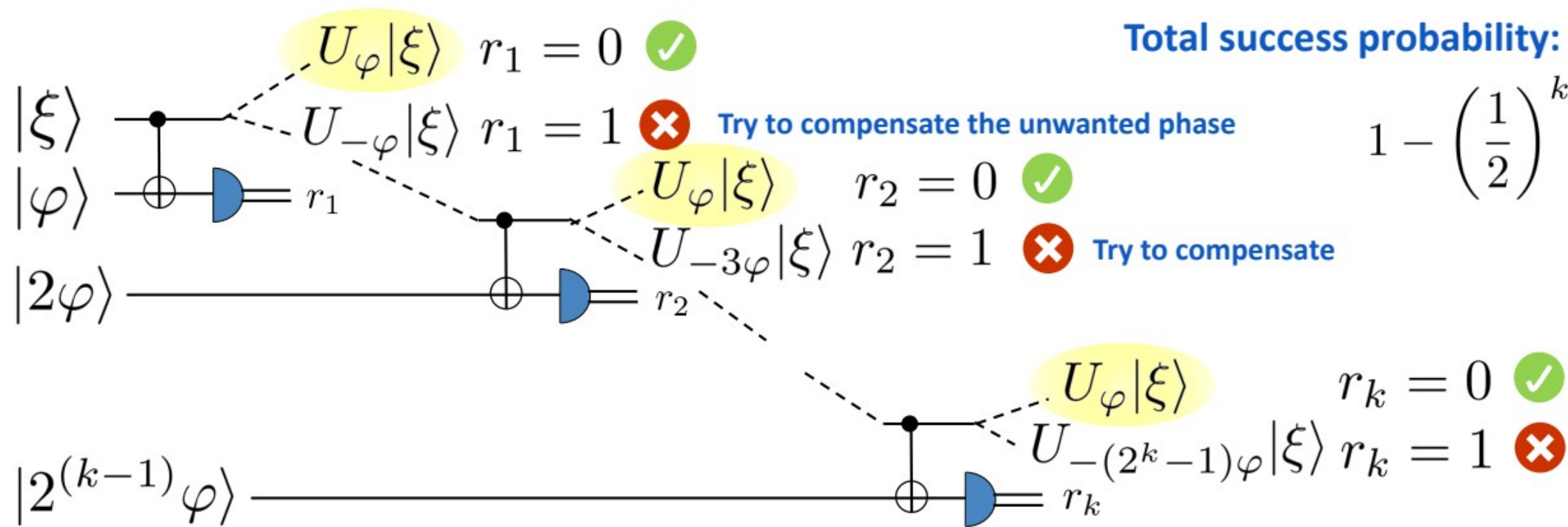
$$U_{\phi} = |0\rangle\langle 0| + e^{i\phi}|1\rangle\langle 1|$$

$$\xi' = U_{\phi} \xi U_{\phi}^*$$

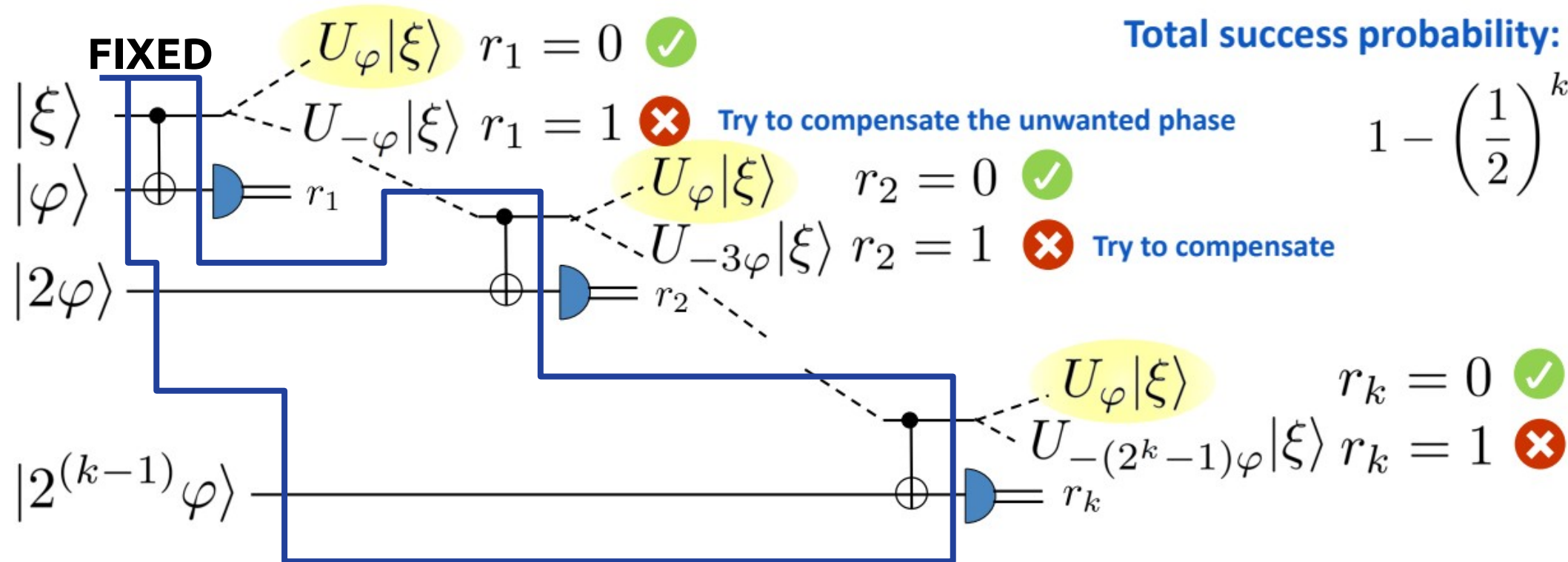
$$\Phi_{\text{program}} = |\phi\rangle |2\phi\rangle \dots |2^{(k-1)}\phi\rangle$$

$$|\phi\rangle = |0\rangle + e^{i\phi}|1\rangle$$

PROGRAMMING THE PHASE



PROGRAMMING THE PHASE



phase stored in

$$\Phi_{\text{program}} = |\varphi\rangle |2\varphi\rangle \dots |2^{(k-1)}\varphi\rangle$$

INVERSE QUESTION

dynamics stored in Φ_{program}

**HOW TO
STORE / LEARN
DYNAMICS?**



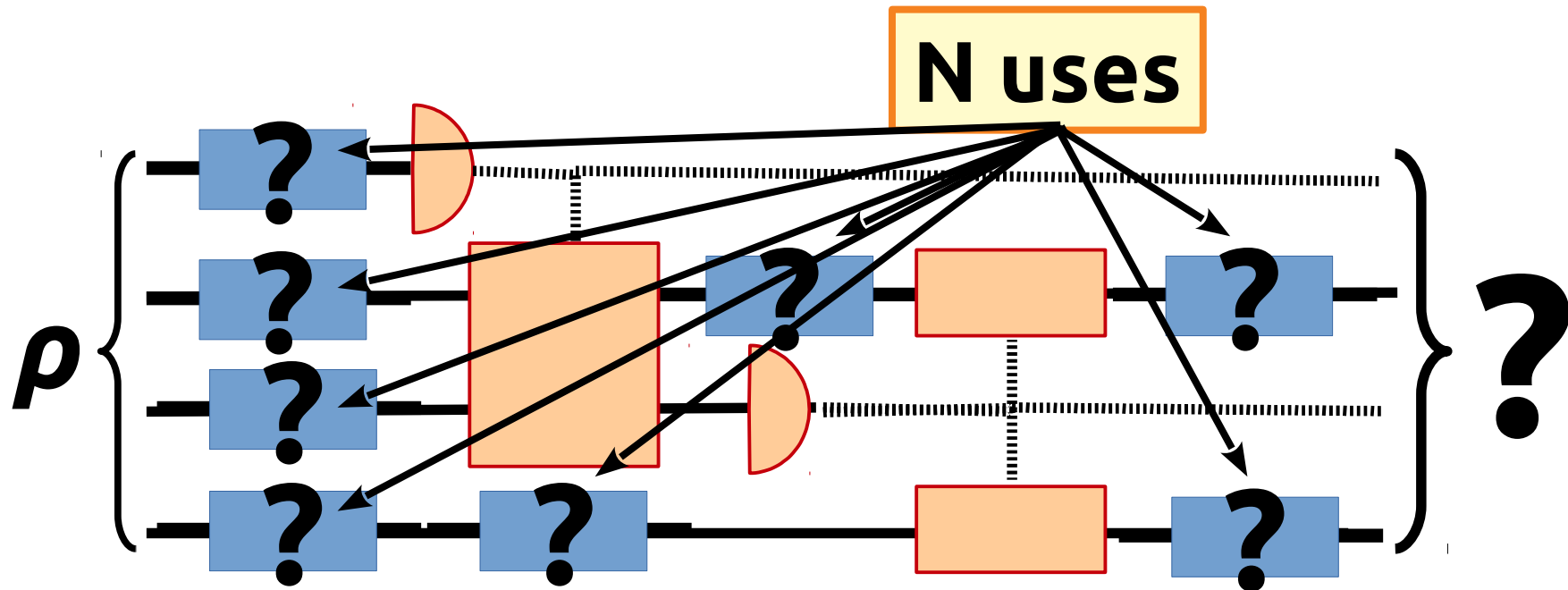
INVERSE QUESTION

dynamics stored in Φ_{program}

HOW TO
PROBABILISTICALLY
STORE / LEARN
UNKNOWN
DYNAMICS?



