

What do quantum computers do?

Daniel J. Bernstein

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“Quantum algorithm”  
means an algorithm that  
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i.e. a sequence of instructions,  
where each instruction is  
in a quantum computer’s  
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**How do we know which  
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Quantum computer type 1 (QC1):  
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e.g.:  $(0, 0, 0)$ .

e.g.:  $(1, 1, 1)$ .

e.g.:  $(0, 1, 1)$ .

Data stored in 64 bits:

a list of 64 elements of  $\{0, 1\}$ .

e.g.:  $(1, 1, 1, 1, 1, 0, 0, 0, 1,$

$0, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0,$

$0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 1,$

$1, 0, 1, 0, 0, 0, 1, 0, 0, 0, 1,$

$0, 0, 1, 1, 1, 0, 0, 1, 0, 0, 0,$

$1, 1, 0, 1, 1, 0, 0, 1, 0, 0, 1)$ .

The state of a quantum computer

Data stored in 3 qubits:

a list of 8 numbers, not all zero.

e.g.:  $(3, 1, 4, 1, 5, 9, 2, 6)$ .

e.g.:  $(-2, 7, -1, 8, 1, -8, -2, 8)$ .

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Data stored in 4 qubits: a list of  
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Data stored in 64 qubits:

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1, 0, 1, 0, 0, 0, 1, 0, 0, 0, 1,

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Data stored in 64 qubits:

a list of  $2^{64}$  numbers, not all zero.

Data stored in 1000 qubits: a list  
of  $2^{1000}$  numbers, not all zero.

## The state of a computer

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(0, 0).

(1, 1).

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stored in 64 bits:

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, 0, 0, 1, 0, 0, 0, 1,

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6

## The state of a quantum computer

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Data stored in 1000 qubits: a list

of  $2^{1000}$  numbers, not all zero.

7

## Measuring

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## Computer

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bits:  
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0, 0, 0, 1,  
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6

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7

## Measuring a quantum

Can simply look at  
Cannot simply look  
of numbers stored

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of numbers stored in  $n$  qubit

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### **Measuring $n$ qubits**

- produces  $n$  bits and
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If  $n$  qubits have state

$(a_0, a_1, \dots, a_{2^n-1})$  then

measurement produces  $q$

with probability  $|a_q|^2 / \sum_r |a_r|^2$ .

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State is then all zeros except 1 at position  $q$ .



State of a quantum computer

Stored in 3 qubits:

8 numbers, not all zero.

(1, 4, 1, 5, 9, 2, 6).

(2, 7, -1, 8, 1, -8, -2, 8).

(0, 0, 0, 0, 1, 0, 0).

Stored in 4 qubits: a list of

numbers, not all zero. e.g.:

(1, 5, 9, 2, 6, 5, 3, 5, 8, 9, 7, 9, 3).

Stored in 64 qubits:

$2^{64}$  numbers, not all zero.

Stored in 1000 qubits: a list

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(1, 1, 1, 1)

Quantum computer

qubits:

numbers, not all zero.

(0, 2, 6).

(1, -8, -2, 8).

(1, 0, 0).

qubits: a list of

numbers, not all zero. e.g.:

(5, 3, 5, 8, 9, 7, 9, 3).

qubits:

numbers, not all zero.

100 qubits: a list

of numbers, not all zero.

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e.g.: Say 3 qubits

(1, 1, 1, 1, 1, 1, 1, 1)

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zero.

(2, 8).

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(9, 7, 9, 3).

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e.g.: Say 3 qubits have state  $(1, 1, 1, 1, 1, 1, 1, 1)$ .

Measurement produces

000 = 0 with probability 1/8;

001 = 1 with probability 1/8;

010 = 2 with probability 1/8;

011 = 3 with probability 1/8;

100 = 4 with probability 1/8;

101 = 5 with probability 1/8;

110 = 6 with probability 1/8;

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“Quantum RNG.”

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“Quantum RNG.”

Warning: Quantum RNGs sold today are measurably biased.

Using a quantum computer

Simply look at a bit.

Simply look at the list

of numbers stored in  $n$  qubits.

Using  $n$  qubits

produces  $n$  bits and

measures the state.

If qubits have state

$(a_0, \dots, a_{2^n-1})$  then

measurement produces  $q$

with probability  $|a_q|^2 / \sum_r |a_r|^2$ .

If you get

all zeros at position  $q$ .

8

e.g.: Say 3 qubits have state  
(1, 1, 1, 1, 1, 1, 1, 1).

Measurement produces

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9

e.g.: Say  
(3, 1, 4, 1, 1, 1, 1, 1)



Quantum computer

is not a bit.

Look at the list

of states in  $n$  qubits.

bits

and

state.

state

then

produces  $q$

$$|a_q|^2 / \sum_r |a_r|^2.$$

eros

on  $q$ .

e.g.: Say 3 qubits have state  
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Measurement produces

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e.g.: Say 3 qubits  
(3, 1, 4, 1, 5, 9, 2, 6)

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e.g.: Say 3 qubits have state  
(1, 1, 1, 1, 1, 1, 1, 1).

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ts.

Measurement produces

000 = 0 with probability  $1/8$ ;

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$|r|^2$ .

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e.g.: Say 3 qubits have state  
(3, 1, 4, 1, 5, 9, 2, 6).

Measurement produces

000 = 0 with probability  $9/173$ ;

001 = 1 with probability  $1/173$ ;

010 = 2 with probability  $16/173$ ;

011 = 3 with probability  $1/173$ ;

100 = 4 with probability  $25/173$ ;

101 = 5 with probability  $81/173$ ;

110 = 6 with probability  $4/173$ ;

111 = 7 with probability  $36/173$ .

e.g.: Say 3 qubits have state  
(1, 1, 1, 1, 1, 1, 1, 1).

Measurement produces

000 = 0 with probability  $1/8$ ;

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5 is most likely outcome.

y 3 qubits have state  
(1, 1, 1, 1, 1).

ment produces

with probability  $1/8$ ;

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e.g.: Say  
(0, 0, 0, 0)

have state  
( $\dots$ ).

duces

probability  $1/8$ ;

probability  $1/8$ ;

probability  $1/8$ ;

probability  $1/8$ ;

probability  $1/8$ ;

probability  $1/8$ ;

probability  $1/8$ ;

probability  $1/8$ .

m RNGs sold

bly biased.

e.g.: Say 3 qubits have state  
(3, 1, 4, 1, 5, 9, 2, 6).

Measurement produces

000 = 0 with probability  $9/173$ ;

001 = 1 with probability  $1/173$ ;

010 = 2 with probability  $16/173$ ;

011 = 3 with probability  $1/173$ ;

100 = 4 with probability  $25/173$ ;

101 = 5 with probability  $81/173$ ;

110 = 6 with probability  $4/173$ ;

111 = 7 with probability  $36/173$ .

5 is most likely outcome.

e.g.: Say 3 qubits  
(0, 0, 0, 0, 0, 1, 0, 0

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 010 = 2 with probability  $16/173$ ;  
 011 = 3 with probability  $1/173$ ;  
 100 = 4 with probability  $25/173$ ;  
 101 = 5 with probability  $81/173$ ;  
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 011 = 3 with probability 0;  
 100 = 4 with probability 0;  
 101 = 5 with probability 1;  
 110 = 6 with probability 0;  
 111 = 7 with probability 0.

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 110 = 6 with probability  $4/173$ ;  
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(0, 0, 0, 0, 0, 1, 0, 0).

Measurement produces

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 001 = 1 with probability 0;  
 010 = 2 with probability 0;  
 011 = 3 with probability 0;  
 100 = 4 with probability 0;  
 101 = 5 with probability 1;  
 110 = 6 with probability 0;  
 111 = 7 with probability 0.

5 is guaranteed outcome.

y 3 qubits have state  
(1, 5, 9, 2, 6).

Measurement produces

- with probability  $9/173$ ;
- with probability  $1/173$ ;
- with probability  $16/173$ ;
- with probability  $1/173$ ;
- with probability  $25/173$ ;
- with probability  $81/173$ ;
- with probability  $4/173$ ;
- with probability  $36/173$ .

most likely outcome.

e.g.: Say 3 qubits have state  
(0, 0, 0, 0, 0, 1, 0, 0).

Measurement produces

- 000 = 0 with probability 0;
- 001 = 1 with probability 0;
- 010 = 2 with probability 0;
- 011 = 3 with probability 0;
- 100 = 4 with probability 0;
- 101 = 5 with probability 1;
- 110 = 6 with probability 0;
- 111 = 7 with probability 0.

5 is guaranteed outcome.

NOT gate

NOT<sub>0</sub> gate

(3, 1, 4, 1)

(1, 3, 1, 4)

have state  
)

duces

probability  $9/173$ ;

probability  $1/173$ ;

probability  $16/173$ ;

probability  $1/173$ ;

probability  $25/173$ ;

probability  $81/173$ ;

probability  $4/173$ ;

probability  $36/173$ .

outcome.

e.g.: Say 3 qubits have state  
 $(0, 0, 0, 0, 0, 1, 0, 0)$ .

Measurement produces

$000 = 0$  with probability 0;

$001 = 1$  with probability 0;

$010 = 2$  with probability 0;

$011 = 3$  with probability 0;

$100 = 4$  with probability 0;

$101 = 5$  with probability 1;

$110 = 6$  with probability 0;

$111 = 7$  with probability 0.

5 is guaranteed outcome.

## NOT gates

$\text{NOT}_0$  gate on 3 qubits

$(3, 1, 4, 1, 5, 9, 2, 6)$

$(1, 3, 1, 4, 9, 5, 6, 2)$

e.g.: Say 3 qubits have state  
 $(0, 0, 0, 0, 0, 1, 0, 0)$ .

Measurement produces

000 = 0 with probability 0;

001 = 1 with probability 0;

010 = 2 with probability 0;

011 = 3 with probability 0;

100 = 4 with probability 0;

101 = 5 with probability 1;

110 = 6 with probability 0;

111 = 7 with probability 0.

5 is guaranteed outcome.

## NOT gates

NOT<sub>0</sub> gate on 3 qubits:

$(3, 1, 4, 1, 5, 9, 2, 6) \mapsto$

$(1, 3, 1, 4, 9, 5, 6, 2)$ .

e.g.: Say 3 qubits have state  
 $(0, 0, 0, 0, 0, 1, 0, 0)$ .

Measurement produces

$000 = 0$  with probability 0;

$001 = 1$  with probability 0;

$010 = 2$  with probability 0;

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5 is guaranteed outcome.

## NOT gates

NOT<sub>0</sub> gate on 3 qubits:

$(3, 1, 4, 1, 5, 9, 2, 6) \mapsto$

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e.g.: Say 3 qubits have state  
 $(0, 0, 0, 0, 0, 1, 0, 0)$ .

Measurement produces

$000 = 0$  with probability 0;

$001 = 1$  with probability 0;

$010 = 2$  with probability 0;

$011 = 3$  with probability 0;

$100 = 4$  with probability 0;

$101 = 5$  with probability 1;

$110 = 6$  with probability 0;

$111 = 7$  with probability 0.

5 is guaranteed outcome.

## NOT gates

NOT<sub>0</sub> gate on 3 qubits:

$(3, 1, 4, 1, 5, 9, 2, 6) \mapsto$

$(1, 3, 1, 4, 9, 5, 6, 2)$ .

NOT<sub>0</sub> gate on 4 qubits:

$(3, 1, 4, 1, 5, 9, 2, 6, 5, 3, 5, 8, 9, 7, 9, 3) \mapsto$

$(1, 3, 1, 4, 9, 5, 6, 2, 3, 5, 8, 5, 7, 9, 3, 9)$ .



e.g.: Say 3 qubits have state  
 $(0, 0, 0, 0, 0, 1, 0, 0)$ .

Measurement produces

$000 = 0$  with probability 0;

$001 = 1$  with probability 0;

$010 = 2$  with probability 0;

$011 = 3$  with probability 0;

$100 = 4$  with probability 0;

$101 = 5$  with probability 1;

$110 = 6$  with probability 0;

$111 = 7$  with probability 0.

5 is guaranteed outcome.

## NOT gates

NOT<sub>0</sub> gate on 3 qubits:

$(3, 1, 4, 1, 5, 9, 2, 6) \mapsto$

$(1, 3, 1, 4, 9, 5, 6, 2)$ .

NOT<sub>0</sub> gate on 4 qubits:

$(3, 1, 4, 1, 5, 9, 2, 6, 5, 3, 5, 8, 9, 7, 9, 3) \mapsto$

$(1, 3, 1, 4, 9, 5, 6, 2, 3, 5, 8, 5, 7, 9, 3, 9)$ .

NOT<sub>1</sub> gate on 3 qubits:

$(3, 1, 4, 1, 5, 9, 2, 6) \mapsto$

$(4, 1, 3, 1, 2, 6, 5, 9)$ .

e.g.: Say 3 qubits have state  
 $(0, 0, 0, 0, 0, 1, 0, 0)$ .

Measurement produces

$000 = 0$  with probability 0;

$001 = 1$  with probability 0;

$010 = 2$  with probability 0;

$011 = 3$  with probability 0;

$100 = 4$  with probability 0;

$101 = 5$  with probability 1;

$110 = 6$  with probability 0;

$111 = 7$  with probability 0.

5 is guaranteed outcome.

## NOT gates

NOT<sub>0</sub> gate on 3 qubits:

$(3, 1, 4, 1, 5, 9, 2, 6) \mapsto$

$(1, 3, 1, 4, 9, 5, 6, 2)$ .

NOT<sub>0</sub> gate on 4 qubits:

$(3, 1, 4, 1, 5, 9, 2, 6, 5, 3, 5, 8, 9, 7, 9, 3) \mapsto$

$(1, 3, 1, 4, 9, 5, 6, 2, 3, 5, 8, 5, 7, 9, 3, 9)$ .

NOT<sub>1</sub> gate on 3 qubits:

$(3, 1, 4, 1, 5, 9, 2, 6) \mapsto$

$(4, 1, 3, 1, 2, 6, 5, 9)$ .

NOT<sub>2</sub> gate on 3 qubits:

$(3, 1, 4, 1, 5, 9, 2, 6) \mapsto$

$(5, 9, 2, 6, 3, 1, 4, 1)$ .

y 3 qubits have state  
(0, 0, 1, 0, 0).

ment produces

with probability 0;

with probability 0;

with probability 0;

with probability 0;

with probability 0;

with probability 1;

with probability 0;

with probability 0.

ranted outcome.

## NOT gates

NOT<sub>0</sub> gate on 3 qubits:

(3, 1, 4, 1, 5, 9, 2, 6)  $\mapsto$

(1, 3, 1, 4, 9, 5, 6, 2).

NOT<sub>0</sub> gate on 4 qubits:

(3, 1, 4, 1, 5, 9, 2, 6, 5, 3, 5, 8, 9, 7, 9, 3)  $\mapsto$

(1, 3, 1, 4, 9, 5, 6, 2, 3, 5, 8, 5, 7, 9, 3, 9).

NOT<sub>1</sub> gate on 3 qubits:

(3, 1, 4, 1, 5, 9, 2, 6)  $\mapsto$

(4, 1, 3, 1, 2, 6, 5, 9).

NOT<sub>2</sub> gate on 3 qubits:

(3, 1, 4, 1, 5, 9, 2, 6)  $\mapsto$

(5, 9, 2, 6, 3, 1, 4, 1).

(1, 0, 0,

(0, 1, 0,

(0, 0, 1,

(0, 0, 0,

(0, 0, 0,

(0, 0, 0,

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NOT gates

NOT<sub>0</sub> gate on 3 qubits:

$(3, 1, 4, 1, 5, 9, 2, 6) \mapsto$

$(1, 3, 1, 4, 9, 5, 6, 2)$ .

NOT<sub>0</sub> gate on 4 qubits:

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$(1, 3, 1, 4, 9, 5, 6, 2, 3, 5, 8, 5, 7, 9, 3, 9)$ .

NOT<sub>1</sub> gate on 3 qubits:

$(3, 1, 4, 1, 5, 9, 2, 6) \mapsto$

$(4, 1, 3, 1, 2, 6, 5, 9)$ .

NOT<sub>2</sub> gate on 3 qubits:

$(3, 1, 4, 1, 5, 9, 2, 6) \mapsto$

$(5, 9, 2, 6, 3, 1, 4, 1)$ .

state

$(1, 0, 0, 0, 0, 0, 0, 0)$

$(0, 1, 0, 0, 0, 0, 0, 0)$

$(0, 0, 1, 0, 0, 0, 0, 0)$

$(0, 0, 0, 1, 0, 0, 0, 0)$

$(0, 0, 0, 0, 1, 0, 0, 0)$

$(0, 0, 0, 0, 0, 1, 0, 0)$

$(0, 0, 0, 0, 0, 0, 1, 0)$

$(0, 0, 0, 0, 0, 0, 0, 1)$

Operation on quantum state

NOT<sub>0</sub>, swapping pairs

Operation after measurement

flipping bit 0 of register

Flip: output is not

NOT gates

NOT<sub>0</sub> gate on 3 qubits:

$(3, 1, 4, 1, 5, 9, 2, 6) \mapsto$

$(1, 3, 1, 4, 9, 5, 6, 2)$ .

NOT<sub>0</sub> gate on 4 qubits:

$(3, 1, 4, 1, 5, 9, 2, 6, 5, 3, 5, 8, 9, 7, 9, 3) \mapsto$

$(1, 3, 1, 4, 9, 5, 6, 2, 3, 5, 8, 5, 7, 9, 3, 9)$ .

NOT<sub>1</sub> gate on 3 qubits:

$(3, 1, 4, 1, 5, 9, 2, 6) \mapsto$

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NOT<sub>2</sub> gate on 3 qubits:

$(3, 1, 4, 1, 5, 9, 2, 6) \mapsto$

$(5, 9, 2, 6, 3, 1, 4, 1)$ .

state	measure
$(1, 0, 0, 0, 0, 0, 0, 0)$	000
$(0, 1, 0, 0, 0, 0, 0, 0)$	001
$(0, 0, 1, 0, 0, 0, 0, 0)$	010
$(0, 0, 0, 1, 0, 0, 0, 0)$	011
$(0, 0, 0, 0, 1, 0, 0, 0)$	100
$(0, 0, 0, 0, 0, 1, 0, 0)$	101
$(0, 0, 0, 0, 0, 0, 1, 0)$	110
$(0, 0, 0, 0, 0, 0, 0, 1)$	111

Operation on quantum state

NOT<sub>0</sub>, swapping pairs.

Operation after measurement

flipping bit 0 of result.

Flip: output is not input.

NOT gates

NOT<sub>0</sub> gate on 3 qubits:

(3, 1, 4, 1, 5, 9, 2, 6)  $\mapsto$

(1, 3, 1, 4, 9, 5, 6, 2).

NOT<sub>0</sub> gate on 4 qubits:

(3, 1, 4, 1, 5, 9, 2, 6, 5, 3, 5, 8, 9, 7, 9, 3)  $\mapsto$

(1, 3, 1, 4, 9, 5, 6, 2, 3, 5, 8, 5, 7, 9, 3, 9).

NOT<sub>1</sub> gate on 3 qubits:

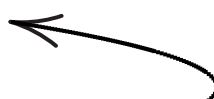

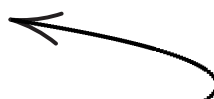

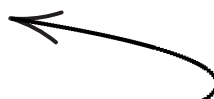

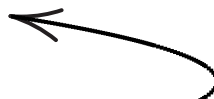

(3, 1, 4, 1, 5, 9, 2, 6)  $\mapsto$

(4, 1, 3, 1, 2, 6, 5, 9).

NOT<sub>2</sub> gate on 3 qubits:

(3, 1, 4, 1, 5, 9, 2, 6)  $\mapsto$

(5, 9, 2, 6, 3, 1, 4, 1).

state	measurement
(1, 0, 0, 0, 0, 0, 0, 0)	000 
(0, 1, 0, 0, 0, 0, 0, 0)	001 
(0, 0, 1, 0, 0, 0, 0, 0)	010 
(0, 0, 0, 1, 0, 0, 0, 0)	011 
(0, 0, 0, 0, 1, 0, 0, 0)	100 
(0, 0, 0, 0, 0, 1, 0, 0)	101 
(0, 0, 0, 0, 0, 0, 1, 0)	110 
(0, 0, 0, 0, 0, 0, 0, 1)	111 

Operation on quantum state:

NOT<sub>0</sub>, swapping pairs.

Operation after measurement:

flipping bit 0 of result.

Flip: output is not input.

tes

ate on 3 qubits:

(1, 5, 9, 2, 6)  $\mapsto$

(4, 9, 5, 6, 2).

ate on 4 qubits:

(5, 9, 2, 6, 5, 3, 5, 8, 9, 7, 9, 3)  $\mapsto$

(9, 5, 6, 2, 3, 5, 8, 5, 7, 9, 3, 9).

ate on 3 qubits:

(1, 5, 9, 2, 6)  $\mapsto$

(1, 2, 6, 5, 9).

ate on 3 qubits:

(1, 5, 9, 2, 6)  $\mapsto$

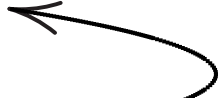

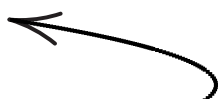


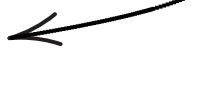
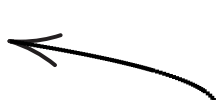
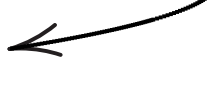
(5, 3, 1, 4, 1).

Control

e.g.  $C_1M$

(3, 1, 4, 1)

(3, 1, 1, 4)

state	measurement
(1, 0, 0, 0, 0, 0, 0, 0)	000 
(0, 1, 0, 0, 0, 0, 0, 0)	001 
(0, 0, 1, 0, 0, 0, 0, 0)	010 
(0, 0, 0, 1, 0, 0, 0, 0)	011 
(0, 0, 0, 0, 1, 0, 0, 0)	100 
(0, 0, 0, 0, 0, 1, 0, 0)	101 
(0, 0, 0, 0, 0, 0, 1, 0)	110 
(0, 0, 0, 0, 0, 0, 0, 1)	111 

Operation on quantum state:

$NOT_0$ , swapping pairs.

Operation after measurement:

flipping bit 0 of result.

Flip: output is not input.

qubits:

)  $\mapsto$

).

qubits:

(3,5,8,9,7,9,3)  $\mapsto$

(5,8,5,7,9,3,9).

qubits:

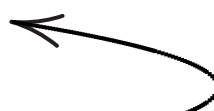






)  $\mapsto$

).

qubits:

)  $\mapsto$

).

state	measurement
(1, 0, 0, 0, 0, 0, 0, 0)	000 
(0, 1, 0, 0, 0, 0, 0, 0)	001 
(0, 0, 1, 0, 0, 0, 0, 0)	010 
(0, 0, 0, 1, 0, 0, 0, 0)	011 
(0, 0, 0, 0, 1, 0, 0, 0)	100 
(0, 0, 0, 0, 0, 1, 0, 0)	101 
(0, 0, 0, 0, 0, 0, 1, 0)	110 
(0, 0, 0, 0, 0, 0, 0, 1)	111

Operation on quantum state:

$\text{NOT}_0$ , swapping pairs.

Operation after measurement:

flipping bit 0 of result.

Flip: output is not input.

## Controlled-NOT (CNOT)

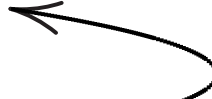





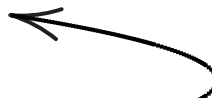

e.g.  $C_1\text{NOT}_0$ :

(3, 1, 4, 1, 5, 9, 2, 6)

(3, 1, 1, 4, 5, 9, 6, 2)



(,3)  $\mapsto$   
(,9).

state	measurement
(1, 0, 0, 0, 0, 0, 0, 0)	000 
(0, 1, 0, 0, 0, 0, 0, 0)	001 
(0, 0, 1, 0, 0, 0, 0, 0)	010 
(0, 0, 0, 1, 0, 0, 0, 0)	011 
(0, 0, 0, 0, 1, 0, 0, 0)	100 
(0, 0, 0, 0, 0, 1, 0, 0)	101 
(0, 0, 0, 0, 0, 0, 1, 0)	110 
(0, 0, 0, 0, 0, 0, 0, 1)	111 

Operation on quantum state:

$\text{NOT}_0$ , swapping pairs.

Operation after measurement:

flipping bit 0 of result.

Flip: output is not input.

## Controlled-NOT (CNOT) gate

e.g.  $C_1\text{NOT}_0$ :

(3, 1, 4, 1, 5, 9, 2, 6)  $\mapsto$

(3, 1, 1, 4, 5, 9, 6, 2).

state	measurement
$(1, 0, 0, 0, 0, 0, 0, 0)$	000
$(0, 1, 0, 0, 0, 0, 0, 0)$	001
$(0, 0, 1, 0, 0, 0, 0, 0)$	010
$(0, 0, 0, 1, 0, 0, 0, 0)$	011
$(0, 0, 0, 0, 1, 0, 0, 0)$	100
$(0, 0, 0, 0, 0, 1, 0, 0)$	101
$(0, 0, 0, 0, 0, 0, 1, 0)$	110
$(0, 0, 0, 0, 0, 0, 0, 1)$	111

Operation on quantum state:

$\text{NOT}_0$ , swapping pairs.

Operation after measurement:

flipping bit 0 of result.

Flip: output is not input.

## Controlled-NOT (CNOT) gates

e.g.  $C_1\text{NOT}_0$ :

$(3, 1, 4, 1, 5, 9, 2, 6) \mapsto$

$(3, 1, 1, 4, 5, 9, 6, 2)$ .

state	measurement
$(1, 0, 0, 0, 0, 0, 0, 0)$	000
$(0, 1, 0, 0, 0, 0, 0, 0)$	001
$(0, 0, 1, 0, 0, 0, 0, 0)$	010
$(0, 0, 0, 1, 0, 0, 0, 0)$	011
$(0, 0, 0, 0, 1, 0, 0, 0)$	100
$(0, 0, 0, 0, 0, 1, 0, 0)$	101
$(0, 0, 0, 0, 0, 0, 1, 0)$	110
$(0, 0, 0, 0, 0, 0, 0, 1)$	111

Operation on quantum state:

$\text{NOT}_0$ , swapping pairs.

Operation after measurement:

flipping bit 0 of result.

Flip: output is not input.

## Controlled-NOT (CNOT) gates

e.g.  $C_1\text{NOT}_0$ :

$(3, 1, 4, 1, 5, 9, 2, 6) \mapsto$

$(3, 1, 1, 4, 5, 9, 6, 2)$ .

Operation after measurement:

flipping bit 0 *if* bit 1 is set; i.e.,

$(q_2, q_1, q_0) \mapsto (q_2, q_1, q_0 \oplus q_1)$ .

state	measurement
$(1, 0, 0, 0, 0, 0, 0, 0)$	000
$(0, 1, 0, 0, 0, 0, 0, 0)$	001
$(0, 0, 1, 0, 0, 0, 0, 0)$	010
$(0, 0, 0, 1, 0, 0, 0, 0)$	011
$(0, 0, 0, 0, 1, 0, 0, 0)$	100
$(0, 0, 0, 0, 0, 1, 0, 0)$	101
$(0, 0, 0, 0, 0, 0, 1, 0)$	110
$(0, 0, 0, 0, 0, 0, 0, 1)$	111

Operation on quantum state:

$\text{NOT}_0$ , swapping pairs.

Operation after measurement:

flipping bit 0 of result.

Flip: output is not input.

## Controlled-NOT (CNOT) gates

e.g.  $C_1\text{NOT}_0$ :

$(3, 1, 4, 1, 5, 9, 2, 6) \mapsto$

$(3, 1, 1, 4, 5, 9, 6, 2)$ .

Operation after measurement:

flipping bit 0 *if* bit 1 is set; i.e.,

$(q_2, q_1, q_0) \mapsto (q_2, q_1, q_0 \oplus q_1)$ .

e.g.  $C_2\text{NOT}_0$ :

$(3, 1, 4, 1, 5, 9, 2, 6) \mapsto$

$(3, 1, 4, 1, 9, 5, 6, 2)$ .

state	measurement
$(1, 0, 0, 0, 0, 0, 0, 0)$	000
$(0, 1, 0, 0, 0, 0, 0, 0)$	001
$(0, 0, 1, 0, 0, 0, 0, 0)$	010
$(0, 0, 0, 1, 0, 0, 0, 0)$	011
$(0, 0, 0, 0, 1, 0, 0, 0)$	100
$(0, 0, 0, 0, 0, 1, 0, 0)$	101
$(0, 0, 0, 0, 0, 0, 1, 0)$	110
$(0, 0, 0, 0, 0, 0, 0, 1)$	111

Operation on quantum state:

$\text{NOT}_0$ , swapping pairs.

Operation after measurement:

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e.g.  $C_1\text{NOT}_0$ :

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$(3, 1, 1, 4, 5, 9, 6, 2)$ .

Operation after measurement:

flipping bit 0 *if* bit 1 is set; i.e.,

$(q_2, q_1, q_0) \mapsto (q_2, q_1, q_0 \oplus q_1)$ .

e.g.  $C_2\text{NOT}_0$ :

$(3, 1, 4, 1, 5, 9, 2, 6) \mapsto$

$(3, 1, 4, 1, 9, 5, 6, 2)$ .

e.g.  $C_0\text{NOT}_2$ :

$(3, 1, 4, 1, 5, 9, 2, 6) \mapsto$

$(3, 9, 4, 6, 5, 1, 2, 1)$ .

state	measurement
$(0, 0, 0, 0, 0)$	000
$(0, 0, 0, 0, 0)$	001
$(0, 0, 0, 0, 0)$	010
$(1, 0, 0, 0, 0)$	011
$(0, 1, 0, 0, 0)$	100
$(0, 0, 1, 0, 0)$	101
$(0, 0, 0, 1, 0)$	110
$(0, 0, 0, 0, 1)$	111

on on quantum state:

swapping pairs.

on after measurement:

bit 0 of result.

output is not input.

## Controlled-NOT (CNOT) gates

e.g.  $C_1\text{NOT}_0$ :

$(3, 1, 4, 1, 5, 9, 2, 6) \mapsto$

$(3, 1, 1, 4, 5, 9, 6, 2)$ .

Operation after measurement:

flipping bit 0 *if* bit 1 is set; i.e.,

$(q_2, q_1, q_0) \mapsto (q_2, q_1, q_0 \oplus q_1)$ .

e.g.  $C_2\text{NOT}_0$ :

$(3, 1, 4, 1, 5, 9, 2, 6) \mapsto$

$(3, 1, 4, 1, 9, 5, 6, 2)$ .

e.g.  $C_0\text{NOT}_2$ :

$(3, 1, 4, 1, 5, 9, 2, 6) \mapsto$

$(3, 9, 4, 6, 5, 1, 2, 1)$ .

## Toffoli g

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controlle

e.g.  $C_2C$

$(3, 1, 4, 1$

$(3, 1, 4, 1$

Controlled-NOT (CNOT) gates

e.g.  $C_1\text{NOT}_0$ :

$(3, 1, 4, 1, 5, 9, 2, 6) \mapsto$

$(3, 1, 1, 4, 5, 9, 6, 2)$ .

Operation after measurement:

flipping bit 0 *if* bit 1 is set; i.e.,

$(q_2, q_1, q_0) \mapsto (q_2, q_1, q_0 \oplus q_1)$ .

e.g.  $C_2\text{NOT}_0$ :

$(3, 1, 4, 1, 5, 9, 2, 6) \mapsto$

$(3, 1, 4, 1, 9, 5, 6, 2)$ .

e.g.  $C_0\text{NOT}_2$ :

$(3, 1, 4, 1, 5, 9, 2, 6) \mapsto$

$(3, 9, 4, 6, 5, 1, 2, 1)$ .

Toffoli gates

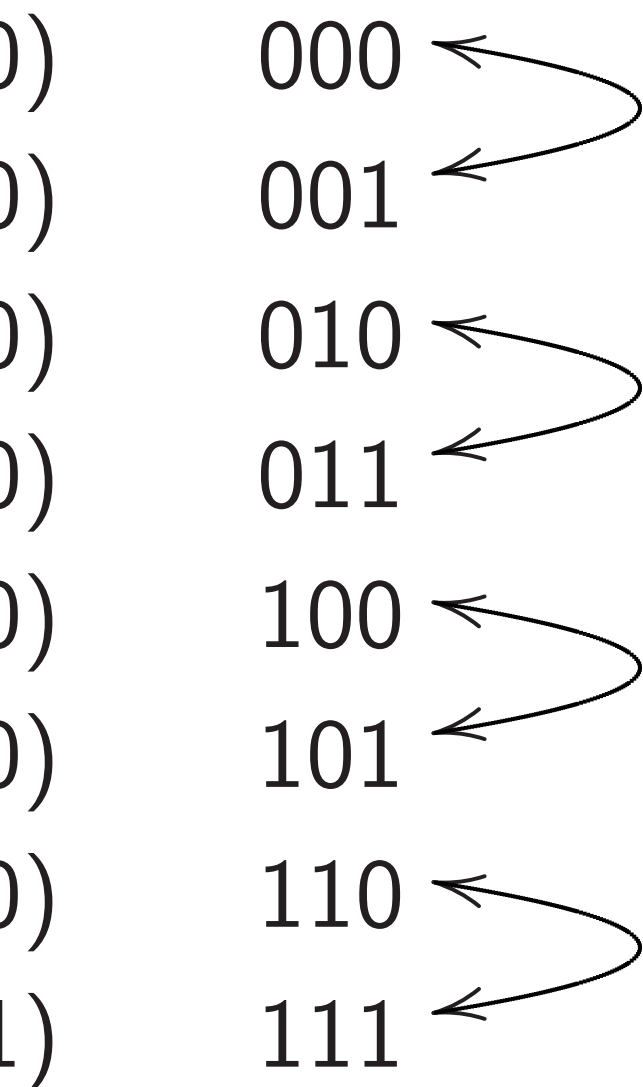
Also known as CC  
controlled-controlled

e.g.  $C_2C_1\text{NOT}_0$ :

$(3, 1, 4, 1, 5, 9, 2, 6)$

$(3, 1, 4, 1, 5, 9, 6, 2)$

measurement



ntum state:

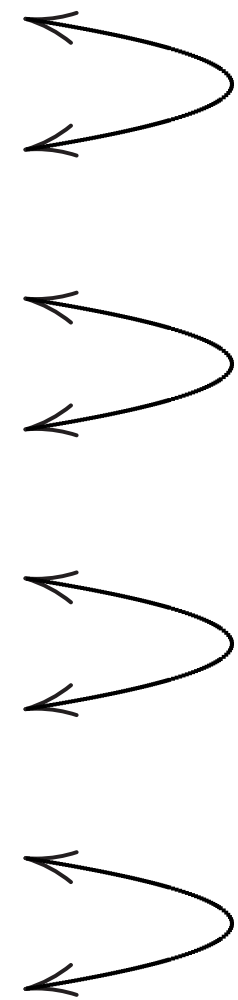
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## Controlled-NOT (CNOT) gates

e.g.  $C_1\text{NOT}_0$ :

$$(3, 1, 4, 1, 5, 9, 2, 6) \mapsto$$

$$(3, 1, 1, 4, 5, 9, 6, 2).$$

Operation after measurement:

flipping bit 0 *if* bit 1 is set; i.e.,

$$(q_2, q_1, q_0) \mapsto (q_2, q_1, q_0 \oplus q_1).$$

e.g.  $C_2\text{NOT}_0$ :

$$(3, 1, 4, 1, 5, 9, 2, 6) \mapsto$$

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e.g.  $C_0\text{NOT}_2$ :

$$(3, 1, 4, 1, 5, 9, 2, 6) \mapsto$$

$$(3, 9, 4, 6, 5, 1, 2, 1).$$

## Toffoli gates

Also known as CCNOT gate

controlled-controlled-NOT g

e.g.  $C_2C_1\text{NOT}_0$ :

$$(3, 1, 4, 1, 5, 9, 2, 6) \mapsto$$

$$(3, 1, 4, 1, 5, 9, 6, 2).$$



Controlled-NOT (CNOT) gates

e.g.  $C_1NOT_0$ :

$(3, 1, 4, 1, 5, 9, 2, 6) \mapsto$

$(3, 1, 1, 4, 5, 9, 6, 2)$ .

Operation after measurement:

flipping bit 0 *if* bit 1 is set; i.e.,

$(q_2, q_1, q_0) \mapsto (q_2, q_1, q_0 \oplus q_1)$ .

e.g.  $C_2NOT_0$ :

$(3, 1, 4, 1, 5, 9, 2, 6) \mapsto$

$(3, 1, 4, 1, 9, 5, 6, 2)$ .

e.g.  $C_0NOT_2$ :

$(3, 1, 4, 1, 5, 9, 2, 6) \mapsto$

$(3, 9, 4, 6, 5, 1, 2, 1)$ .

Toffoli gates

Also known as CCNOT gates:  
controlled-controlled-NOT gates.

e.g.  $C_2C_1NOT_0$ :

$(3, 1, 4, 1, 5, 9, 2, 6) \mapsto$

$(3, 1, 4, 1, 5, 9, 6, 2)$ .

Controlled-NOT (CNOT) gates

e.g.  $C_1NOT_0$ :

$$(3, 1, 4, 1, 5, 9, 2, 6) \mapsto (3, 1, 1, 4, 5, 9, 6, 2).$$

Operation after measurement:

flipping bit 0 *if* bit 1 is set; i.e.,

$$(q_2, q_1, q_0) \mapsto (q_2, q_1, q_0 \oplus q_1).$$

e.g.  $C_2NOT_0$ :

$$(3, 1, 4, 1, 5, 9, 2, 6) \mapsto (3, 1, 4, 1, 9, 5, 6, 2).$$

e.g.  $C_0NOT_2$ :

$$(3, 1, 4, 1, 5, 9, 2, 6) \mapsto (3, 9, 4, 6, 5, 1, 2, 1).$$

Toffoli gates

Also known as CCNOT gates:  
controlled-controlled-NOT gates.

e.g.  $C_2C_1NOT_0$ :

$$(3, 1, 4, 1, 5, 9, 2, 6) \mapsto (3, 1, 4, 1, 5, 9, 6, 2).$$

Operation after measurement:

$$(q_2, q_1, q_0) \mapsto (q_2, q_1, q_0 \oplus q_1 q_2).$$

Controlled-NOT (CNOT) gates

e.g.  $C_1NOT_0$ :

$$(3, 1, 4, 1, 5, 9, 2, 6) \mapsto (3, 1, 1, 4, 5, 9, 6, 2).$$

Operation after measurement:

flipping bit 0 *if* bit 1 is set; i.e.,

$$(q_2, q_1, q_0) \mapsto (q_2, q_1, q_0 \oplus q_1).$$

e.g.  $C_2NOT_0$ :

$$(3, 1, 4, 1, 5, 9, 2, 6) \mapsto (3, 1, 4, 1, 9, 5, 6, 2).$$

e.g.  $C_0NOT_2$ :

$$(3, 1, 4, 1, 5, 9, 2, 6) \mapsto (3, 9, 4, 6, 5, 1, 2, 1).$$

Toffoli gates

Also known as CCNOT gates:

controlled-controlled-NOT gates.

e.g.  $C_2C_1NOT_0$ :

$$(3, 1, 4, 1, 5, 9, 2, 6) \mapsto (3, 1, 4, 1, 5, 9, 6, 2).$$

Operation after measurement:

$$(q_2, q_1, q_0) \mapsto (q_2, q_1, q_0 \oplus q_1 q_2).$$

e.g.  $C_0C_1NOT_2$ :

$$(3, 1, 4, 1, 5, 9, 2, 6) \mapsto (3, 1, 4, 6, 5, 9, 2, 1).$$

Controlled-NOT (CNOT) gatesCNOT<sub>0</sub>:(1, 5, 9, 2, 6)  $\mapsto$ 

(4, 5, 9, 6, 2).

Operation after measurement:

bit 0 *if* bit 1 is set; i.e., $(q_2, q_1, q_0) \mapsto (q_2, q_1, q_0 \oplus q_1)$ .CNOT<sub>1</sub>:(1, 5, 9, 2, 6)  $\mapsto$ 

(1, 9, 5, 6, 2).

CNOT<sub>2</sub>:(1, 5, 9, 2, 6)  $\mapsto$ 

(5, 5, 1, 2, 1).

Toffoli gatesAlso known as CCNOT gates:  
controlled-controlled-NOT gates.e.g. C<sub>2</sub>C<sub>1</sub>NOT<sub>0</sub>:(3, 1, 4, 1, 5, 9, 2, 6)  $\mapsto$ 

(3, 1, 4, 1, 5, 9, 6, 2).

Operation after measurement:

 $(q_2, q_1, q_0) \mapsto (q_2, q_1, q_0 \oplus q_1 q_2)$ .e.g. C<sub>0</sub>C<sub>1</sub>NOT<sub>2</sub>:(3, 1, 4, 1, 5, 9, 2, 6)  $\mapsto$ 

(3, 1, 4, 6, 5, 9, 2, 1).

More shCombined  
to build

CNOT) gates)  $\mapsto$ 

).

Measurement:

t 1 is set; i.e.,

,  $q_1, q_0 \oplus q_1$ ).)  $\mapsto$ 

).

)  $\mapsto$ 

).

Toffoli gates

Also known as CCNOT gates:  
controlled-controlled-NOT gates.

e.g.  $C_2C_1NOT_0$ : $(3, 1, 4, 1, 5, 9, 2, 6) \mapsto$  $(3, 1, 4, 1, 5, 9, 6, 2)$ .

Operation after measurement:

 $(q_2, q_1, q_0) \mapsto (q_2, q_1, q_0 \oplus q_1 q_2)$ .e.g.  $C_0C_1NOT_2$ : $(3, 1, 4, 1, 5, 9, 2, 6) \mapsto$  $(3, 1, 4, 6, 5, 9, 2, 1)$ .More shuffling

Combine NOT, CNOT,  
to build other permutations.

Toffoli gates

Also known as CCNOT gates:  
controlled-controlled-NOT gates.

e.g.  $C_2C_1NOT_0$ :

$(3, 1, 4, 1, 5, 9, 2, 6) \mapsto$

$(3, 1, 4, 1, 5, 9, 6, 2)$ .

Operation after measurement:

$(q_2, q_1, q_0) \mapsto (q_2, q_1, q_0 \oplus q_1 q_2)$ .

e.g.  $C_0C_1NOT_2$ :

$(3, 1, 4, 1, 5, 9, 2, 6) \mapsto$

$(3, 1, 4, 6, 5, 9, 2, 1)$ .

More shuffling

Combine NOT, CNOT, Toffoli  
to build other permutations.

Toffoli gates

Also known as CCNOT gates:  
controlled-controlled-NOT gates.

e.g.  $C_2C_1NOT_0$ :

$(3, 1, 4, 1, 5, 9, 2, 6) \mapsto$

$(3, 1, 4, 1, 5, 9, 6, 2)$ .

Operation after measurement:

$(q_2, q_1, q_0) \mapsto (q_2, q_1, q_0 \oplus q_1 q_2)$ .

e.g.  $C_0C_1NOT_2$ :

$(3, 1, 4, 1, 5, 9, 2, 6) \mapsto$

$(3, 1, 4, 6, 5, 9, 2, 1)$ .

More shuffling

Combine NOT, CNOT, Toffoli  
to build other permutations.

## Toffoli gates

Also known as CCNOT gates:  
controlled-controlled-NOT gates.

e.g.  $C_2C_1NOT_0$ :

$(3, 1, 4, 1, 5, 9, 2, 6) \mapsto$

$(3, 1, 4, 1, 5, 9, 6, 2)$ .

