

# Tracking Quantum Circuits By Polynomials

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<sup>1</sup>Includes joint work with Amlan Chakrabarti, U. Calcutta, and prospectively students. . .

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- Central notion: **Quantum Circuits**.
- Hardly the only player—quantum adiabatic machines are closer to built and quantum communication systems are already deployed.

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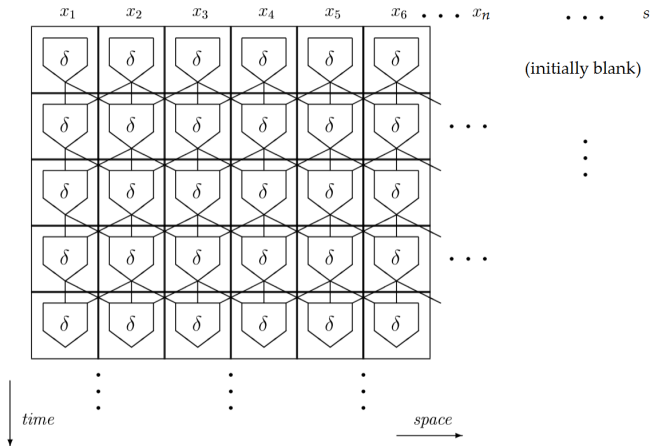
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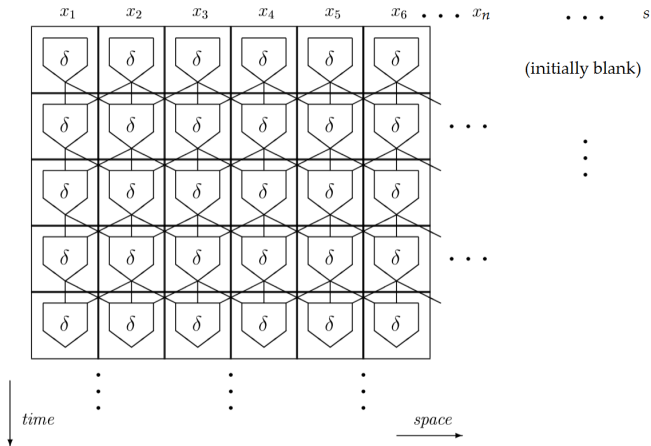
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- Bits have no common identity across wires, but we can create a universal layout for Boolean circuits in which they *do* retain identity...

## Turing “Cue Bits”



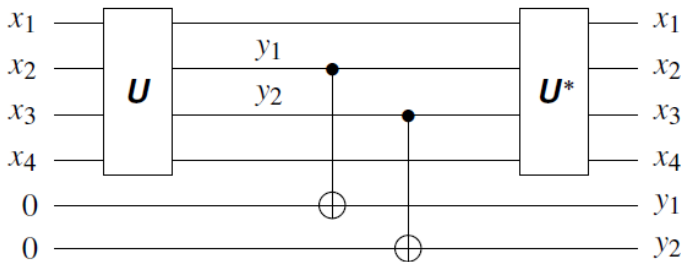
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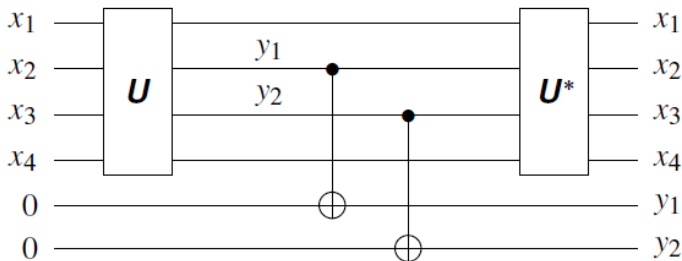


Space  $s$ , so  $n - s$  “ancillary” cells. Can also be made *reversible*.

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Must be reversible. In certain circumstances values  $y_1, y_2, \dots$  can be copied to extra lines as “ $f(x)$ .” Then reversing the gates re-creates the input  $x$  so the whole mapping is invertible (“copy-uncompute trick”).

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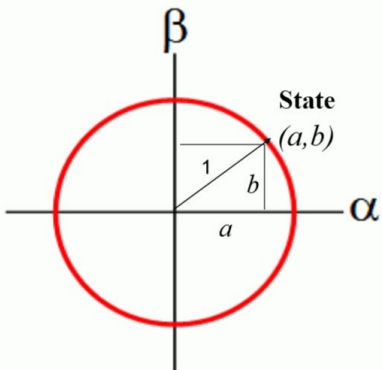
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- Under the hood are  $S = 2^s$  complex entries of a unit **state vector**.