LEARNING THE PROCESS

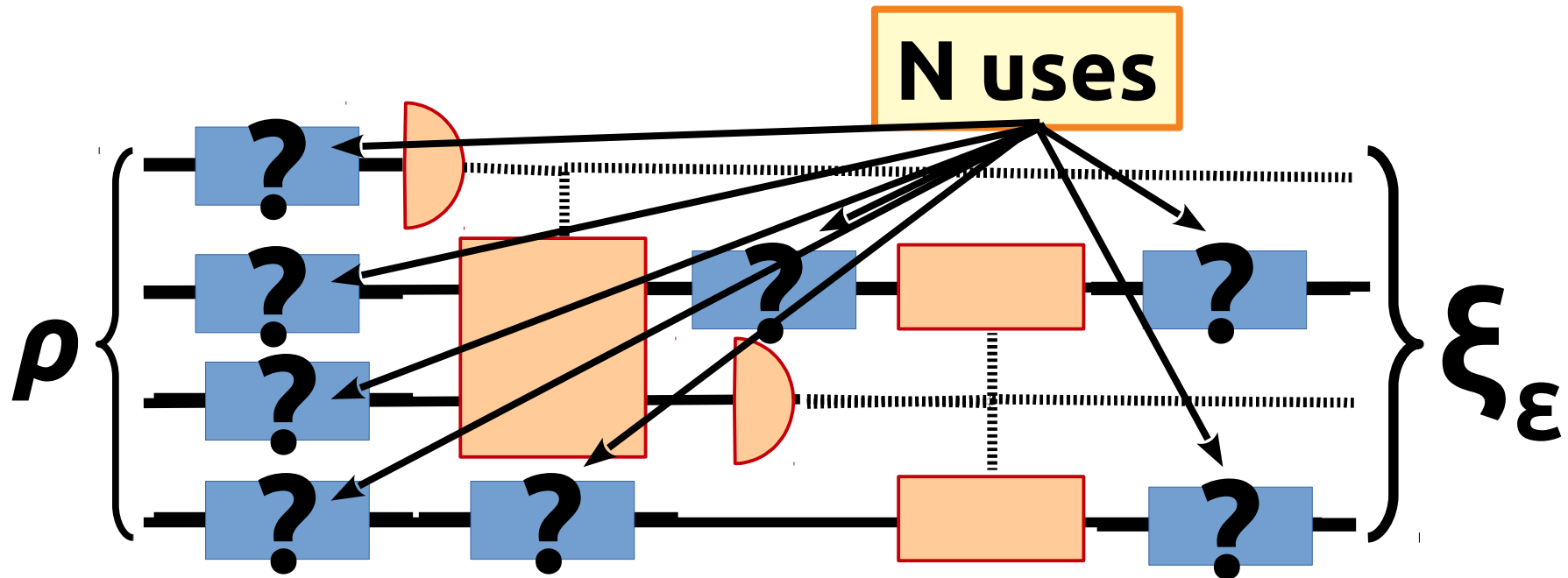


result of learning

→ a program simulating the process

→ stored in ~~bits~~ **qubits**

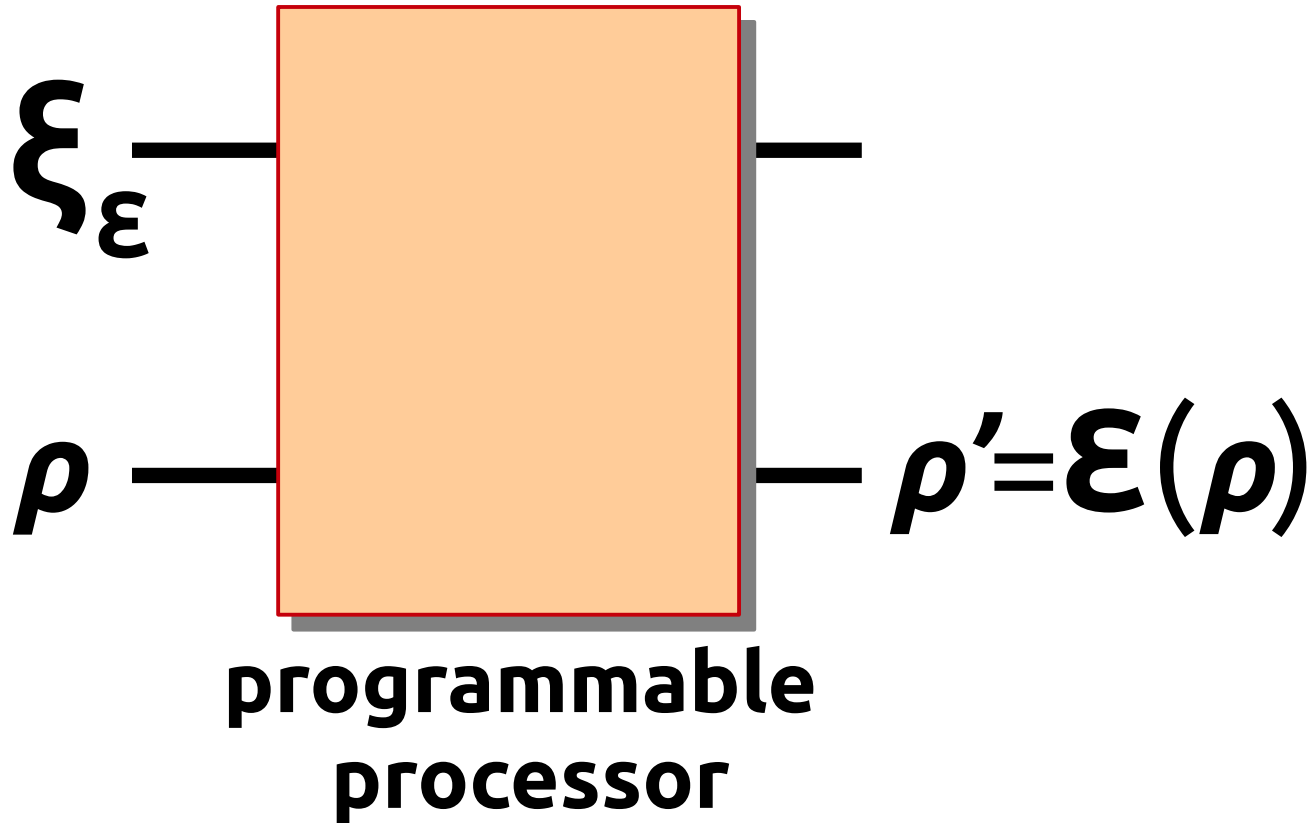
LEARNING THE PROCESS



Storing in quantum memory

$$\mathcal{E} \rightarrow \xi_\epsilon$$

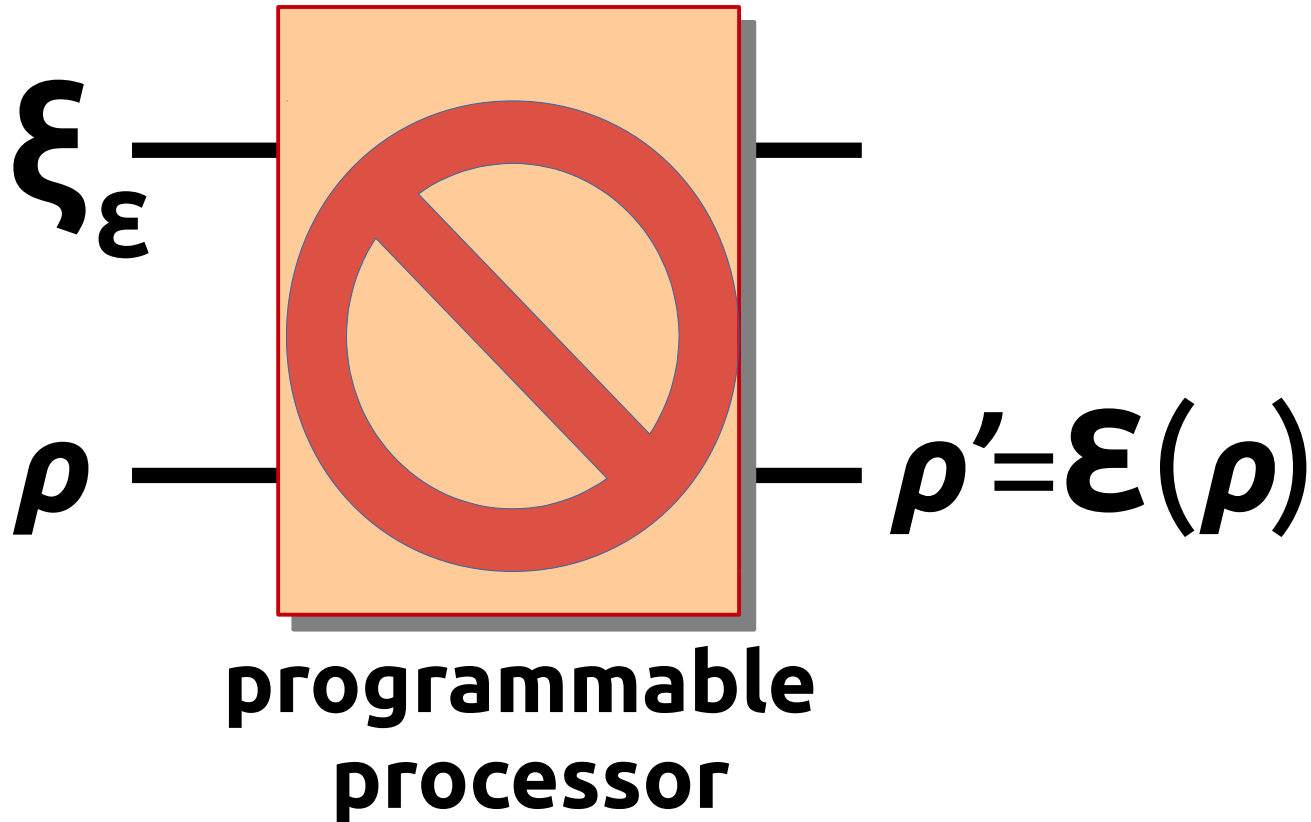
UNDONE LEARNING



$N \rightarrow 1$ RETRIEVAL

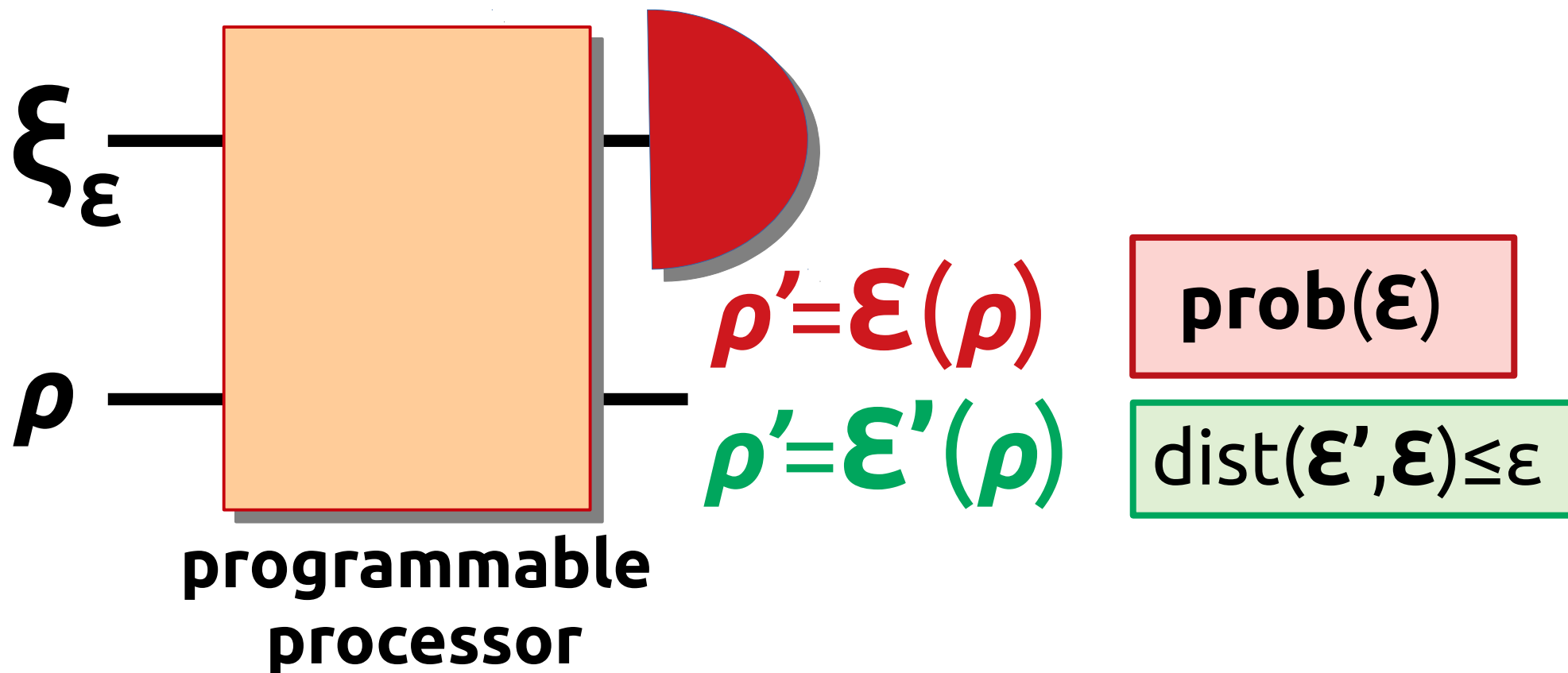
UNDONE LEARNING

NO PROGRAMMING



N → 1 RETRIEVAL

UNDONE LEARNING **PROBABILISTICALLY** **APPROXIMATIVELY**

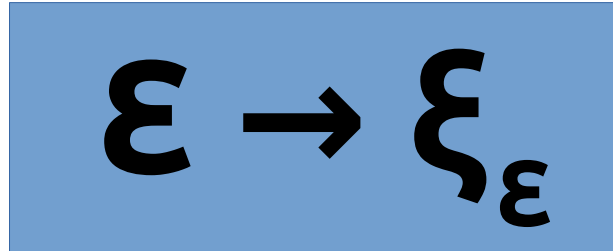


? N → 1 RETRIEVAL ?

QUANTUM LEARNING