Operation after measurement:

$(q_2, q_1, q_0) \mapsto (q_2, q_1, q_0 \oplus q_1 q_2)$ .

e.g.  $C_0C_1NOT_2$ :

$(3, 1, 4, 1, 5, 9, 2, 6) \mapsto$

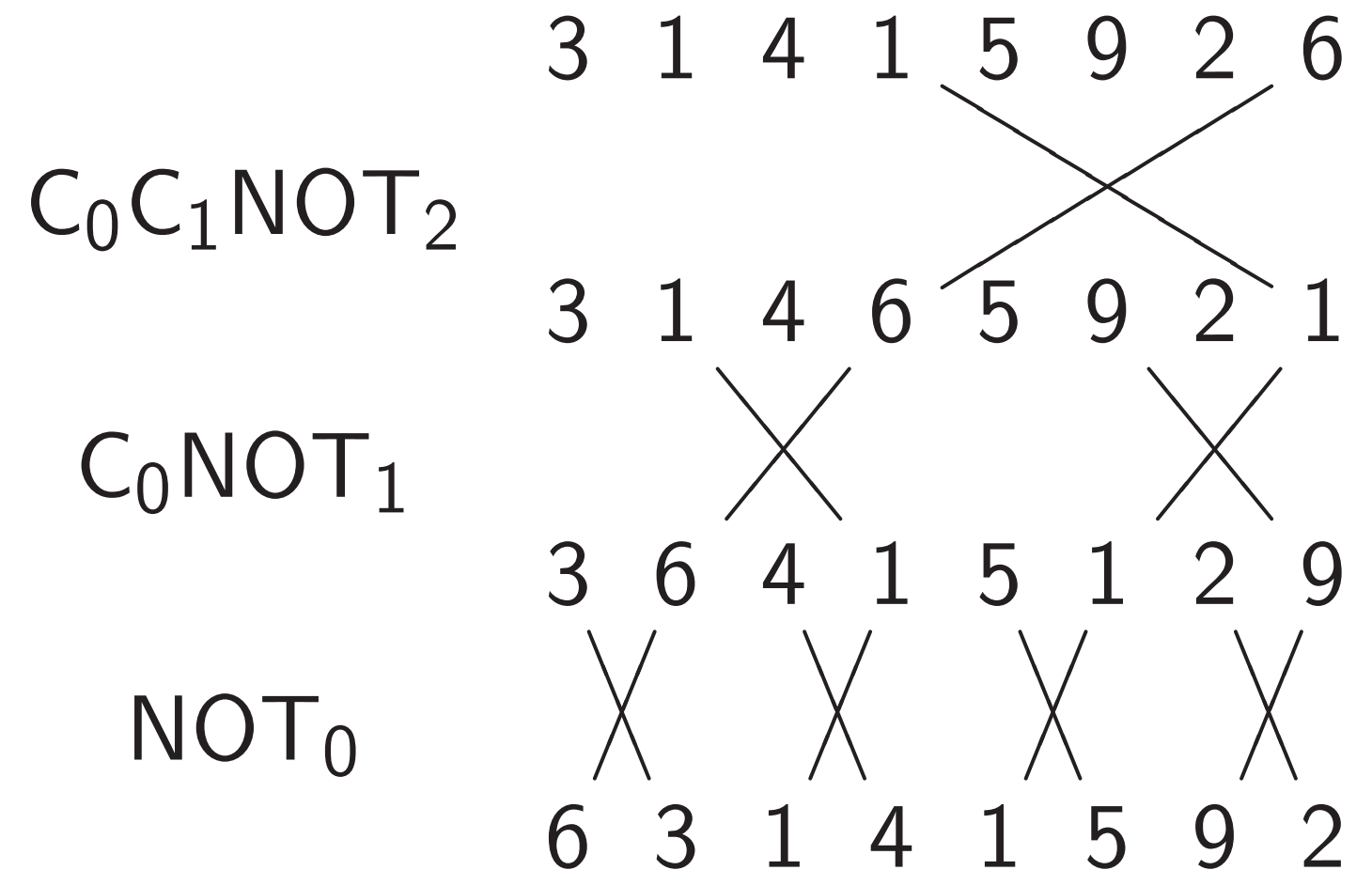
$(3, 1, 4, 6, 5, 9, 2, 1)$ .

## More shuffling

Combine NOT, CNOT, Toffoli  
to build other permutations.

e.g. series of gates to

rotate 8 positions by distance 1:





gates

own as CCNOT gates:

ed-controlled-NOT gates.

$C_1NOT_0$ :

$(1, 5, 9, 2, 6) \mapsto$

$(1, 5, 9, 6, 2)$ .

on after measurement:

$(q_0) \mapsto (q_2, q_1, q_0 \oplus q_1 q_2)$ .

$C_1NOT_2$ :

$(1, 5, 9, 2, 6) \mapsto$

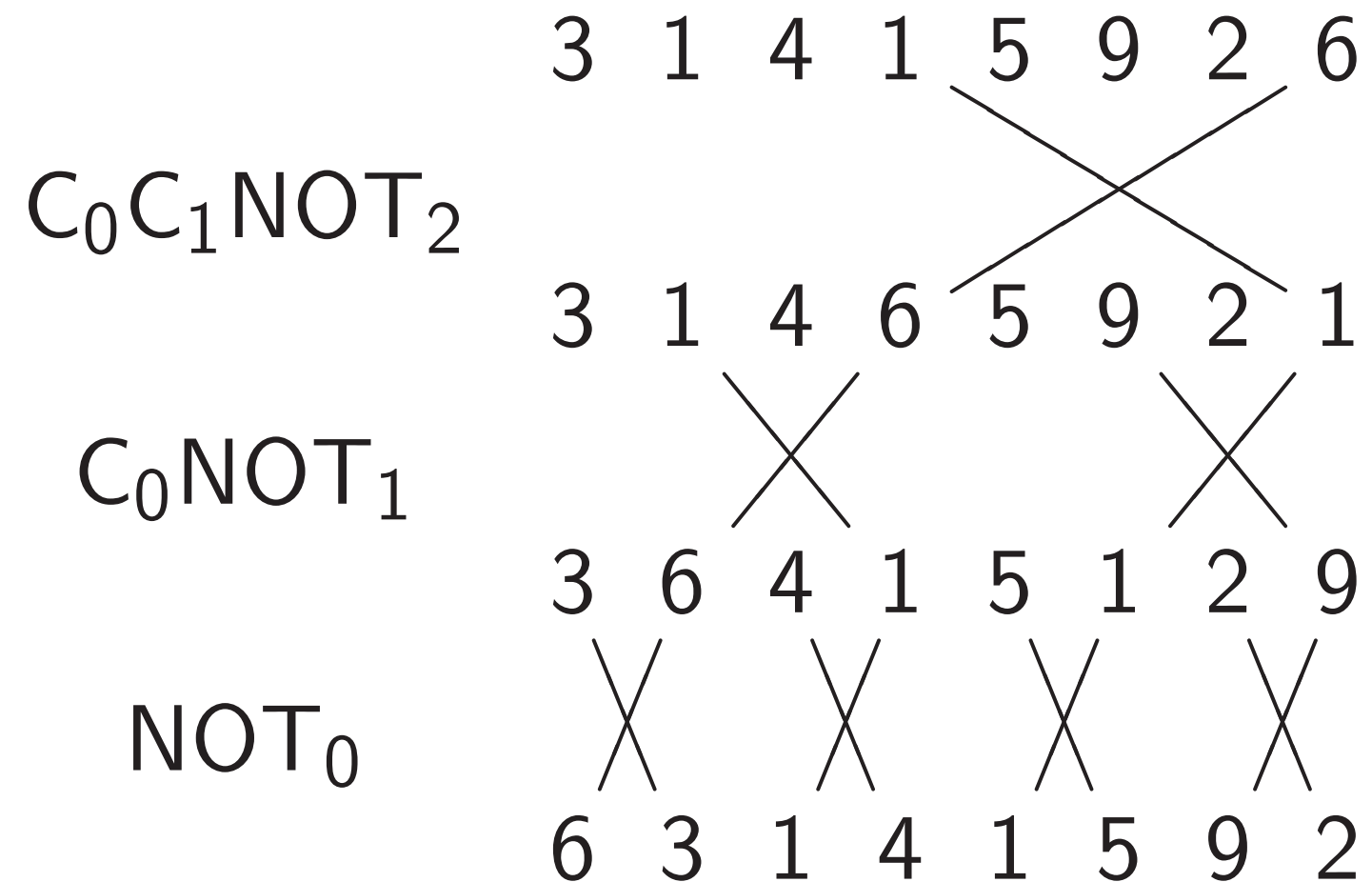
$(5, 5, 9, 2, 1)$ .

## More shuffling

Combine NOT, CNOT, Toffoli to build other permutations.

e.g. series of gates to

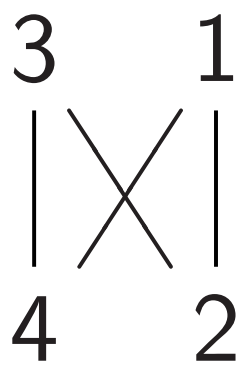
rotate 8 positions by distance 1:



Hadama

Hadama

$(a, b) \mapsto$

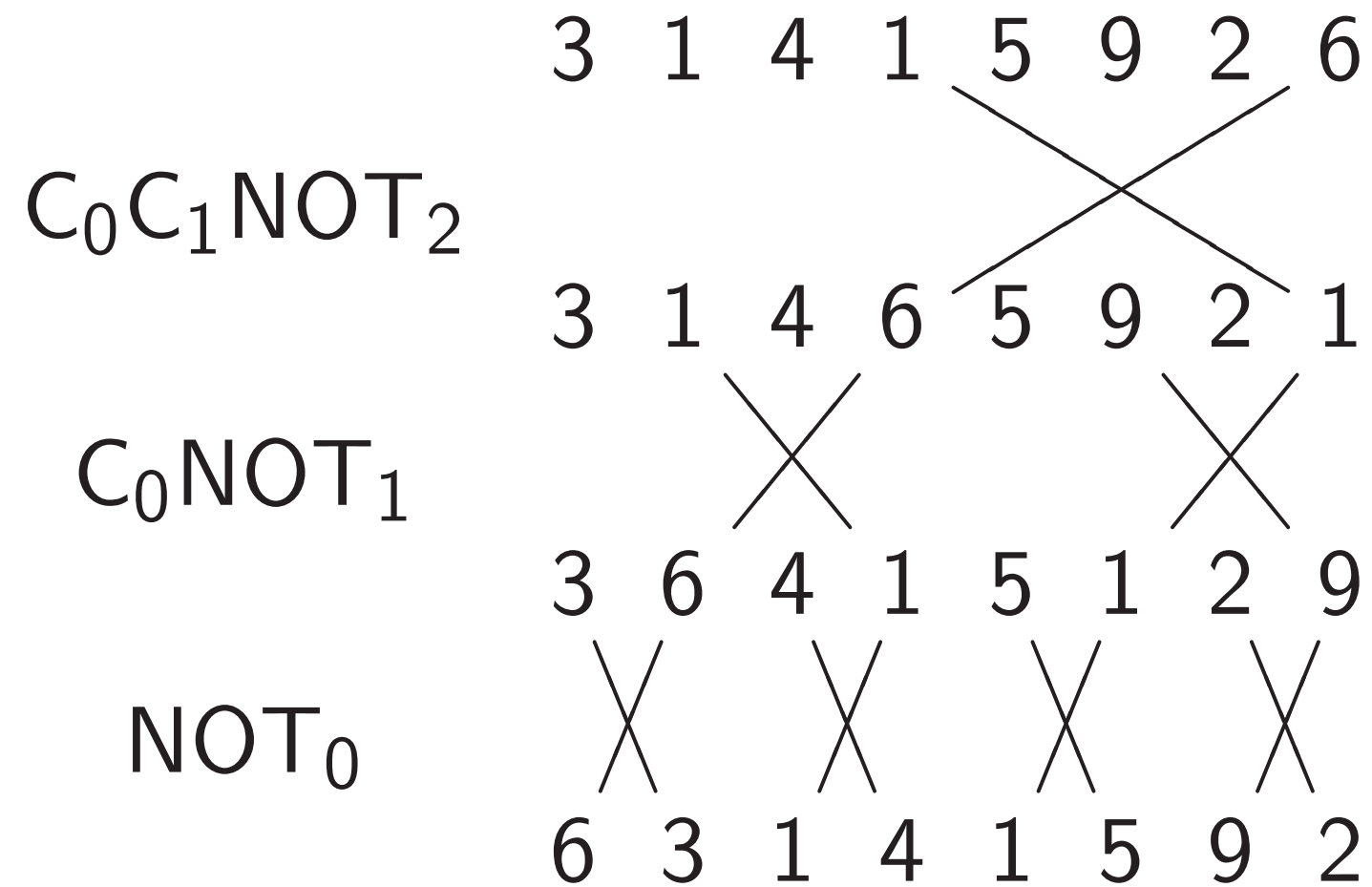


More shuffling

Combine NOT, CNOT, Toffoli to build other permutations.

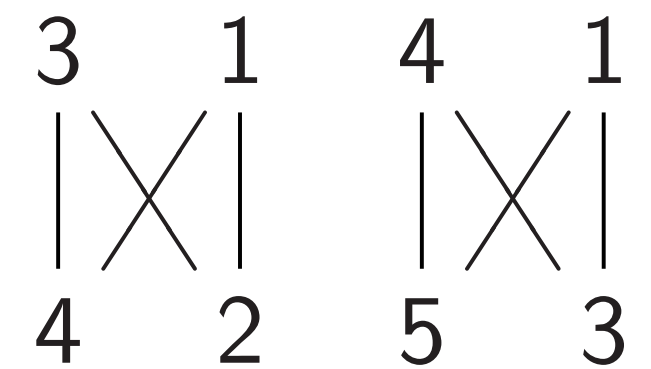
e.g. series of gates to

rotate 8 positions by distance 1:

Hadamard gates

Hadamard<sub>0</sub>:

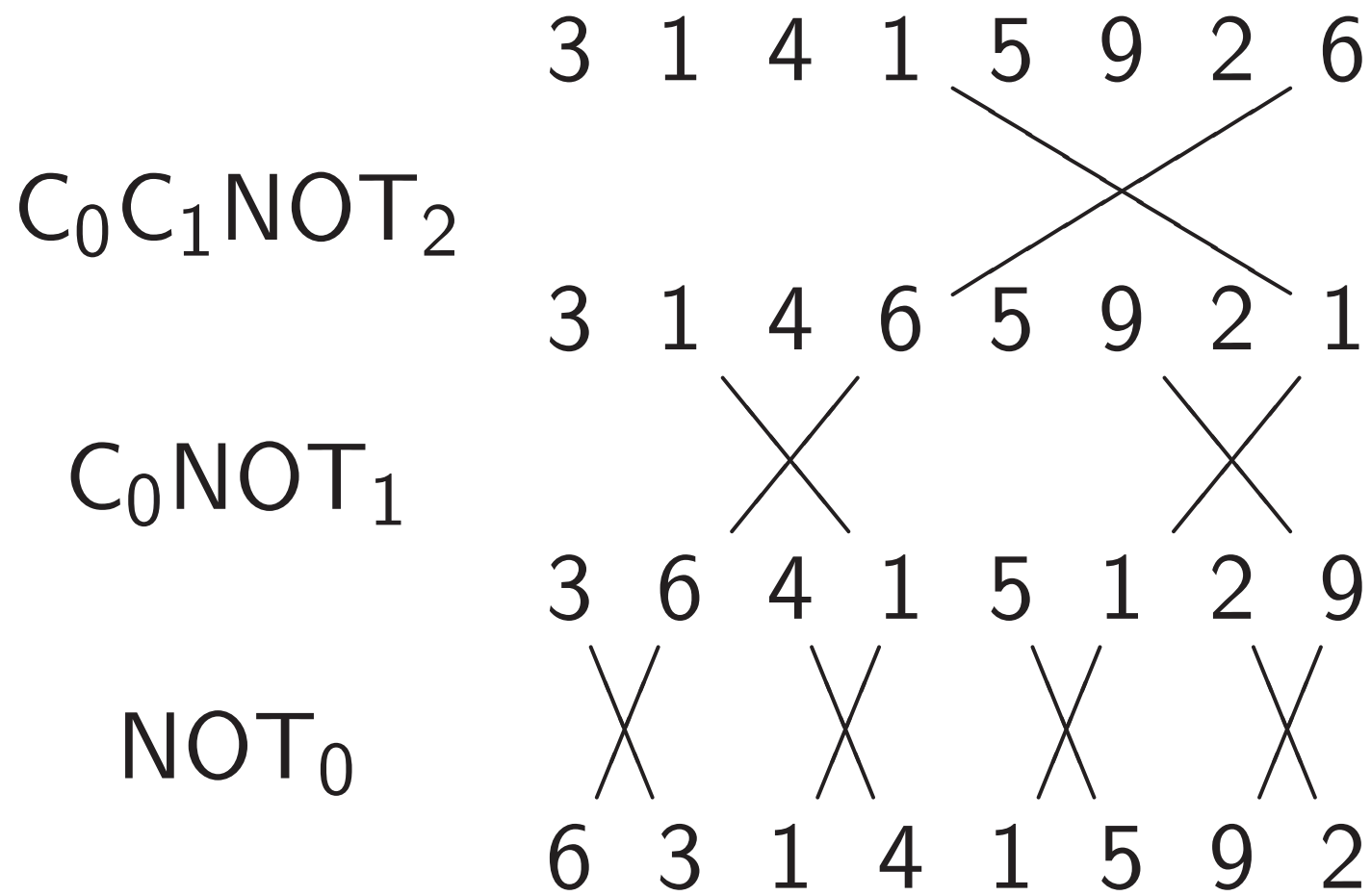
$$(a, b) \mapsto (a + b, a - b)$$



## More shuffling

Combine NOT, CNOT, Toffoli gates to build other permutations.

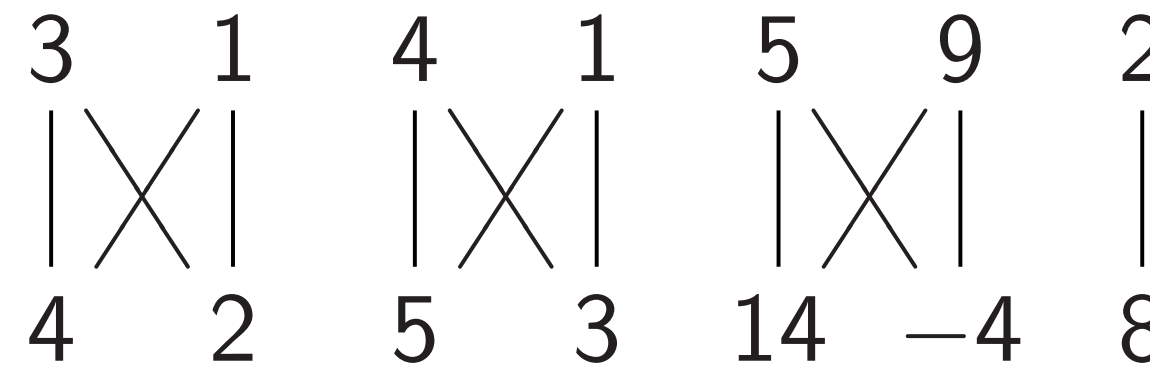
e.g. series of gates to rotate 8 positions by distance 1:



## Hadamard gates

Hadamard<sub>0</sub>:

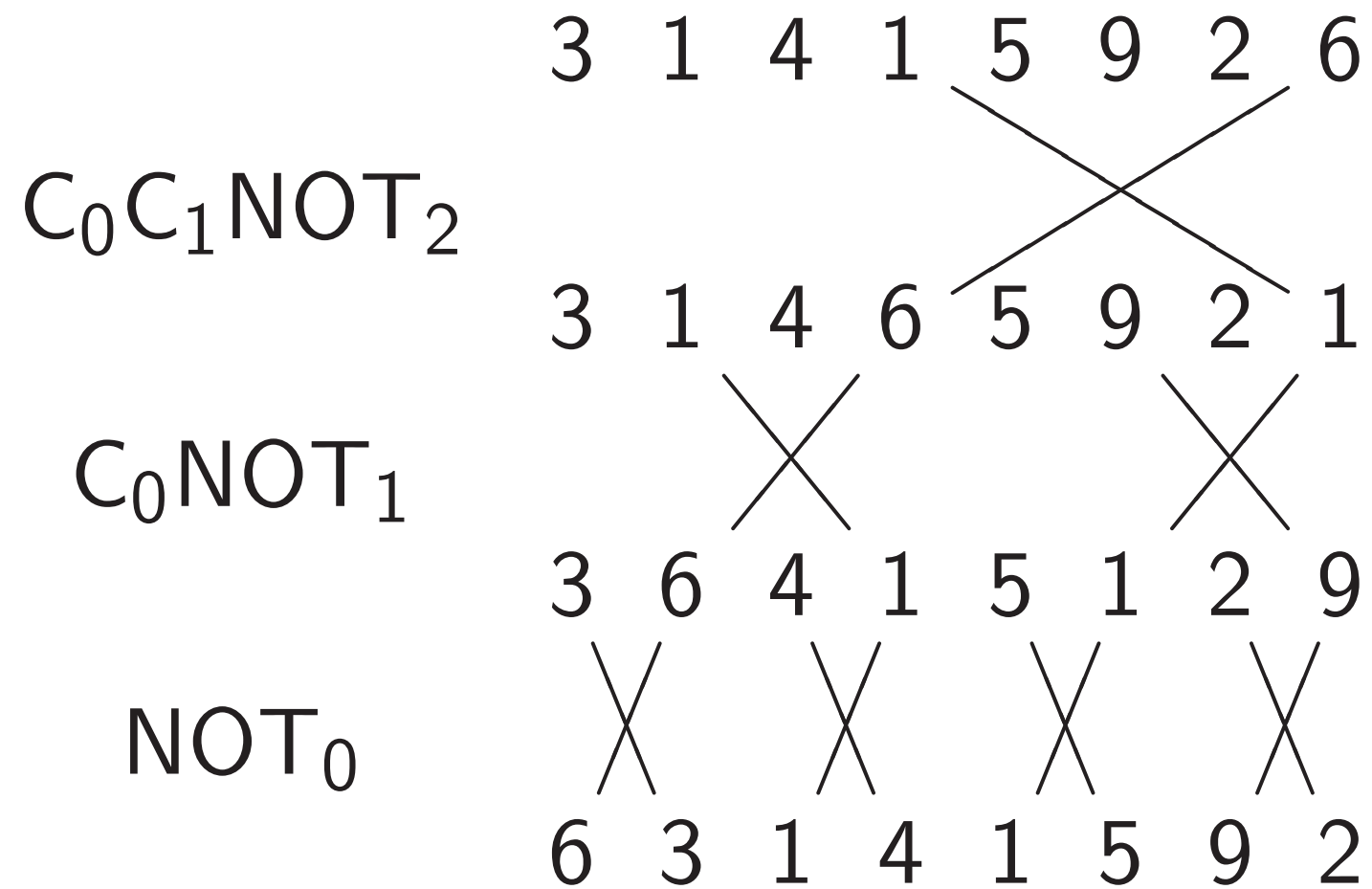
$$(a, b) \mapsto (a + b, a - b).$$



More shuffling

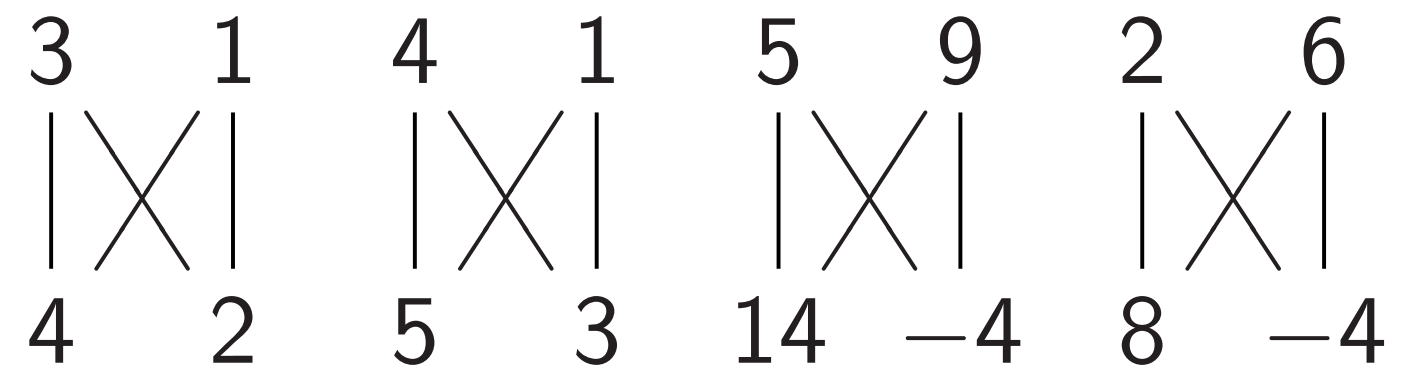
Combine NOT, CNOT, Toffoli to build other permutations.

e.g. series of gates to rotate 8 positions by distance 1:

Hadamard gates

Hadamard<sub>0</sub>:

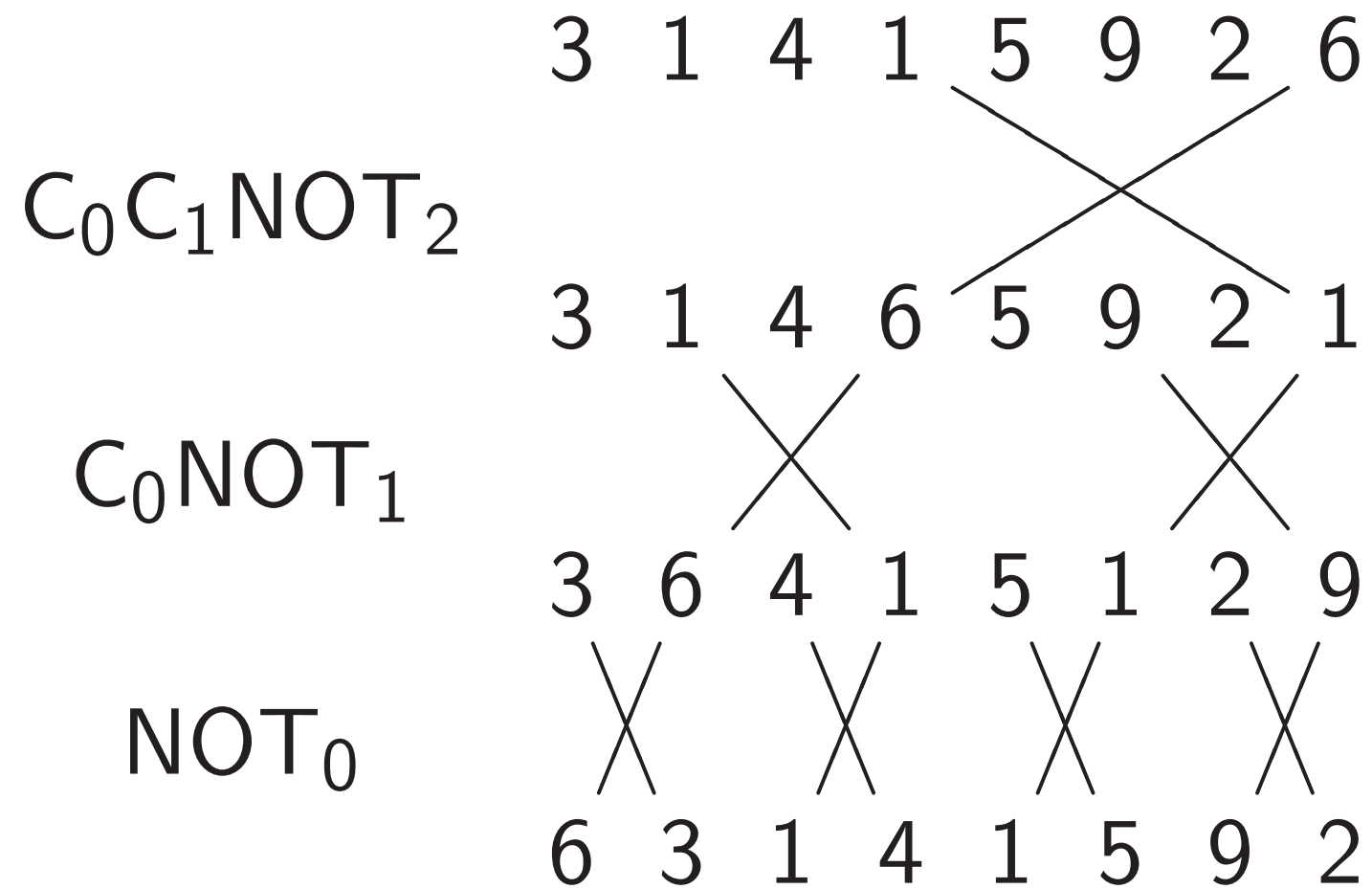
$$(a, b) \mapsto (a + b, a - b).$$



More shuffling

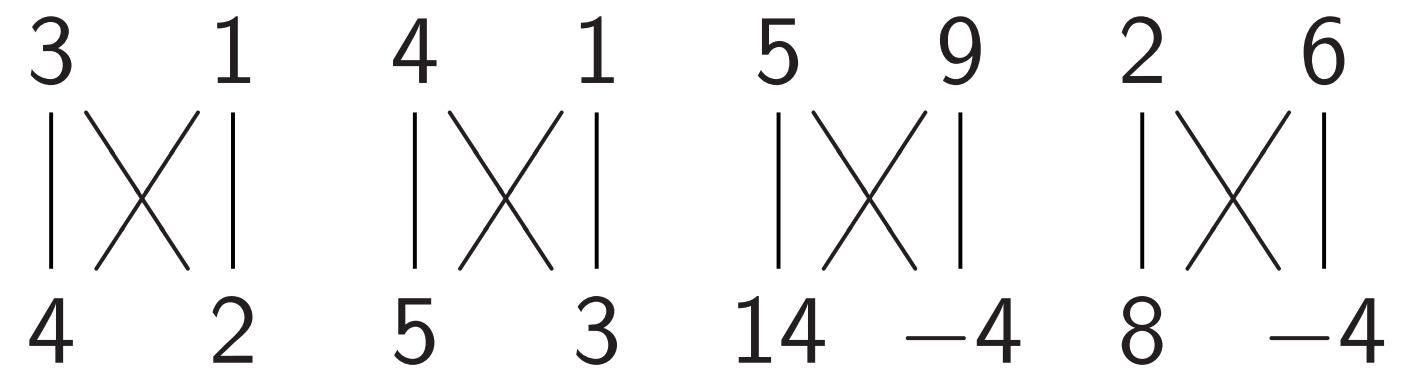
Combine NOT, CNOT, Toffoli to build other permutations.

e.g. series of gates to rotate 8 positions by distance 1:

Hadamard gates

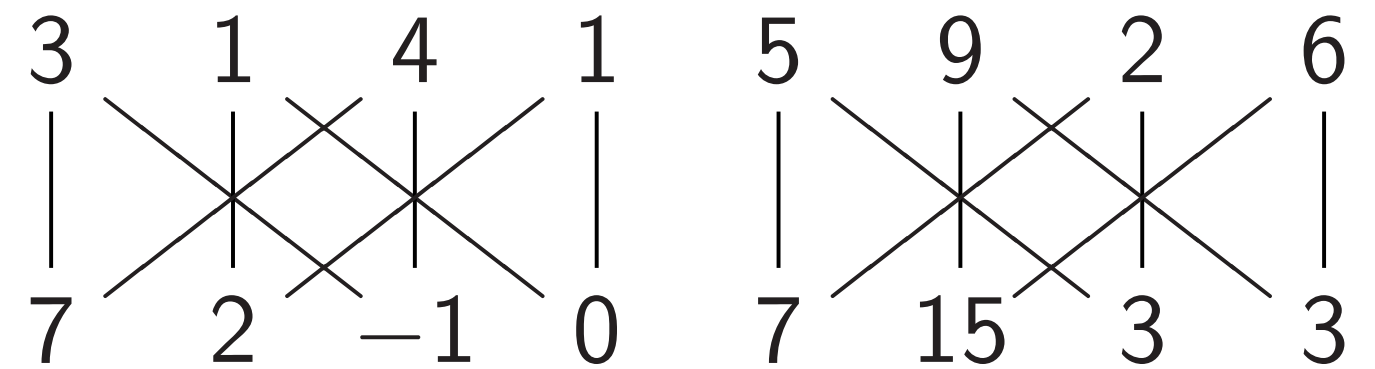
Hadamard<sub>0</sub>:

$$(a, b) \mapsto (a + b, a - b).$$



Hadamard<sub>1</sub>:

$$(a, b, c, d) \mapsto (a + c, b + d, a - c, b - d).$$



Shuffling

NOT, CNOT, Toffoli  
other permutations.

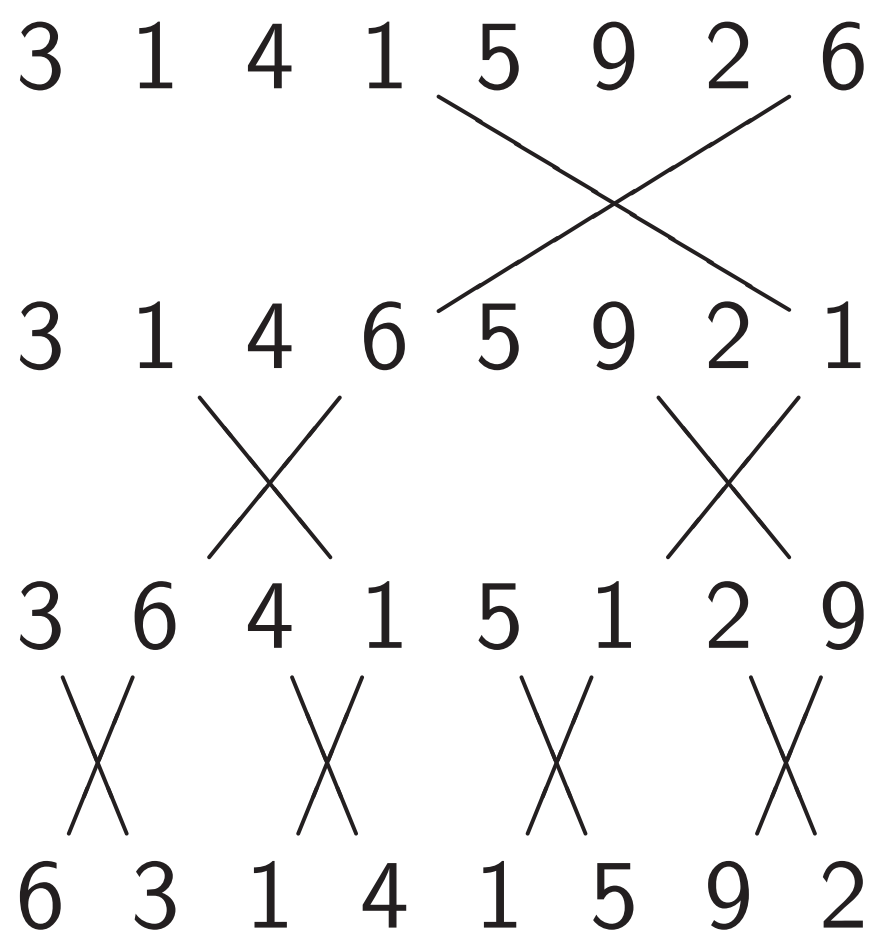
sequences of gates to

positions by distance 1:

$T_2$

$T_1$

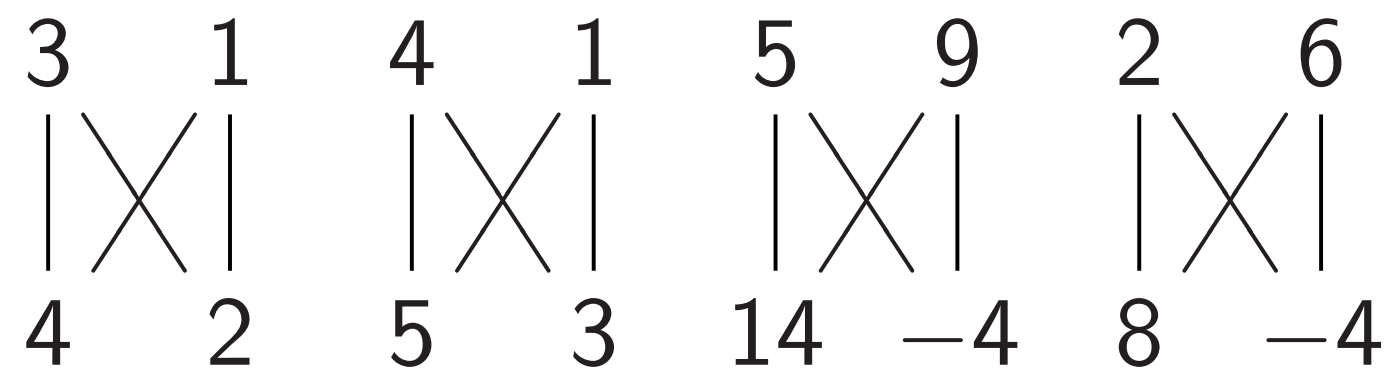
$T_0$



Hadamard gates

Hadamard<sub>0</sub>:

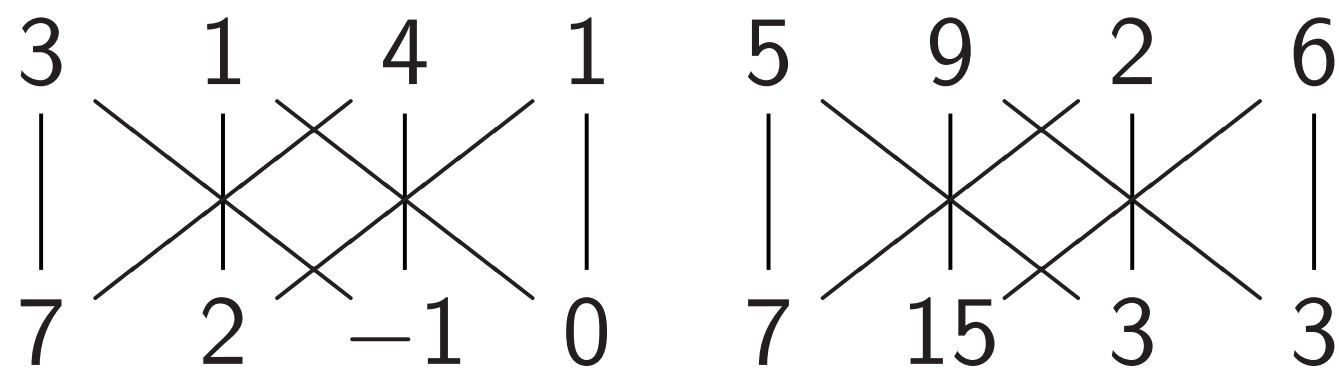
$$(a, b) \mapsto (a + b, a - b).$$



Hadamard<sub>1</sub>:

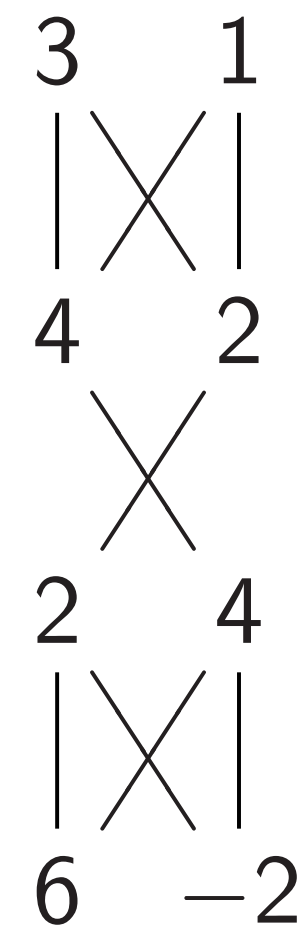
$$(a, b, c, d) \mapsto$$

$$(a + c, b + d, a - c, b - d).$$



Some Hadamard

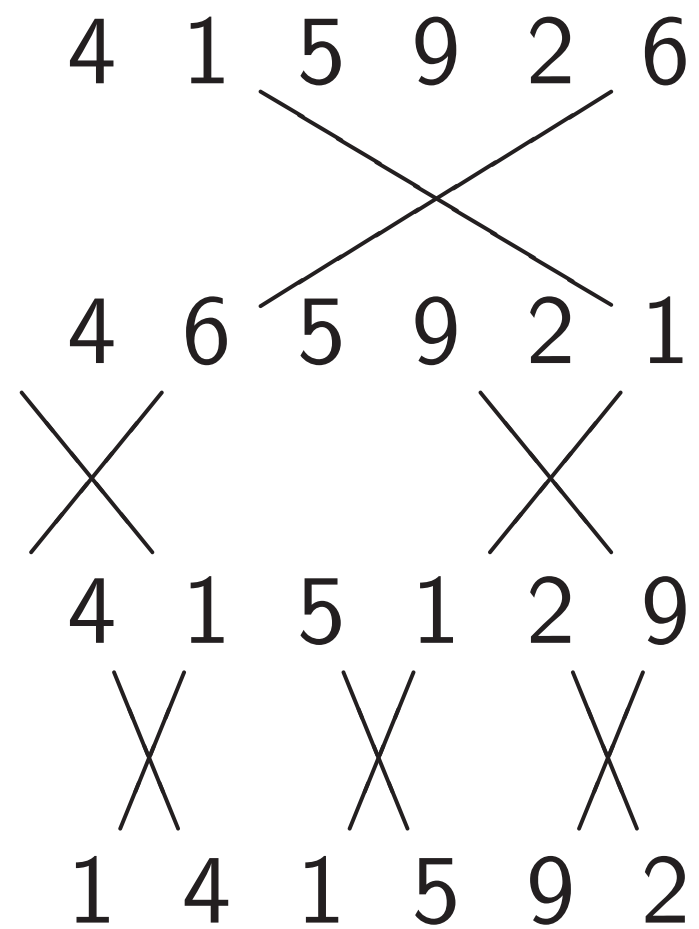
Hadamard



NOT, Toffoli  
permutations.

to

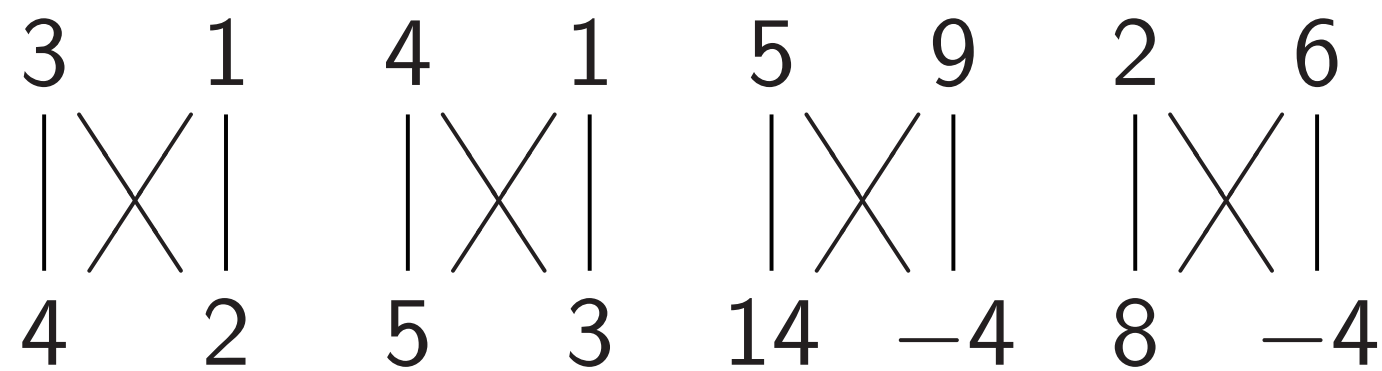
by distance 1:



## Hadamard gates

Hadamard<sub>0</sub>:

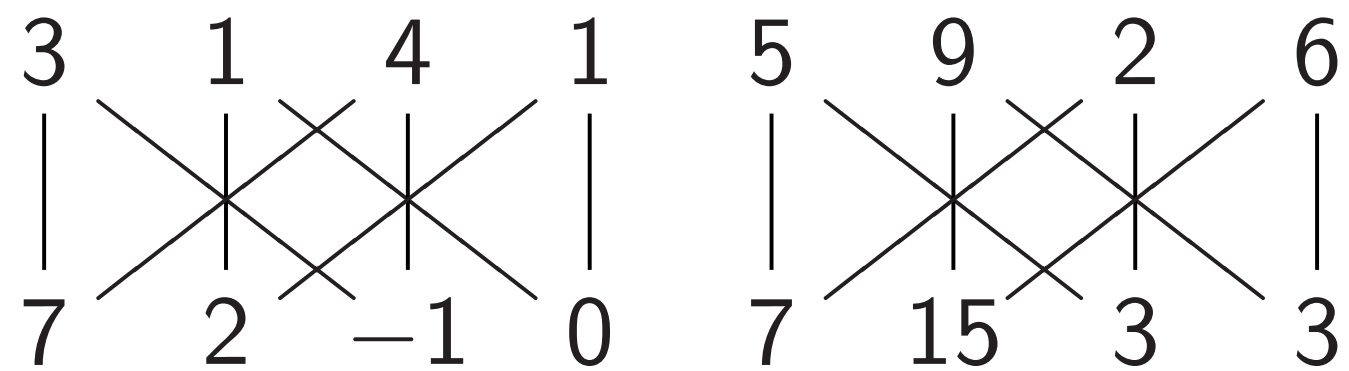
$$(a, b) \mapsto (a + b, a - b).$$



Hadamard<sub>1</sub>:

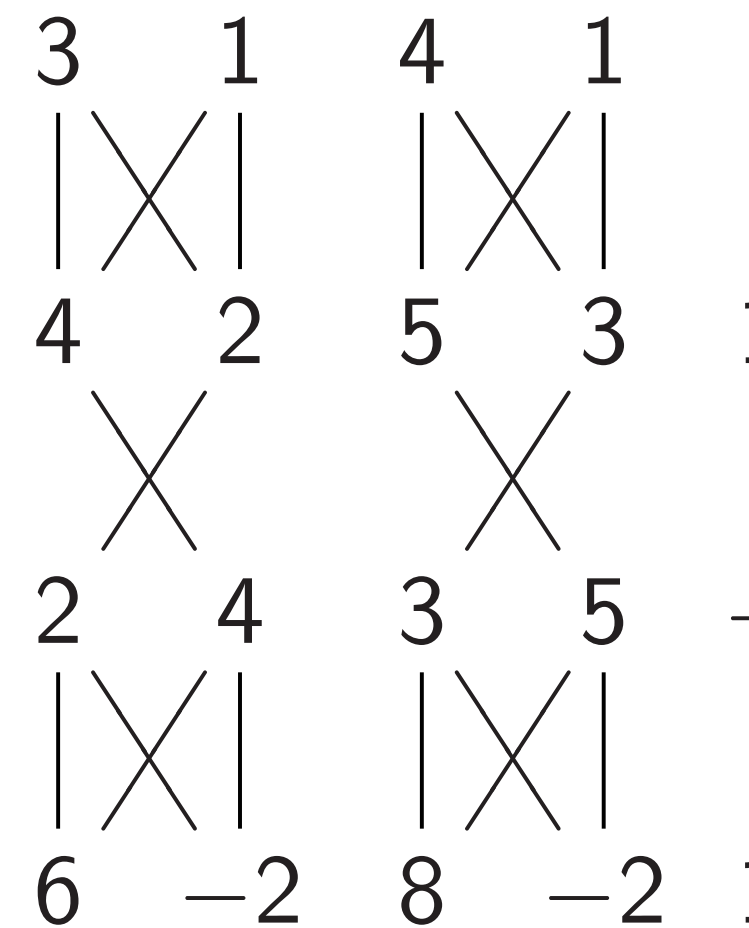
$$(a, b, c, d) \mapsto$$

$$(a + c, b + d, a - c, b - d).$$



## Some Hadamard a

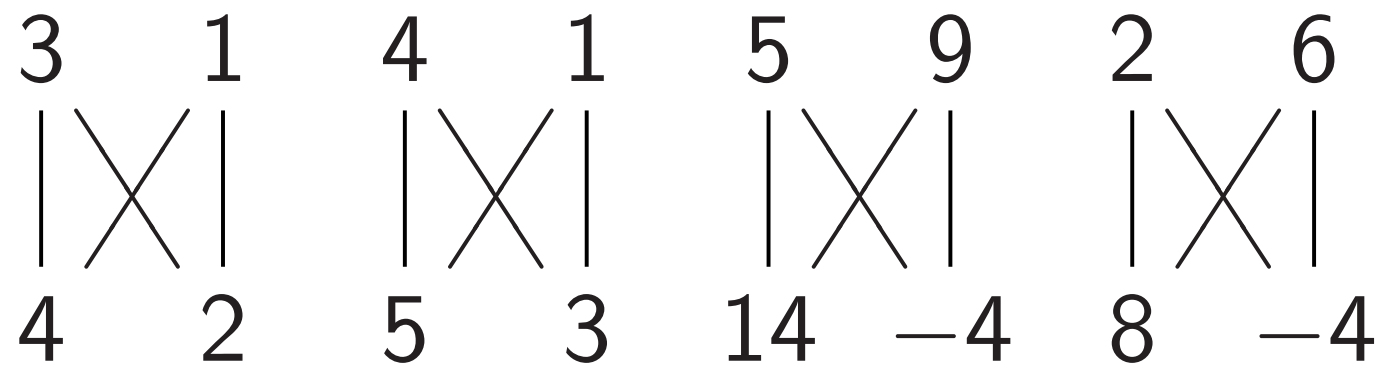
Hadamard<sub>0</sub>, NOT



Hadamard gates

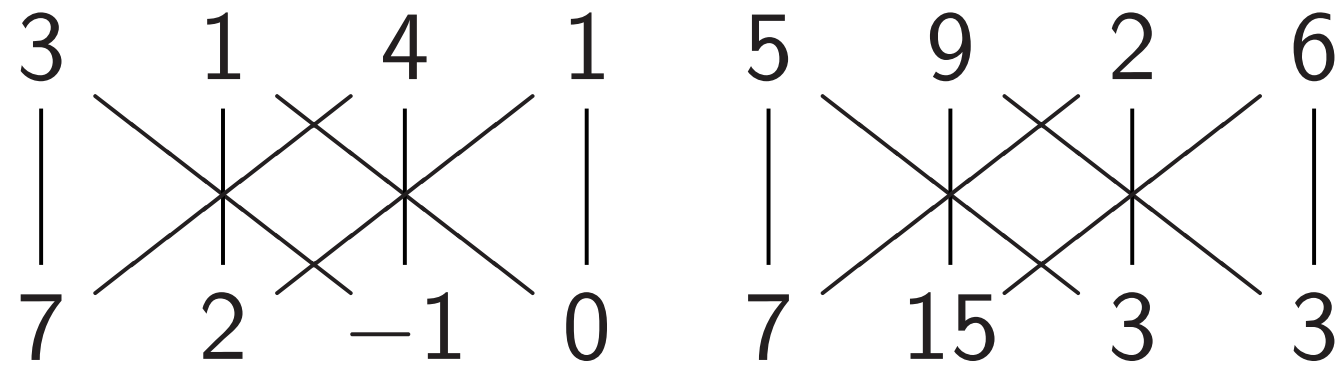
Hadamard<sub>0</sub>:

$$(a, b) \mapsto (a + b, a - b).$$



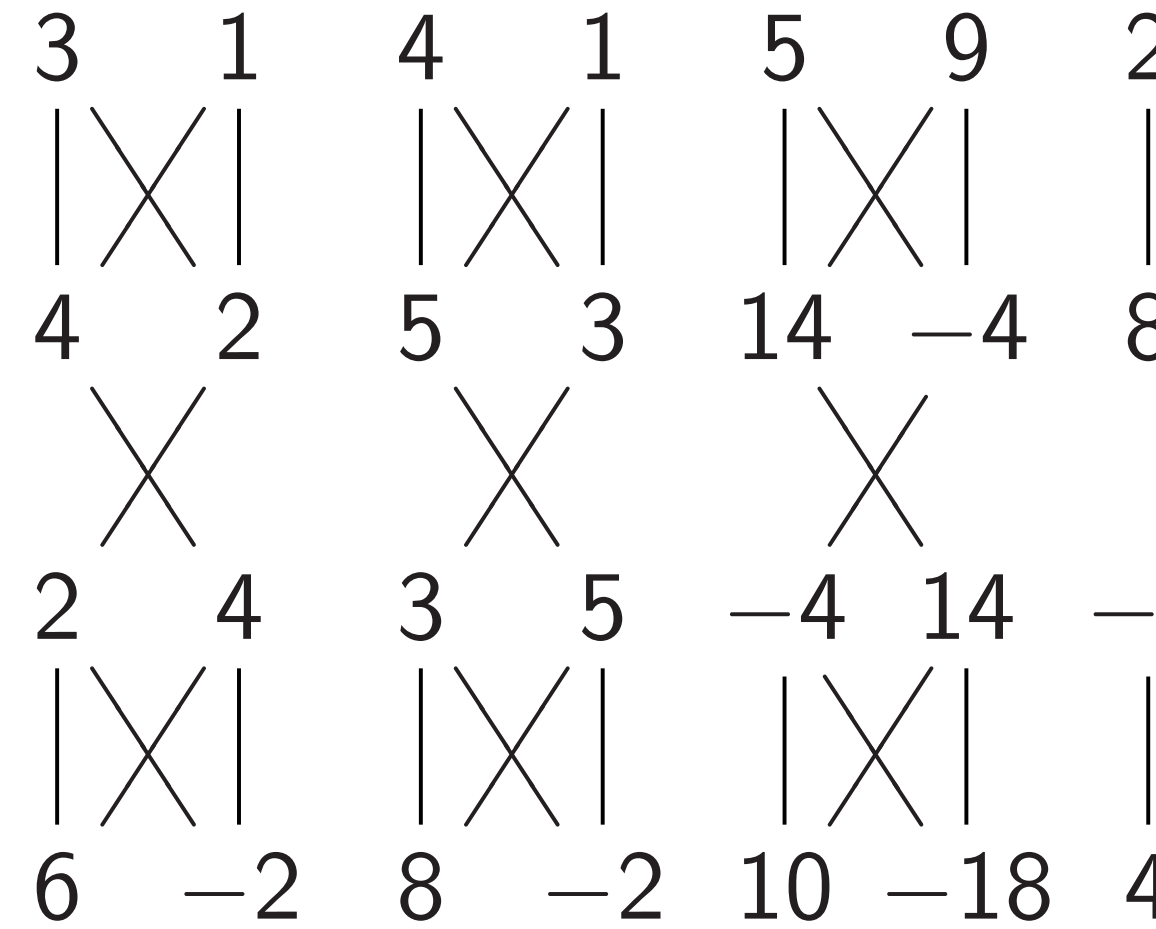
Hadamard<sub>1</sub>:

$$(a, b, c, d) \mapsto (a + c, b + d, a - c, b - d).$$



Some Hadamard application

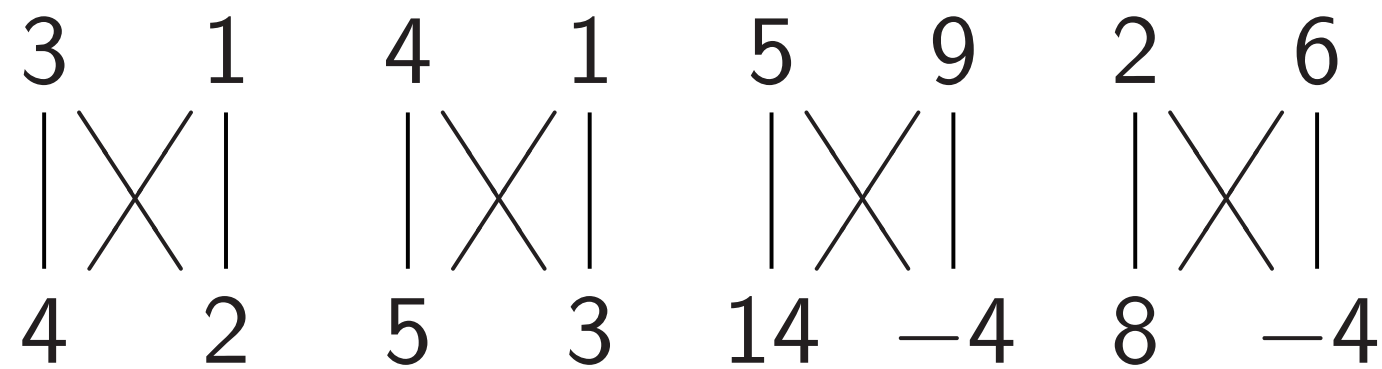
Hadamard<sub>0</sub>, NOT<sub>0</sub>, Hadamard<sub>0</sub>





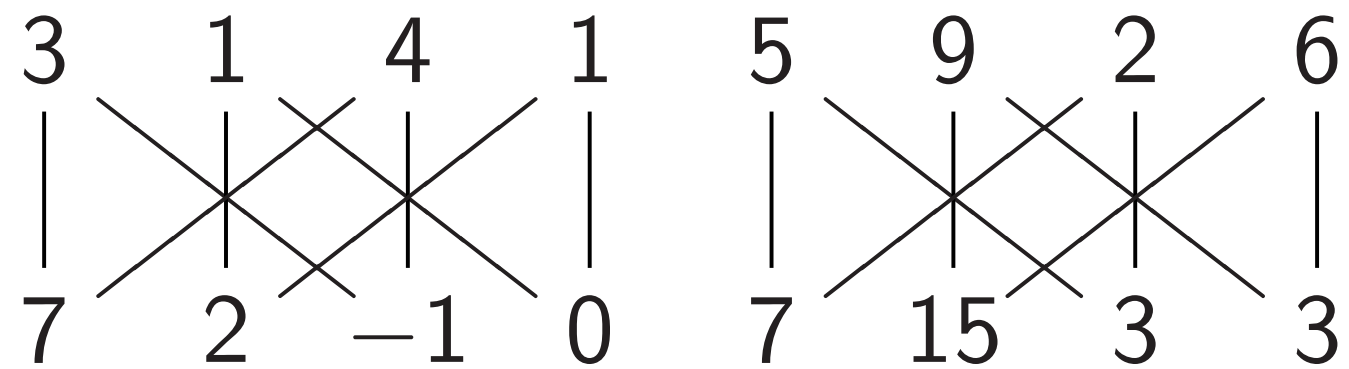
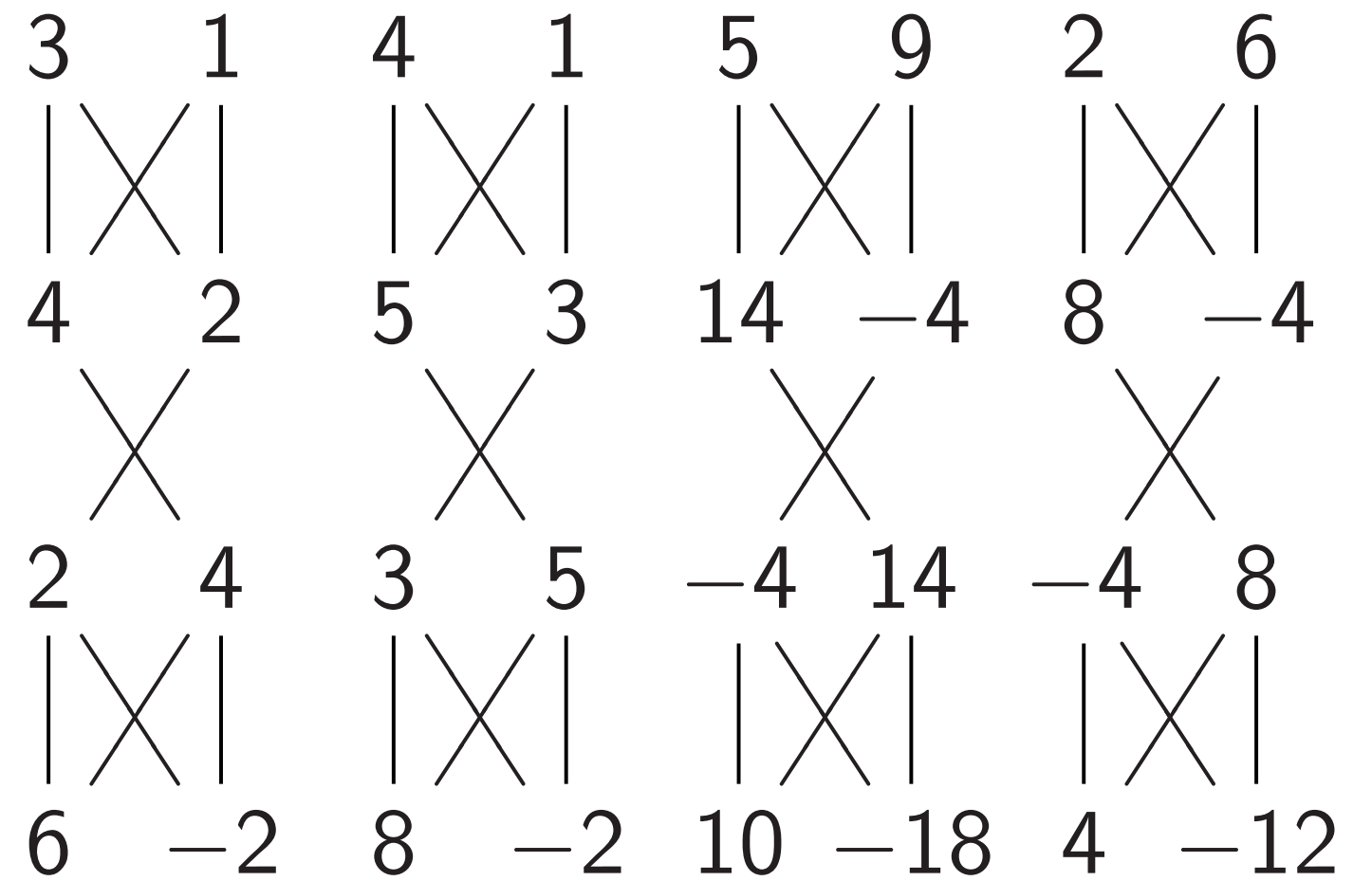
Hadamard gatesHadamard<sub>0</sub>:

$$(a, b) \mapsto (a + b, a - b).$$

Hadamard<sub>1</sub>:

$$(a, b, c, d) \mapsto$$

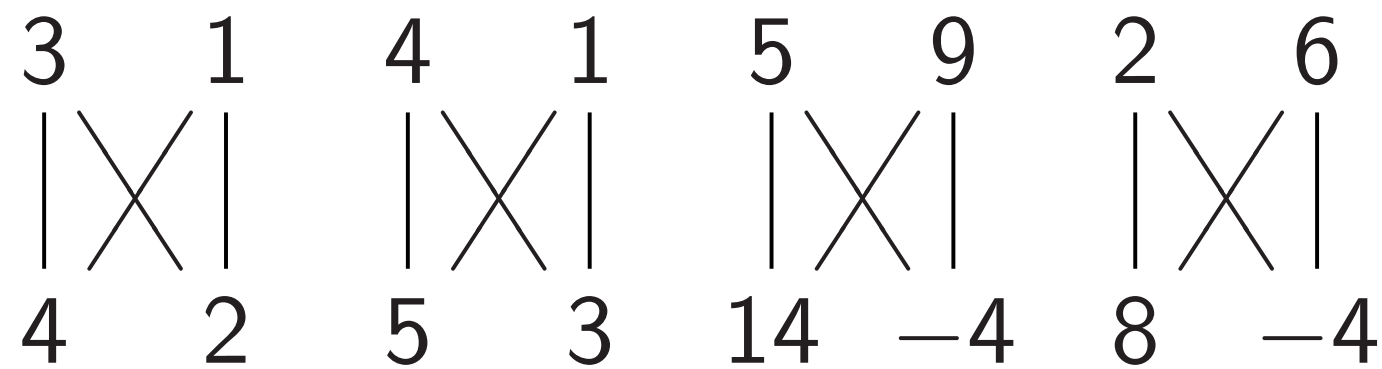
$$(a + c, b + d, a - c, b - d).$$

Some Hadamard applicationsHadamard<sub>0</sub>, NOT<sub>0</sub>, Hadamard<sub>0</sub>:

## Hadamard gates

Hadamard<sub>0</sub>:

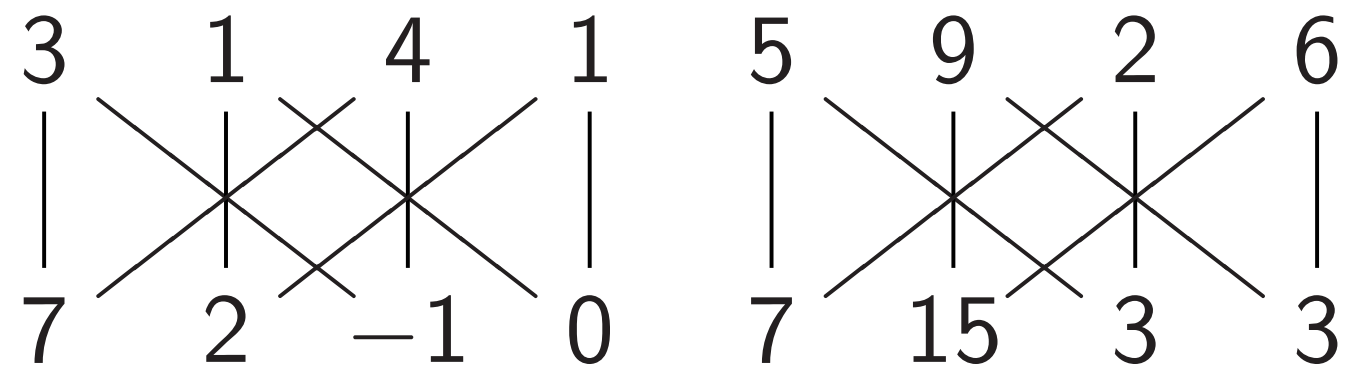
$$(a, b) \mapsto (a + b, a - b).$$



Hadamard<sub>1</sub>:

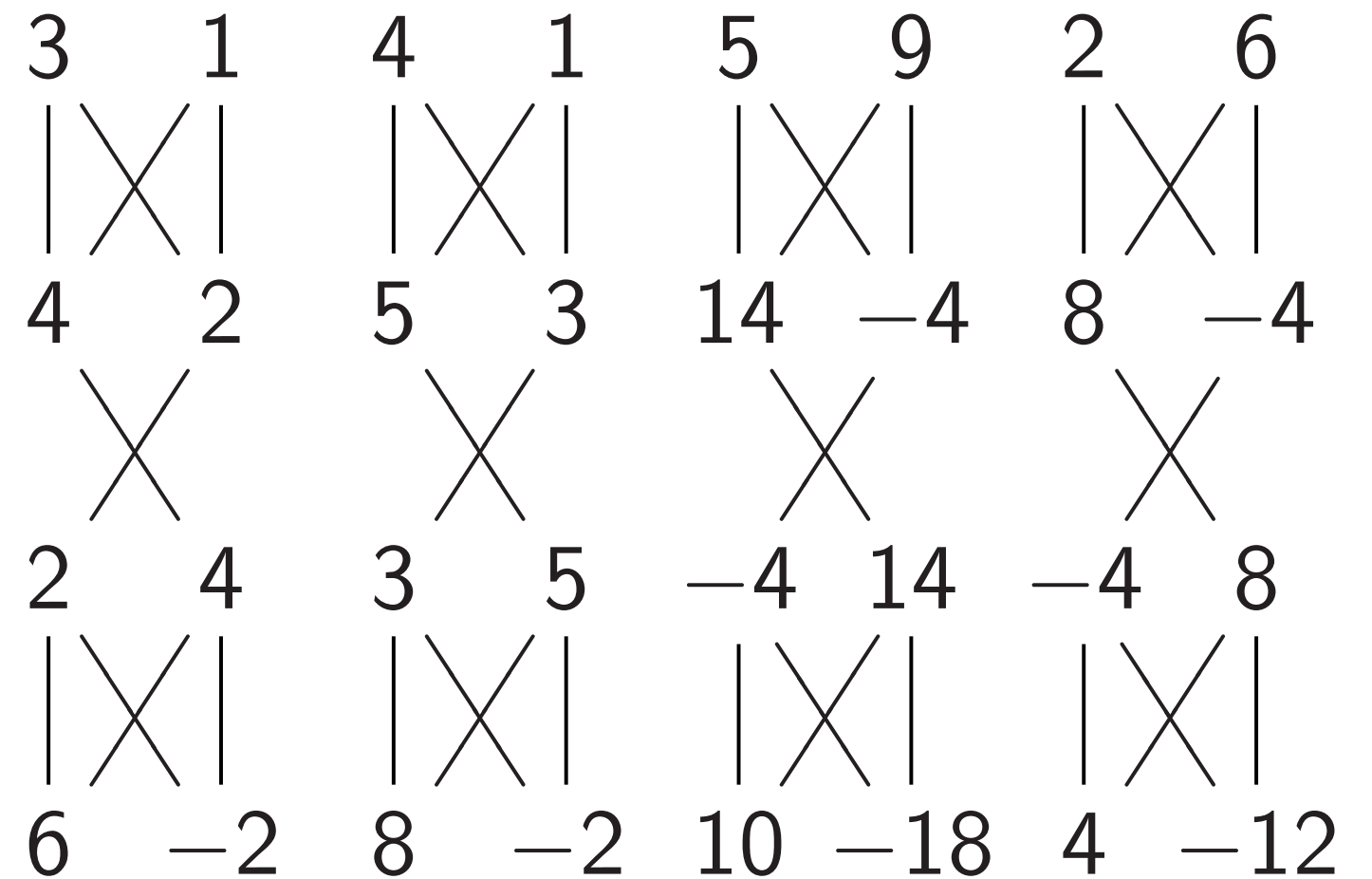
$$(a, b, c, d) \mapsto$$

$$(a + c, b + d, a - c, b - d).$$



## Some Hadamard applications

Hadamard<sub>0</sub>, NOT<sub>0</sub>, Hadamard<sub>0</sub>:

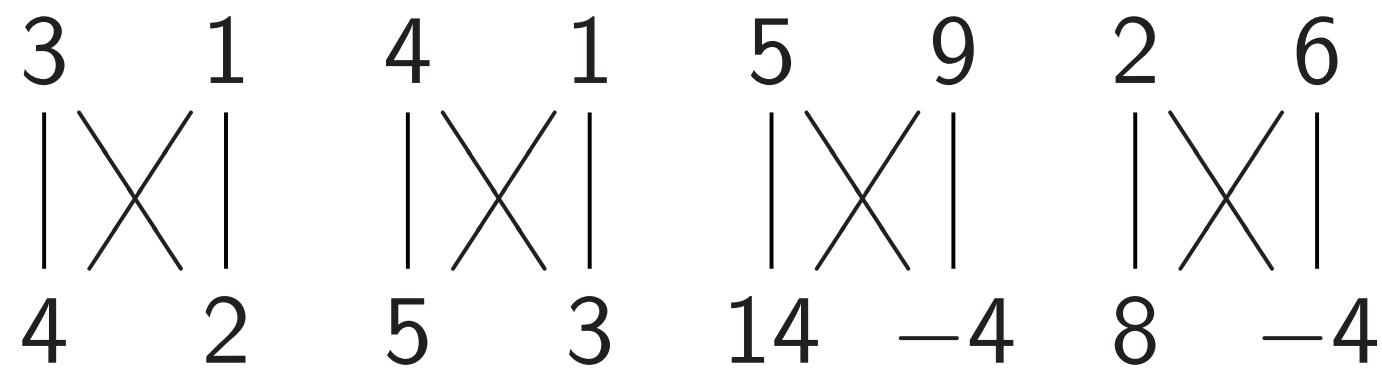


“Multiply each amplitude by 2.”  
This is not physically observable.

## Hadamard gates

Hadamard<sub>0</sub>:

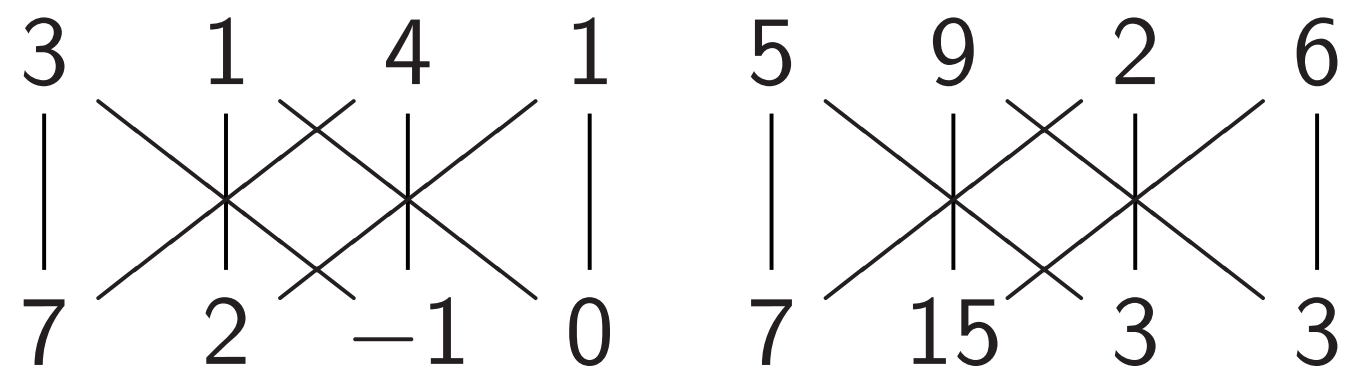
$$(a, b) \mapsto (a + b, a - b).$$



Hadamard<sub>1</sub>:

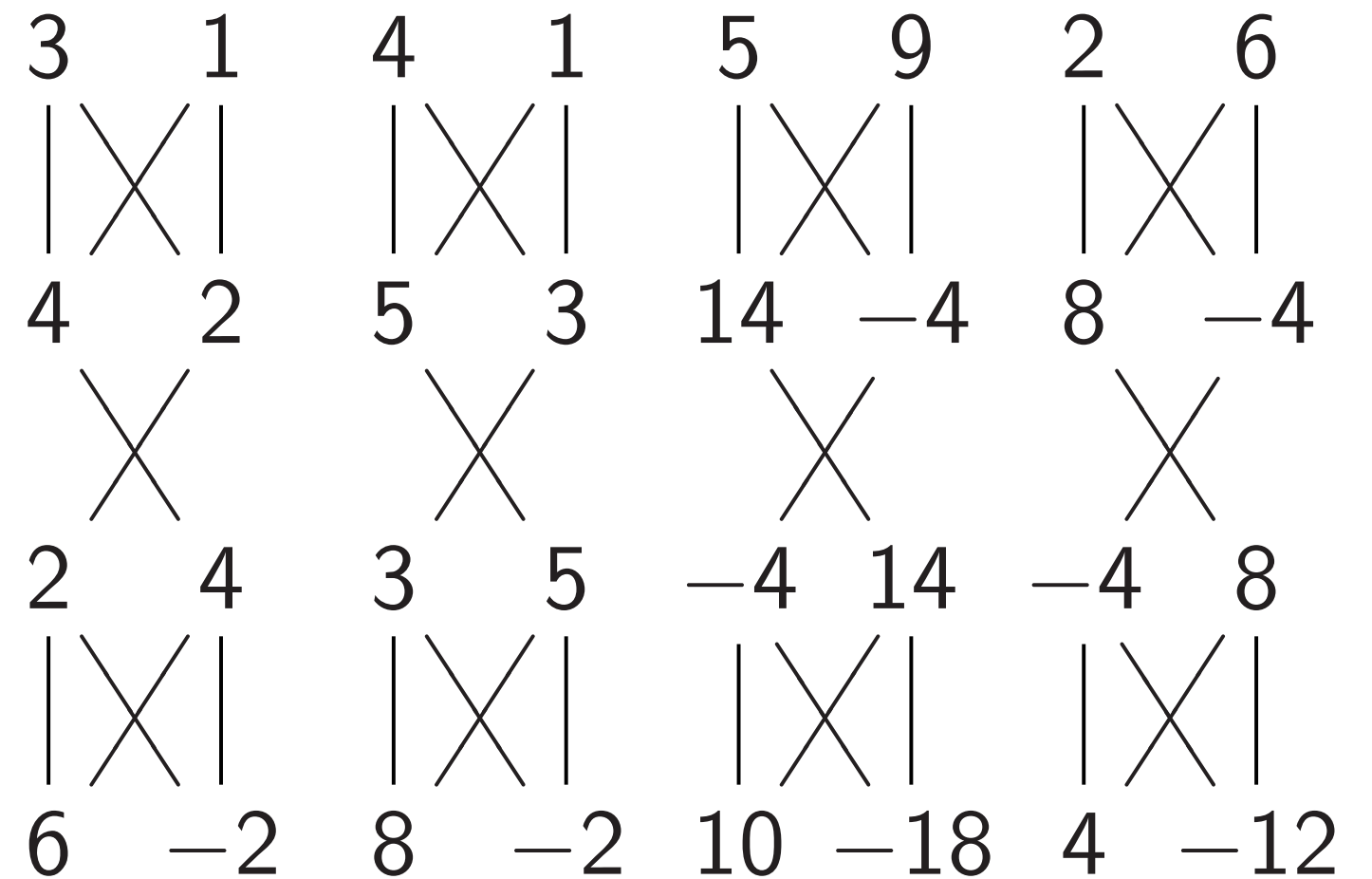
$$(a, b, c, d) \mapsto$$

$$(a + c, b + d, a - c, b - d).$$



## Some Hadamard applications

Hadamard<sub>0</sub>, NOT<sub>0</sub>, Hadamard<sub>0</sub>:



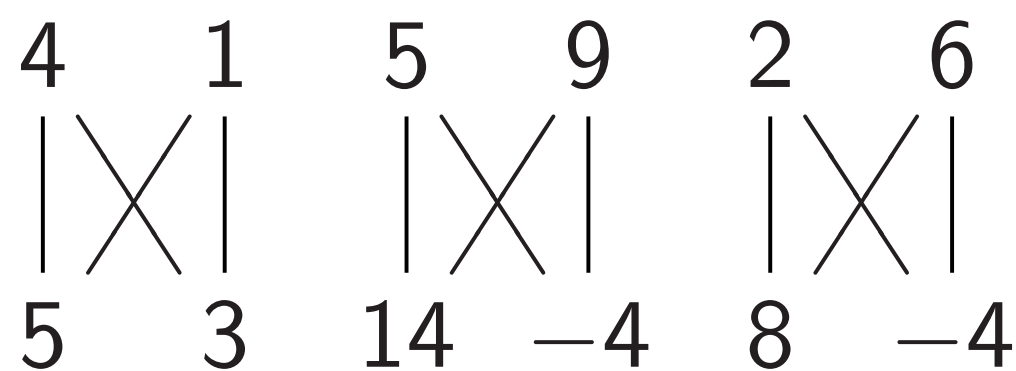
“Multiply each amplitude by 2.”  
This is not physically observable.

“Negate amplitude if  $q_0$  is set.”  
No effect on measuring *now*.

rd gates

rd<sub>0</sub>:

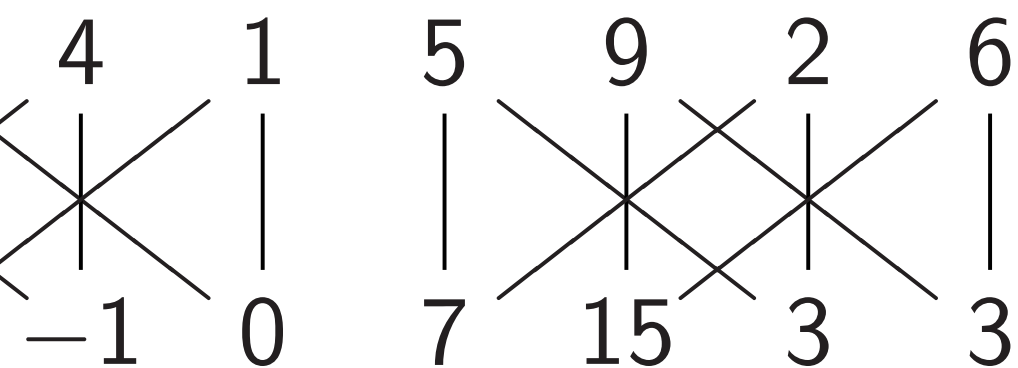
$(a + b, a - b)$ .



rd<sub>1</sub>:

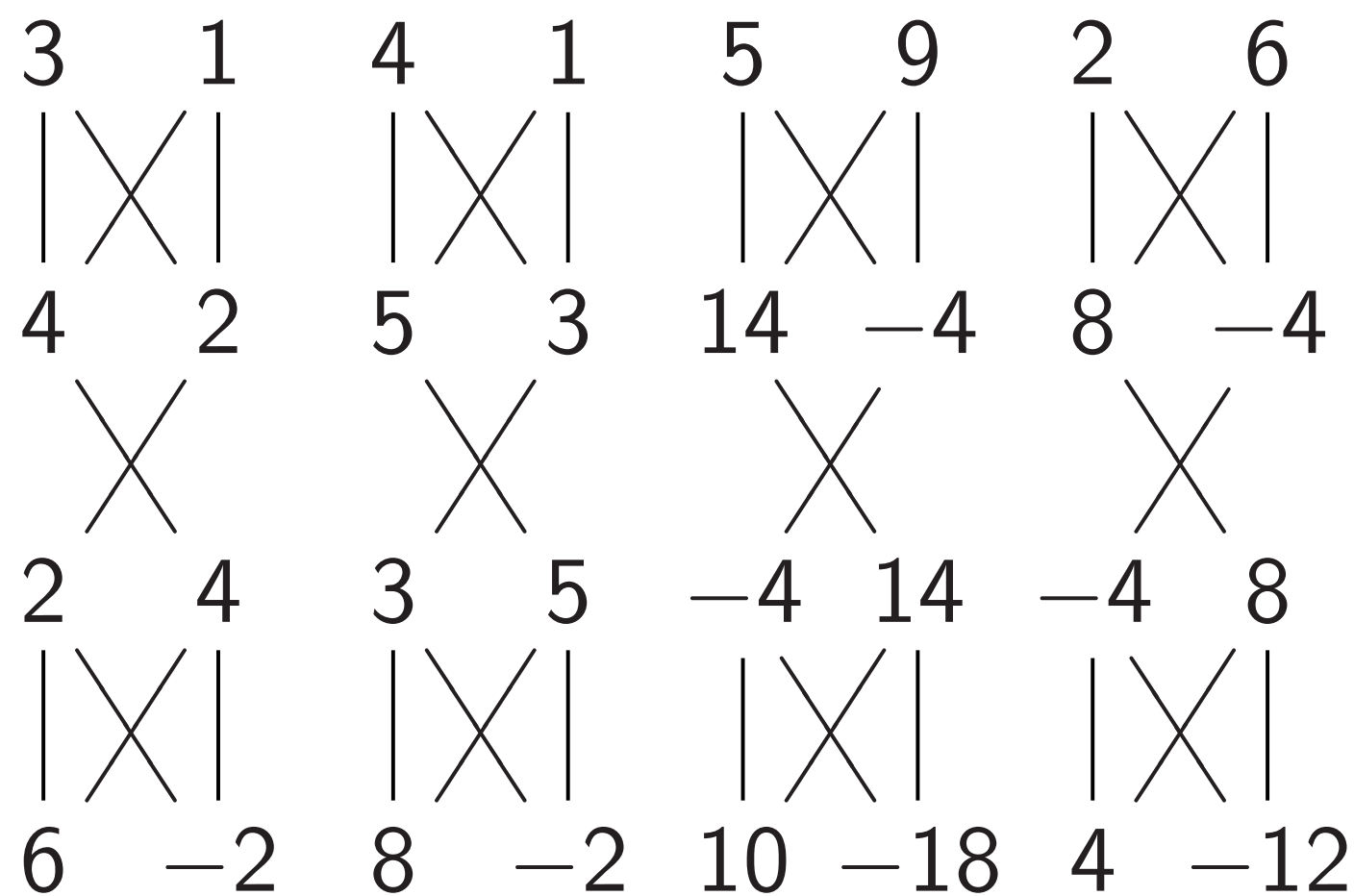
$d) \mapsto$

$(a + d, a - c, b - d)$ .



Some Hadamard applications

Hadamard<sub>0</sub>, NOT<sub>0</sub>, Hadamard<sub>0</sub>:



“Multiply each amplitude by 2.”

This is not physically observable.

“Negate amplitude if  $q_0$  is set.”

No effect on measuring *now*.

Fancier

“Negate

Assumes

$C_0C_1N$

Hadama

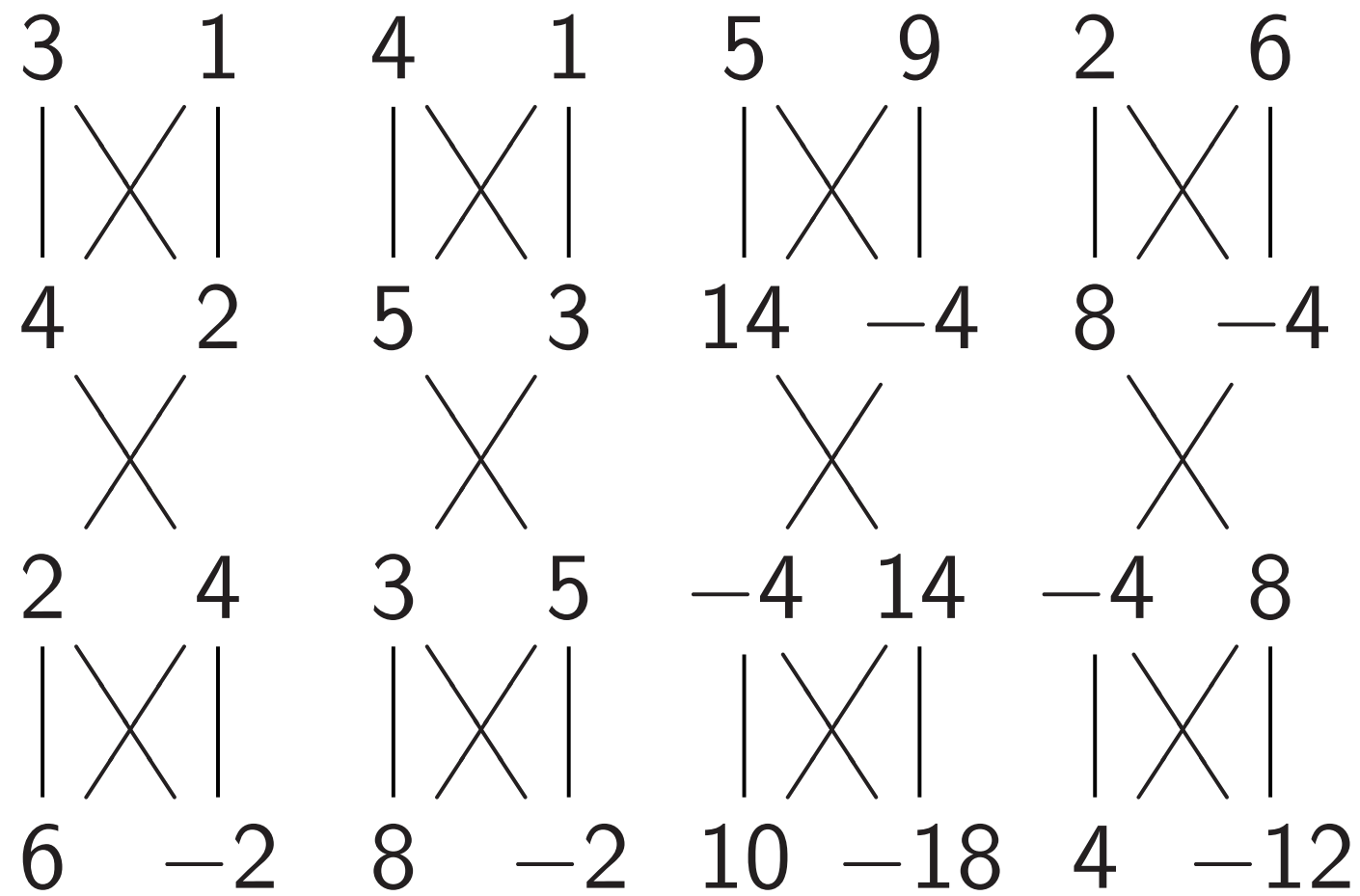
NOT

Hadama

$C_0C_1N$

## Some Hadamard applications

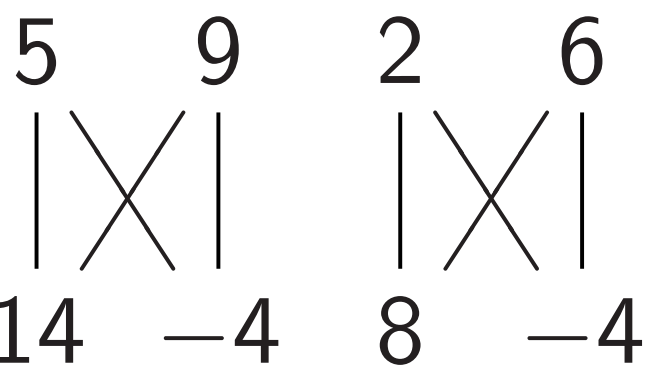
Hadamard<sub>0</sub>, NOT<sub>0</sub>, Hadamard<sub>0</sub>:



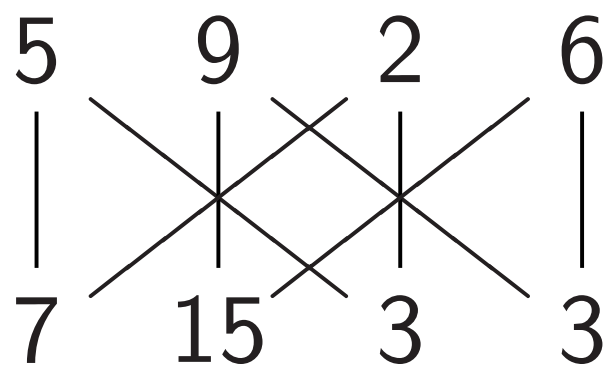
“Multiply each amplitude by 2.”  
This is not physically observable.