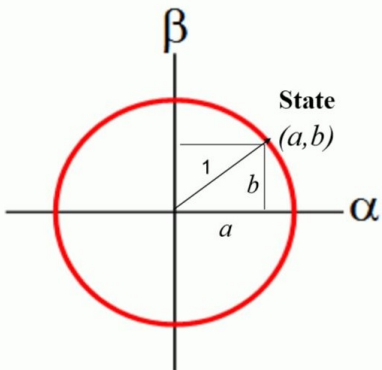
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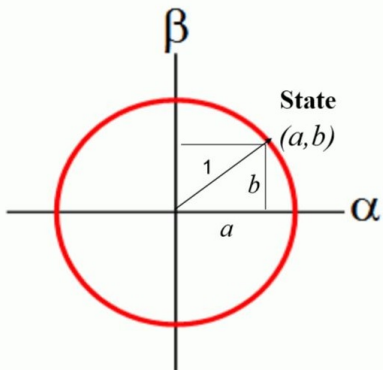
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- A *qutrit* yields 3 results and is  $(a, b, c)$  where  $|a|^2 + |b|^2 + |c|^2 = 1$ .

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- Only non-permutation gate needed for universality.
- But also common:  $Y = \begin{bmatrix} 0 & -i \\ i & 0 \end{bmatrix}$ ,  $Z = \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}$ ,  $S = \begin{bmatrix} 1 & 0 \\ 0 & i \end{bmatrix}$ .

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With  $k = 2$  qubits,  $K = 4$ . “Controlled Not” showing quantum coordinates:

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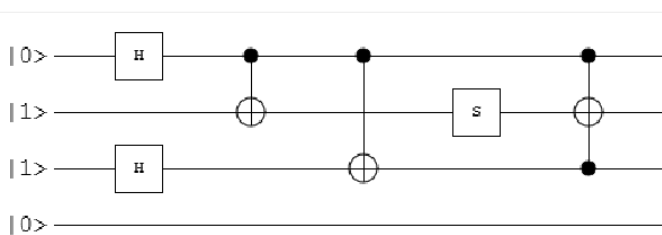
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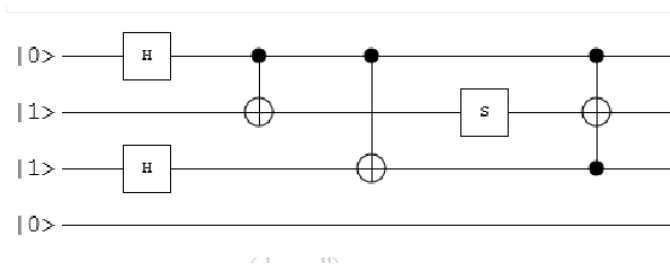
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- $\text{H} + \text{CNOT}$  is not quantum universal; it recognizes a proper subclass of P.



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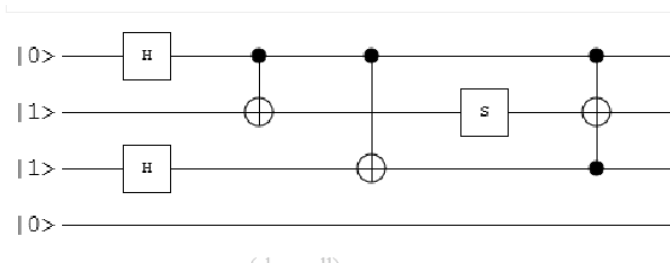


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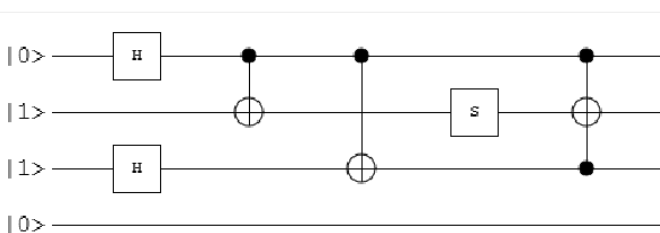
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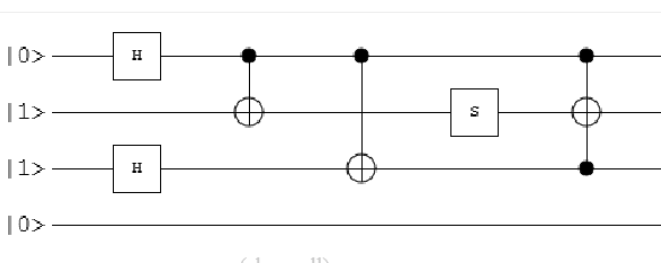
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- 4 After the  $S$  in stage 4, a TOF with controls on 1,3 and target on 2. The whole  $C$  computes a unitary  $U_C$ .

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# Input-Output and Measurement

- Input:  $E_x = e_{x_0} \otimes e_{x_1} \otimes \dots \otimes e_{x_n} \otimes e_0^{\otimes (s-n)}$ .
- Output: A state vector  $(z_0, \dots, z_{S-1})$ ,  $S = 2^s$ .
- Measure all lines: For any outcome  $b \in \{0, 1\}^s$ ,  
 $\Pr[C(x) \rightarrow b] = |z_b|^2 = |\langle E_x U_C e_b \rangle|^2$ .
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- Call this amplitude  $z_d$  as  $A[C(x) \mapsto d]$ .