Storing process in quantum memory.

N uses



Retrieval through programming.

M applications

either approximate, or probabilistic



APPROXIMATE Q LEARNING

Optimal strategy for unitary channels

MEASURE-AND-ROTATE

APPROXIMATE Q LEARNING

Optimal strategy for unitary channels

MEASURE-AND-ROTATE

- optimal learning = optimal estimation
- storing is classical

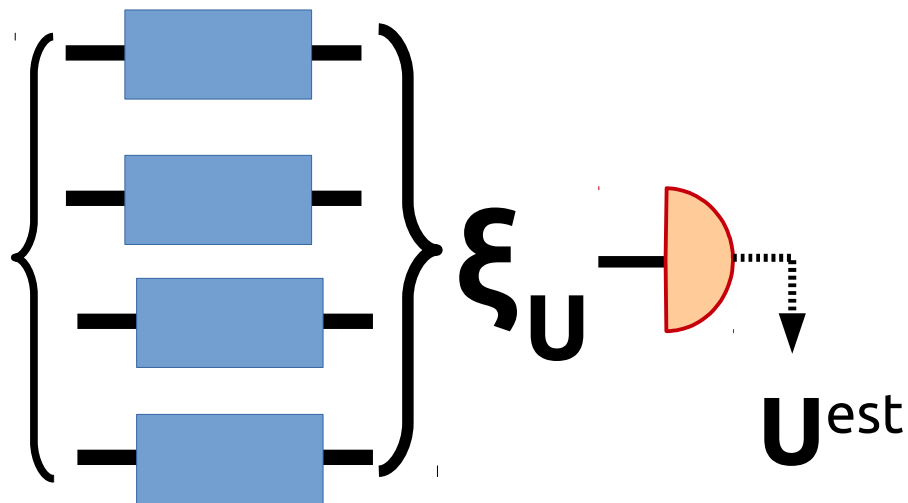
APPROXIMATE Q LEARNING

Optimal strategy for unitary channels

→ optimal state

$$|\Psi\rangle = \bigotimes_{j \in IRR} \sqrt{p_j/d_j} |I_j\rangle\rangle$$

$$|I_j\rangle\rangle \in H_j \otimes H_j$$



→ optimal POVM (continuous)