“Negate amplitude if  $q_0$  is set.”  
No effect on measuring *now*.

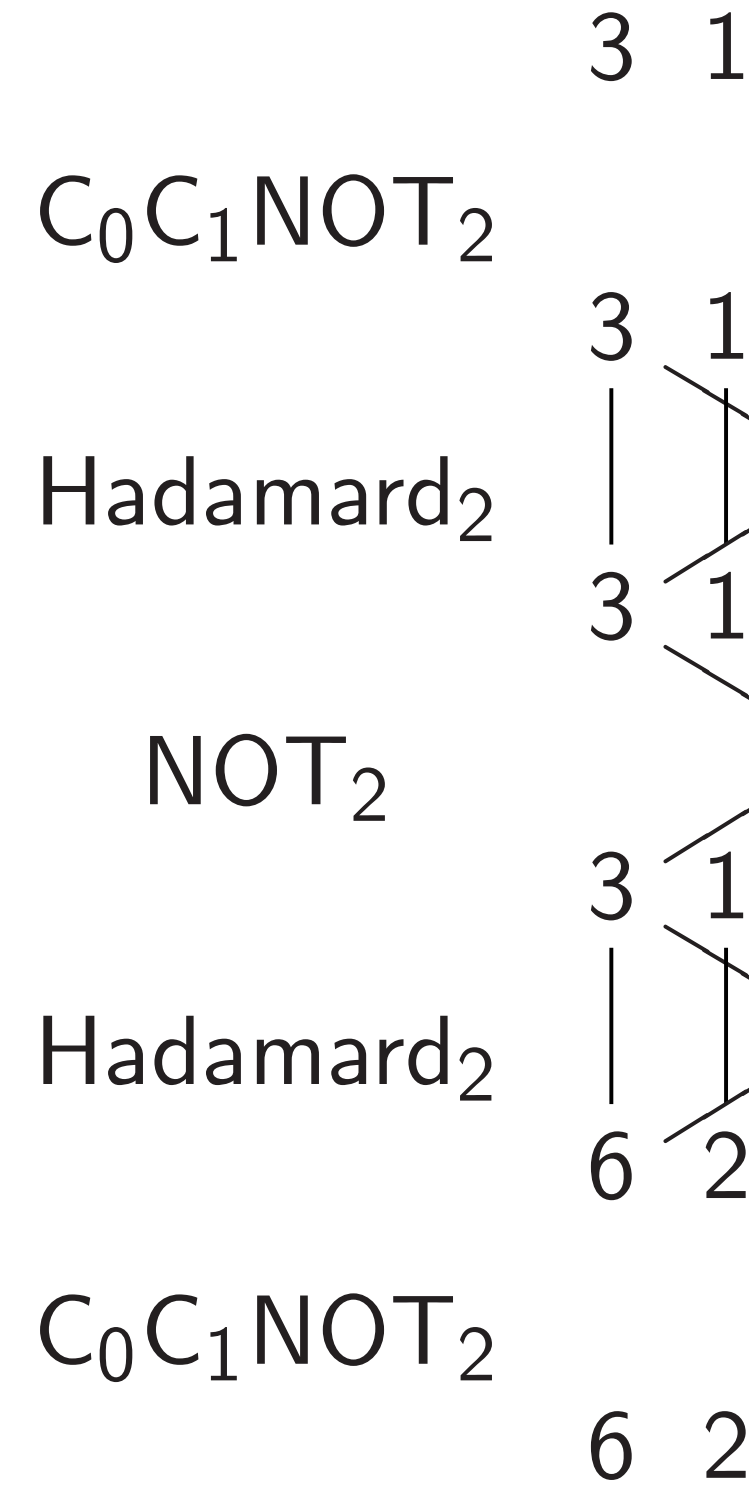
– *b*).



*c*, *b* – *d*).

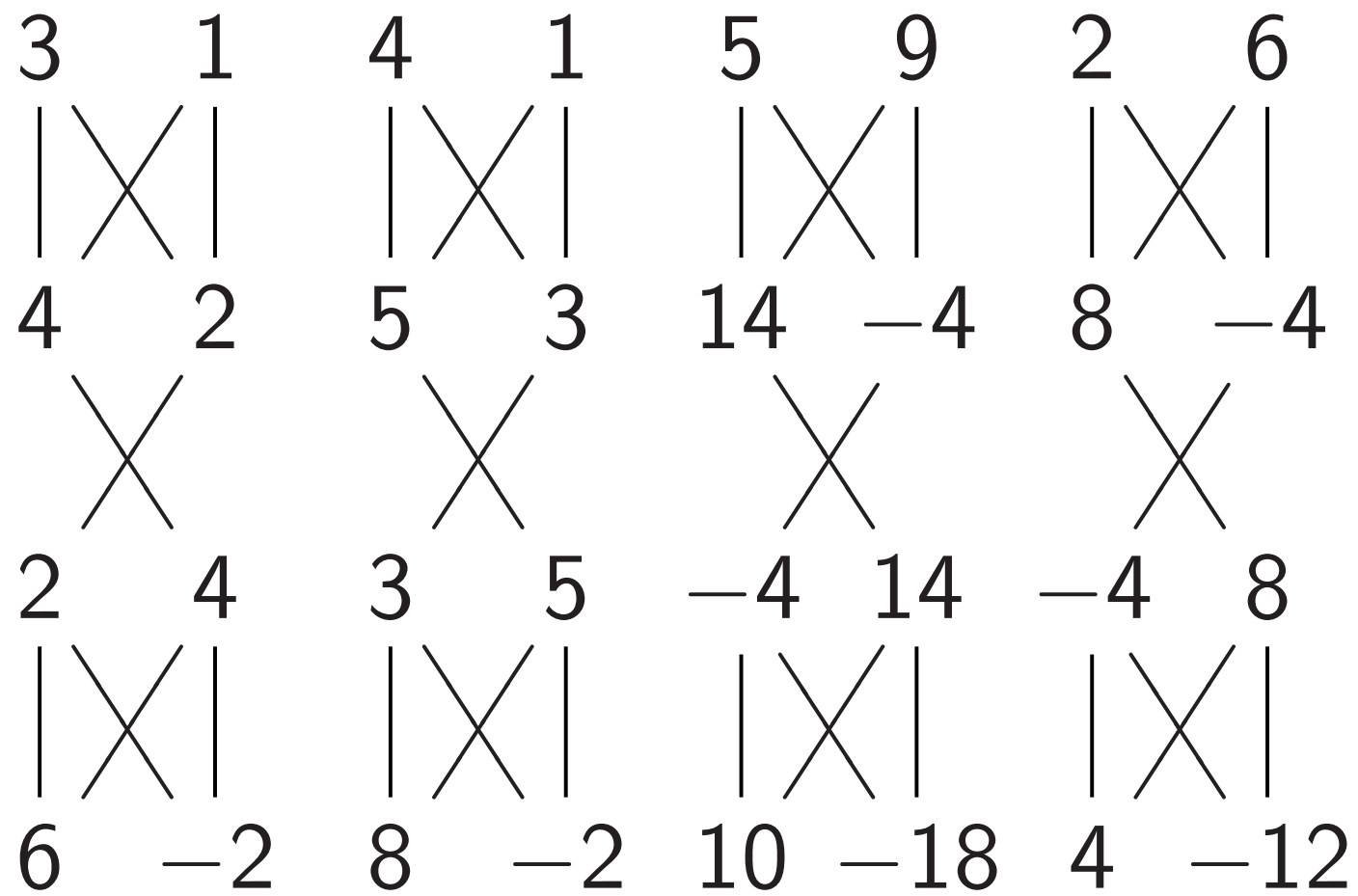


Fancier example:  
“Negate amplitude  
Assumes  $q_2 = 0$ :



# Some Hadamard applications

Hadamard<sub>0</sub>, NOT<sub>0</sub>, Hadamard<sub>0</sub>:

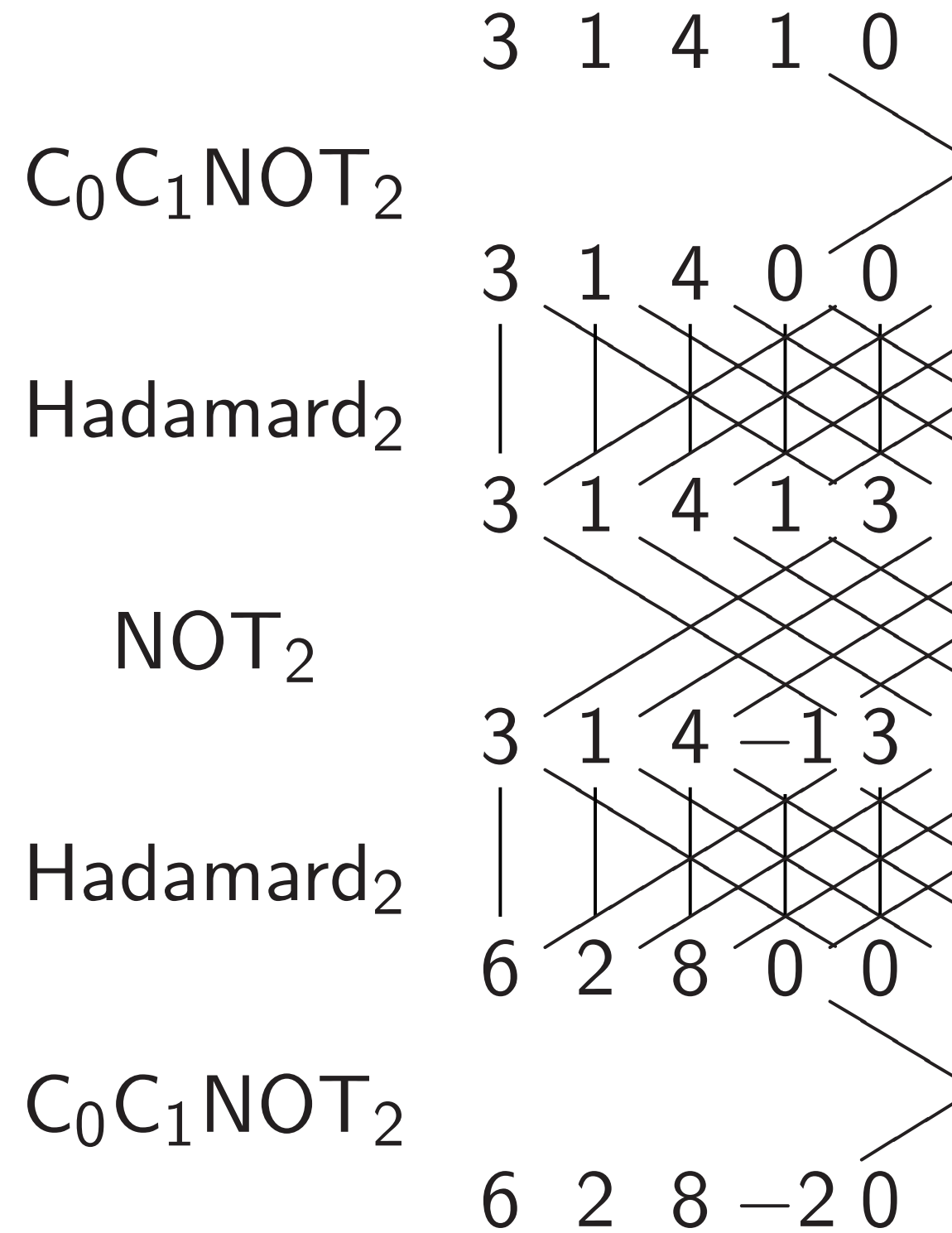


“Multiply each amplitude by 2.”  
This is not physically observable.

“Negate amplitude if  $q_0$  is set.”  
No effect on measuring *now*.

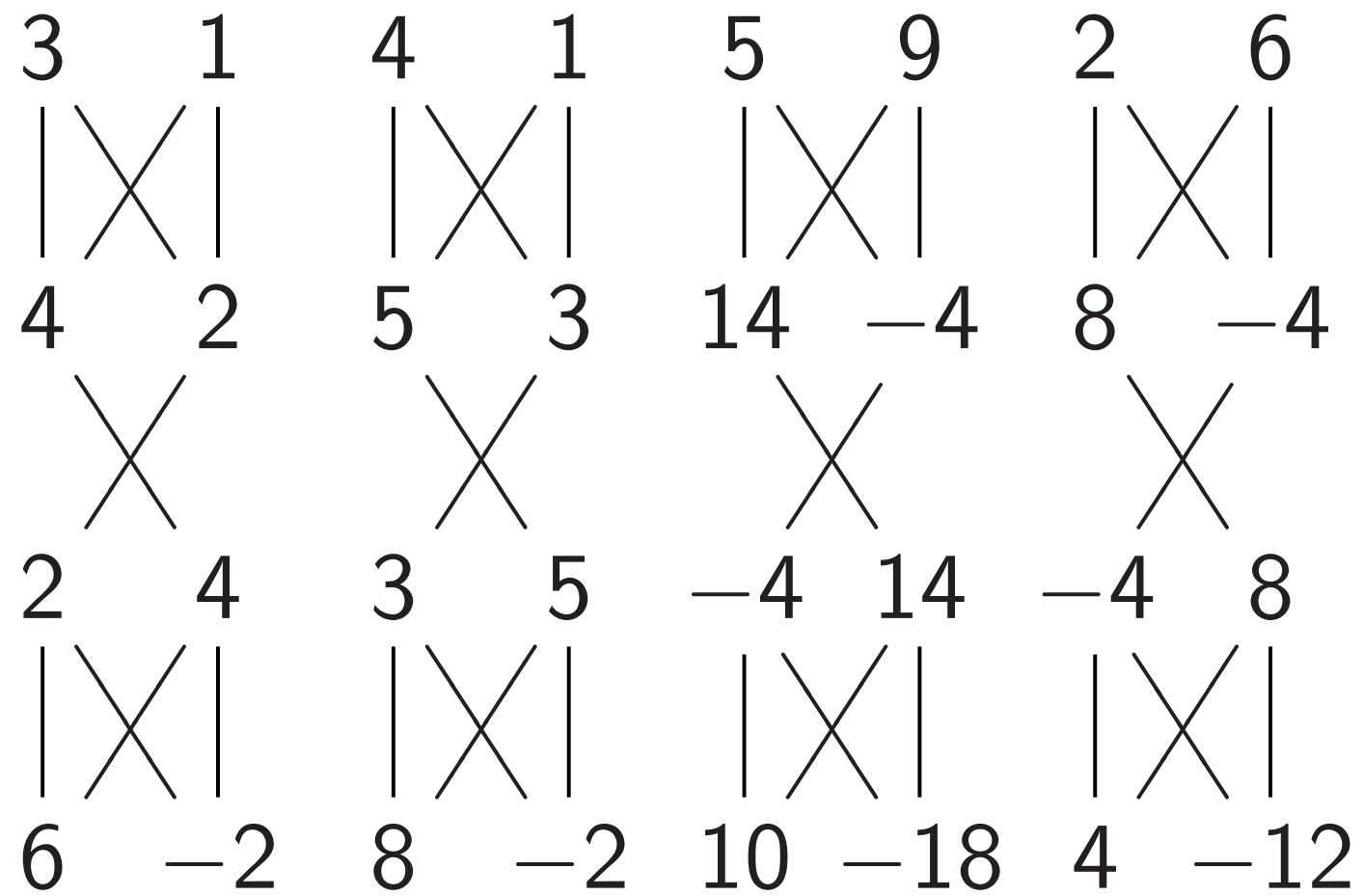
Fancier example:

“Negate amplitude if  $q_0 q_1$  is set.”  
Assumes  $q_2 = 0$ : “ancilla” qubit



## Some Hadamard applications

Hadamard<sub>0</sub>, NOT<sub>0</sub>, Hadamard<sub>0</sub>:

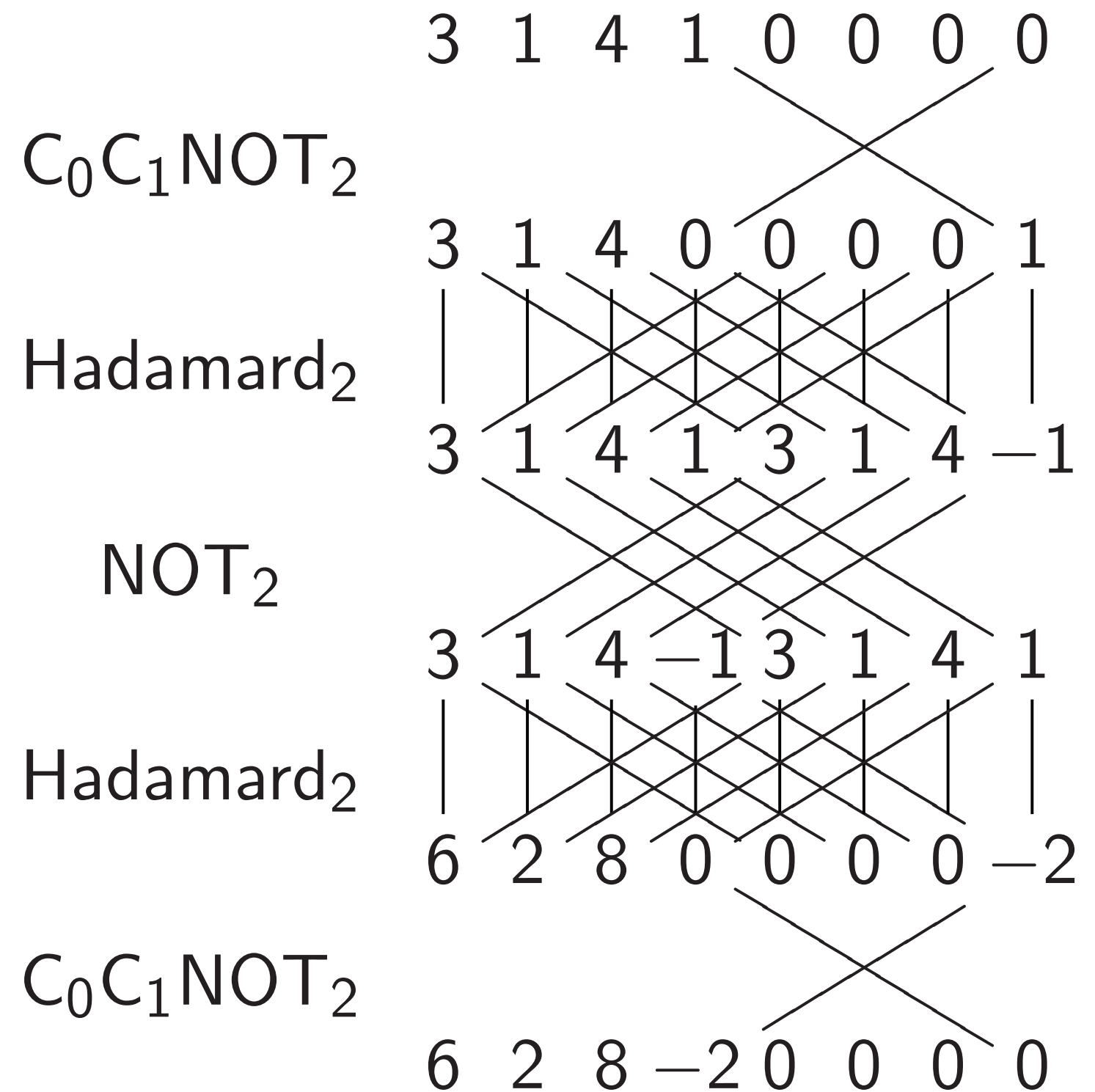


“Multiply each amplitude by 2.”  
This is not physically observable.

“Negate amplitude if  $q_0$  is set.”  
No effect on measuring *now*.

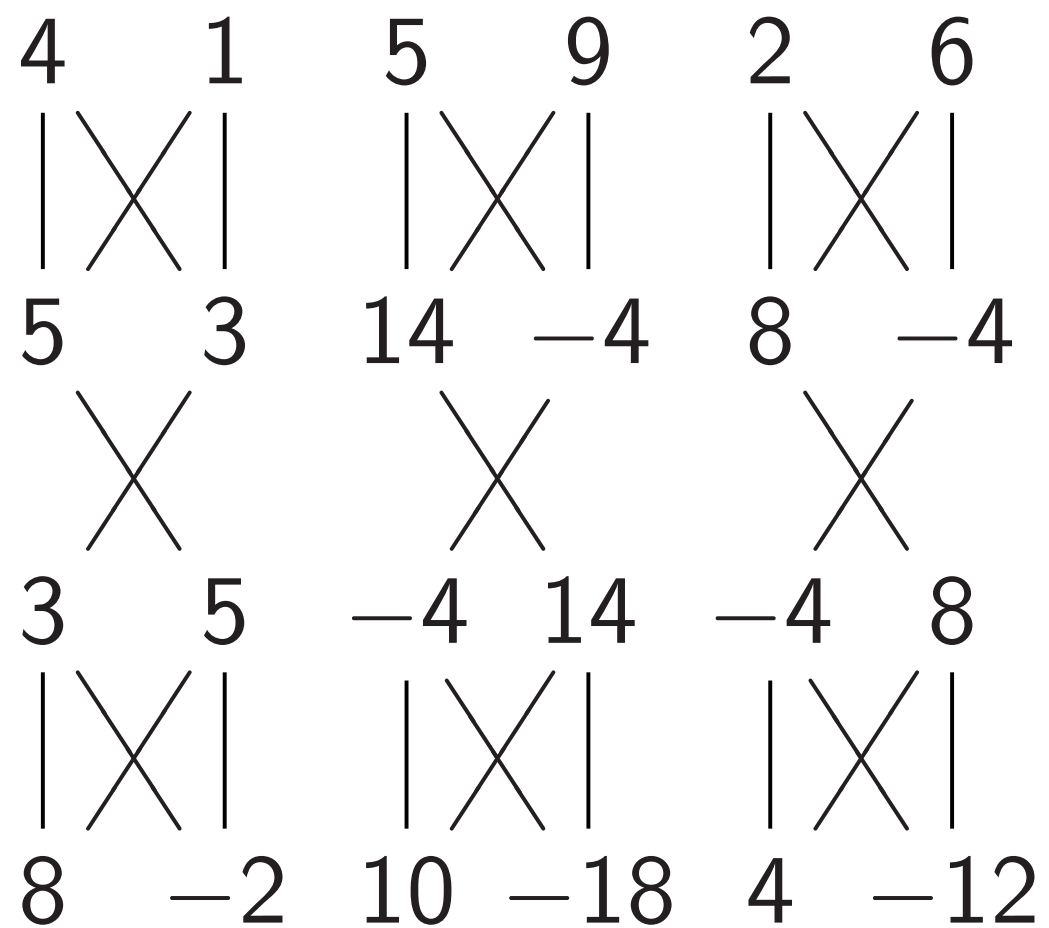
Fancier example:

“Negate amplitude if  $q_0q_1$  is set.”  
Assumes  $q_2 = 0$ : “ancilla” qubit.



## Hadamard applications

$\text{NOT}_0$ ,  $\text{NOT}_1$ ,  $\text{Hadamard}_0$ :



“Multiply each amplitude by 2.”  
 “Not physically observable.”

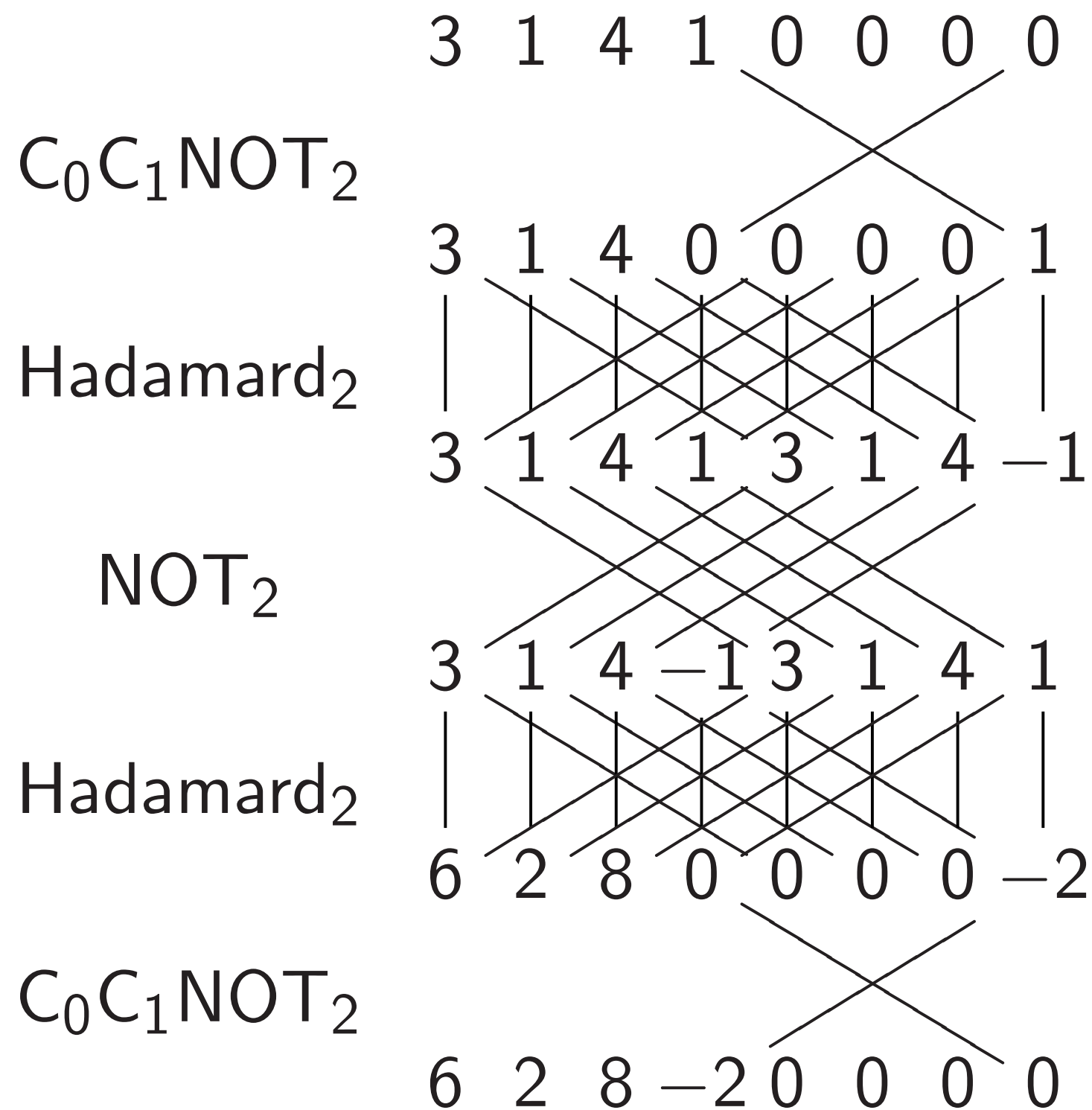
“Negate amplitude if  $q_0$  is set.”  
 “Start on measuring *now*.”

## Fancier example:

“Negate amplitude if  $q_0 q_1$  is set.”

Assumes  $q_2 = 0$ : “ancilla” qubit.

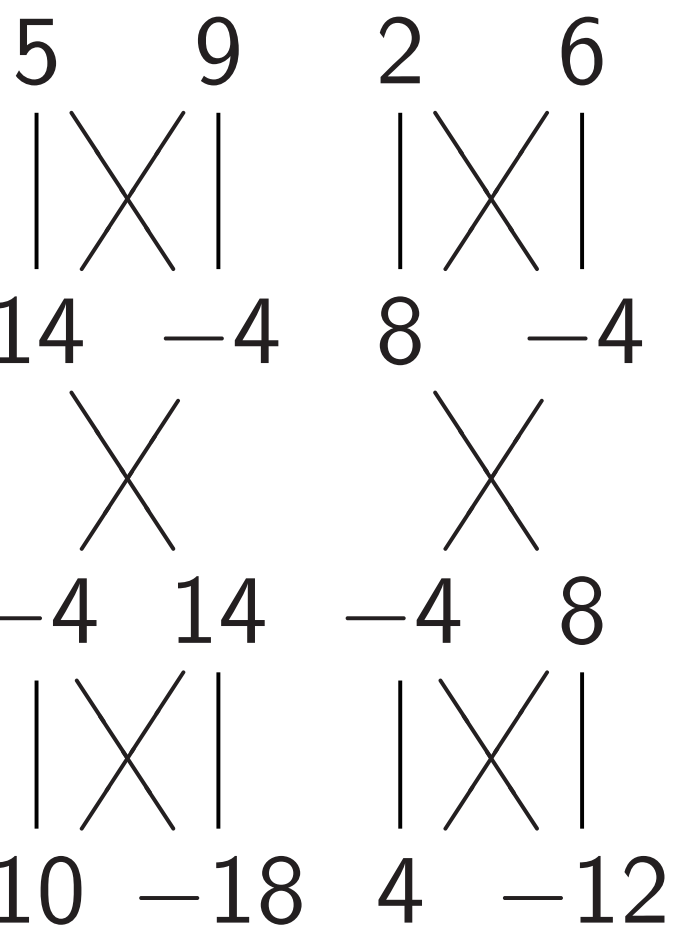
Affects  
 amplitude  
 (3, 1, 4, 1)





Applications

$C_0$ , Hadamard<sub>0</sub>:



Amplitude by 2."

ally observable.

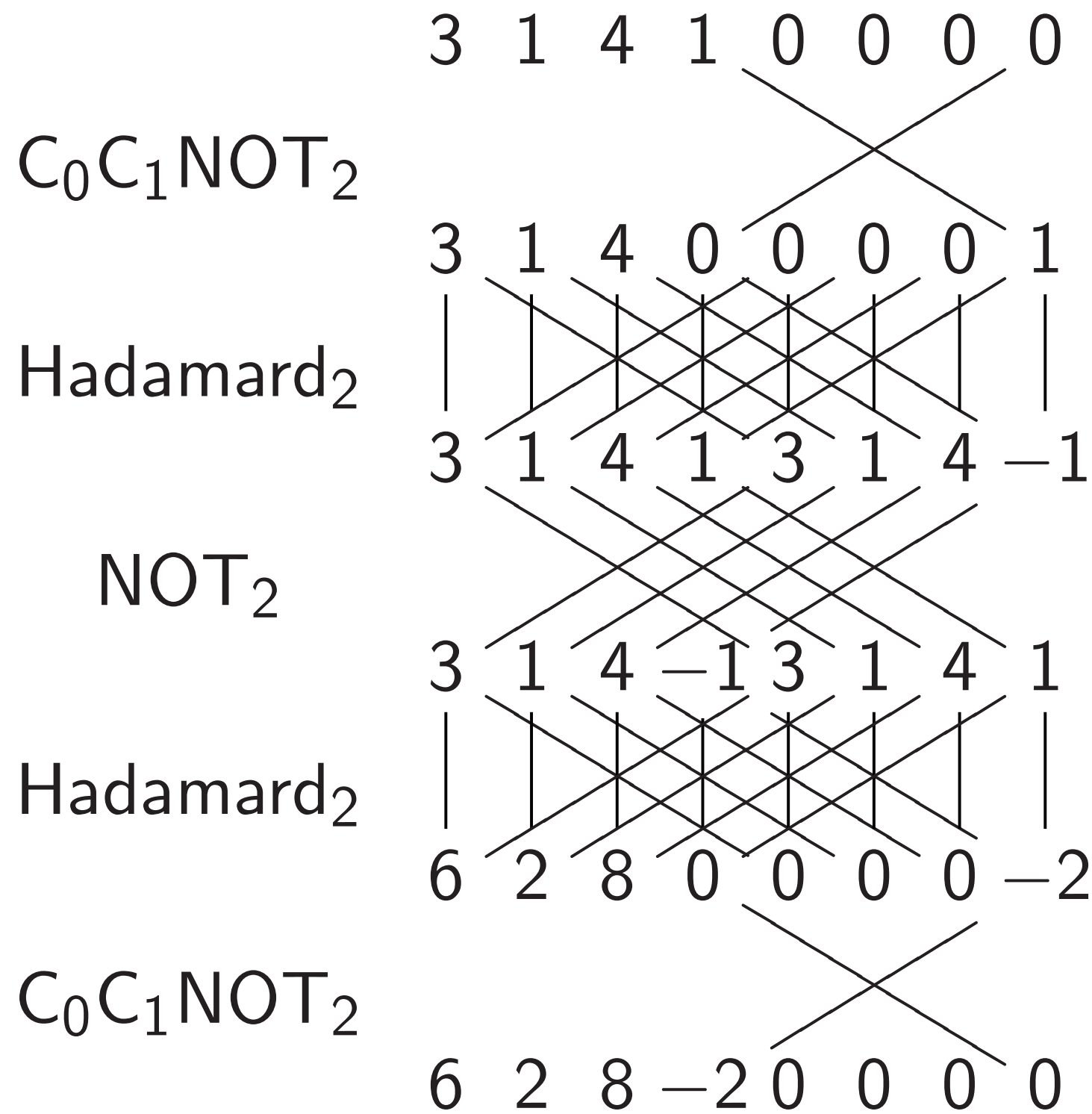
e if  $q_0$  is set."

uring *now*.

Fancier example:

"Negate amplitude if  $q_0q_1$  is set."

Assumes  $q_2 = 0$ : "ancilla" qubit.



Affects measurement

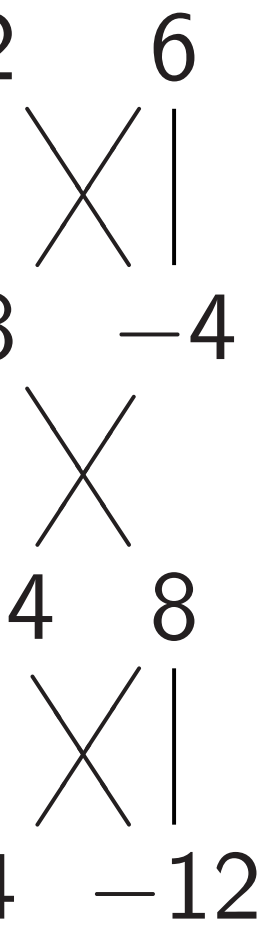
amplitude around

$(3, 1, 4, 1) \mapsto (1.5,$

Fancier example:

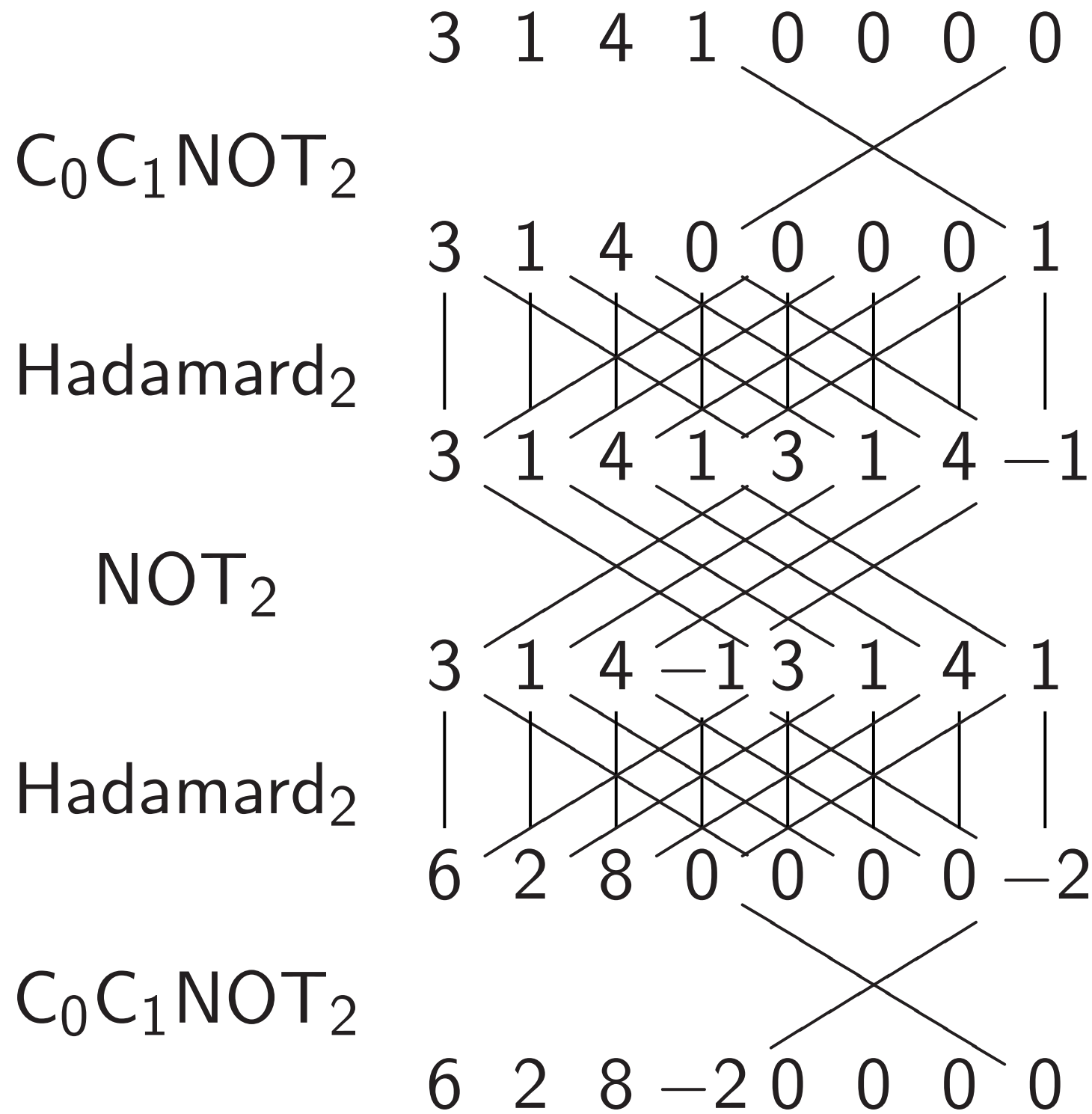
“Negate amplitude if  $q_0q_1$  is set.”

Assumes  $q_2 = 0$ : “ancilla” qubit.



by 2.”  
able.  
et.”  
.

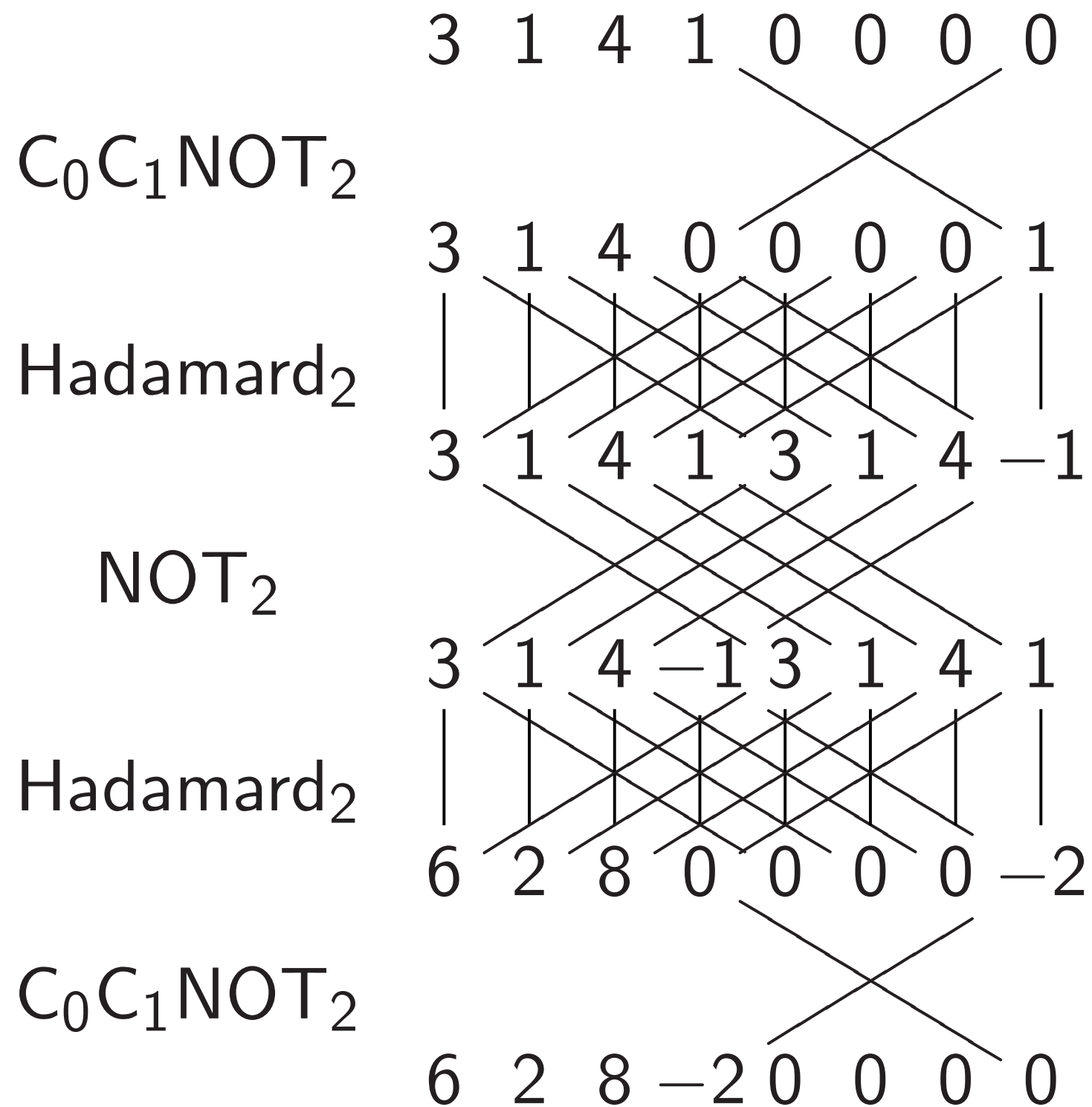
Affects measurements: “Negate amplitude around its average”  
 $(3, 1, 4, 1) \mapsto (1.5, 3.5, 0.5, 3)$



Fancier example:

“Negate amplitude if  $q_0q_1$  is set.”

Assumes  $q_2 = 0$ : “ancilla” qubit.