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A language  $L$  belongs to BQP if there are poly-time uniform quantum circuits  $C_n$  for each  $n$  such that for all  $n$  and inputs  $x \in \{0, 1\}^n$ , designating qubit 1 for yes/no output:

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- Goal: calculate the *amplitude* of  $C_n(x) \mapsto 1$ , explicitly or implicitly.
- But no classical way known without paying  $\approx 2^n$  time overhead as for factoring.

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- But in many cases, **#SAT solvers** may be effective.
- Computation by solver more “offline” than managing  $2^n$ -sized arrays.
- $P_C$  may yield other algebraic and physical info about  $C$ , including about entanglement.

# The Basic Theorem

Theorem (Dawson et al.. 2004, implicitly before?)

*Given  $C$  built from TOF gates and  $h$ -many H gates, we can efficiently compute a polynomial  $P_C(x_1, \dots, x_n, y_1, \dots, y_h, z_1, \dots, z_r)$  and a constant  $R$  (here,  $R = \sqrt{2^h}$ ) such that for all  $x \in \{0, 1\}^n$  and  $z \in \{0, 1\}^r$ ,*

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- Hence cannot approximate the difference by approximating each term. Need exact solution counting.

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## How it Works: The “Quantum Grille”

Consider a general  $2 \times 2$  matrix  $A$ . Assign an indicator variable  $u$  to its input and  $y$  to its output:

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- Every qubit at every stage has a well-defined *local* “algebraic value.”



## How it Works: Two-Qubit Gate

Now let  $A$  be a general  $4 \times 4$  matrix. Assign indicator variables  $u_1, u_2$  to the two incoming qubits and  $y_1, y_2$  to their outgoing selves:

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	$(1 - y_1)(1 - y_2)$	$(1 - y_1)y_2$	$y_1(1 - y_2)$	$y_1y_2$
$(1 - u_1)(1 - u_2)$	$a_{11}$	$a_{12}$	$a_{13}$	$a_{14}$
$(1 - u_1)u_2$	$a_{21}$	$a_{22}$	$a_{23}$	$a_{24}$
$u_1(1 - u_2)$	$a_{31}$	$a_{32}$	$a_{33}$	$a_{34}$
$u_1u_2$	$a_{41}$	$a_{42}$	$a_{43}$	$a_{44}$

- Similar for  $8 \times 8$  etc. Initially there are a lot of terms.
- However, with substitution for permutation gates, the entire polynomial collapses to the constant 1!
- The effect on  $P_C$  is then how the substituted terms are input to subsequent Hadamard and other kinds of gates.

## How it Works: Qutrit Case

Let qutrit values be  $0, 1, -1$  indexed by the basis vectors  $e_0 = (1, 0, 0)^T$ ,  $e_1 = (0, 1, 0)^T$ , and  $e_2 = (0, 0, 1)^T$ .

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	$2 - 2y^2$	$y^2 + y$	$y^2 - y$
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Now  $P_A$  has 36 terms. But it simplifies for certain matrices:

$$A = \frac{1}{\sqrt{3}} \begin{bmatrix} 1 & 1 & 1 \\ 1 & \omega & \omega^2 \\ 1 & \omega^2 & \omega \end{bmatrix}; \quad P_A = 4 + 2\sqrt{3}iuy + 2u^2y^2.$$

Here  $\omega = \frac{1}{2}(-1 + \sqrt{3}i)$ , and the “ $H_3$  multiplier” is  $2/\sqrt{3}$ .

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- Whereas  $c = 0$  zeroes all such terms, so the only effect is to inflate  $R$ .

# Additive Extension

## Theorem

Given any  $C$  of minphase  $K$ , we can efficiently compute a polynomial  $Q_C(x_1, \dots, x_n, y_1, \dots, y_h, z_1, \dots, z_r, w_1, \dots, w_t)$  over  $\mathbb{Z}_K$  and a constant  $R$  such that for all  $x \in \{0, 1\}^n$  and  $z \in \{0, 1\}^r$ ,

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where  $Q_C = \sum_{\text{gates } g} q_g + \sum_{\text{constraints } c} q_c$  has bounded degree.

Thus we can do all calculations using  $Q_C$  over  $\mathbb{Z}_K$ .

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To enforce a desired output value  $z_i$  on qubit  $i$  with final term  $u_i$ :

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Open: replace  $K$  by  $\log K$  in the time? Affirmative for  $A[C(x) \mapsto z]$ .

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- Hence [CCLL] gives poly-time simulation by solution counting in  $\mathbb{Z}_4$ .

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