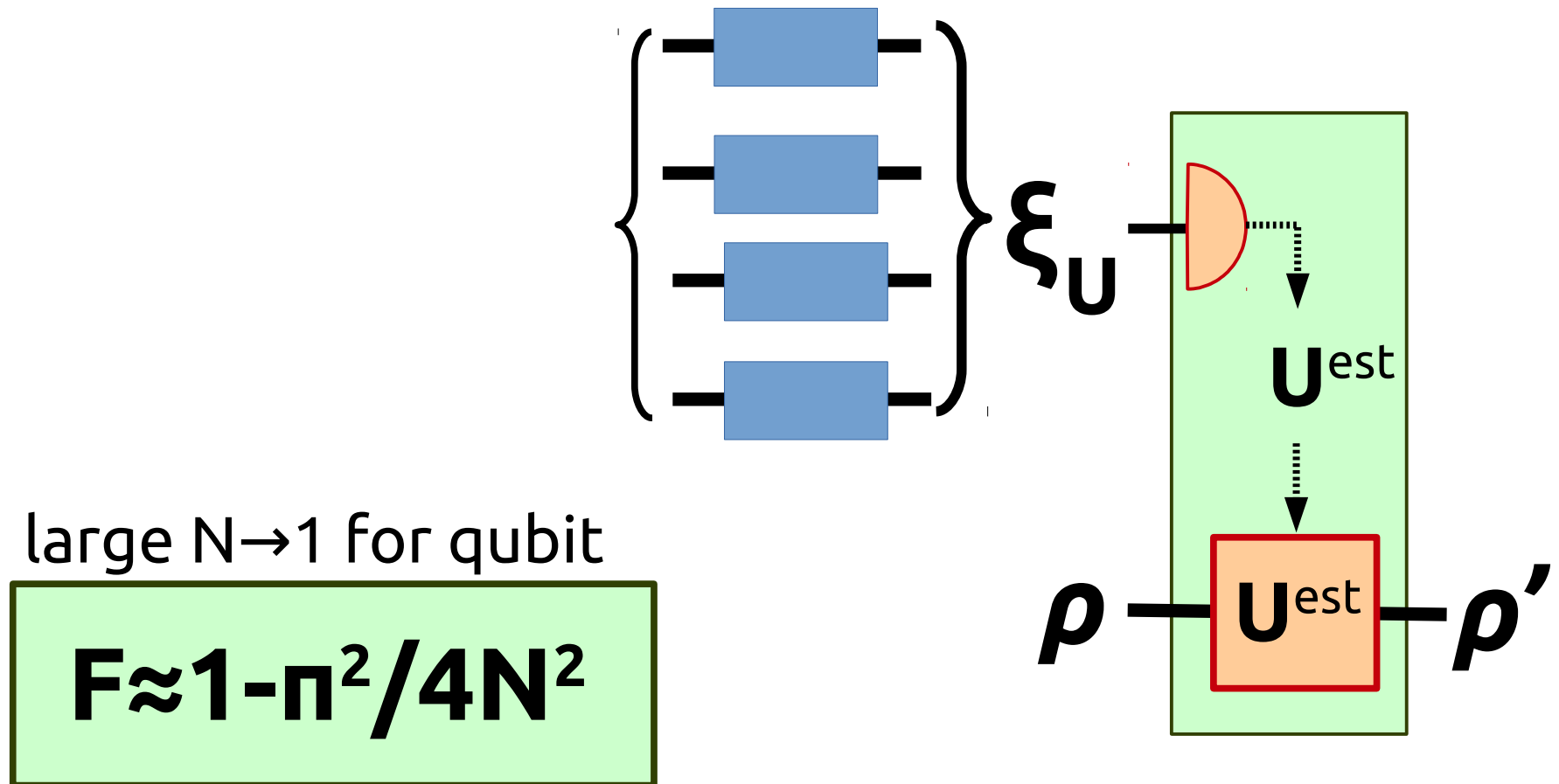
$$E_{U_{est}} = |\eta_{U_{est}}\rangle\langle\eta_{U_{est}}|$$

$$|\eta_{U_{est}}\rangle = \bigotimes_{j \in IRR} \sqrt{d_j} |U_j\rangle\rangle$$