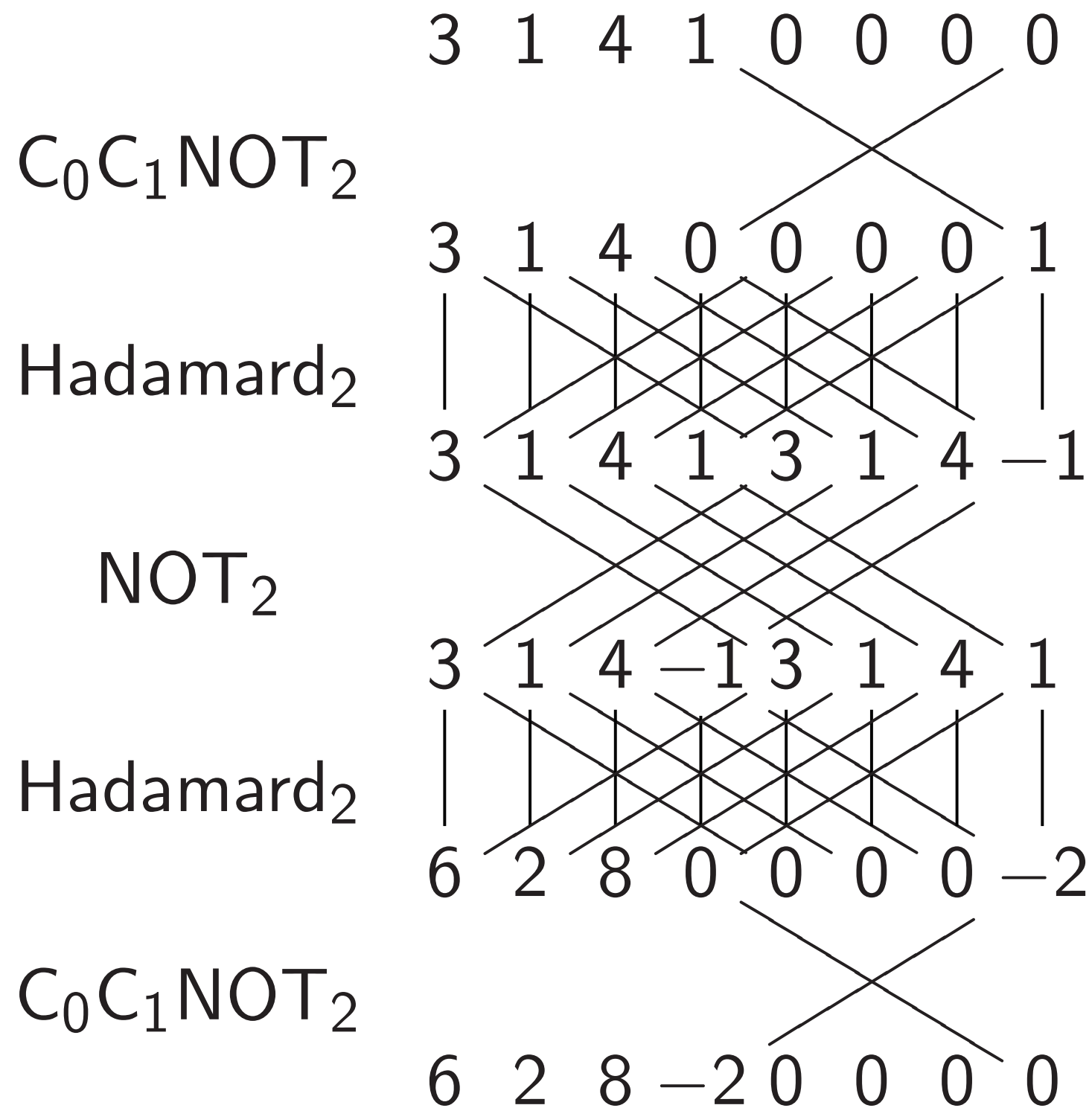
Affects measurements: “Negate amplitude around its average.”

$(3, 1, 4, 1) \mapsto (1.5, 3.5, 0.5, 3.5)$ .

Fancier example:

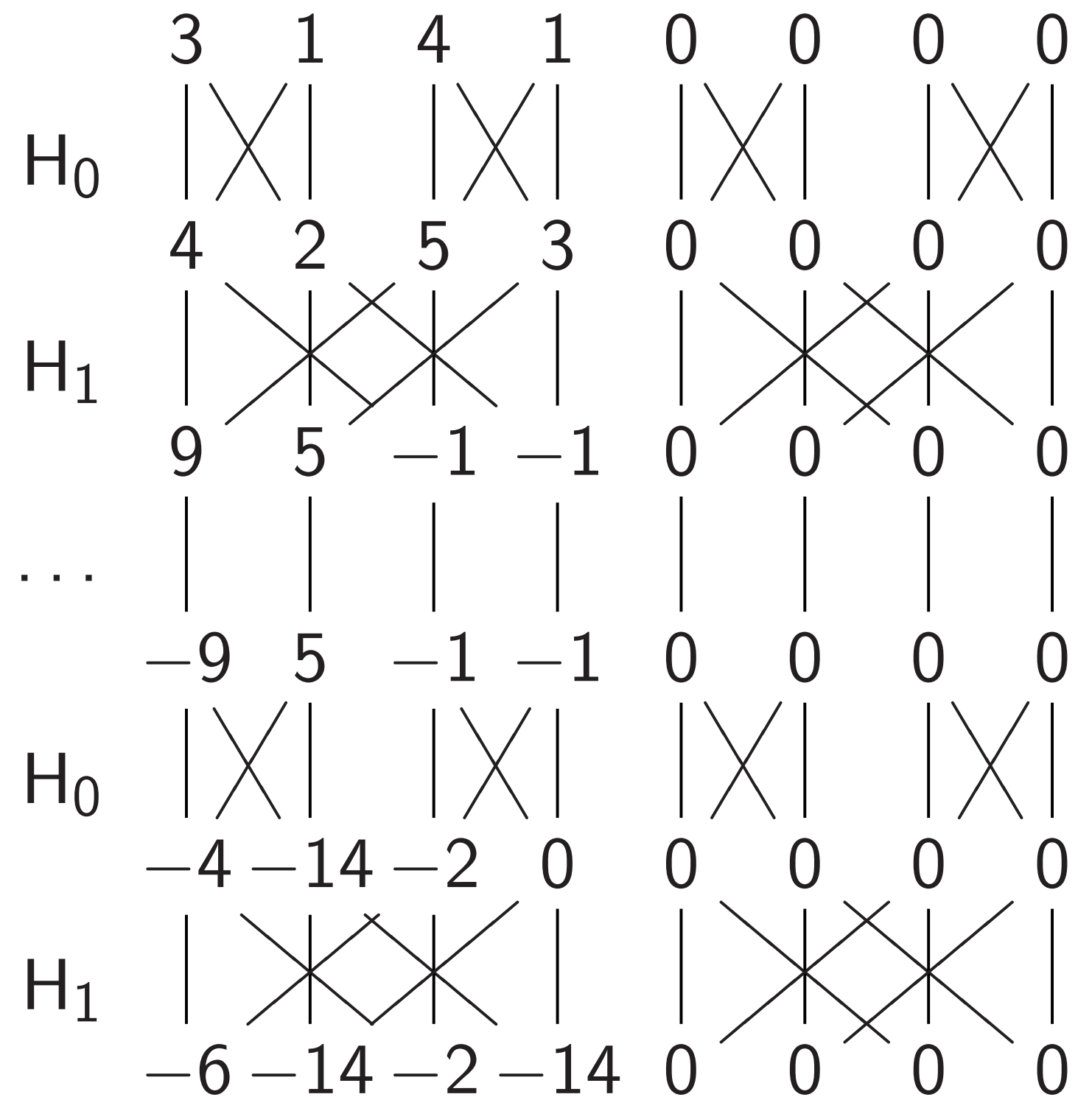
“Negate amplitude if  $q_0q_1$  is set.”

Assumes  $q_2 = 0$ : “ancilla” qubit.



Affects measurements: “Negate amplitude around its average.”

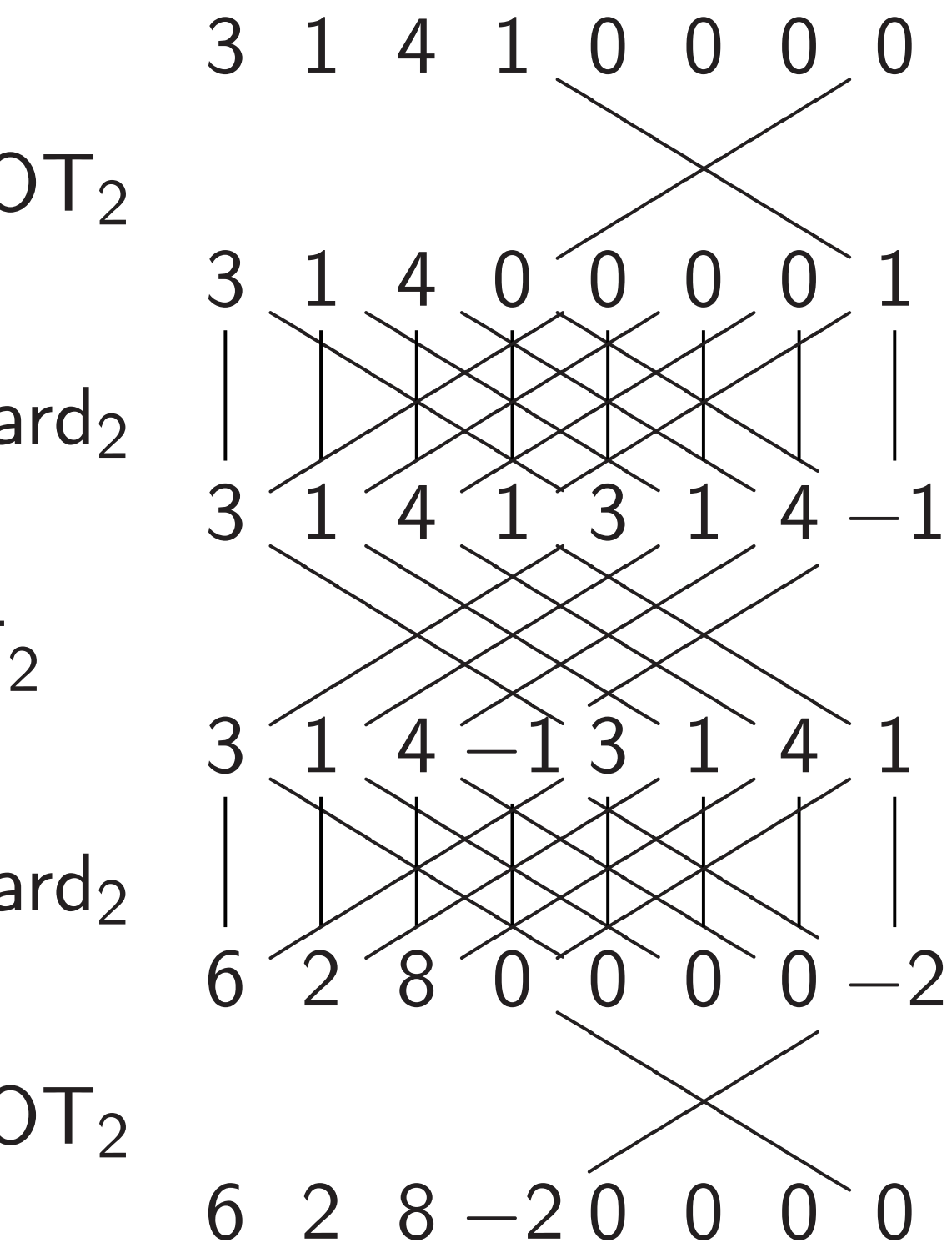
$(3, 1, 4, 1) \mapsto (1.5, 3.5, 0.5, 3.5)$ .



example:

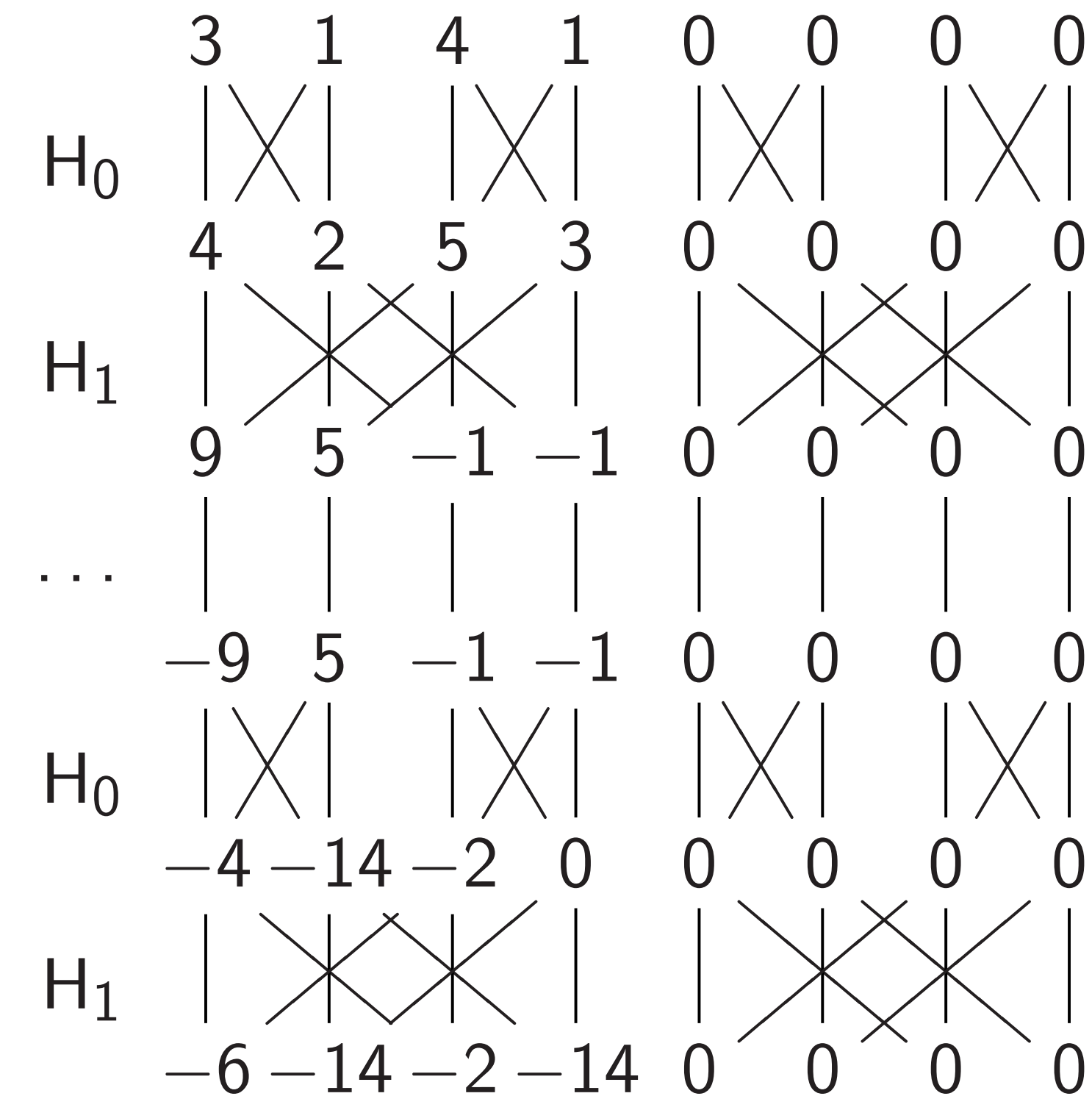
amplitude if  $q_0q_1$  is set."

$q_2 = 0$ : "ancilla" qubit.



Affects measurements: "Negate amplitude around its average."

$(3, 1, 4, 1) \mapsto (1.5, 3.5, 0.5, 3.5)$ .



Simon's

Step 1.

1, 0, 0,

0, 0, 0,

0, 0, 0,

0, 0, 0,

0, 0, 0,

0, 0, 0,

0, 0, 0,

0, 0, 0,







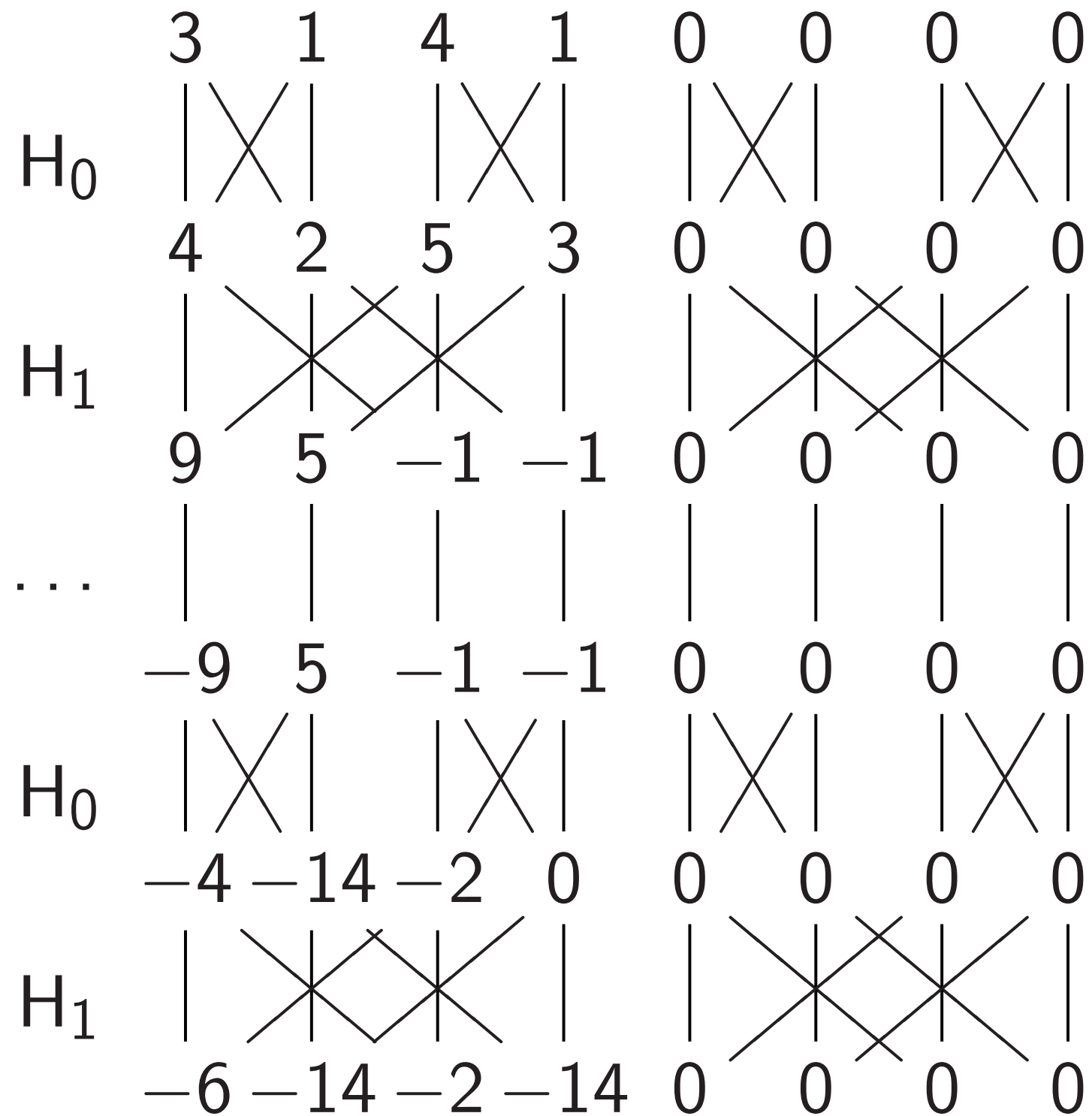






Affects measurements: “Negate amplitude around its average.”

$(3, 1, 4, 1) \mapsto (1.5, 3.5, 0.5, 3.5)$ .



## Simon's algorithm

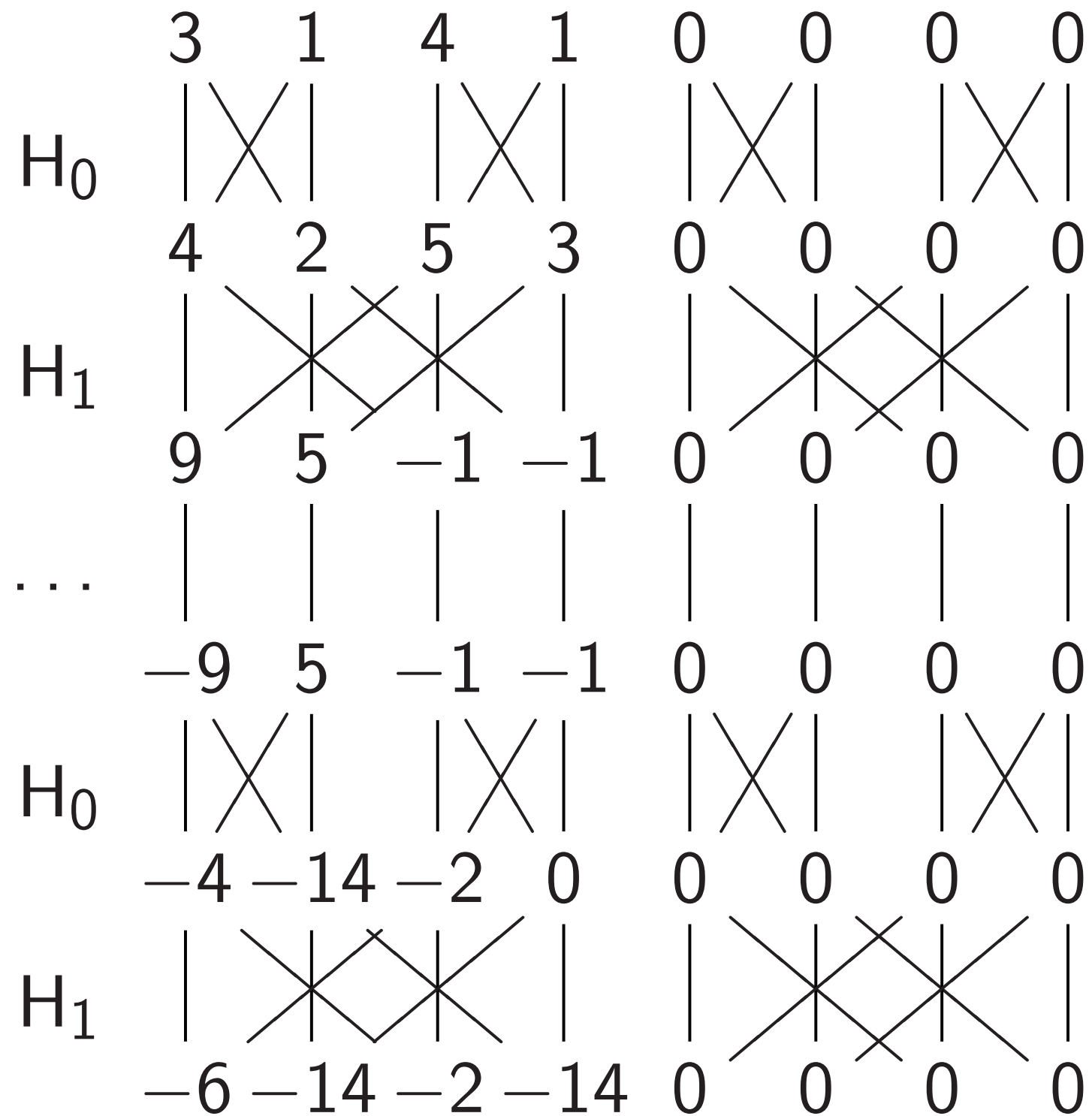
Step 4. Hadamard<sub>2</sub>:

**1, 1, 1, 1, 1, 1, 1, 1,**  
 0, 0, 0, 0, 0, 0, 0, 0,  
 0, 0, 0, 0, 0, 0, 0, 0,  
 0, 0, 0, 0, 0, 0, 0, 0,  
 0, 0, 0, 0, 0, 0, 0, 0,  
 0, 0, 0, 0, 0, 0, 0, 0,  
 0, 0, 0, 0, 0, 0, 0, 0,  
 0, 0, 0, 0, 0, 0, 0, 0.

Each column is a parallel universe.

Affects measurements: “Negate amplitude around its average.”

$(3, 1, 4, 1) \mapsto (1.5, 3.5, 0.5, 3.5)$ .



## Simon's algorithm

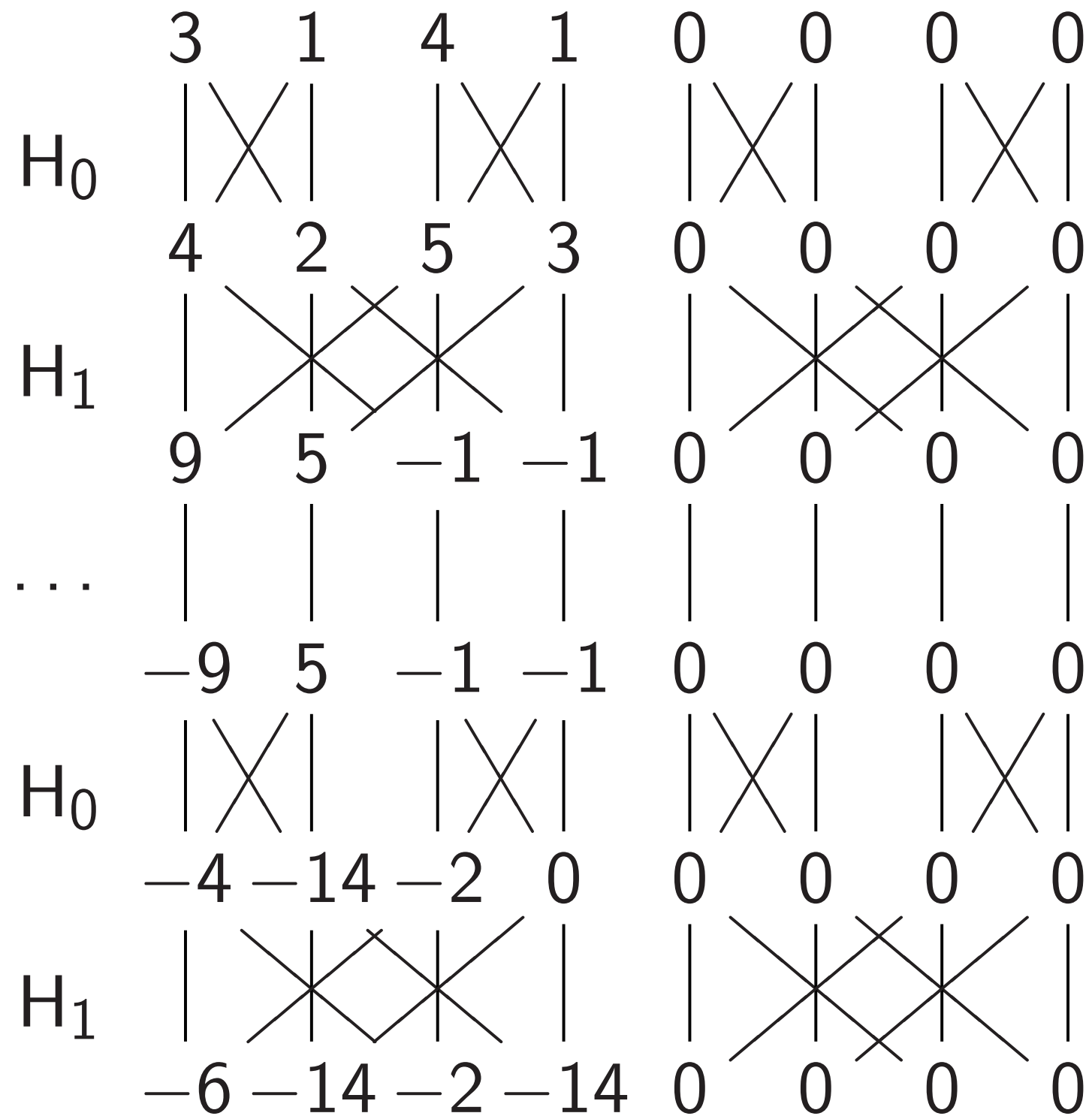
Step 5.  $C_0NOT_3$ :

$1, 0, 1, 0, 1, 0, 1, 0,$   
 $0, 1, 0, 1, 0, 1, 0, 1,$   
 $0, 0, 0, 0, 0, 0, 0, 0,$   
 $0, 0, 0, 0, 0, 0, 0, 0,$   
 $0, 0, 0, 0, 0, 0, 0, 0,$   
 $0, 0, 0, 0, 0, 0, 0, 0,$   
 $0, 0, 0, 0, 0, 0, 0, 0,$   
 $0, 0, 0, 0, 0, 0, 0, 0.$

Each column is a parallel universe performing its own computations.

Affects measurements: “Negate amplitude around its average.”

$$(3, 1, 4, 1) \mapsto (1.5, 3.5, 0.5, 3.5).$$



## Simon's algorithm

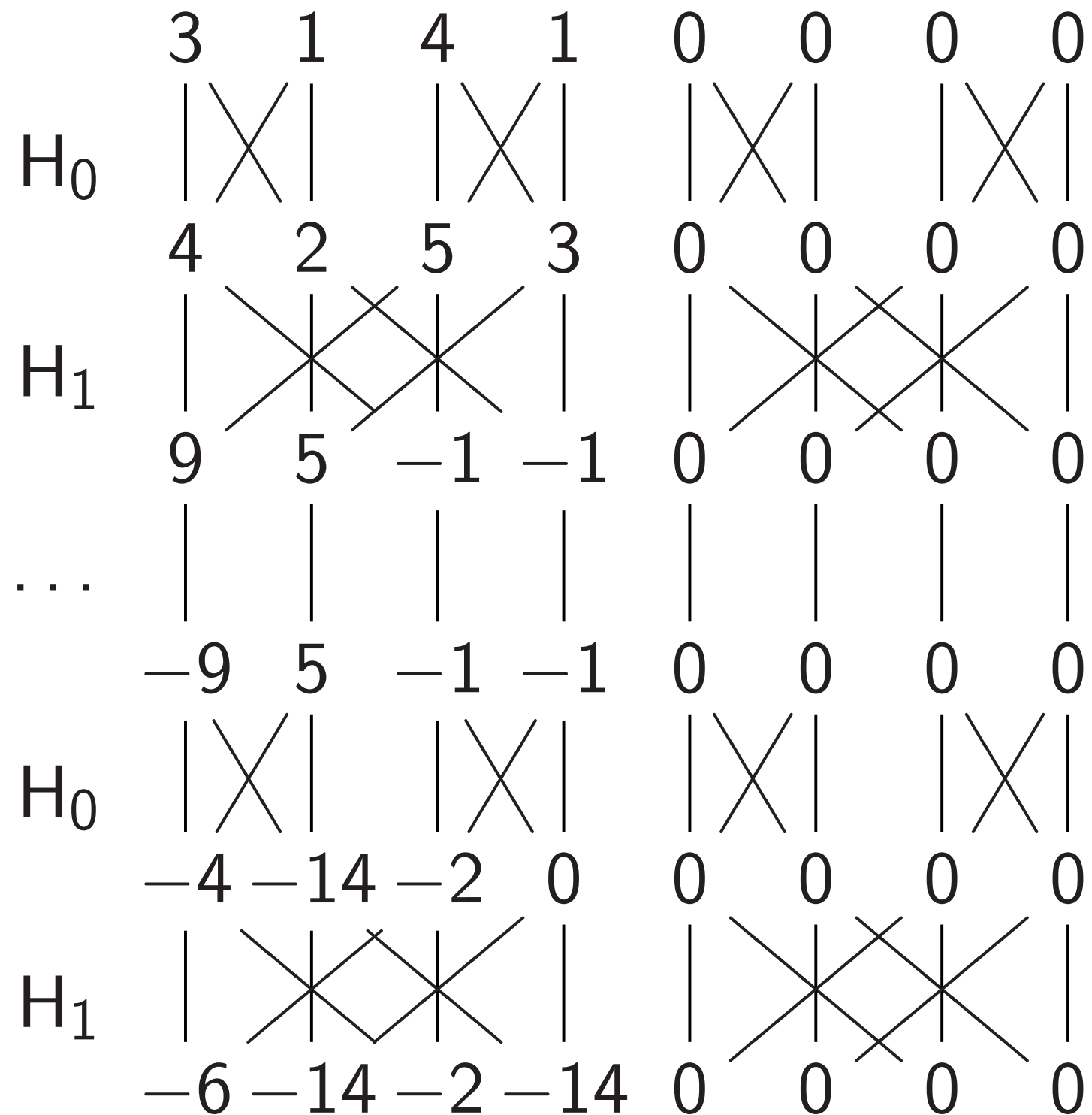
Step 5b. More shuffling:

$1, 0, 0, 0, 1, 0, 0, 0,$   
 $0, 1, 0, 0, 0, 1, 0, 0,$   
 $0, 0, 0, 0, 0, 0, 0, 0,$   
 $0, 0, 0, 0, 0, 0, 0, 0,$   
 $0, 0, 1, 0, 0, 0, 1, 0,$   
 $0, 0, 0, 1, 0, 0, 0, 1,$   
 $0, 0, 0, 0, 0, 0, 0, 0,$   
 $0, 0, 0, 0, 0, 0, 0, 0.$

Each column is a parallel universe performing its own computations.

Affects measurements: “Negate amplitude around its average.”

$$(3, 1, 4, 1) \mapsto (1.5, 3.5, 0.5, 3.5).$$



## Simon's algorithm

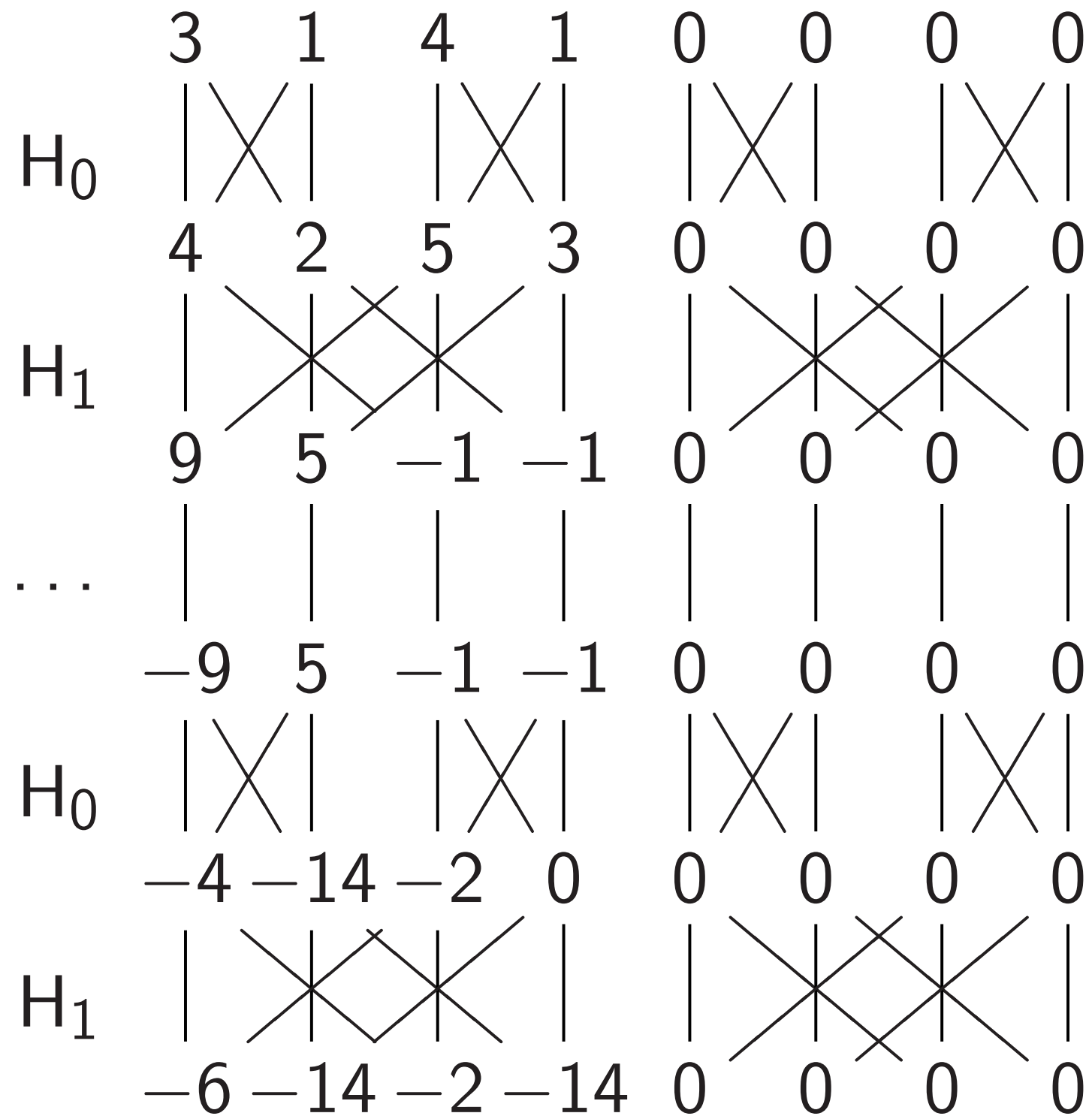
Step 5c. More shuffling:

1, 0, 0, 0, 0, 0, 0, 0,  
 0, 1, 0, 0, 0, 0, 0, 0,  
 0, 0, 0, 0, 1, 0, 0, 0,  
 0, 0, 0, 0, 0, 1, 0, 0,  
 0, 0, 1, 0, 0, 0, 0, 0,  
 0, 0, 0, 1, 0, 0, 0, 0,  
 0, 0, 0, 0, 0, 0, 1, 0,  
 0, 0, 0, 0, 0, 0, 0, 1.

Each column is a parallel universe performing its own computations.

Affects measurements: “Negate amplitude around its average.”

$$(3, 1, 4, 1) \mapsto (1.5, 3.5, 0.5, 3.5).$$



## Simon's algorithm

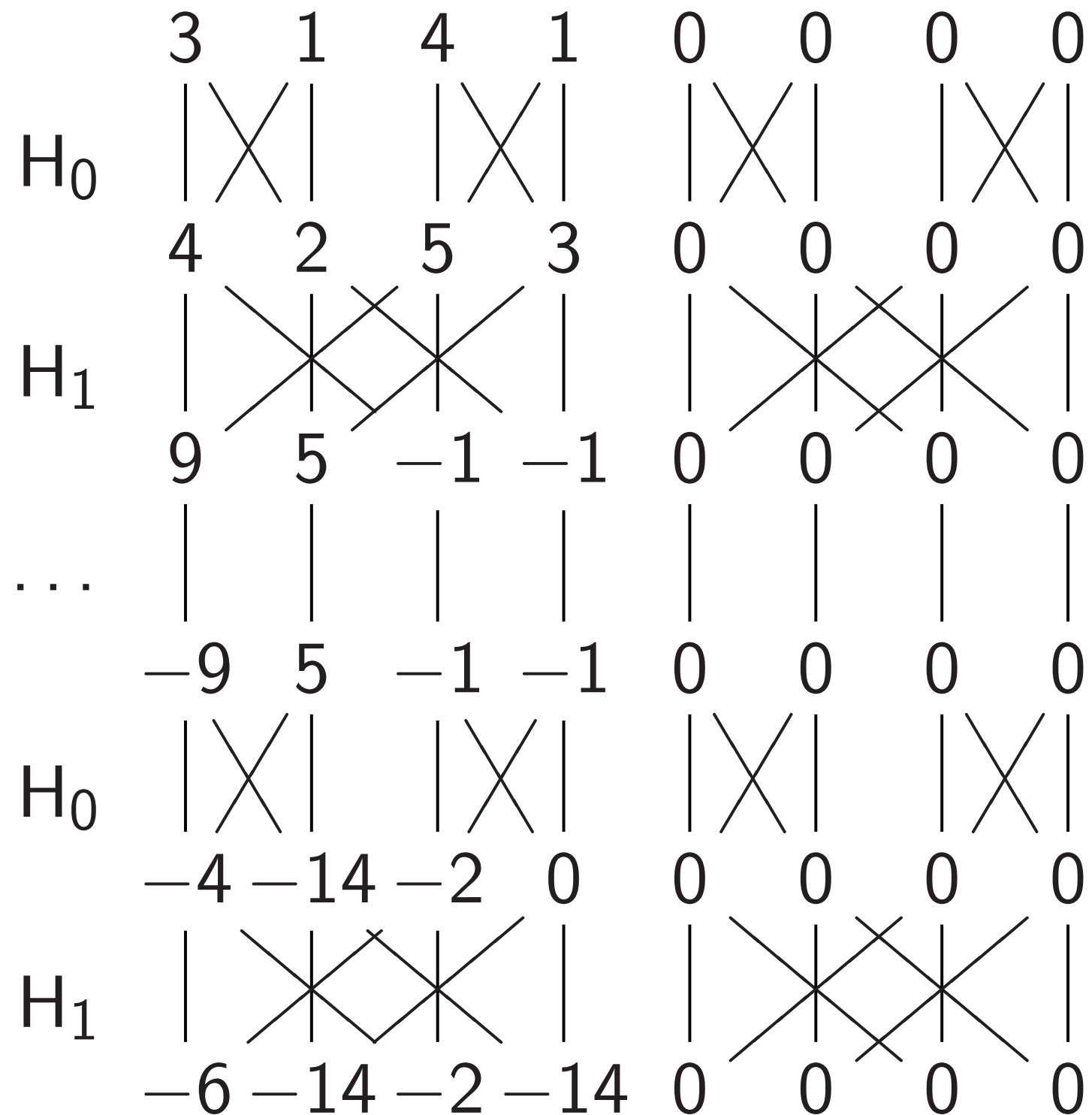
Step 5d. More shuffling:

1, 0, 0, 0, 0, 0, 0, 0,  
 0, 0, 0, 0, 0, 1, 0, 0,  
 0, 0, 0, 0, 1, 0, 0, 0,  
 0, 1, 0, 0, 0, 0, 0, 0,  
 0, 0, 1, 0, 0, 0, 0, 0,  
 0, 0, 0, 0, 0, 0, 0, 1,  
 0, 0, 0, 0, 0, 0, 1, 0,  
 0, 0, 0, 1, 0, 0, 0, 0.

Each column is a parallel universe performing its own computations.

Affects measurements: “Negate amplitude around its average.”

$$(3, 1, 4, 1) \mapsto (1.5, 3.5, 0.5, 3.5).$$



## Simon's algorithm

Step 5e. More shuffling:

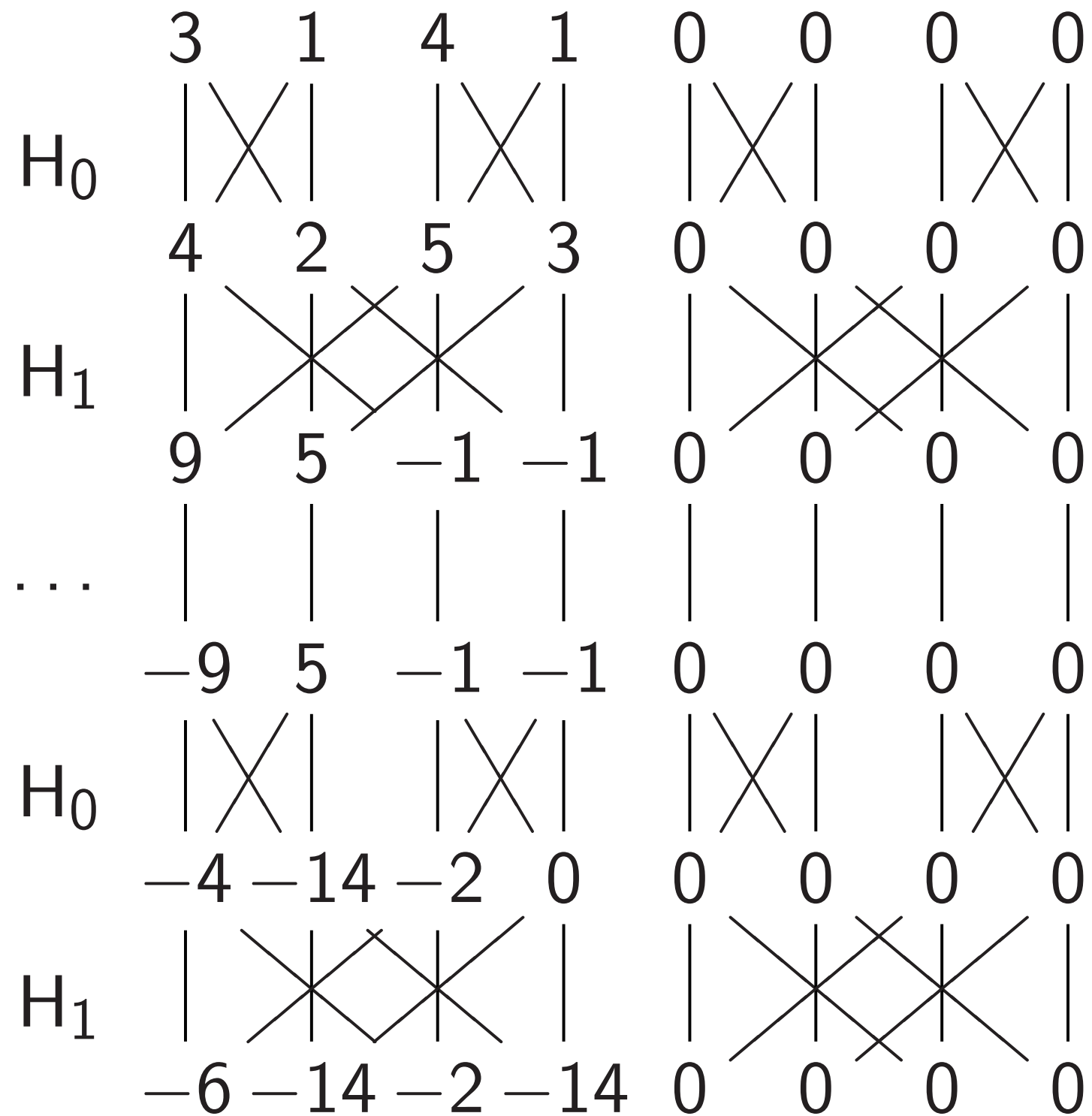
1, 0, 0, 0, 0, 0, 0, 0,  
 0, 0, 0, 0, 0, 1, 0, 0,  
 0, 0, 0, 0, 1, 0, 0, 0,  
 0, 1, 0, 0, 0, 0, 0, 0,  
 0, 0, 1, 0, 0, 0, 0, 1,  
 0, 0, 0, 0, 0, 0, 0, 0,  
 0, 0, 0, 1, 0, 0, 1, 0,  
 0, 0, 0, 0, 0, 0, 0, 0.

Each column is a parallel universe performing its own computations.



Affects measurements: “Negate amplitude around its average.”

$$(3, 1, 4, 1) \mapsto (1.5, 3.5, 0.5, 3.5).$$



## Simon's algorithm

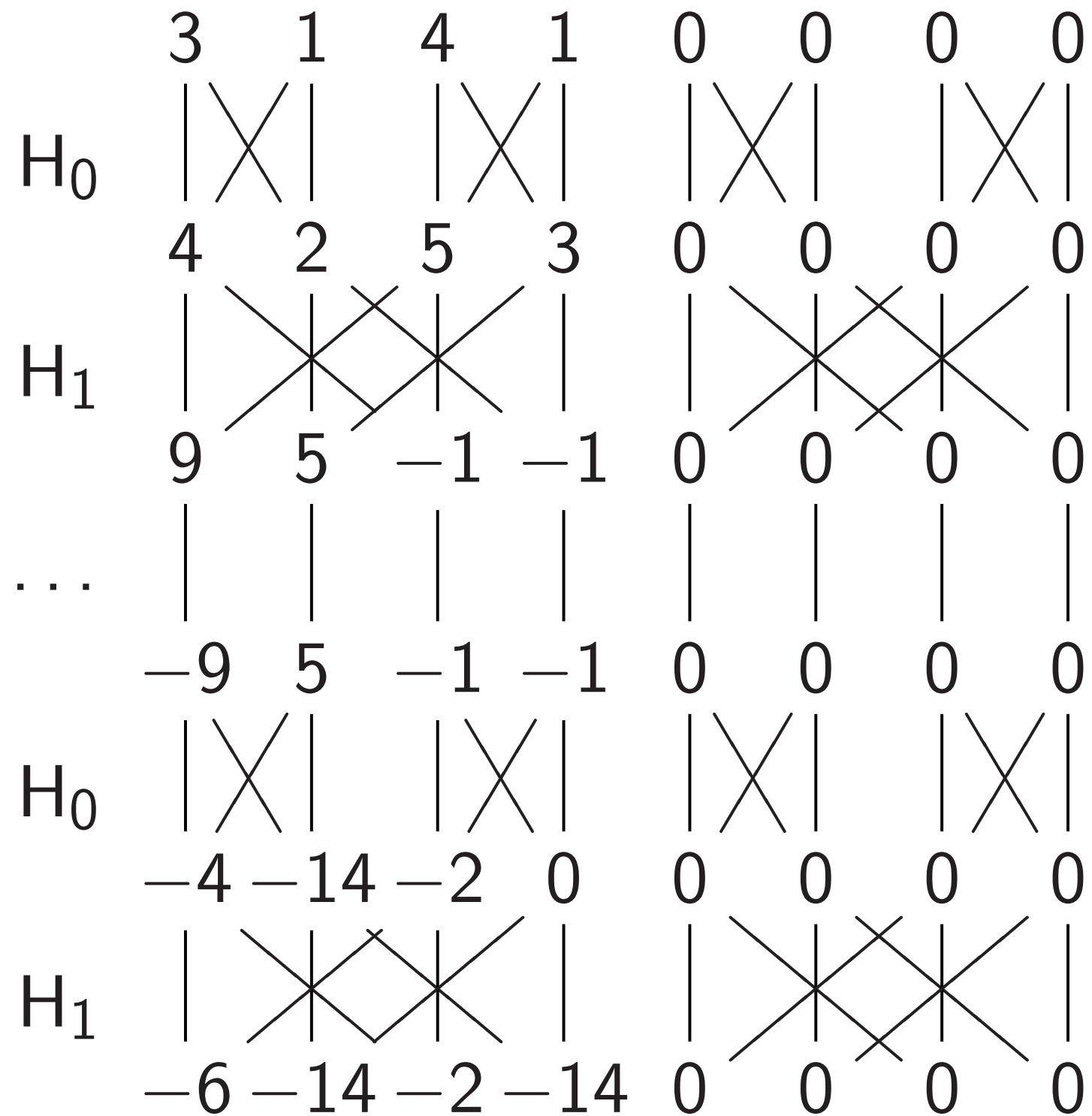
Step 5f. More shuffling:

$0, 0, 0, 0, 0, 1, 0, 0,$   
 $1, 0, 0, 0, 0, 0, 0, 0,$   
 $0, 1, 0, 0, 0, 0, 0, 0,$   
 $0, 0, 0, 0, 1, 0, 0, 0,$   
 $0, 0, 0, 0, 0, 0, 0, 0,$   
 $0, 0, 1, 0, 0, 0, 0, 1,$   
 $0, 0, 0, 0, 0, 0, 0, 0,$   
 $0, 0, 0, 1, 0, 0, 1, 0.$

Each column is a parallel universe performing its own computations.

Affects measurements: “Negate amplitude around its average.”

$$(3, 1, 4, 1) \mapsto (1.5, 3.5, 0.5, 3.5).$$



## Simon's algorithm

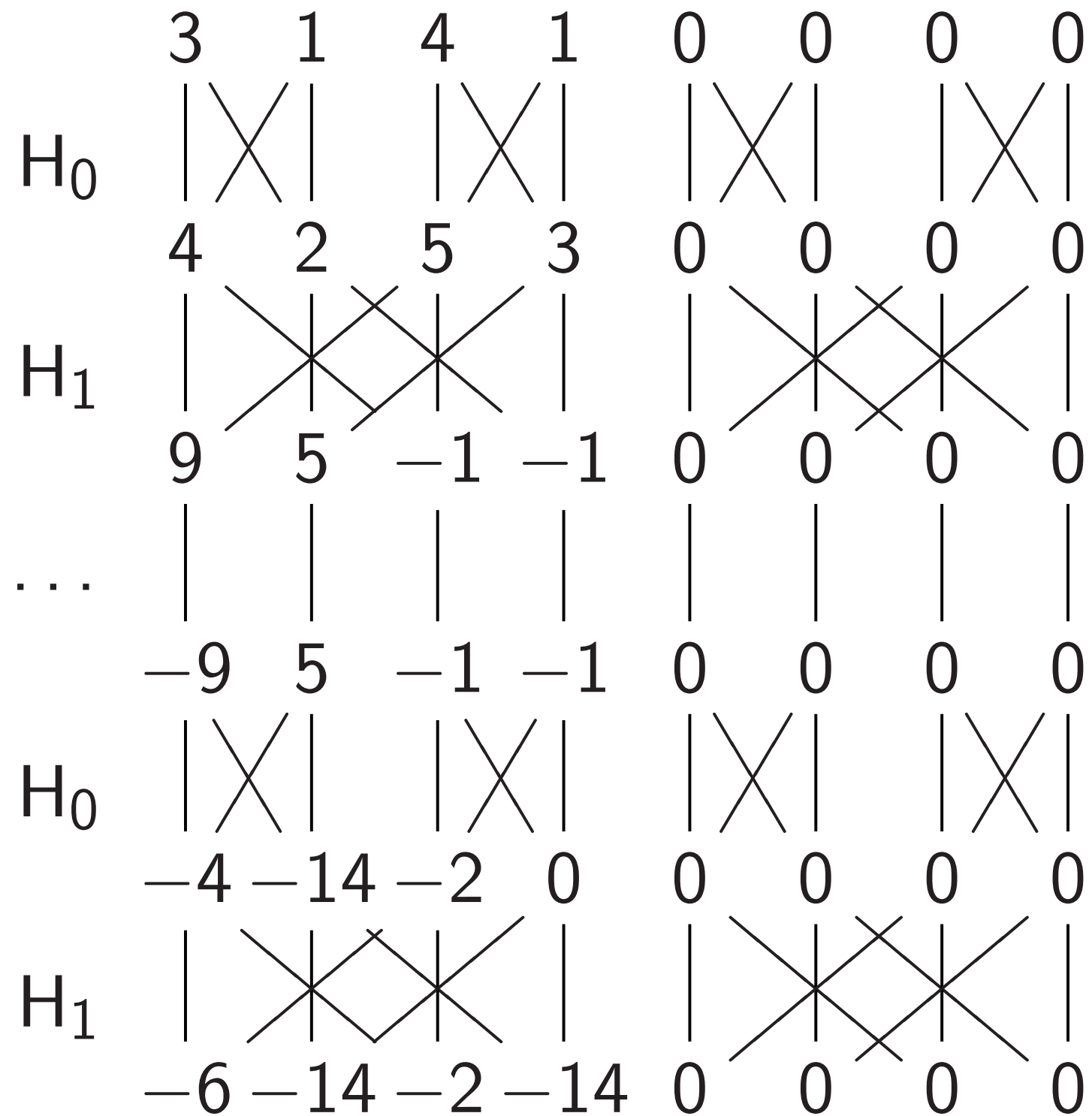
Step 5g. More shuffling:

0, 1, 0, 0, 0, 0, 0, 0,  
 0, 0, 0, 0, 1, 0, 0, 0,  
 0, 0, 0, 0, 0, 1, 0, 0,  
 1, 0, 0, 0, 0, 0, 0, 0,  
 0, 0, 0, 0, 0, 0, 0, 0,  
 0, 0, 0, 1, 0, 0, 1, 0,  
 0, 0, 0, 0, 0, 0, 0, 0,  
 0, 0, 1, 0, 0, 0, 0, 1.

Each column is a parallel universe performing its own computations.

Affects measurements: “Negate amplitude around its average.”

$$(3, 1, 4, 1) \mapsto (1.5, 3.5, 0.5, 3.5).$$



## Simon's algorithm

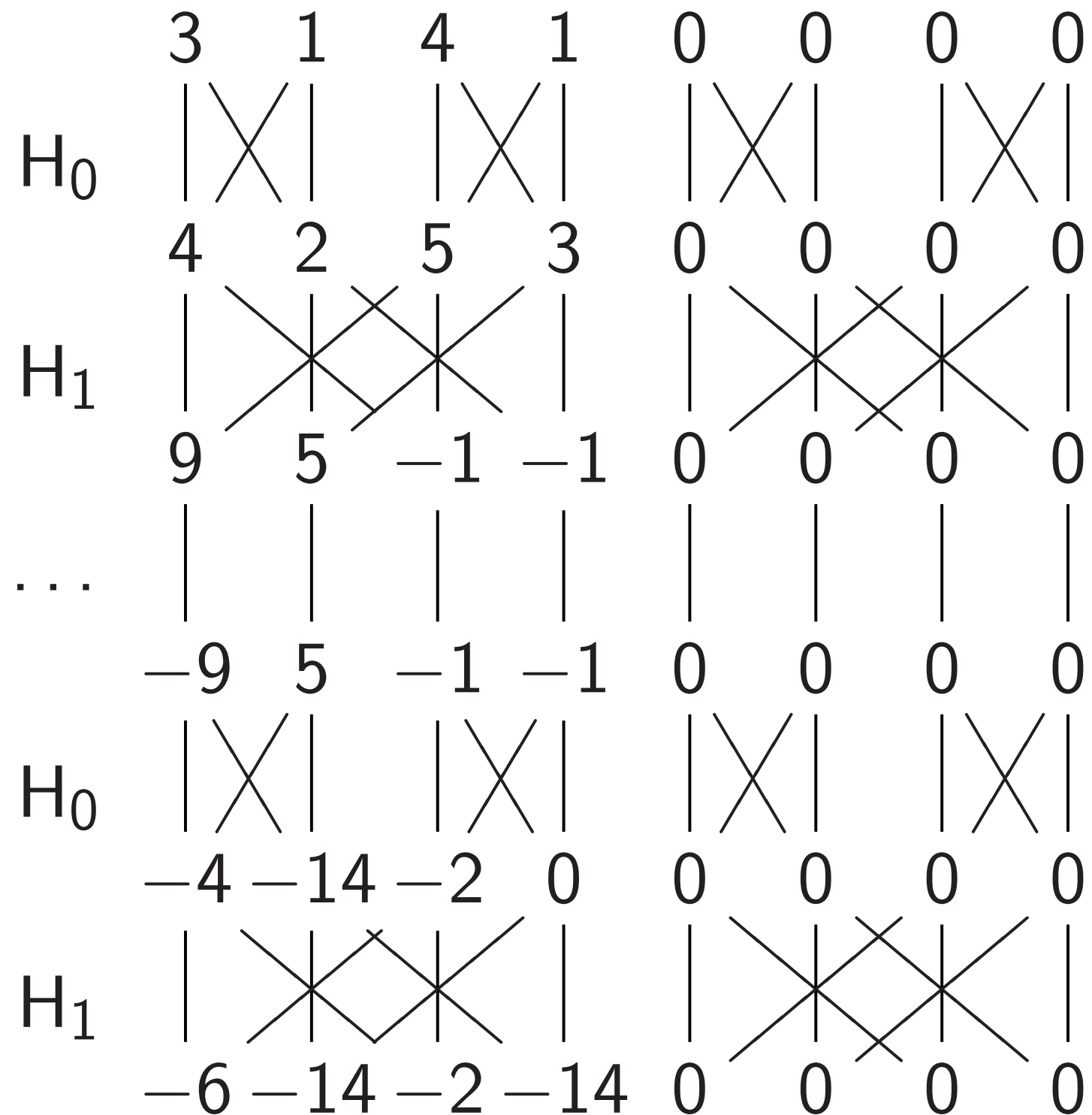
Step 5h. More shuffling:

0, 0, 0, 0, 0, 0, 0, 0,  
 0, 0, 0, 1, 0, 0, 1, 0,  
 0, 0, 0, 0, 0, 0, 0, 0,  
 0, 0, 1, 0, 0, 0, 0, 1,  
 0, 1, 0, 0, 0, 0, 0, 0,  
 0, 0, 0, 0, 1, 0, 0, 0,  
 0, 0, 0, 0, 0, 1, 0, 0,  
 1, 0, 0, 0, 0, 0, 0, 0.

Each column is a parallel universe performing its own computations.

Affects measurements: “Negate amplitude around its average.”

$$(3, 1, 4, 1) \mapsto (1.5, 3.5, 0.5, 3.5).$$



## Simon's algorithm

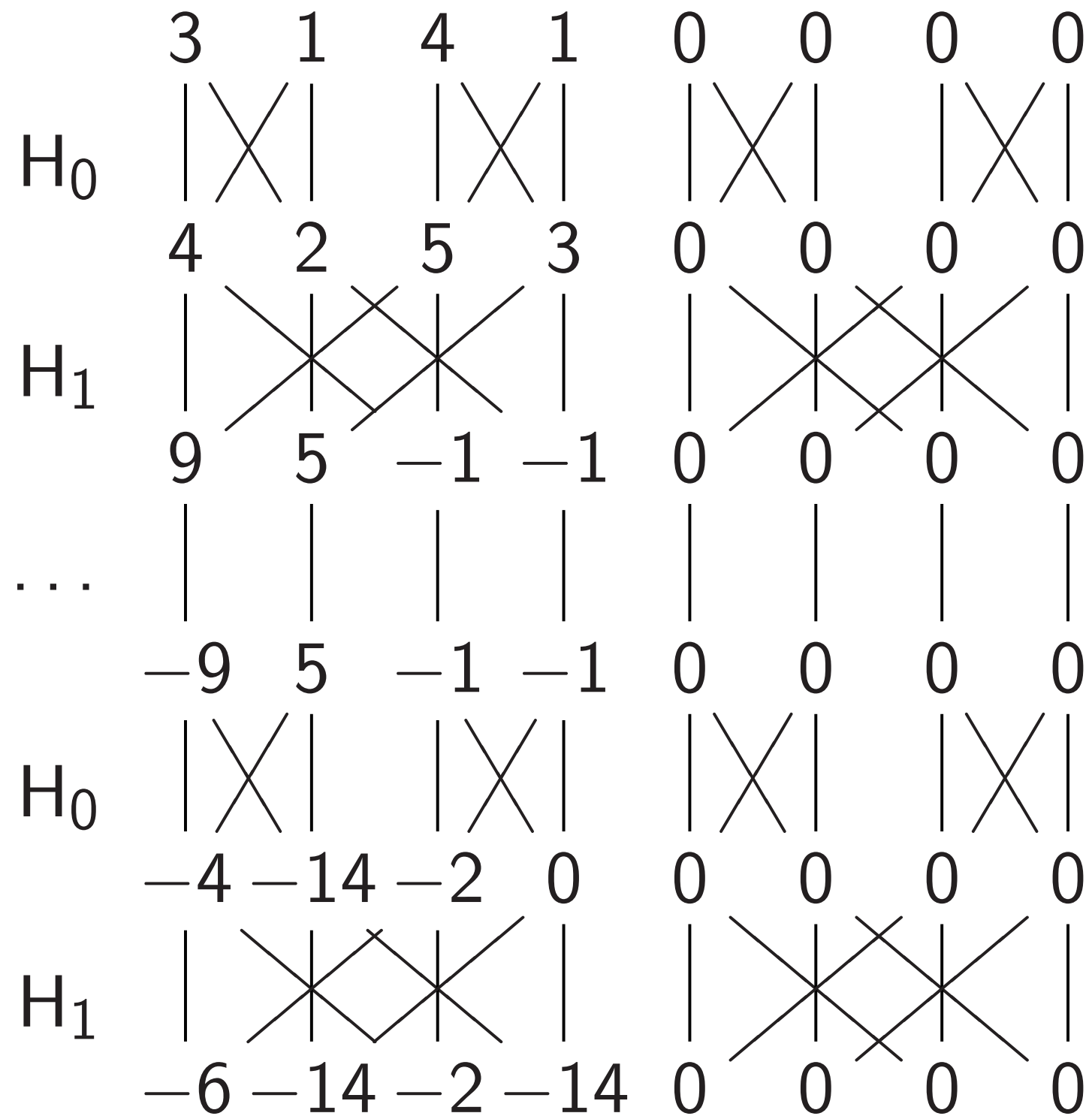
Step 5i. More shuffling:

$0, 0, 0, 0, 0, 0, 1, 0,$   
 $0, 0, 0, 1, 0, 0, 0, 0,$   
 $0, 0, 0, 0, 0, 0, 0, 1,$   
 $0, 0, 1, 0, 0, 0, 0, 0,$   
 $0, 1, 0, 0, 0, 0, 0, 0,$   
 $0, 0, 0, 0, 1, 0, 0, 0,$   
 $0, 0, 0, 0, 0, 1, 0, 0,$   
 $1, 0, 0, 0, 0, 0, 0, 0.$

Each column is a parallel universe performing its own computations.

Affects measurements: “Negate amplitude around its average.”

$$(3, 1, 4, 1) \mapsto (1.5, 3.5, 0.5, 3.5).$$



## Simon's algorithm

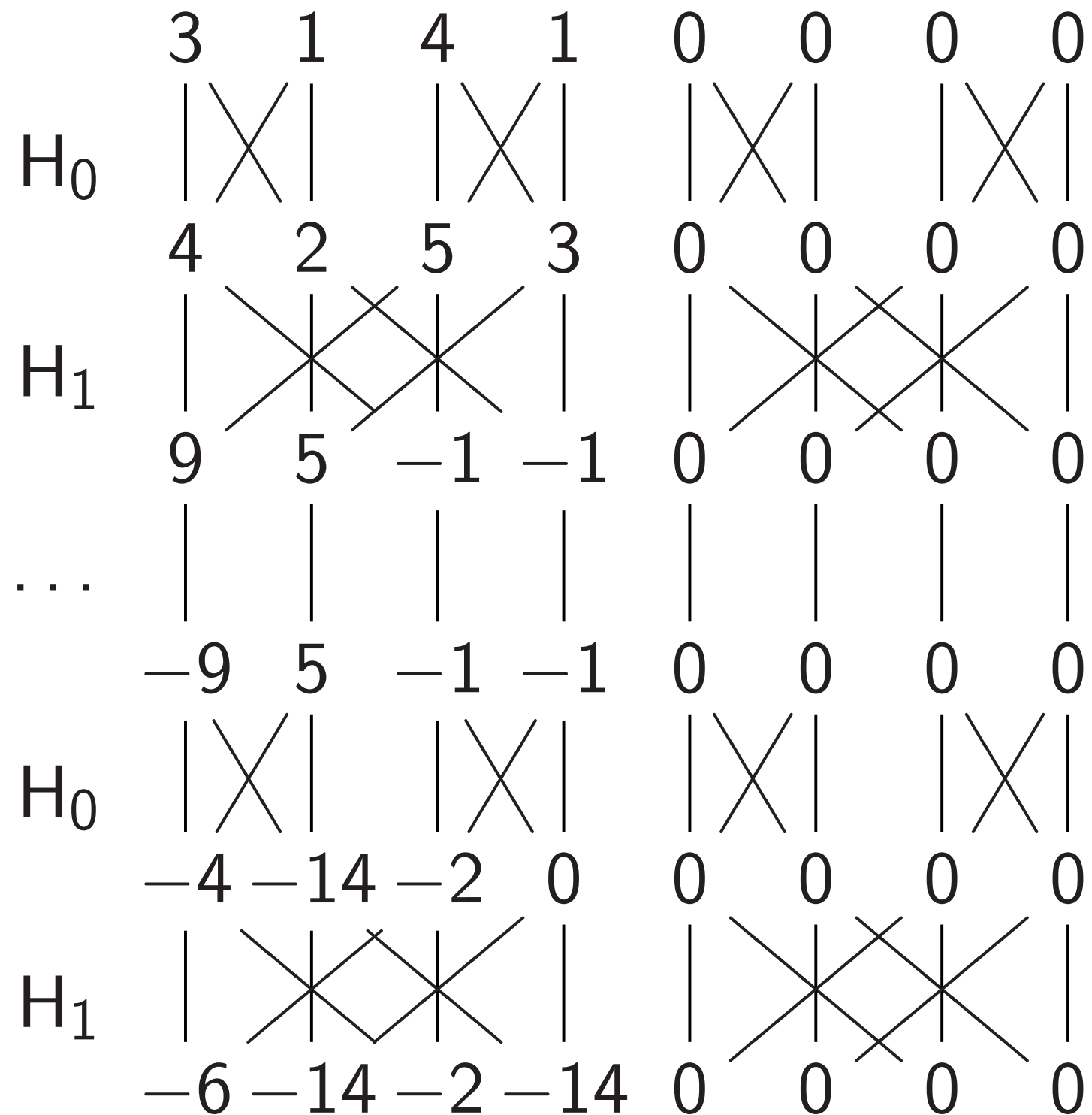
Step 5j. Final shuffling:

0, 0, 0, 0, 0, 0, 0, 0,  
 0, 0, 0, 1, 0, 0, 1, 0,  
 0, 0, 0, 0, 0, 0, 0, 0,  
 0, 0, 1, 0, 0, 0, 0, 1,  
 0, 1, 0, 0, 1, 0, 0, 0,  
 0, 0, 0, 0, 0, 0, 0, 0,  
 0, 0, 0, 0, 0, 0, 0, 0,  
 1, 0, 0, 0, 0, 1, 0, 0.

Each column is a parallel universe performing its own computations.

Affects measurements: “Negate amplitude around its average.”

$$(3, 1, 4, 1) \mapsto (1.5, 3.5, 0.5, 3.5).$$



## Simon's algorithm

Step 5j. Final shuffling:

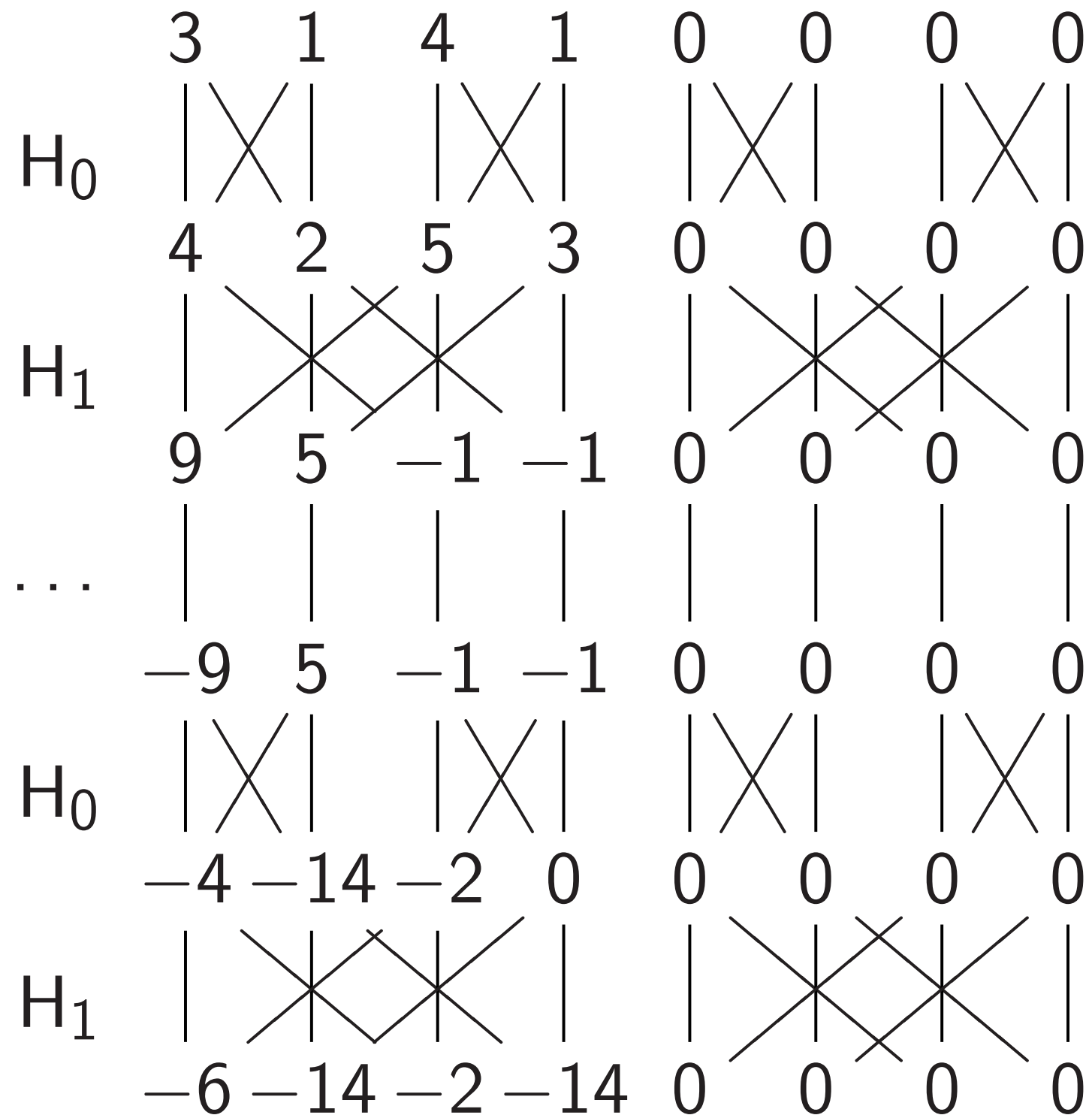
0, 0, 0, 0, 0, 0, 0, 0,  
 0, 0, 0, 1, 0, 0, 1, 0,  
 0, 0, 0, 0, 0, 0, 0, 0,  
 0, 0, 1, 0, 0, 0, 0, 1,  
 0, 1, 0, 0, 1, 0, 0, 0,  
 0, 0, 0, 0, 0, 0, 0, 0,  
 0, 0, 0, 0, 0, 0, 0, 0,  
 1, 0, 0, 0, 0, 1, 0, 0.

Each column is a parallel universe performing its own computations.

Surprise:  $u$  and  $u \oplus 101$  match.

Affects measurements: “Negate amplitude around its average.”

$$(3, 1, 4, 1) \mapsto (1.5, 3.5, 0.5, 3.5).$$



## Simon's algorithm

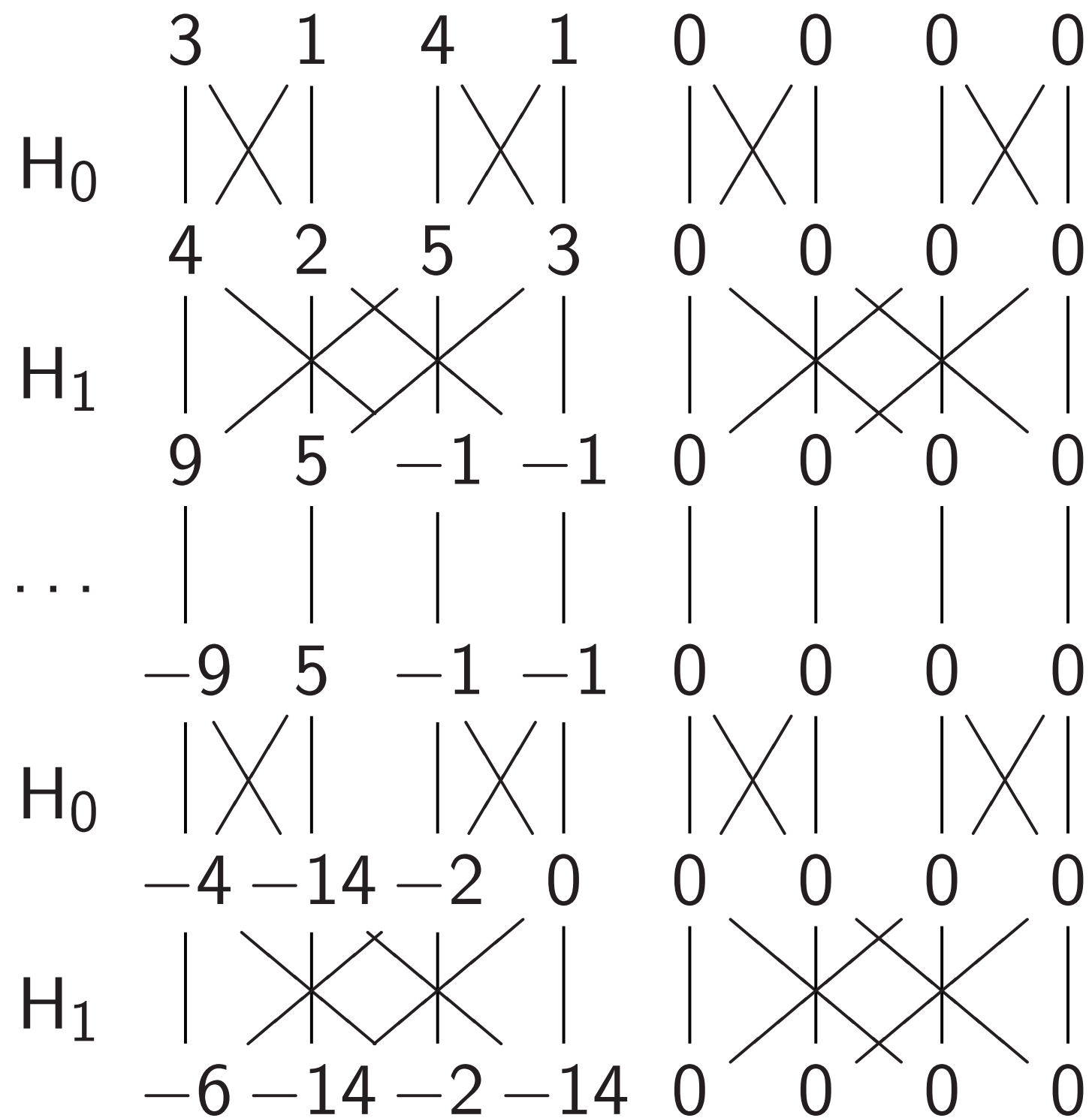
Step 6. Hadamard<sub>0</sub>:

0, 0, 0, 0, 0, 0, 0, 0,  
 0, 0, 1,  $\bar{1}$ , 0, 0, 1, 1,  
 0, 0, 0, 0, 0, 0, 0, 0,  
 0, 0, 1, 1, 0, 0, 1,  $\bar{1}$ ,  
 1,  $\bar{1}$ , 0, 0, 1, 1, 0, 0,  
 0, 0, 0, 0, 0, 0, 0, 0,  
 0, 0, 0, 0, 0, 0, 0, 0,  
 1, 1, 0, 0, 1,  $\bar{1}$ , 0, 0.

Notation:  $\bar{1}$  means  $-1$ .

Affects measurements: “Negate amplitude around its average.”

$$(3, 1, 4, 1) \mapsto (1.5, 3.5, 0.5, 3.5).$$



## Simon's algorithm

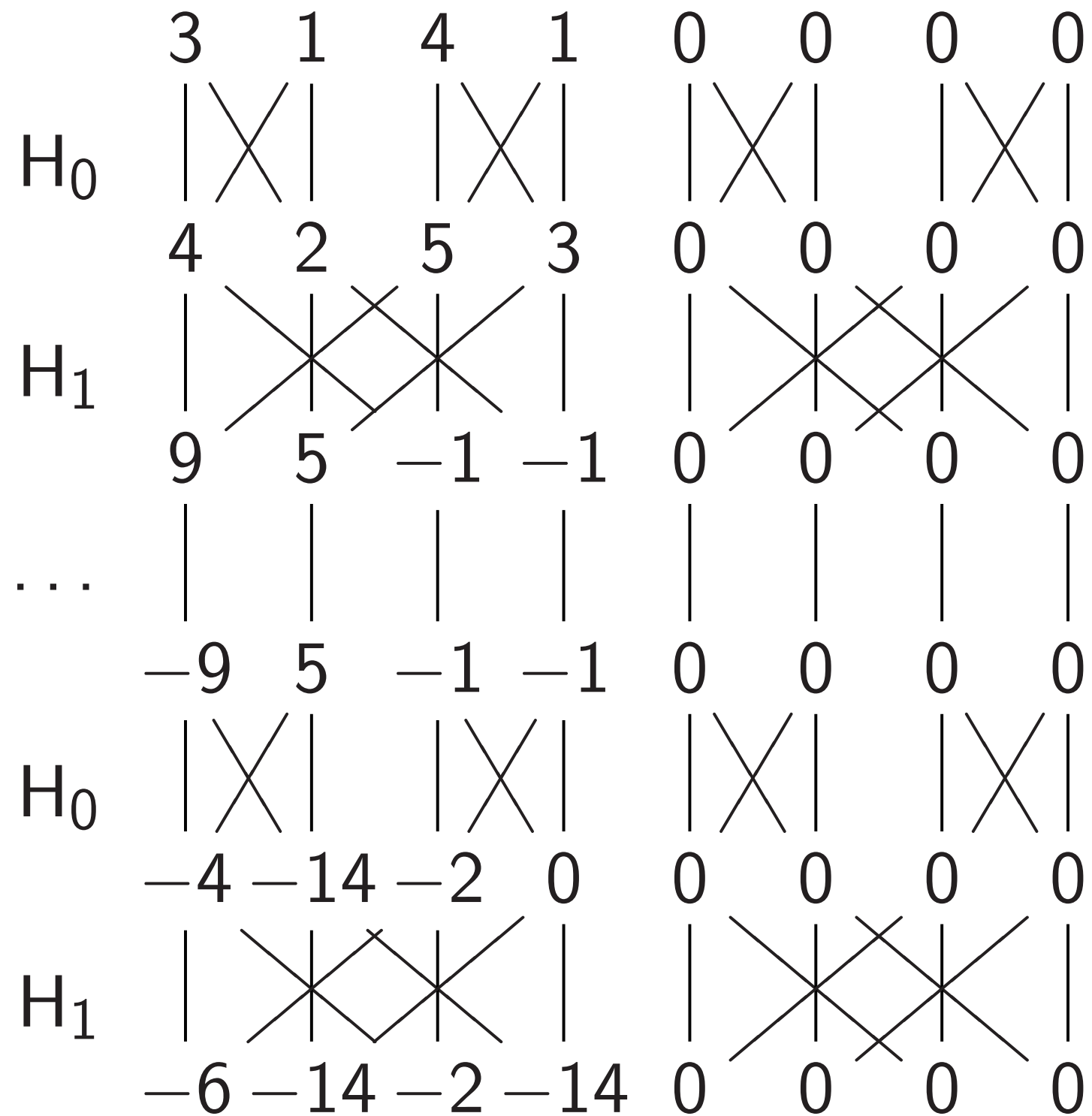
Step 7. Hadamard<sub>1</sub>:

0, 0, 0, 0, 0, 0, 0, 0,  
**1,  $\bar{1}$ ,  $\bar{1}$ , 1, 1, 1,  $\bar{1}$ ,  $\bar{1}$ ,**  
 0, 0, 0, 0, 0, 0, 0, 0,  
**1, 1,  $\bar{1}$ ,  $\bar{1}$ , 1,  $\bar{1}$ ,  $\bar{1}$ , 1,**  
**1,  $\bar{1}$ , 1,  $\bar{1}$ , 1, 1, 1, 1,**  
 0, 0, 0, 0, 0, 0, 0, 0,  
 0, 0, 0, 0, 0, 0, 0, 0,  
**1, 1, 1, 1, 1,  $\bar{1}$ , 1,  $\bar{1}$ .**



Affects measurements: “Negate amplitude around its average.”

$$(3, 1, 4, 1) \mapsto (1.5, 3.5, 0.5, 3.5).$$



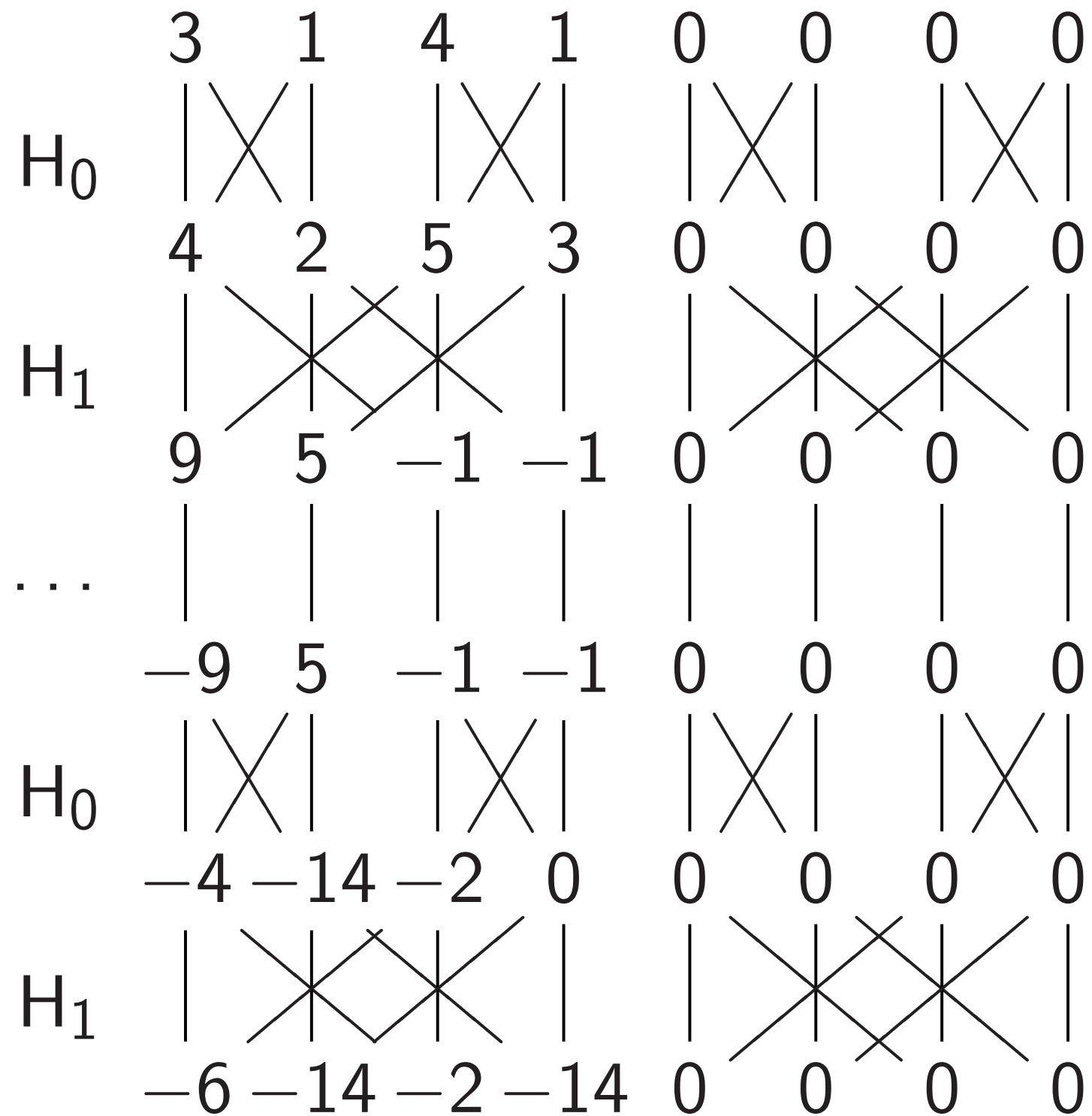
## Simon's algorithm

Step 8. Hadamard<sub>2</sub>:

0, 0, 0, 0, 0, 0, 0, 0,  
 2, 0,  $\bar{2}$ , 0, 0,  $\bar{2}$ , 0, 2,  
 0, 0, 0, 0, 0, 0, 0, 0,  
 2, 0,  $\bar{2}$ , 0, 0, 2, 0,  $\bar{2}$ ,  
 2, 0, 2, 0, 0,  $\bar{2}$ , 0,  $\bar{2}$ ,  
 0, 0, 0, 0, 0, 0, 0, 0,  
 0, 0, 0, 0, 0, 0, 0, 0,  
 2, 0, 2, 0, 0, 2, 0, 2.