APPROXIMATE Q LEARNING

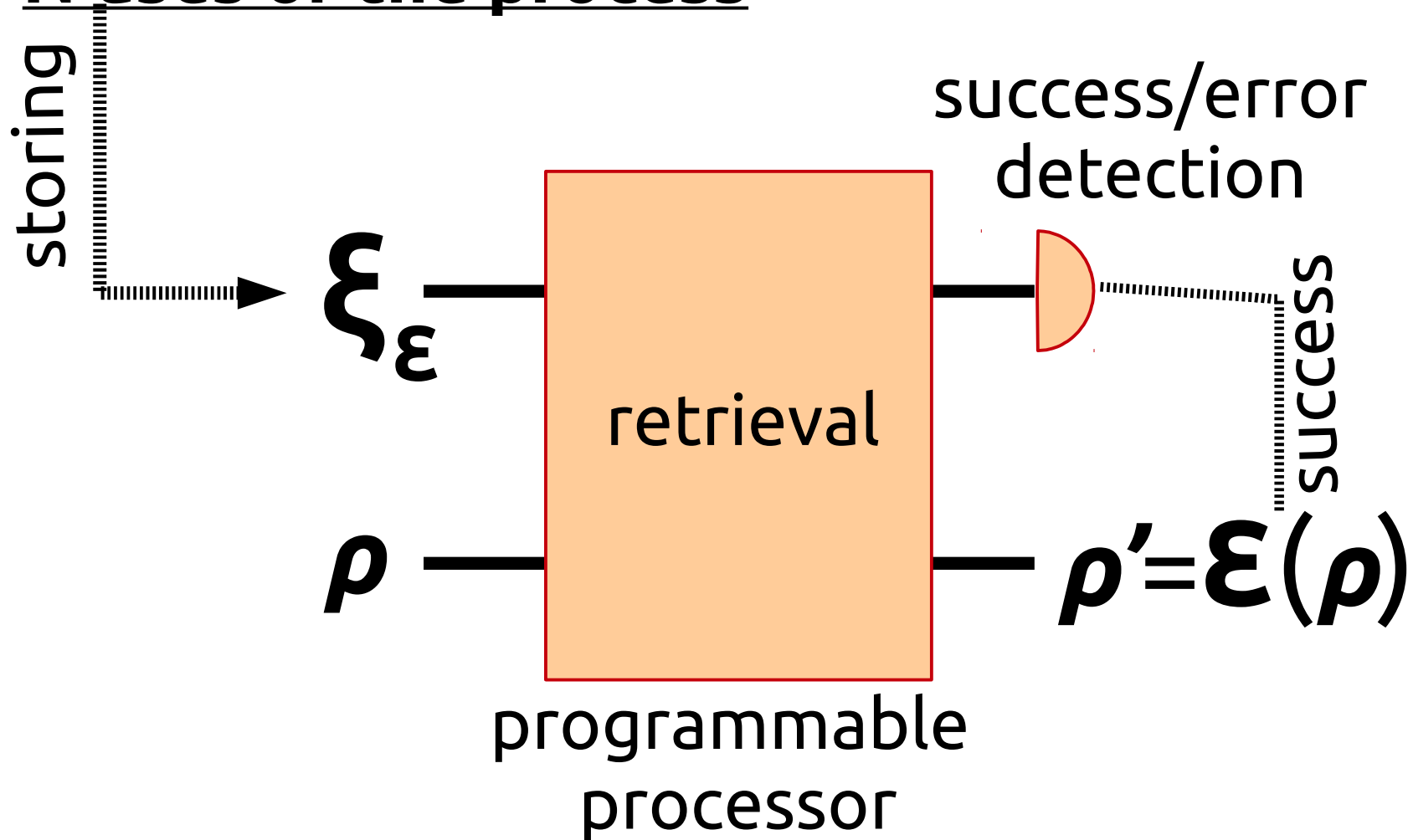
Optimal strategy for unitary channels



PROBABILISTIC Q LEARNING

→ PROBABILISTIC PERFORMANCE

→ N uses of the process

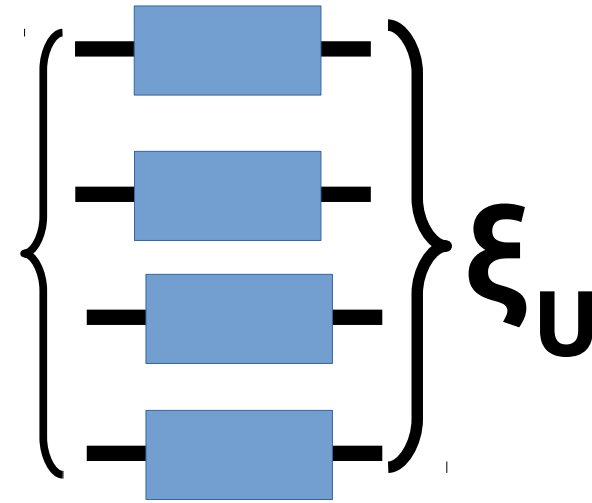


PROBABILISTIC Q LEARNING

Optimal strategy for unitary channels

→ optimal storing

$$|\Psi\rangle = \bigotimes_{j \in IRR} \sqrt{p_j/d_j} |I_j\rangle\rangle$$

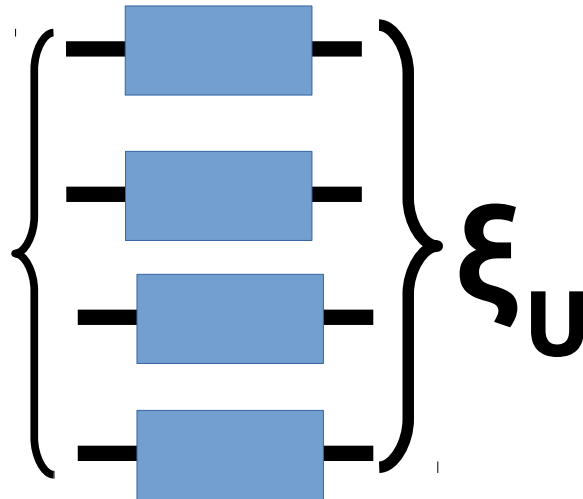


PROBABILISTIC Q LEARNING

Optimal strategy for unitary channels

→ optimal storing

$$|\Psi\rangle = \bigotimes_{j \in IRR} \sqrt{p_j/d_j} |I_j\rangle\rangle$$



→ optimal retrieval

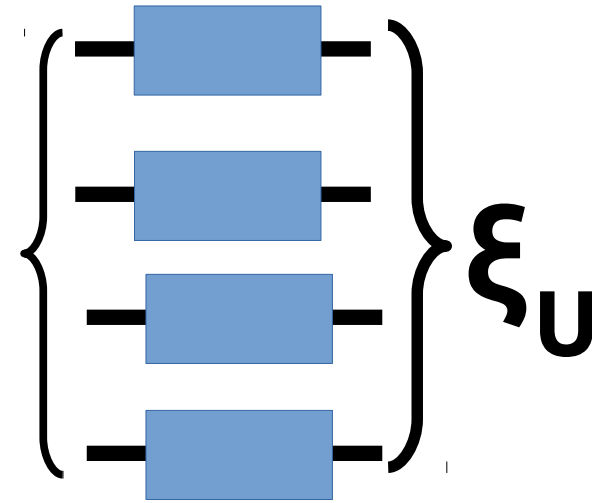
~~MEASURE-AND-ROTATE~~

PROBABILISTIC Q LEARNING

Optimal strategy for unitary channels

→ optimal storing

$$|\Psi\rangle = \bigotimes_{j \in IRR} \sqrt{p_j/d_j} |I_j\rangle\rangle$$



→ optimal retrieval

QUANTUM

PROBABILISTIC Q LEARNING

1 → 1 case

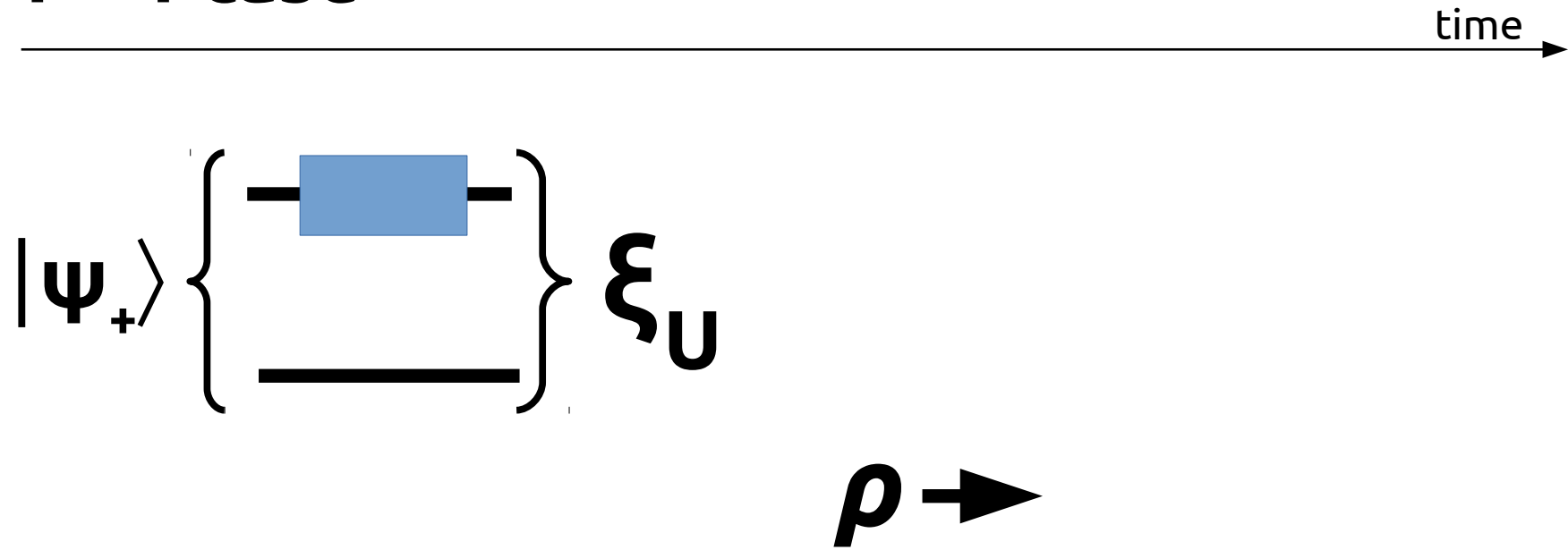
time



$\rho \rightarrow$

PROBABILISTIC Q LEARNING

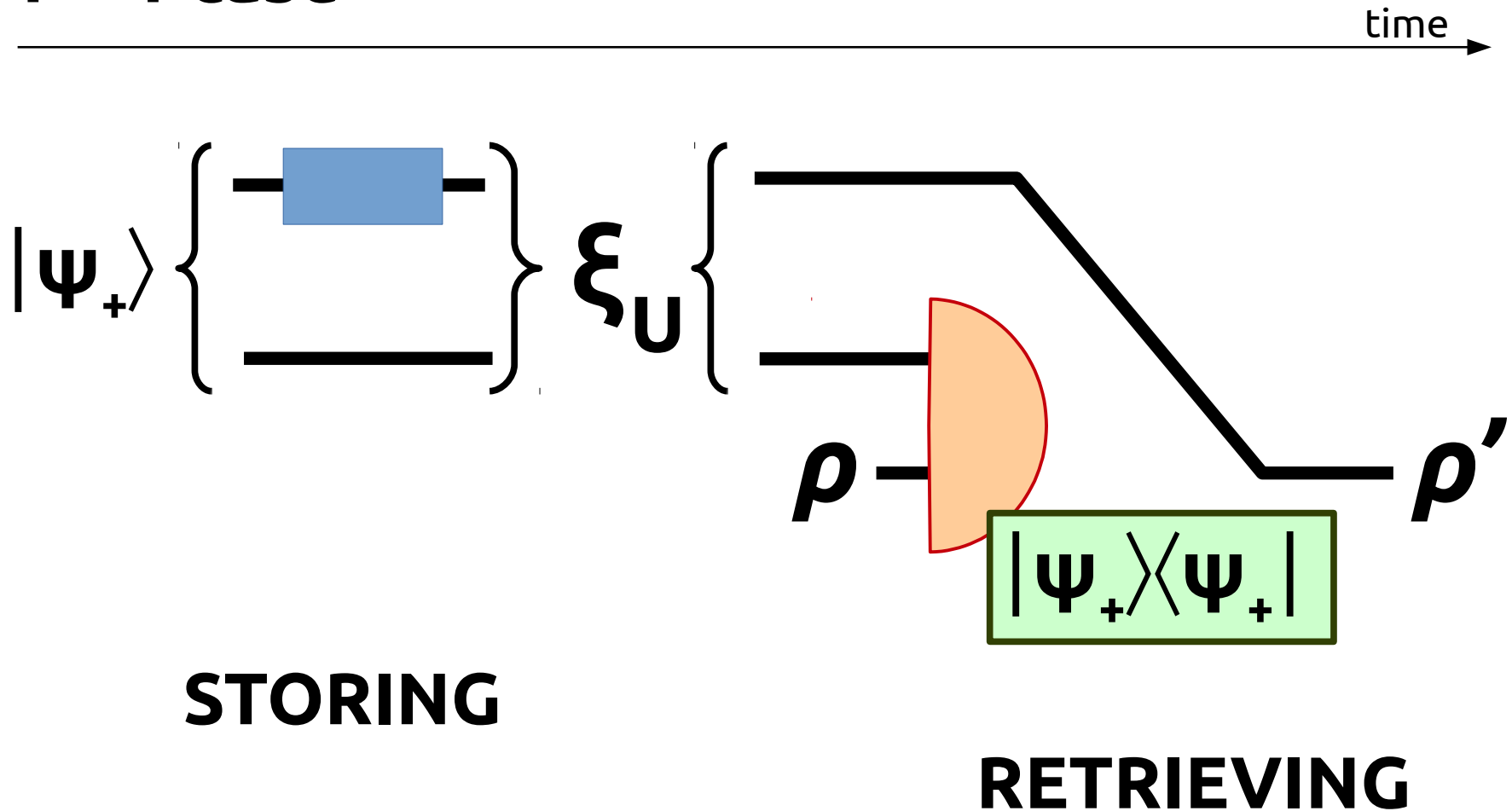
1 → 1 case



STORING

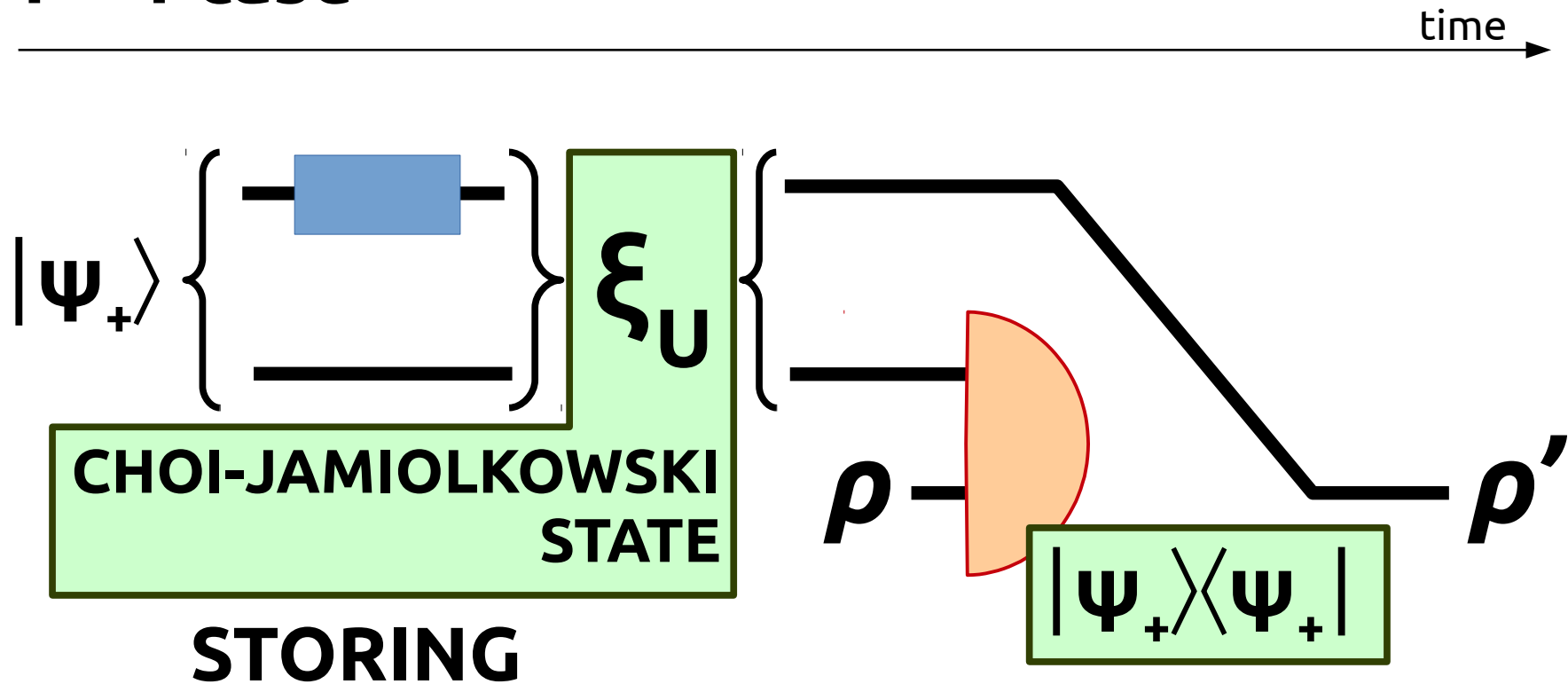
PROBABILISTIC Q LEARNING

1 → 1 case



PROBABILISTIC Q LEARNING

1 → 1 case



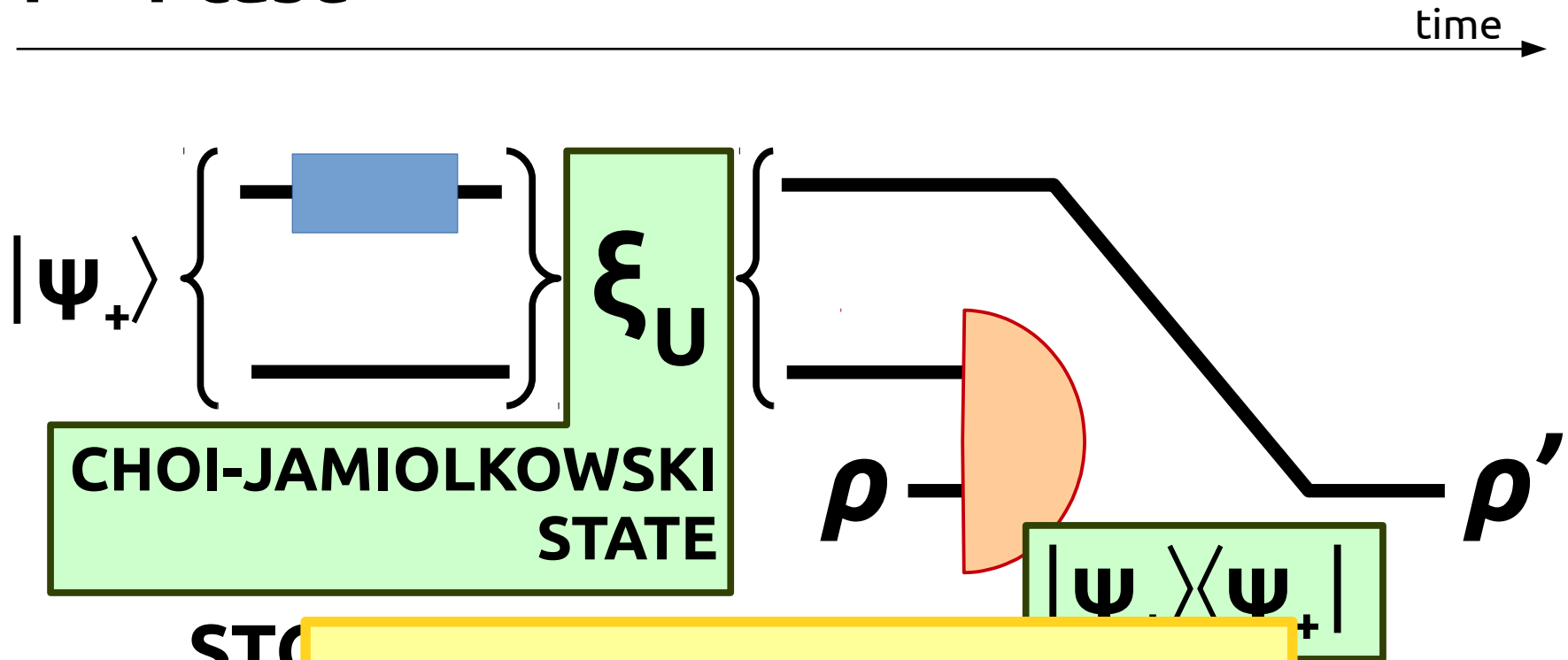
INCOMPLETE

RETRIEVING

TELEPORTATION

PROBABILISTIC Q LEARNING

1 → 1 case



STO

$P_{\text{success}} = 1/d^2$

INCO

VING

TELEPORTATION

PROBABILISTIC Q LEARNING

Optimal $N \rightarrow 1$ for unitary channels

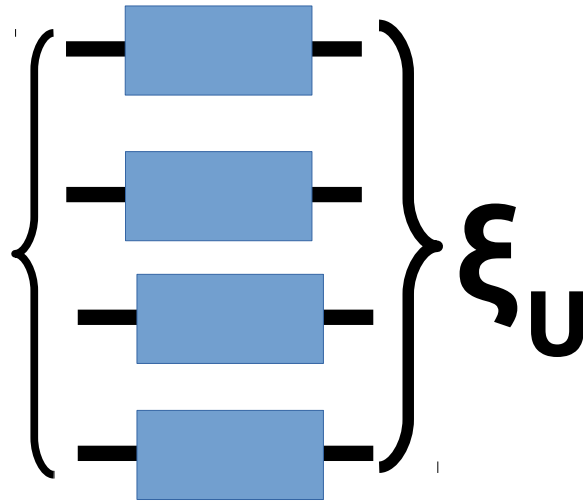
$$P_{\text{success}} = \frac{N}{N-1+d^2}$$