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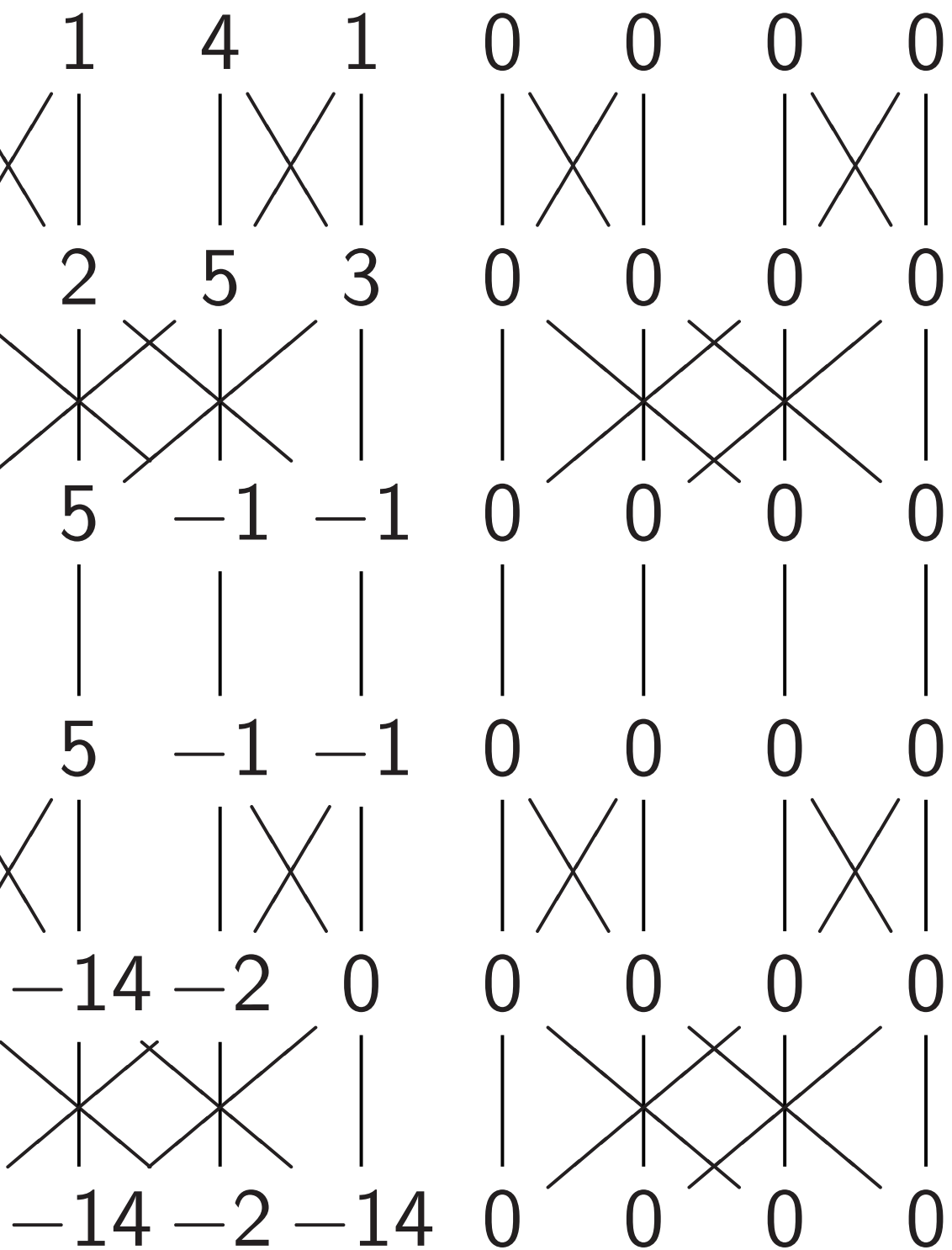
Step 8. Hadamard<sub>2</sub>:

$$\begin{array}{cccccccc}
 0, & 0, & 0, & 0, & 0, & 0, & 0, & 0, \\
 2, & 0, & \bar{2}, & 0, & 0, & \bar{2}, & 0, & 2, \\
 0, & 0, & 0, & 0, & 0, & 0, & 0, & 0, \\
 2, & 0, & \bar{2}, & 0, & 0, & 2, & 0, & \bar{2}, \\
 2, & 0, & 2, & 0, & 0, & \bar{2}, & 0, & \bar{2}, \\
 0, & 0, & 0, & 0, & 0, & 0, & 0, & 0, \\
 0, & 0, & 0, & 0, & 0, & 0, & 0, & 0, \\
 2, & 0, & 2, & 0, & 0, & 2, & 0, & 2.
 \end{array}$$

Step 9: Measure. Obtain some information about the surprise: a random vector orthogonal to 101.

measurements: “Negate  
around its average.”

$(1) \mapsto (1.5, 3.5, 0.5, 3.5)$ .



## Simon's algorithm

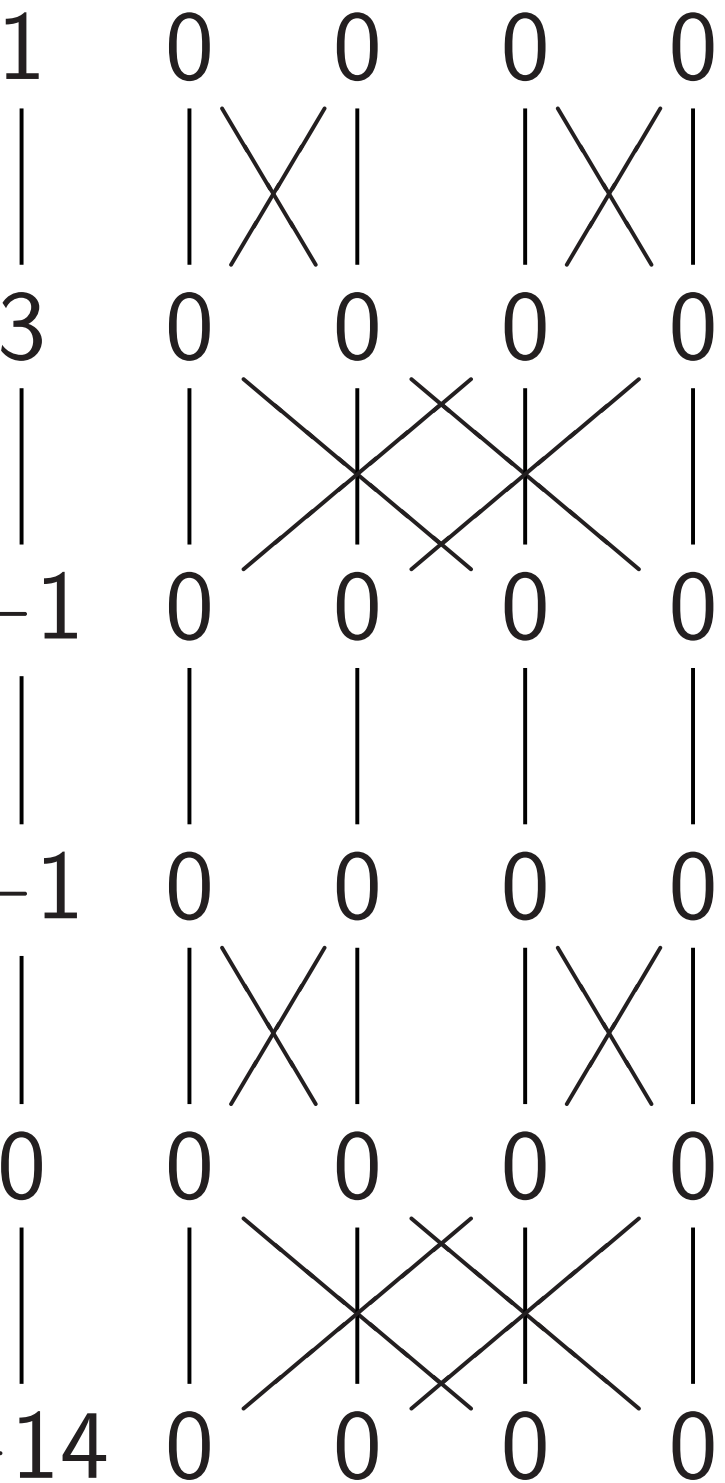
Repeat t

Step 8. Hadamard<sub>2</sub>:

$0, 0, 0, 0, 0, 0, 0, 0,$   
 $2, 0, \bar{2}, 0, 0, \bar{2}, 0, 2,$   
 $0, 0, 0, 0, 0, 0, 0, 0,$   
 $2, 0, \bar{2}, 0, 0, 2, 0, \bar{2},$   
 $2, 0, 2, 0, 0, \bar{2}, 0, \bar{2},$   
 $0, 0, 0, 0, 0, 0, 0, 0,$   
 $0, 0, 0, 0, 0, 0, 0, 0,$   
 $2, 0, 2, 0, 0, 2, 0, 2.$

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(3.5, 0.5, 3.5).



## Simon's algorithm

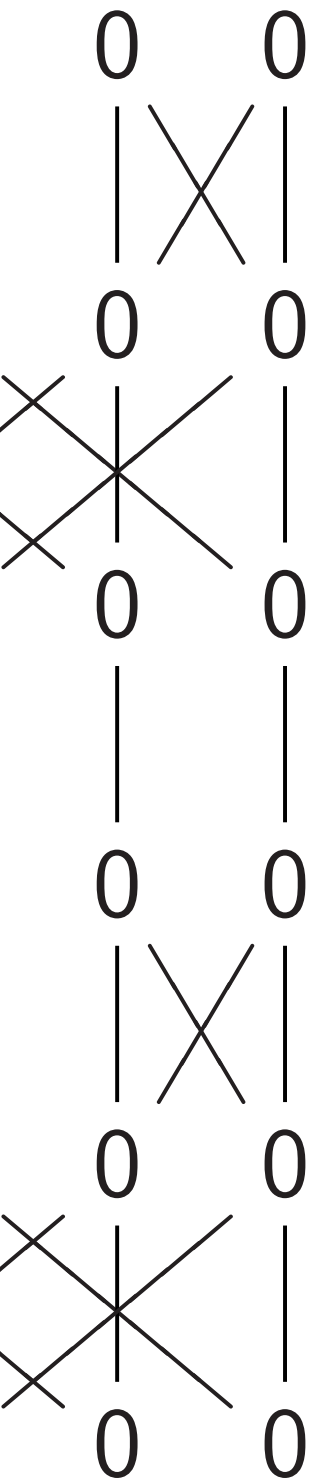
Repeat to figure o

Step 8. Hadamard<sub>2</sub>:

0, 0, 0, 0, 0, 0, 0, 0,  
 $2$ , 0,  $\bar{2}$ , 0, 0,  $\bar{2}$ , 0,  $2$ ,  
 0, 0, 0, 0, 0, 0, 0, 0,  
 $2$ , 0,  $\bar{2}$ , 0, 0,  $2$ , 0,  $\bar{2}$ ,  
 $2$ , 0,  $2$ , 0, 0,  $\bar{2}$ , 0,  $\bar{2}$ ,  
 0, 0, 0, 0, 0, 0, 0, 0,  
 0, 0, 0, 0, 0, 0, 0, 0,  
 $2$ , 0,  $2$ , 0, 0,  $2$ , 0,  $2$ .

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gate  
e.”  
.5).



## Simon's algorithm

Step 8. Hadamard<sub>2</sub>:

0, 0, 0, 0, 0, 0, 0, 0,  
 $\sqrt{2}$ , 0,  $\sqrt{2}$ , 0, 0,  $\sqrt{2}$ , 0,  $\sqrt{2}$ ,  
 0, 0, 0, 0, 0, 0, 0, 0,  
 $\sqrt{2}$ , 0,  $\sqrt{2}$ , 0, 0,  $\sqrt{2}$ , 0,  $\sqrt{2}$ ,  
 $\sqrt{2}$ , 0,  $\sqrt{2}$ , 0, 0,  $\sqrt{2}$ , 0,  $\sqrt{2}$ ,  
 0, 0, 0, 0, 0, 0, 0, 0,  
 0, 0, 0, 0, 0, 0, 0, 0,  
 $\sqrt{2}$ , 0,  $\sqrt{2}$ , 0, 0,  $\sqrt{2}$ , 0,  $\sqrt{2}$ .

Step 9: Measure. Obtain some information about the surprise: a random vector orthogonal to 101.

Repeat to figure out 101.

Simon's algorithm

Step 8. Hadamard<sub>2</sub>:

0, 0, 0, 0, 0, 0, 0, 0,

2, 0,  $\bar{2}$ , 0, 0,  $\bar{2}$ , 0, 2,

0, 0, 0, 0, 0, 0, 0, 0,

2, 0,  $\bar{2}$ , 0, 0, 2, 0,  $\bar{2}$ ,

2, 0, 2, 0, 0,  $\bar{2}$ , 0,  $\bar{2}$ ,

0, 0, 0, 0, 0, 0, 0, 0,

0, 0, 0, 0, 0, 0, 0, 0,

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 0, 0, 0, 0, 0, 0, 0, 0,  
 2, 0,  $\bar{2}$ , 0, 0, 2, 0,  $\bar{2}$ ,  
 2, 0, 2, 0, 0,  $\bar{2}$ , 0,  $\bar{2}$ ,  
 0, 0, 0, 0, 0, 0, 0, 0,  
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Generalize Step 5 to any function  $u \mapsto f(u)$  with  $f(u) = f(u \oplus s)$ .

“Usually” algorithm figures out  $s$ .

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 2, 0,  $\bar{2}$ , 0, 0,  $\bar{2}$ , 0, 2,  
 0, 0, 0, 0, 0, 0, 0, 0,  
 2, 0,  $\bar{2}$ , 0, 0, 2, 0,  $\bar{2}$ ,  
 2, 0, 2, 0, 0,  $\bar{2}$ , 0,  $\bar{2}$ ,  
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Many spectacular applications.



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 0, 0, 0, 0, 0, 0, 0, 0,  
 2, 0,  $\bar{2}$ , 0, 0, 2, 0,  $\bar{2}$ ,  
 2, 0, 2, 0, 0,  $\bar{2}$ , 0,  $\bar{2}$ ,  
 0, 0, 0, 0, 0, 0, 0, 0,  
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 2, 0,  $\bar{2}$ , 0, 0, 2, 0,  $\bar{2}$ ,  
 2, 0, 2, 0, 0,  $\bar{2}$ , 0,  $\bar{2}$ ,  
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Easy to compute discrete logs.

algorithmHadamard<sub>2</sub>:

0, 0, 0, 0, 0,  
 0, 0,  $\bar{2}$ , 0, 2,  
 0, 0, 0, 0, 0,  
 0, 0, 2, 0,  $\bar{2}$ ,  
 0, 0,  $\bar{2}$ , 0,  $\bar{2}$ ,  
 0, 0, 0, 0, 0,  
 0, 0, 0, 0, 0,  
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Grover’s

Assume:  
 has  $f(s)$

Tradition  
 compute  
 hope to  
 Success  
 until #i

$d_2:$   
 $(0, 0,$   
 $(0, 2,$   
 $(0, 0,$   
 $(0, \overline{2},$   
 $(0, \overline{2},$   
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Assume: unique  $s \in \{0, 1\}^n$  has  $f(s) = 0$ .

Traditional algorithm to find  $s$  compute  $f$  for many inputs, hope to find output 0.

Success probability is very low until #inputs approaches  $2^n$ .

Repeat to figure out 101.

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Repeat Step 1 + Step 2  
about  $0.58 \cdot 2^{0.5n}$  times.

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With high probability this finds  $s$ .

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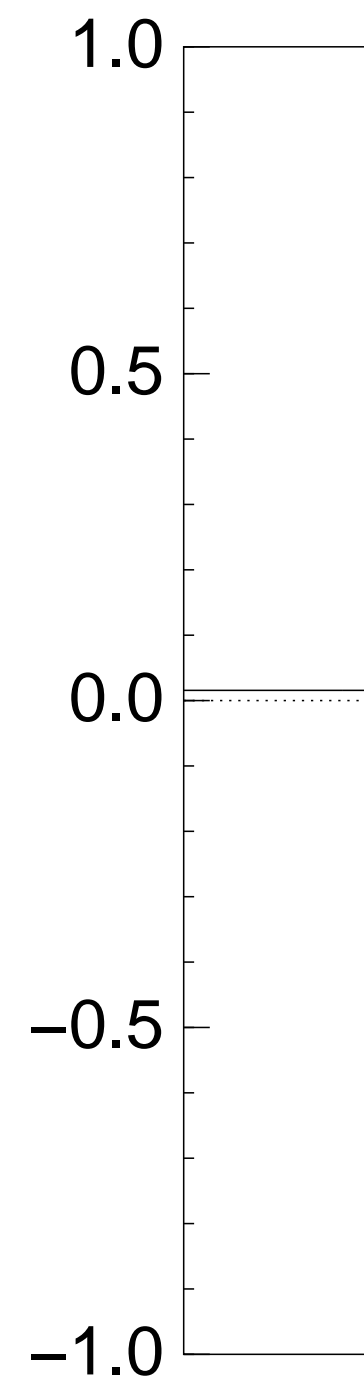
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Normaliz  
 for an ex  
 after 0 s





$\in \{0, 1\}^n$

algorithm to find  $s$ :

any inputs,

at 0.

probability is very low

approaches  $2^n$ .

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evaluations of  $f$ .

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at this

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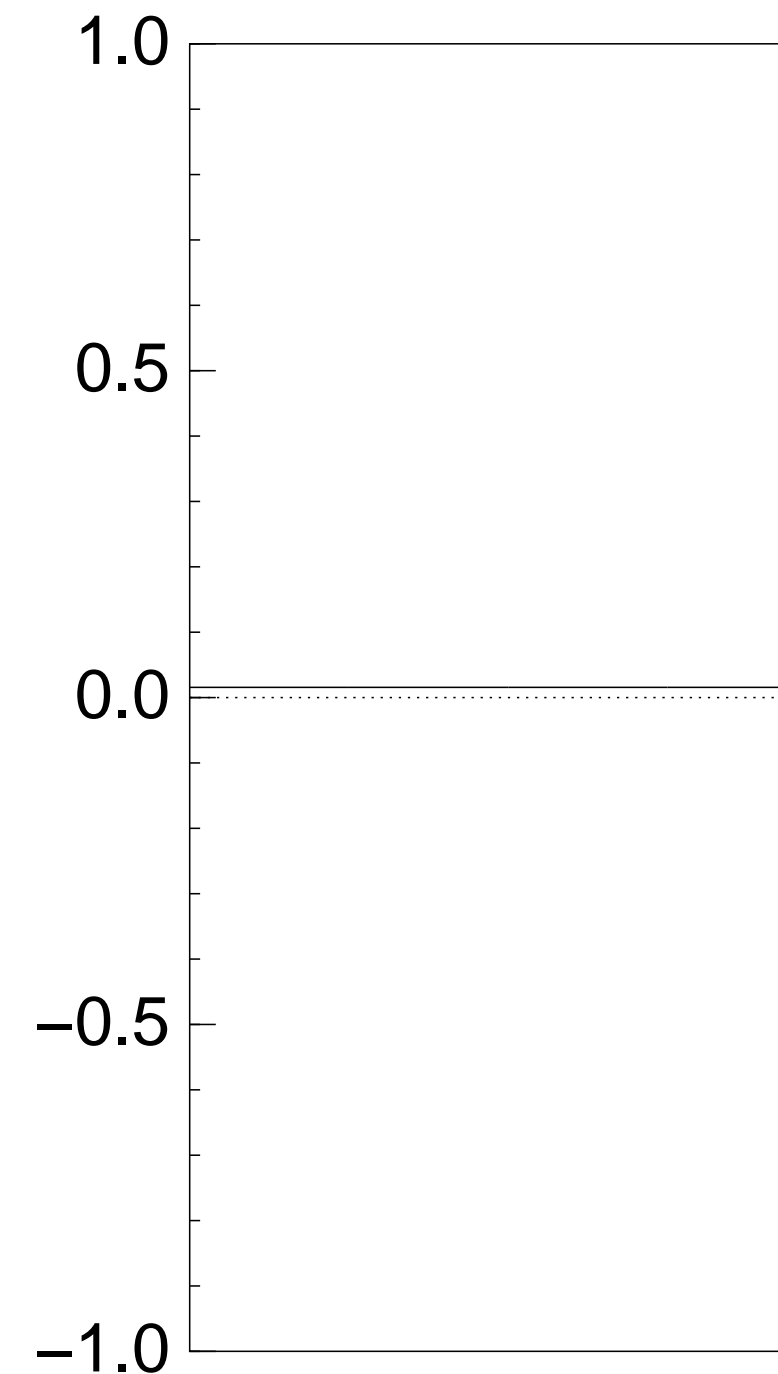
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about  $0.5\pi \cdot 2^{0.5n}$  times.

Measure the  $n$  qubits.

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Normalized graph for an example with after 0 steps:



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Negate  $a$  around its average.

This is also fast.

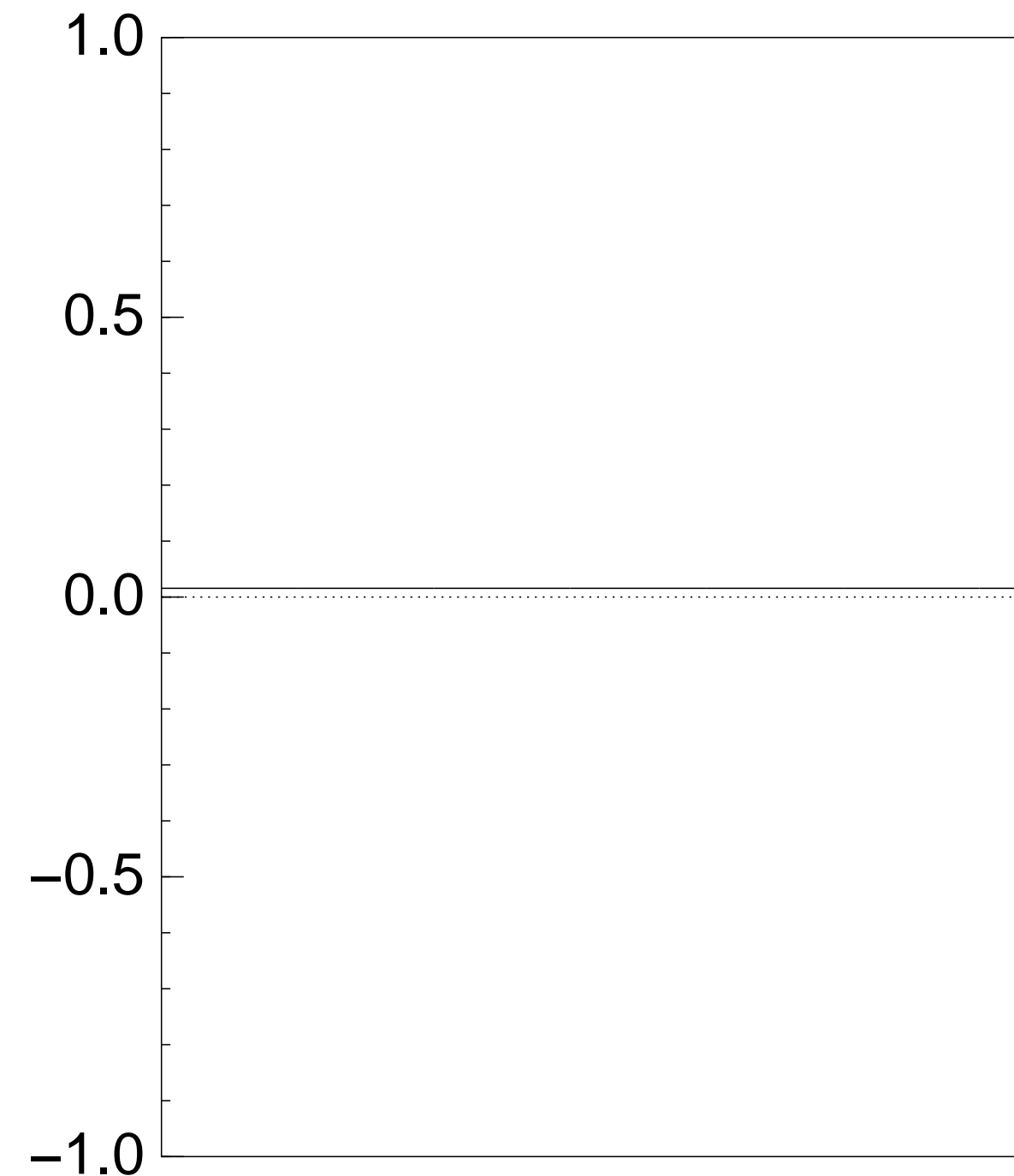
Repeat Step 1 + Step 2

about  $0.58 \cdot 2^{0.5n}$  times.

Measure the  $n$  qubits.

With high probability this finds  $s$ .

Normalized graph of  $q \mapsto a_q$  for an example with  $n = 12$  after 0 steps:



Start from uniform superposition over all  $n$ -bit strings  $q$ .

Step 1: Set  $a \leftarrow b$  where

$$b_q = -a_q \text{ if } f(q) = 0,$$

$$b_q = a_q \text{ otherwise.}$$

This is fast.

Step 2: “Grover diffusion”.

Negate  $a$  around its average.

This is also fast.

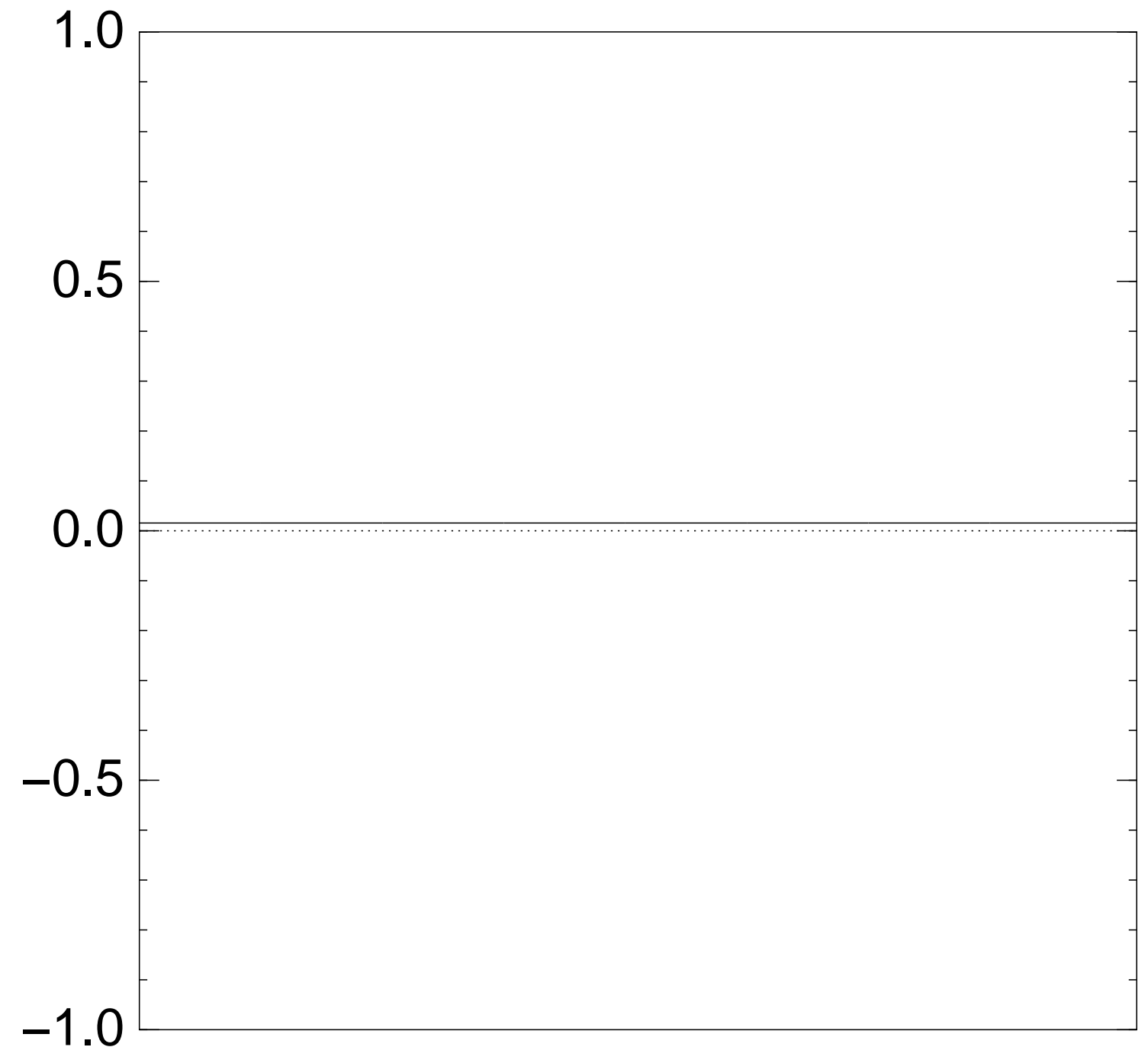
Repeat Step 1 + Step 2

about  $0.58 \cdot 2^{0.5n}$  times.

Measure the  $n$  qubits.

With high probability this finds  $s$ .

Normalized graph of  $q \mapsto a_q$   
for an example with  $n = 12$   
after 0 steps:



Start from uniform superposition over all  $n$ -bit strings  $q$ .

Step 1: Set  $a \leftarrow b$  where

$$b_q = -a_q \text{ if } f(q) = 0,$$

$$b_q = a_q \text{ otherwise.}$$

This is fast.

Step 2: “Grover diffusion”.

Negate  $a$  around its average.

This is also fast.

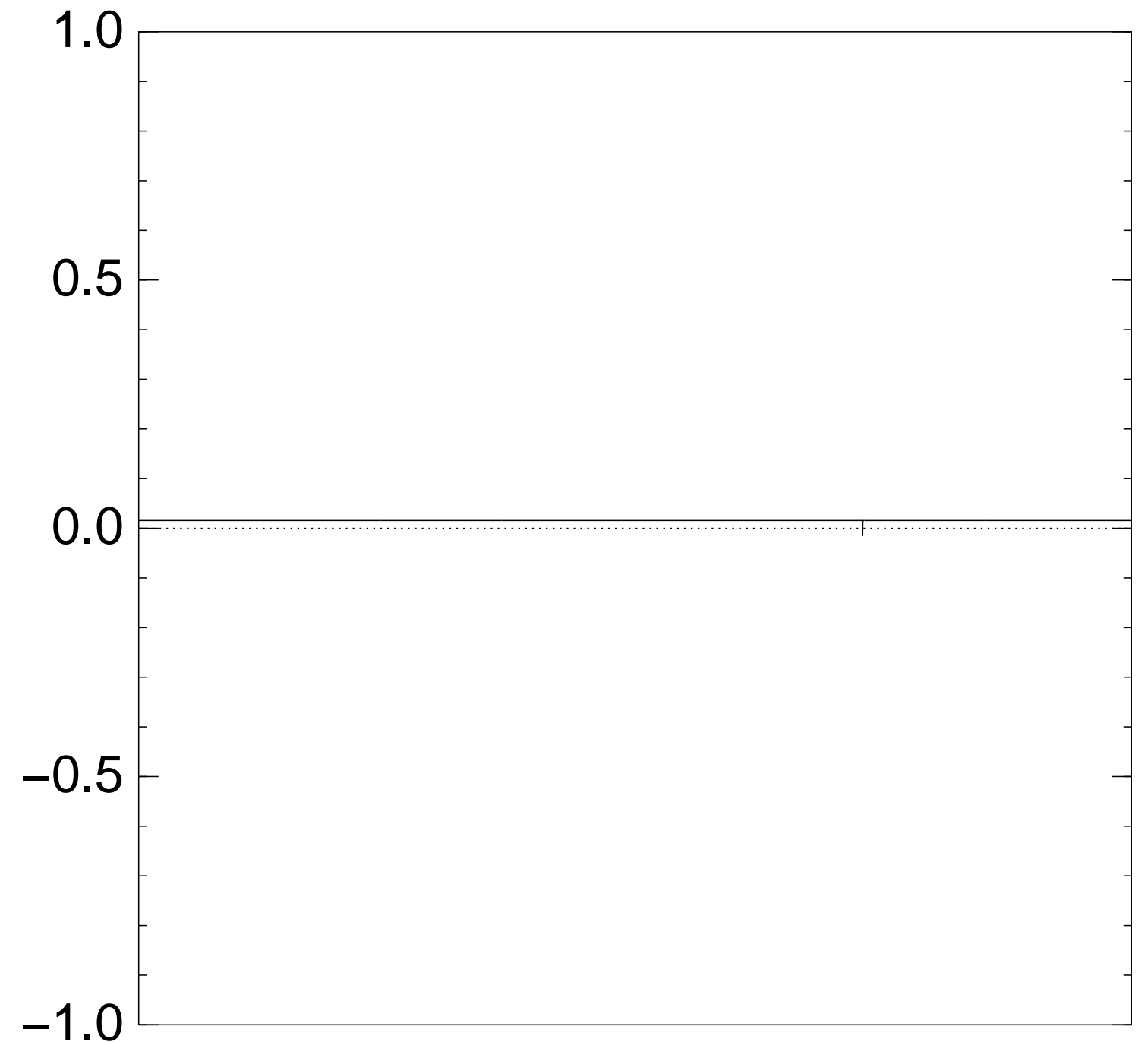
Repeat Step 1 + Step 2

about  $0.58 \cdot 2^{0.5n}$  times.

Measure the  $n$  qubits.

With high probability this finds  $s$ .

Normalized graph of  $q \mapsto a_q$   
for an example with  $n = 12$   
after Step 1:



Start from uniform superposition over all  $n$ -bit strings  $q$ .

Step 1: Set  $a \leftarrow b$  where

$$b_q = -a_q \text{ if } f(q) = 0,$$

$$b_q = a_q \text{ otherwise.}$$

This is fast.

Step 2: “Grover diffusion”.

Negate  $a$  around its average.

This is also fast.

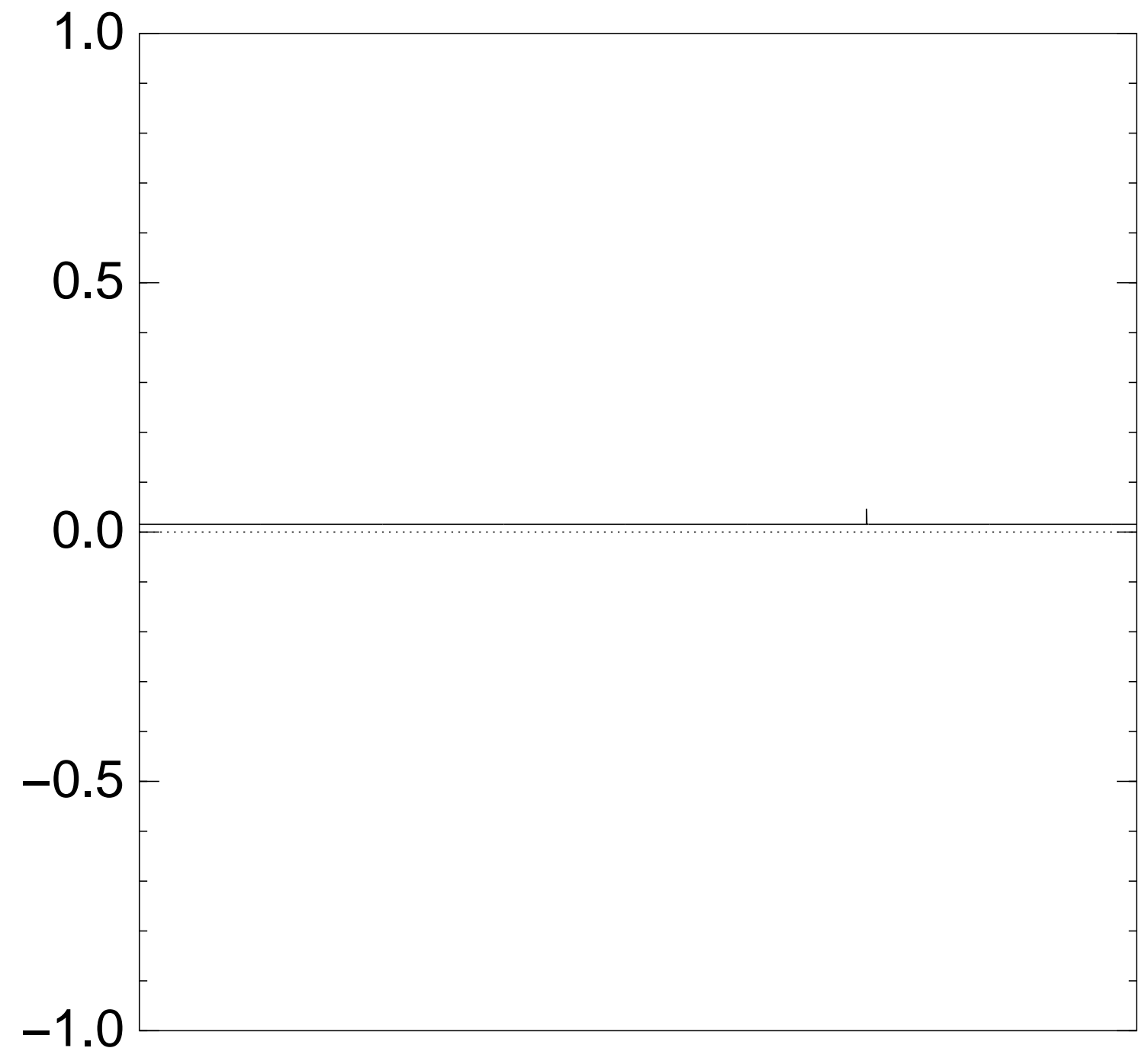
Repeat Step 1 + Step 2

about  $0.58 \cdot 2^{0.5n}$  times.

Measure the  $n$  qubits.

With high probability this finds  $s$ .

Normalized graph of  $q \mapsto a_q$   
for an example with  $n = 12$   
after Step 1 + Step 2:



Start from uniform superposition over all  $n$ -bit strings  $q$ .

Step 1: Set  $a \leftarrow b$  where

$$b_q = -a_q \text{ if } f(q) = 0,$$

$$b_q = a_q \text{ otherwise.}$$

This is fast.

Step 2: “Grover diffusion”.

Negate  $a$  around its average.

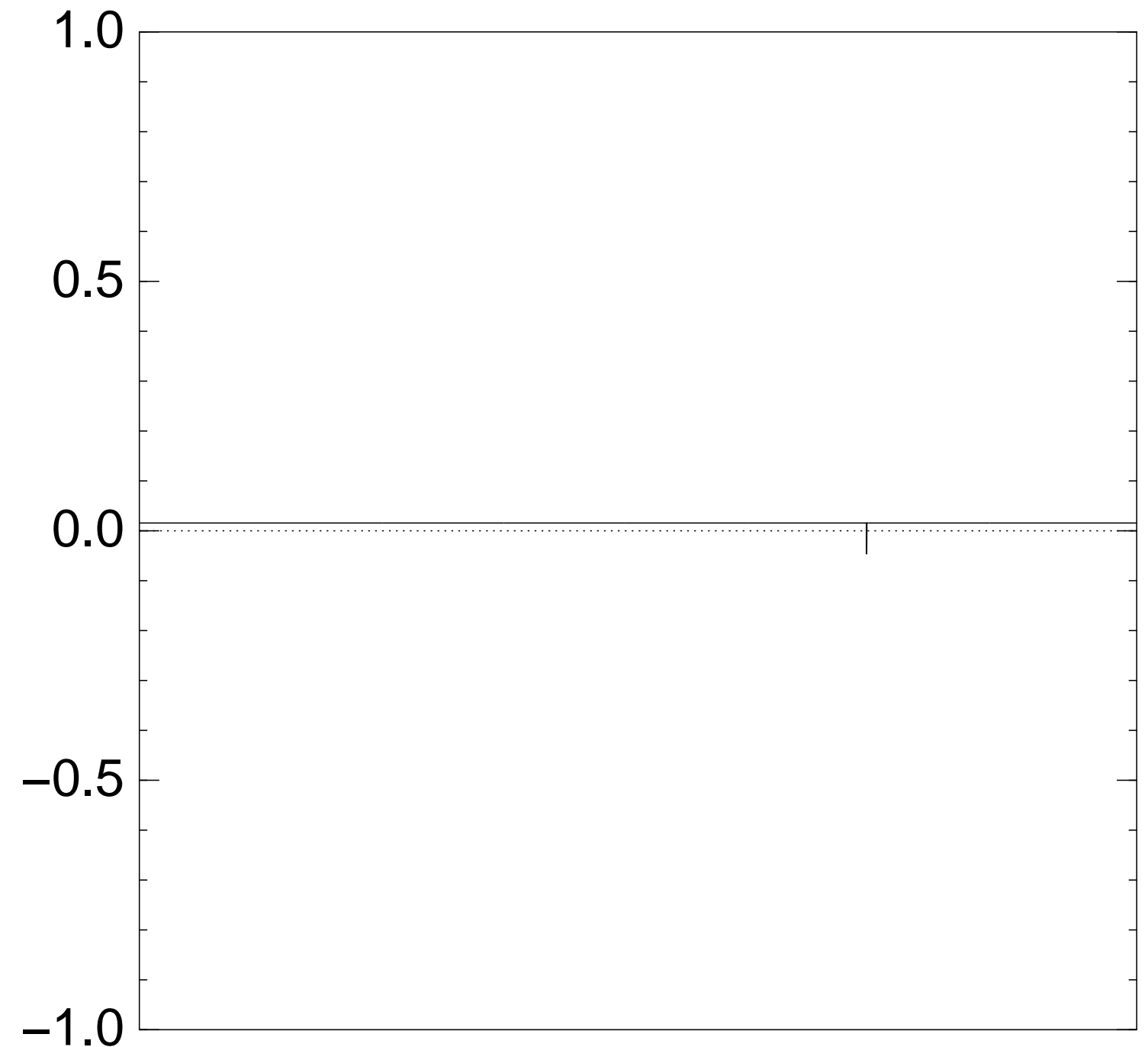
This is also fast.

Repeat Step 1 + Step 2 about  $0.58 \cdot 2^{0.5n}$  times.

Measure the  $n$  qubits.

With high probability this finds  $s$ .

Normalized graph of  $q \mapsto a_q$  for an example with  $n = 12$  after Step 1 + Step 2 + Step 1:



Start from uniform superposition over all  $n$ -bit strings  $q$ .

Step 1: Set  $a \leftarrow b$  where

$$b_q = -a_q \text{ if } f(q) = 0,$$

$$b_q = a_q \text{ otherwise.}$$

This is fast.

Step 2: “Grover diffusion”.

Negate  $a$  around its average.

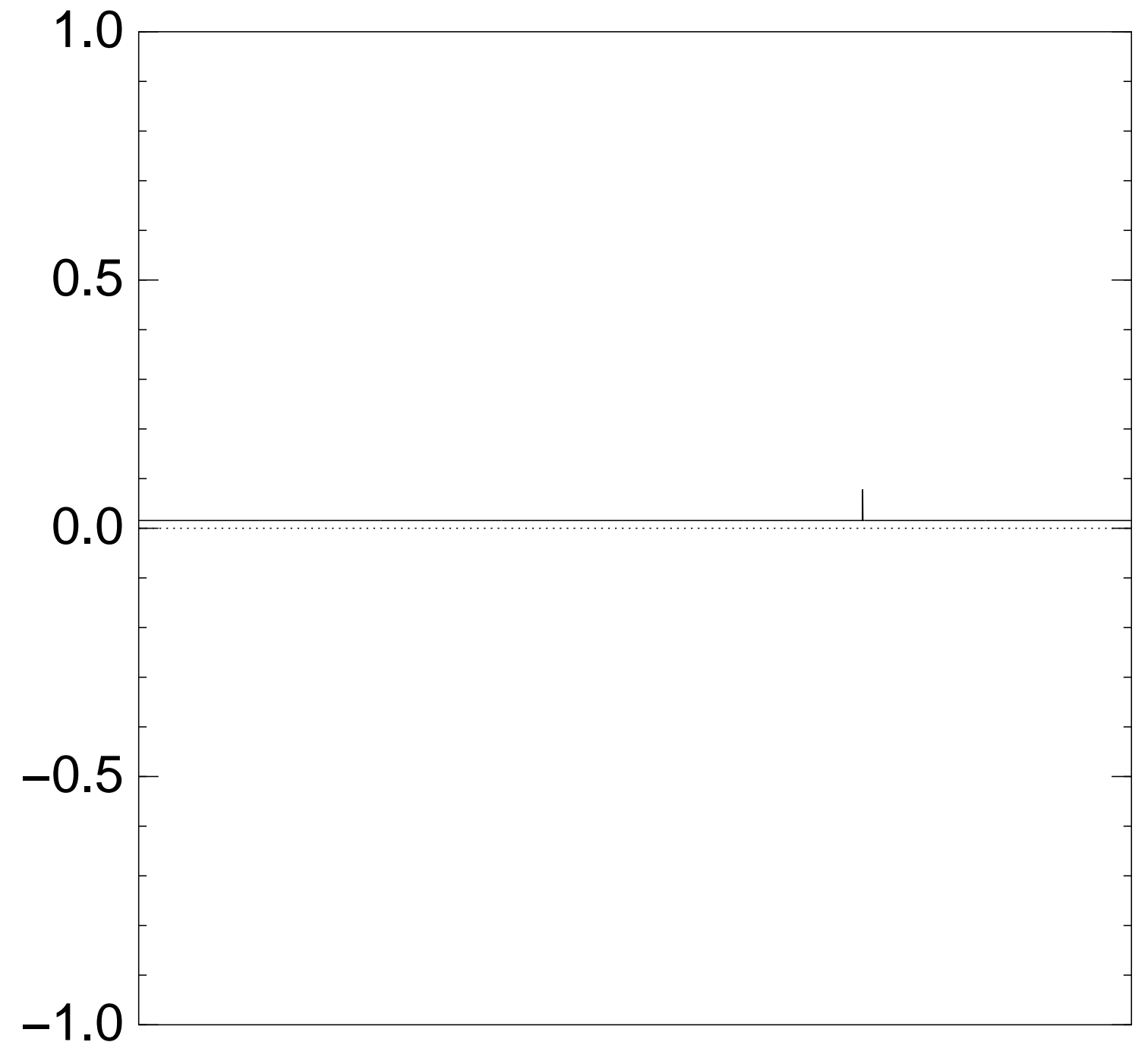
This is also fast.

Repeat Step 1 + Step 2 about  $0.58 \cdot 2^{0.5n}$  times.

Measure the  $n$  qubits.

With high probability this finds  $s$ .

Normalized graph of  $q \mapsto a_q$  for an example with  $n = 12$  after  $2 \times (\text{Step 1} + \text{Step 2})$ :



Start from uniform superposition over all  $n$ -bit strings  $q$ .