PROBABILISTIC Q LEARNING

Optimal $N \rightarrow 1$ for unitary channels

→ optimal storing

$$|\Psi\rangle = \bigotimes_{j \in IRR} \sqrt{p_j/d} |1_j\rangle\rangle$$



PROBABILISTIC Q LEARNING*

Optimal $N \rightarrow 1$ for unitary channels

- optimal retrieval (quantum comb formalism)
- reduction to linear programming problem

$$\begin{aligned} & \underset{\mu_J, p_j}{\text{maximize}} && \lambda = \sum_J d_J^3 \mu_J, \\ & \text{subject to} && 0 \leq d_J \mu_J \leq \frac{p_j}{d_j^2} \quad \forall j \in j_{JJ} \quad \forall J \\ & && p_j \geq 0 \quad \sum_j p_j = 1, \end{aligned}$$

- combinatorial identity for permutation group

$$\sum_j (c_j - r_j)^2 \frac{h_J}{h_j} = N - 1$$

- result is analytical

RELATED RESULTS

Retrieval of inverse of U (undo) for qubits

- the same success probability
- difference only in retrieval
- probabilistic alignment of reference frame

RELATED RESULTS

Trade-off for probabilistic processors

→ retrieval part provides best known bounds (exponentially better than before) on the memory size as a function of failure probability f

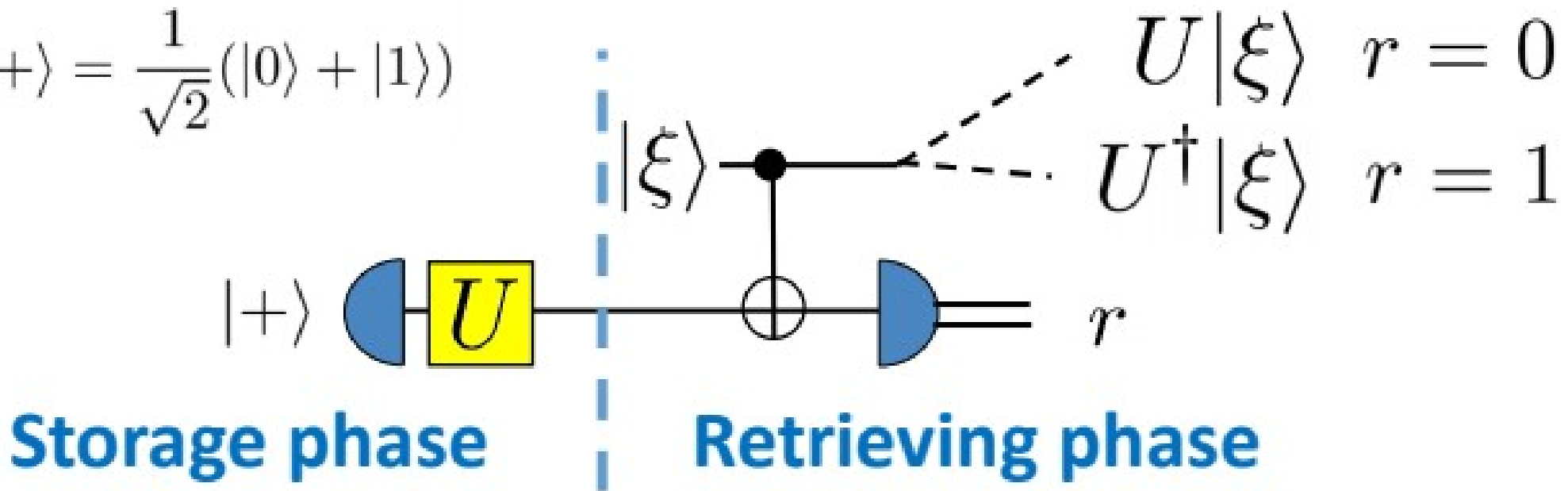
$$\dim H_{\text{mem}} = \sum_{j \in IRR} d_j^2 = \binom{N-1+d^2}{N}$$

$$\propto (1/f)^{(d^2-1)}$$

PROBABILISTIC LEARNING OF PHASE

$$U_\varphi = |0\rangle\langle 0| + e^{i\varphi}|1\rangle\langle 1|$$

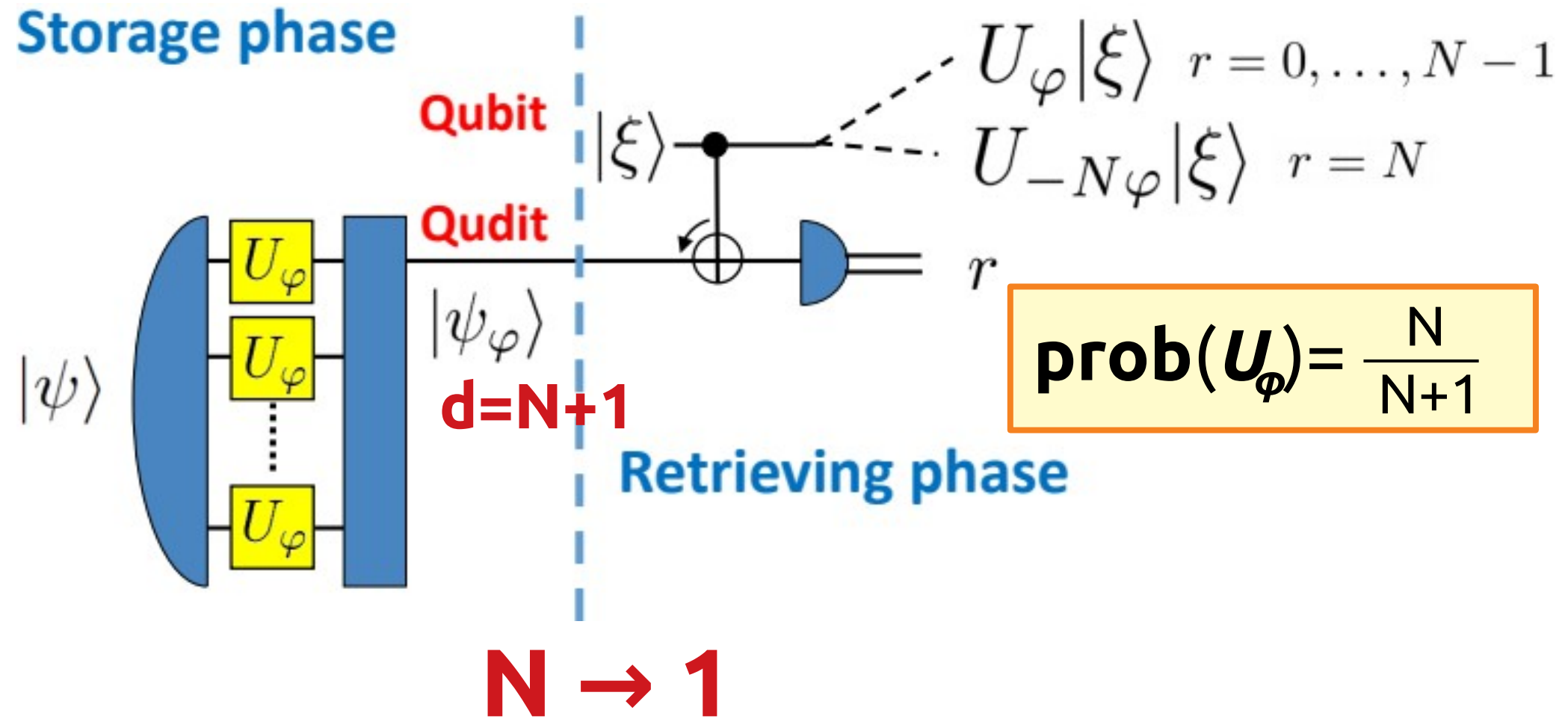
$$|+\rangle = \frac{1}{\sqrt{2}}(|0\rangle + |1\rangle)$$



$$P_{\text{success}} = 1/2$$

1 → 1

PROBABILISTIC LEARNING OF PHASE



PROBABILISTIC LEARNING OF PHASE

$$N = (2^k - 1) \rightarrow 1$$

$$\text{prob}(U_\phi) = 1 - \frac{1}{2^k}$$

PROBABILISTIC LEARNING OF PHASE

$$N = (2^k - 1) \rightarrow 1$$

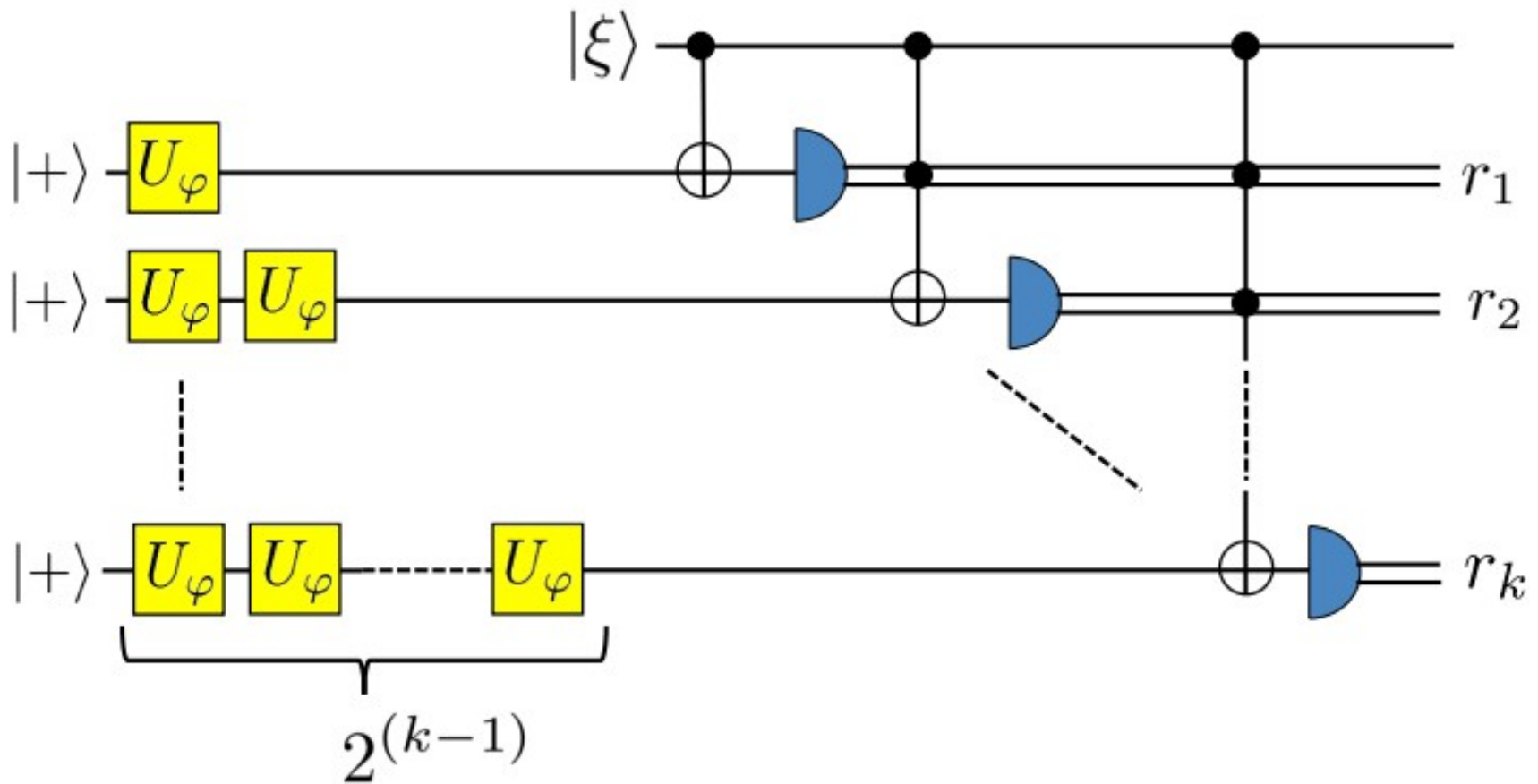
$$\text{prob}(U_\varphi) = 1 - \frac{1}{2^k}$$

same as Vidal-Masanes-Cirac

PROBABILISTIC LEARNING OF PHASE

$$N = (2^k - 1) \rightarrow 1$$

$$\text{prob}(U_\varphi) = 1 - \frac{1}{2^k}$$



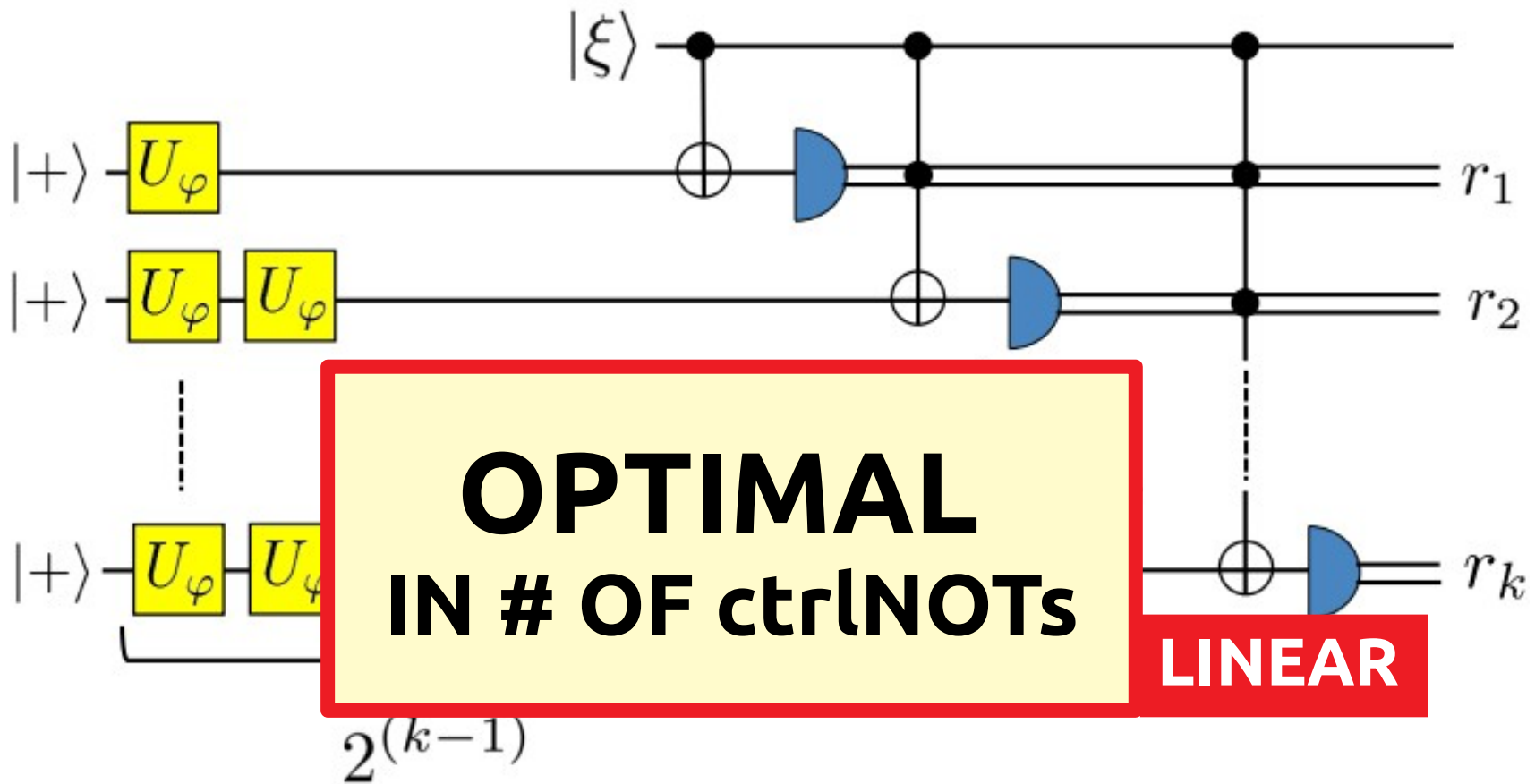
storing

Vidal-Masanes-Cirac

PROBABILISTIC LEARNING OF PHASE

$$N = (2^k - 1) \rightarrow 1$$

$$\text{prob}(U_\varphi) = 1 - \frac{1}{2^k}$$



storing

Vidal-Masanes-Cirac

SUMMARY

store sampling **N** times
retrieve probabilistically **1** use

$$P_{\text{success}}^{\text{universal}} = \frac{N}{N-1+d^2}$$

$$P_{\text{success}}^{\text{phase}}(U_{\varphi}) = \frac{N}{N+1}$$

Probabilistic learning is teleportation, hence, quantum.

joint work with
Michal Sedlák, Alessandro Bisio, Mário Ziman



Slovak Academy of Sciences, Bratislava, Slovakia
Masaryk University, Brno, Czech Republic
University of Pavia, Pavia, Italy

M. Sedlák, A. Bisio, M.Ziman: Optimal probabilistic storage and retrieval of unitary channels, Phys. Rev. Lett. 122, 170502 (2019)
S. Ramgoolam, M. Sedlák: Quantum Information Processing and Composite Quantum Fields, J. High Energ. Phys. 2019, 170 (2019)
M. Sedlák, A. Bisio: On some new hook-content identities, Journal of Algebraic Combinatorics, <https://doi.org/10.1007/s10801-019-00931-5>
M. Sedlák, M. Ziman: Probabilistic storage and retrieval of qubit phase gates, Physical Review A 102, 032618 (2020) [arXiv:2008.09555]

THANK YOU FOR YOUR ATTENTION