Step 1: Set  $a \leftarrow b$  where

$$b_q = -a_q \text{ if } f(q) = 0,$$

$$b_q = a_q \text{ otherwise.}$$

This is fast.

Step 2: “Grover diffusion”.

Negate  $a$  around its average.

This is also fast.

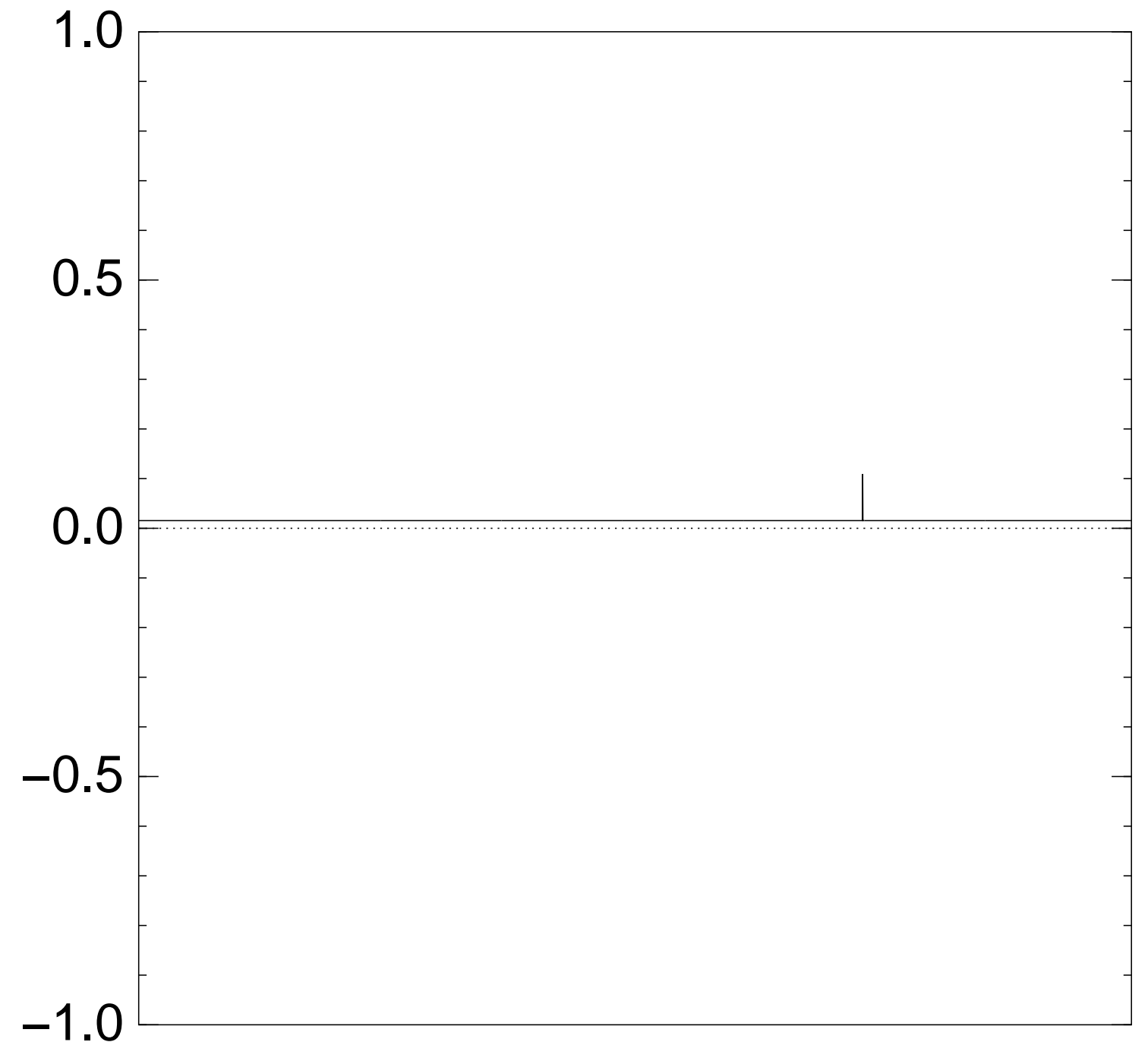
Repeat Step 1 + Step 2

about  $0.58 \cdot 2^{0.5n}$  times.

Measure the  $n$  qubits.

With high probability this finds  $s$ .

Normalized graph of  $q \mapsto a_q$  for an example with  $n = 12$  after  $3 \times (\text{Step 1} + \text{Step 2})$ :





Start from uniform superposition over all  $n$ -bit strings  $q$ .

Step 1: Set  $a \leftarrow b$  where

$$b_q = -a_q \text{ if } f(q) = 0,$$

$$b_q = a_q \text{ otherwise.}$$

This is fast.

Step 2: “Grover diffusion”.

Negate  $a$  around its average.

This is also fast.

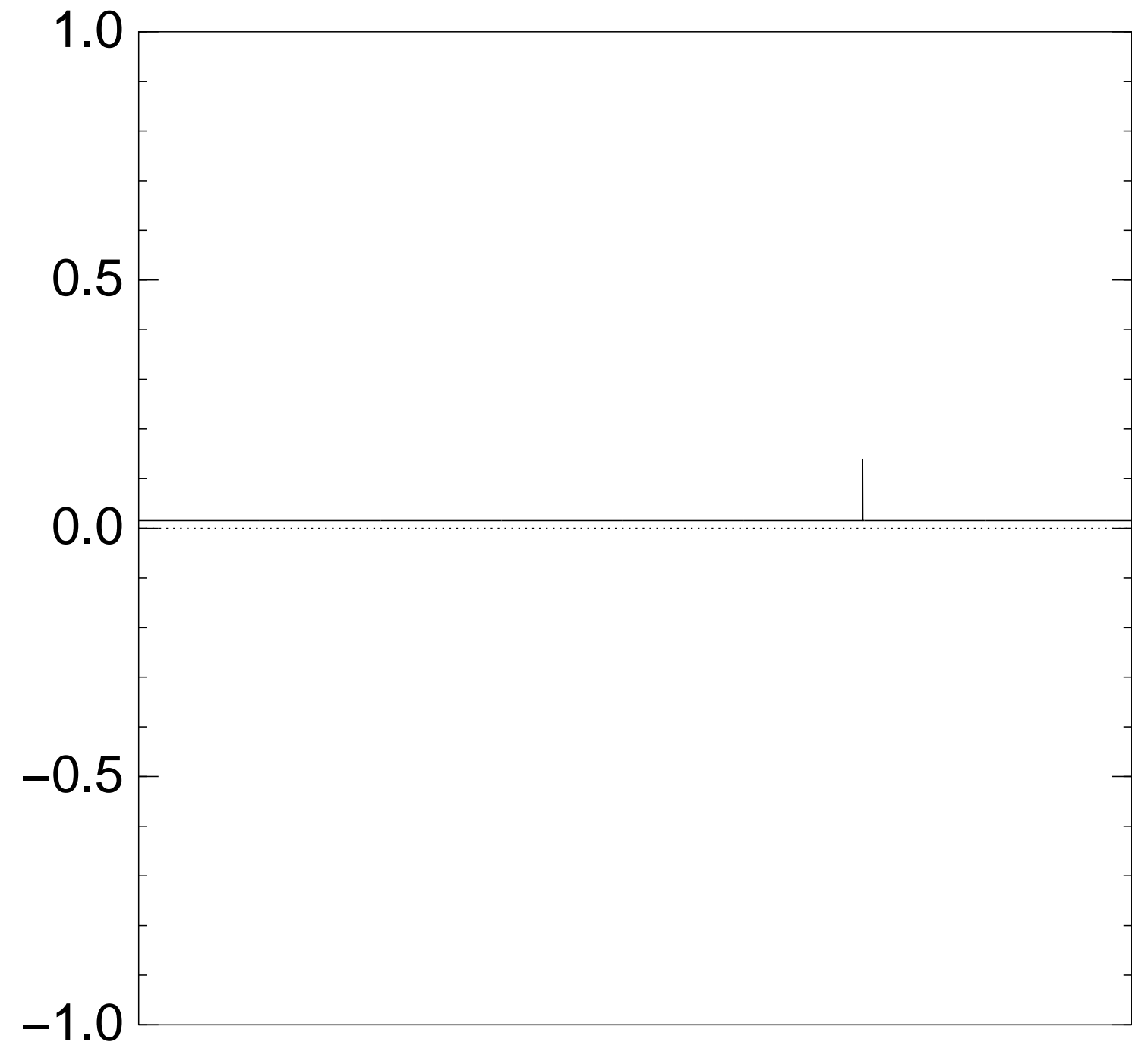
Repeat Step 1 + Step 2

about  $0.58 \cdot 2^{0.5n}$  times.

Measure the  $n$  qubits.

With high probability this finds  $s$ .

Normalized graph of  $q \mapsto a_q$  for an example with  $n = 12$  after  $4 \times$  (Step 1 + Step 2):



Start from uniform superposition over all  $n$ -bit strings  $q$ .

Step 1: Set  $a \leftarrow b$  where

$$b_q = -a_q \text{ if } f(q) = 0,$$

$$b_q = a_q \text{ otherwise.}$$

This is fast.

Step 2: “Grover diffusion”.

Negate  $a$  around its average.

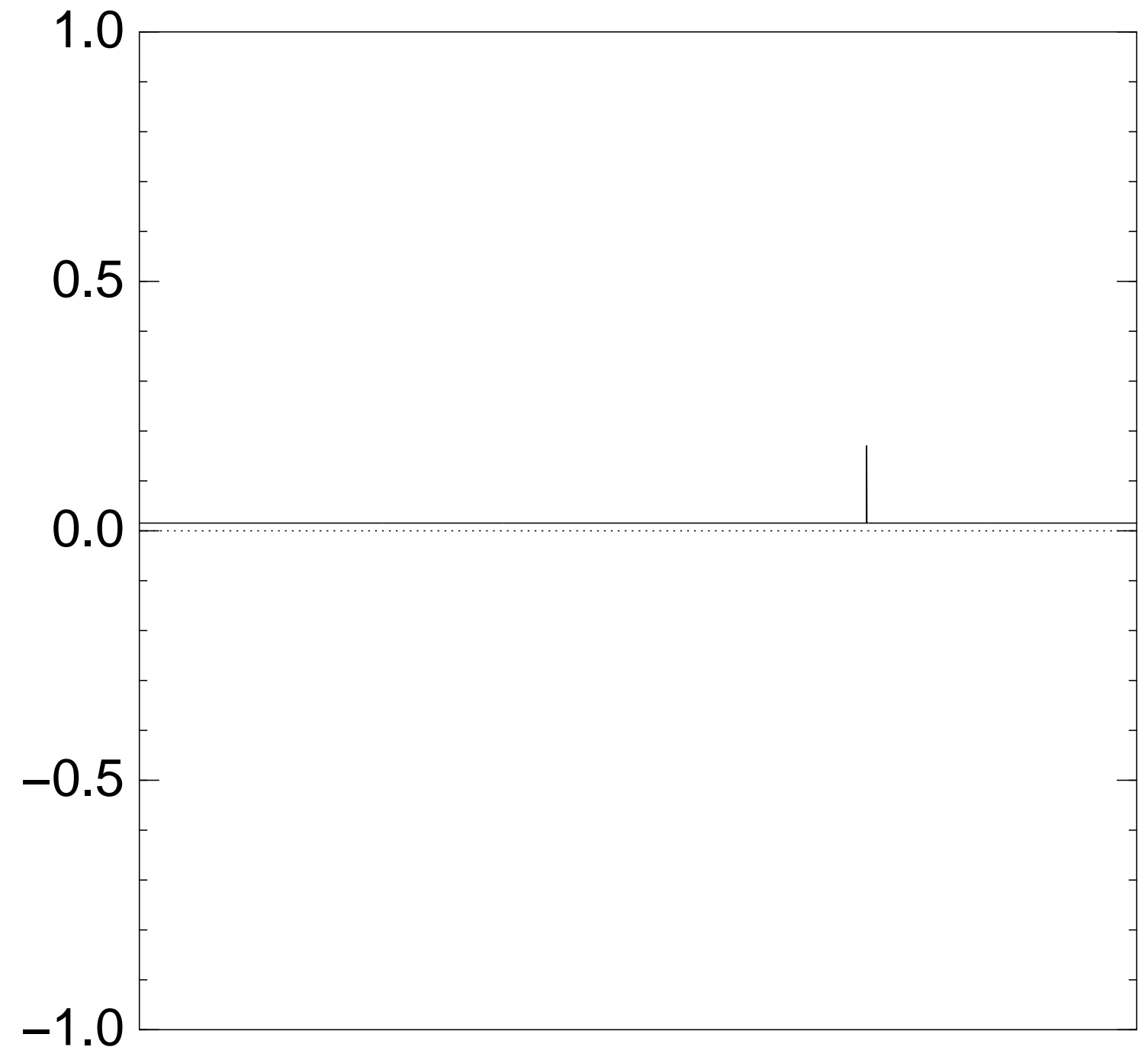
This is also fast.

Repeat Step 1 + Step 2 about  $0.58 \cdot 2^{0.5n}$  times.

Measure the  $n$  qubits.

With high probability this finds  $s$ .

Normalized graph of  $q \mapsto a_q$  for an example with  $n = 12$  after  $5 \times$  (Step 1 + Step 2):



Start from uniform superposition over all  $n$ -bit strings  $q$ .

Step 1: Set  $a \leftarrow b$  where

$$b_q = -a_q \text{ if } f(q) = 0,$$

$$b_q = a_q \text{ otherwise.}$$

This is fast.

Step 2: “Grover diffusion”.

Negate  $a$  around its average.

This is also fast.

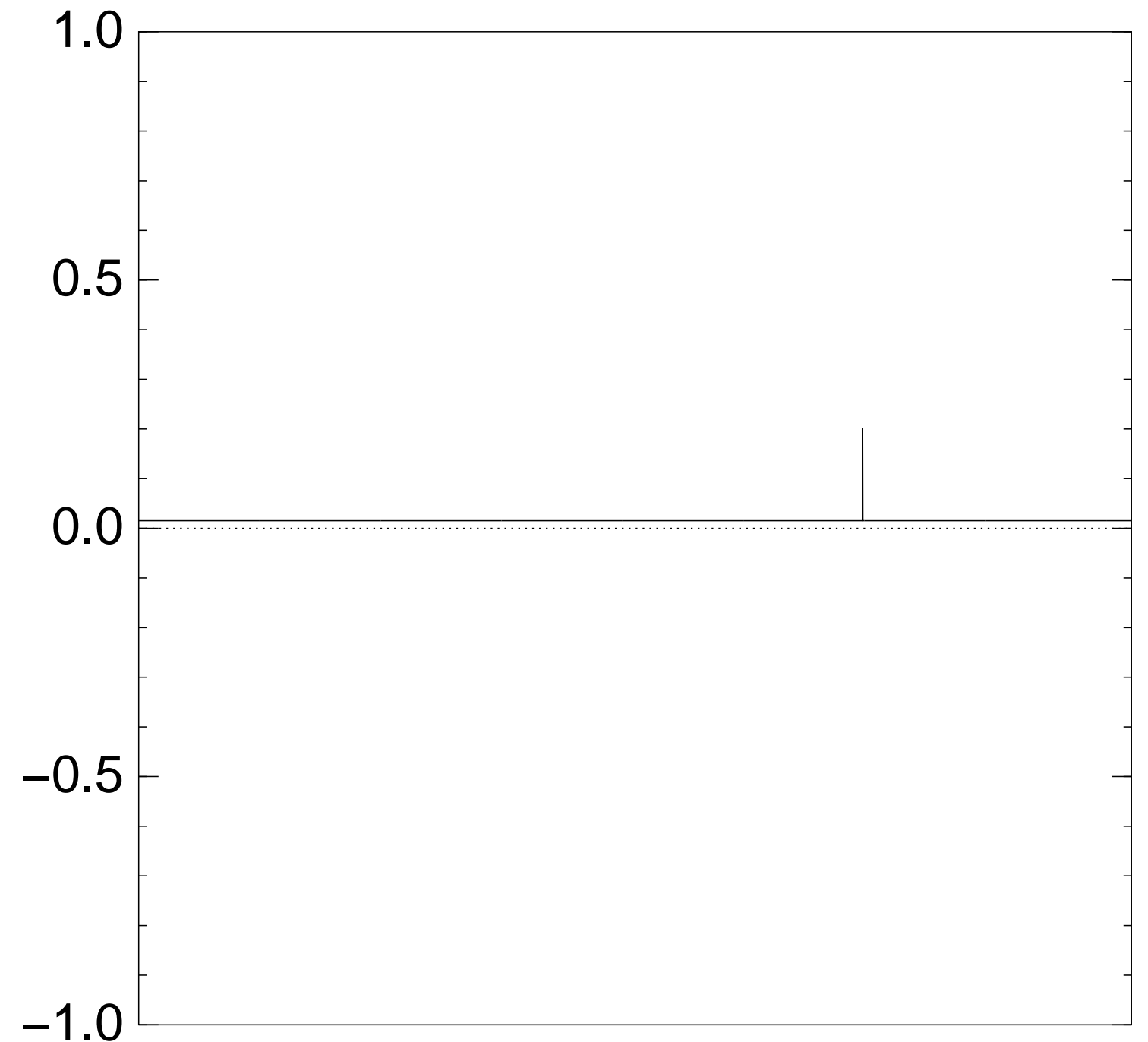
Repeat Step 1 + Step 2

about  $0.58 \cdot 2^{0.5n}$  times.

Measure the  $n$  qubits.

With high probability this finds  $s$ .

Normalized graph of  $q \mapsto a_q$  for an example with  $n = 12$  after  $6 \times (\text{Step 1} + \text{Step 2})$ :



Start from uniform superposition over all  $n$ -bit strings  $q$ .

Step 1: Set  $a \leftarrow b$  where

$$b_q = -a_q \text{ if } f(q) = 0,$$

$$b_q = a_q \text{ otherwise.}$$

This is fast.

Step 2: “Grover diffusion”.

Negate  $a$  around its average.

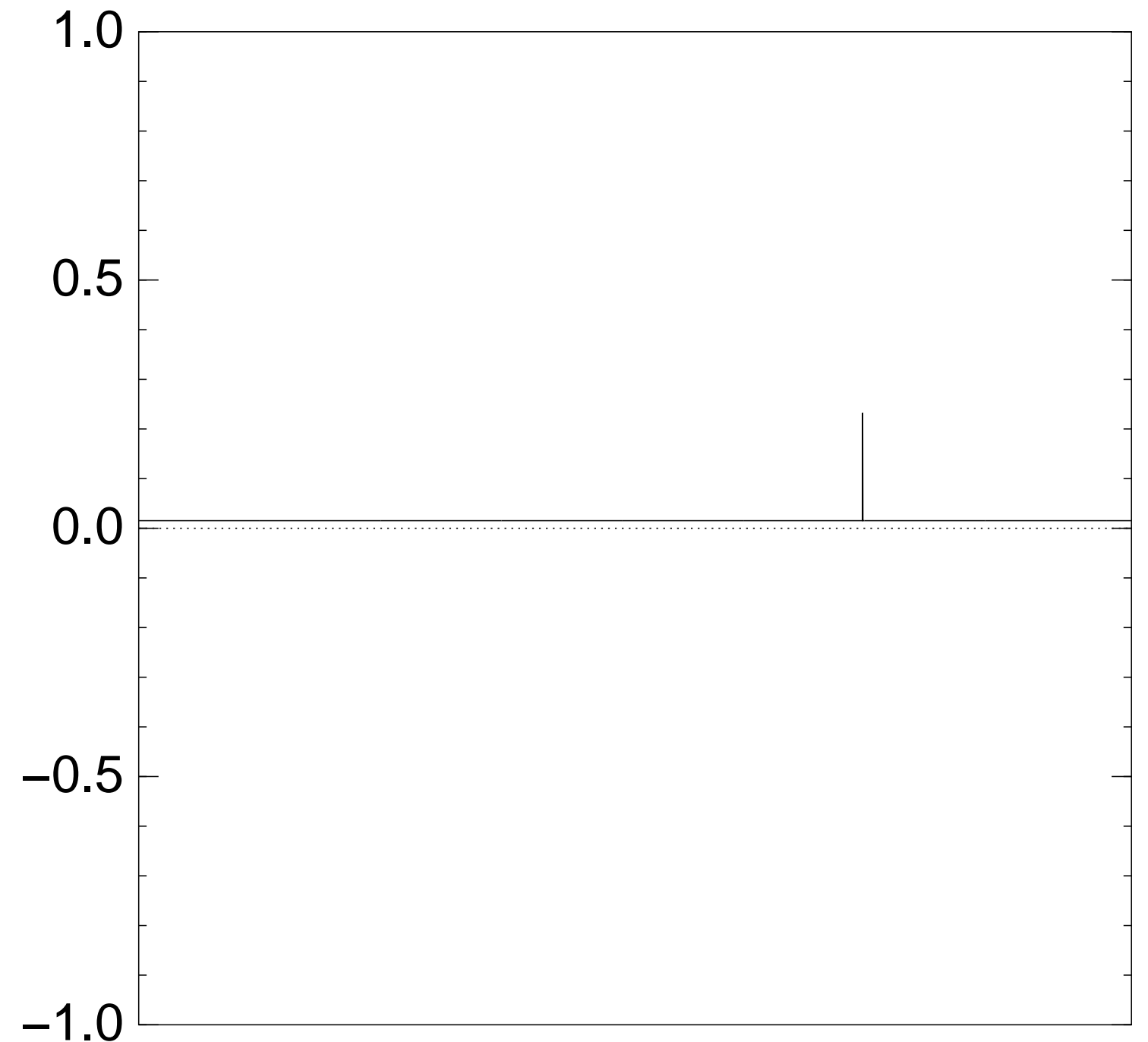
This is also fast.

Repeat Step 1 + Step 2 about  $0.58 \cdot 2^{0.5n}$  times.

Measure the  $n$  qubits.

With high probability this finds  $s$ .

Normalized graph of  $q \mapsto a_q$  for an example with  $n = 12$  after  $7 \times$  (Step 1 + Step 2):



Start from uniform superposition over all  $n$ -bit strings  $q$ .

Step 1: Set  $a \leftarrow b$  where

$$b_q = -a_q \text{ if } f(q) = 0,$$

$$b_q = a_q \text{ otherwise.}$$

This is fast.

Step 2: “Grover diffusion”.

Negate  $a$  around its average.

This is also fast.

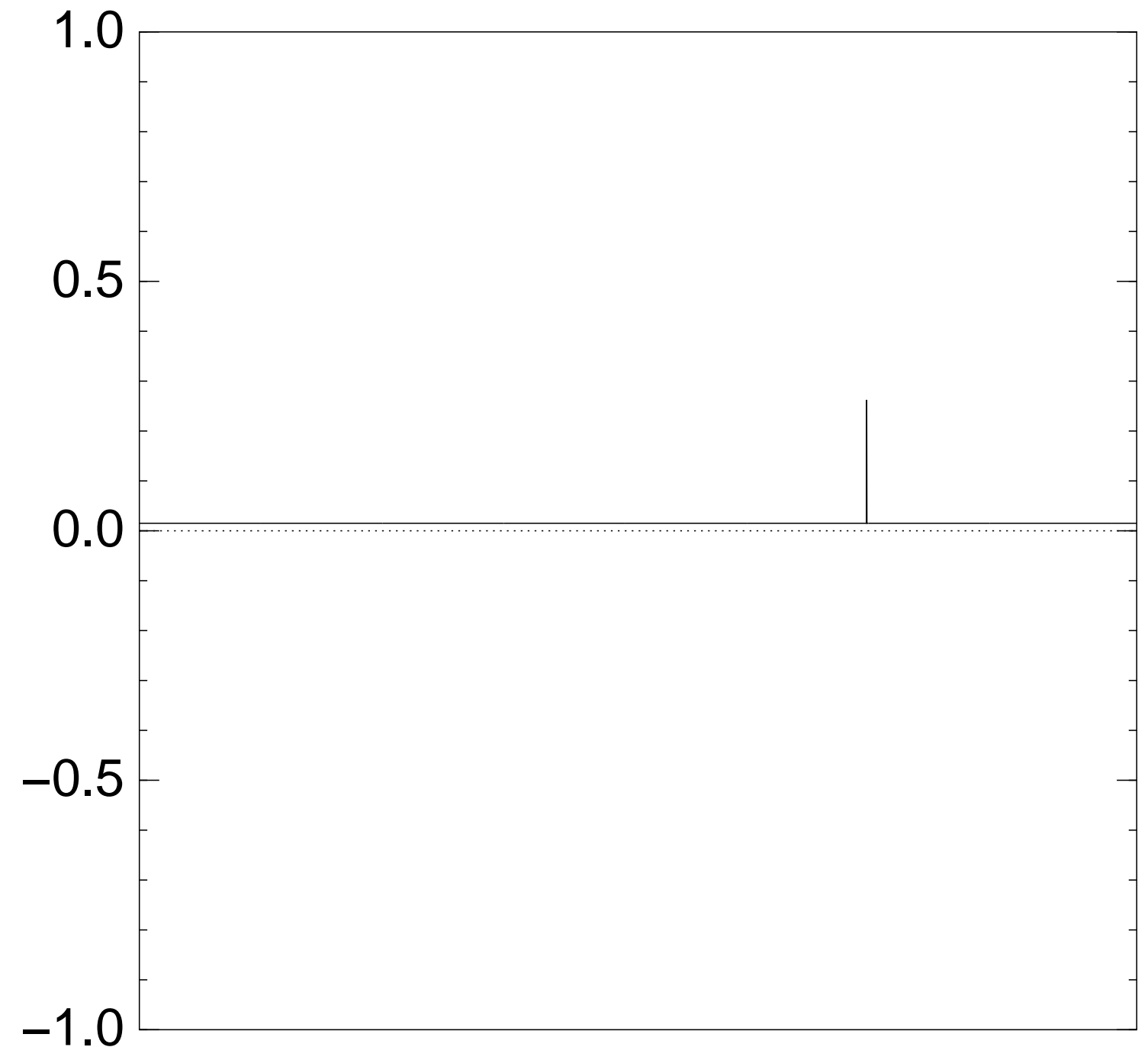
Repeat Step 1 + Step 2

about  $0.58 \cdot 2^{0.5n}$  times.

Measure the  $n$  qubits.

With high probability this finds  $s$ .

Normalized graph of  $q \mapsto a_q$  for an example with  $n = 12$  after  $8 \times (\text{Step 1} + \text{Step 2})$ :



Start from uniform superposition over all  $n$ -bit strings  $q$ .

Step 1: Set  $a \leftarrow b$  where

$$b_q = -a_q \text{ if } f(q) = 0,$$

$$b_q = a_q \text{ otherwise.}$$

This is fast.

Step 2: “Grover diffusion”.

Negate  $a$  around its average.

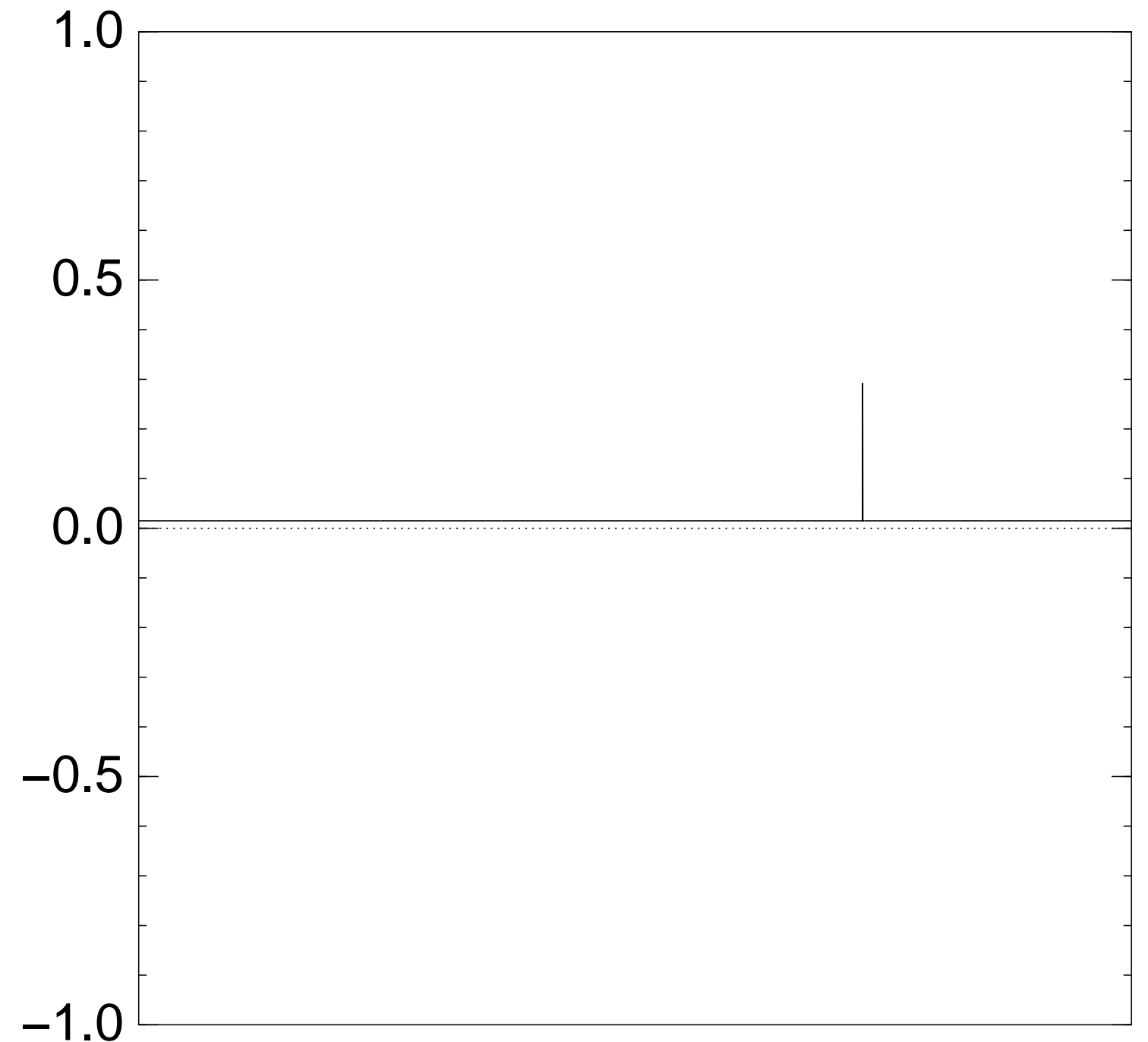
This is also fast.

Repeat Step 1 + Step 2 about  $0.58 \cdot 2^{0.5n}$  times.

Measure the  $n$  qubits.

With high probability this finds  $s$ .

Normalized graph of  $q \mapsto a_q$  for an example with  $n = 12$  after  $9 \times (\text{Step 1} + \text{Step 2})$ :



Start from uniform superposition over all  $n$ -bit strings  $q$ .

Step 1: Set  $a \leftarrow b$  where

$$b_q = -a_q \text{ if } f(q) = 0,$$

$$b_q = a_q \text{ otherwise.}$$

This is fast.

Step 2: “Grover diffusion”.

Negate  $a$  around its average.

This is also fast.

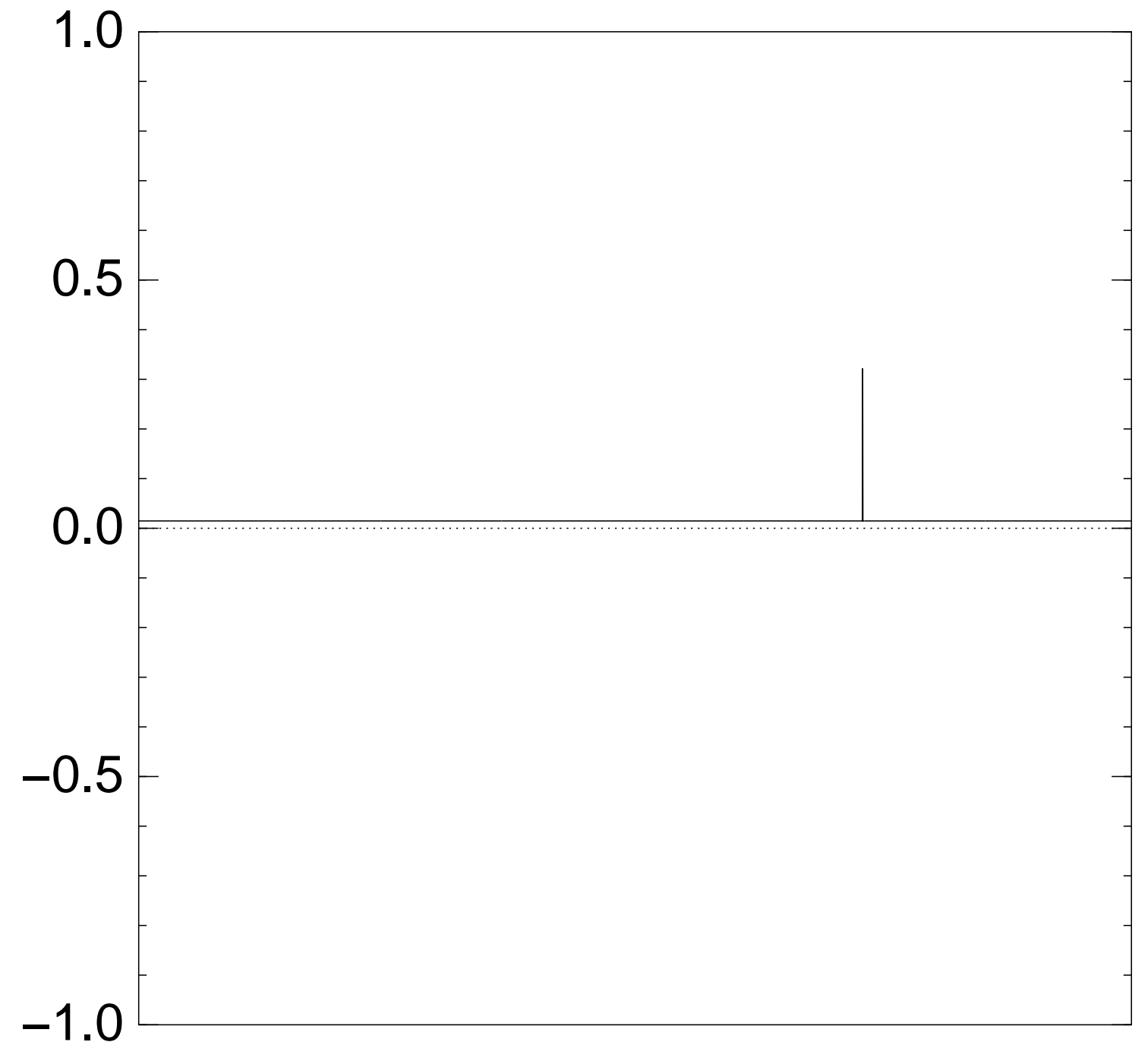
Repeat Step 1 + Step 2

about  $0.58 \cdot 2^{0.5n}$  times.

Measure the  $n$  qubits.

With high probability this finds  $s$ .

Normalized graph of  $q \mapsto a_q$  for an example with  $n = 12$  after  $10 \times (\text{Step 1} + \text{Step 2})$ :



Start from uniform superposition over all  $n$ -bit strings  $q$ .

Step 1: Set  $a \leftarrow b$  where

$$b_q = -a_q \text{ if } f(q) = 0,$$

$$b_q = a_q \text{ otherwise.}$$

This is fast.

Step 2: “Grover diffusion”.

Negate  $a$  around its average.

This is also fast.

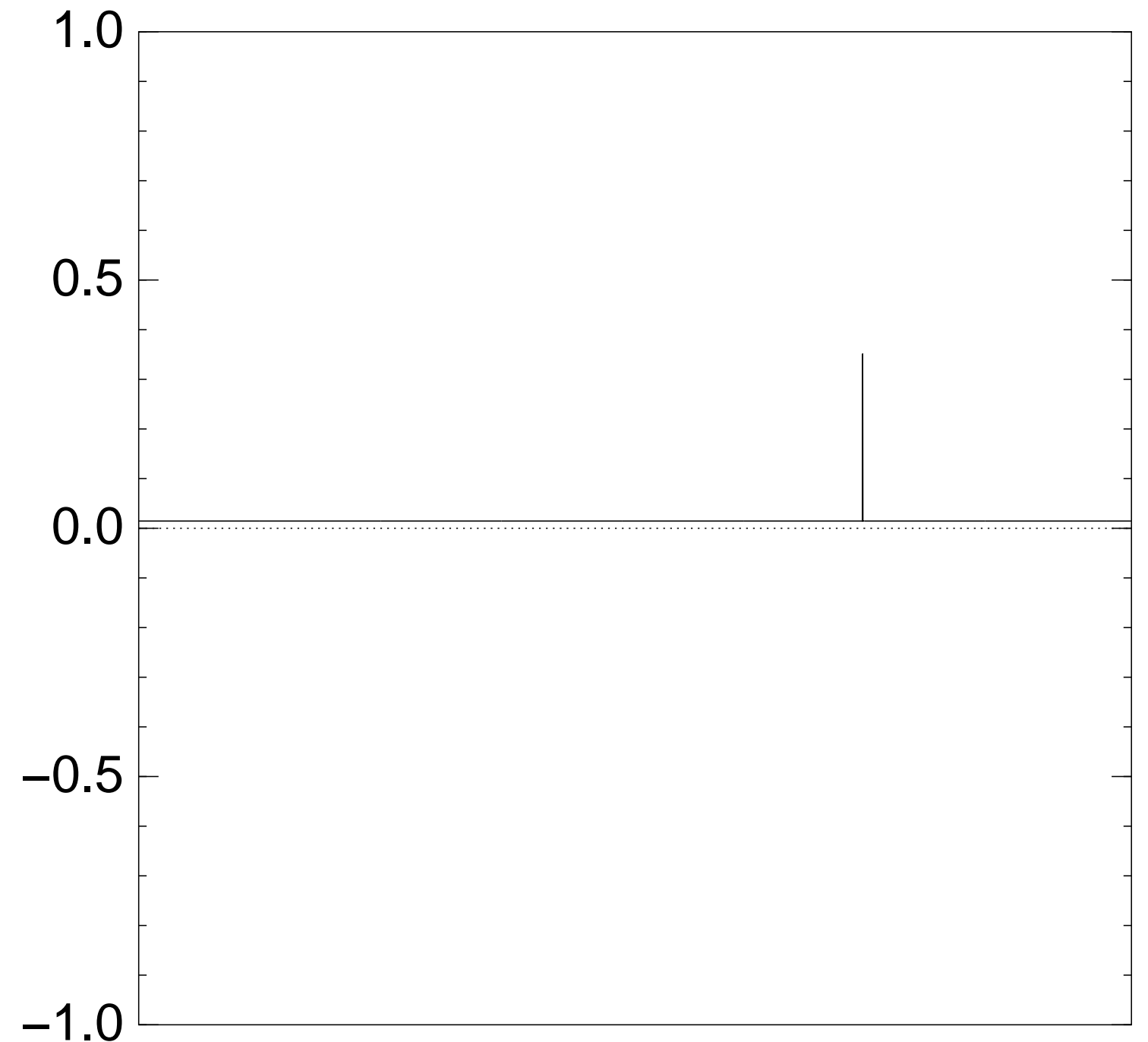
Repeat Step 1 + Step 2

about  $0.58 \cdot 2^{0.5n}$  times.

Measure the  $n$  qubits.

With high probability this finds  $s$ .

Normalized graph of  $q \mapsto a_q$  for an example with  $n = 12$  after  $11 \times (\text{Step 1} + \text{Step 2})$ :





Start from uniform superposition over all  $n$ -bit strings  $q$ .

Step 1: Set  $a \leftarrow b$  where

$$b_q = -a_q \text{ if } f(q) = 0,$$

$$b_q = a_q \text{ otherwise.}$$

This is fast.

Step 2: “Grover diffusion”.

Negate  $a$  around its average.

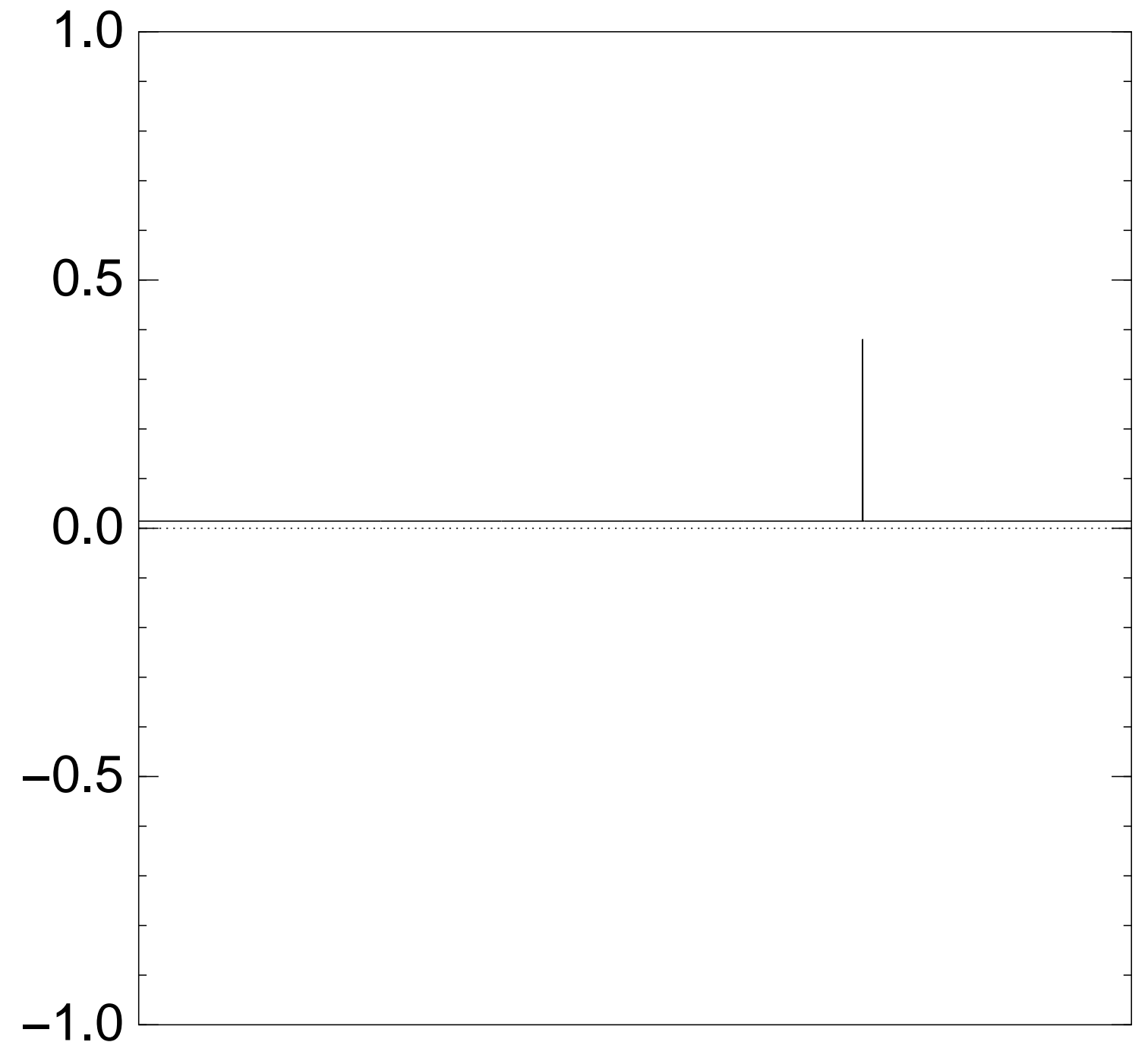
This is also fast.

Repeat Step 1 + Step 2 about  $0.58 \cdot 2^{0.5n}$  times.

Measure the  $n$  qubits.

With high probability this finds  $s$ .

Normalized graph of  $q \mapsto a_q$  for an example with  $n = 12$  after  $12 \times (\text{Step 1} + \text{Step 2})$ :



Start from uniform superposition over all  $n$ -bit strings  $q$ .

Step 1: Set  $a \leftarrow b$  where

$$b_q = -a_q \text{ if } f(q) = 0,$$

$$b_q = a_q \text{ otherwise.}$$

This is fast.

Step 2: “Grover diffusion”.

Negate  $a$  around its average.

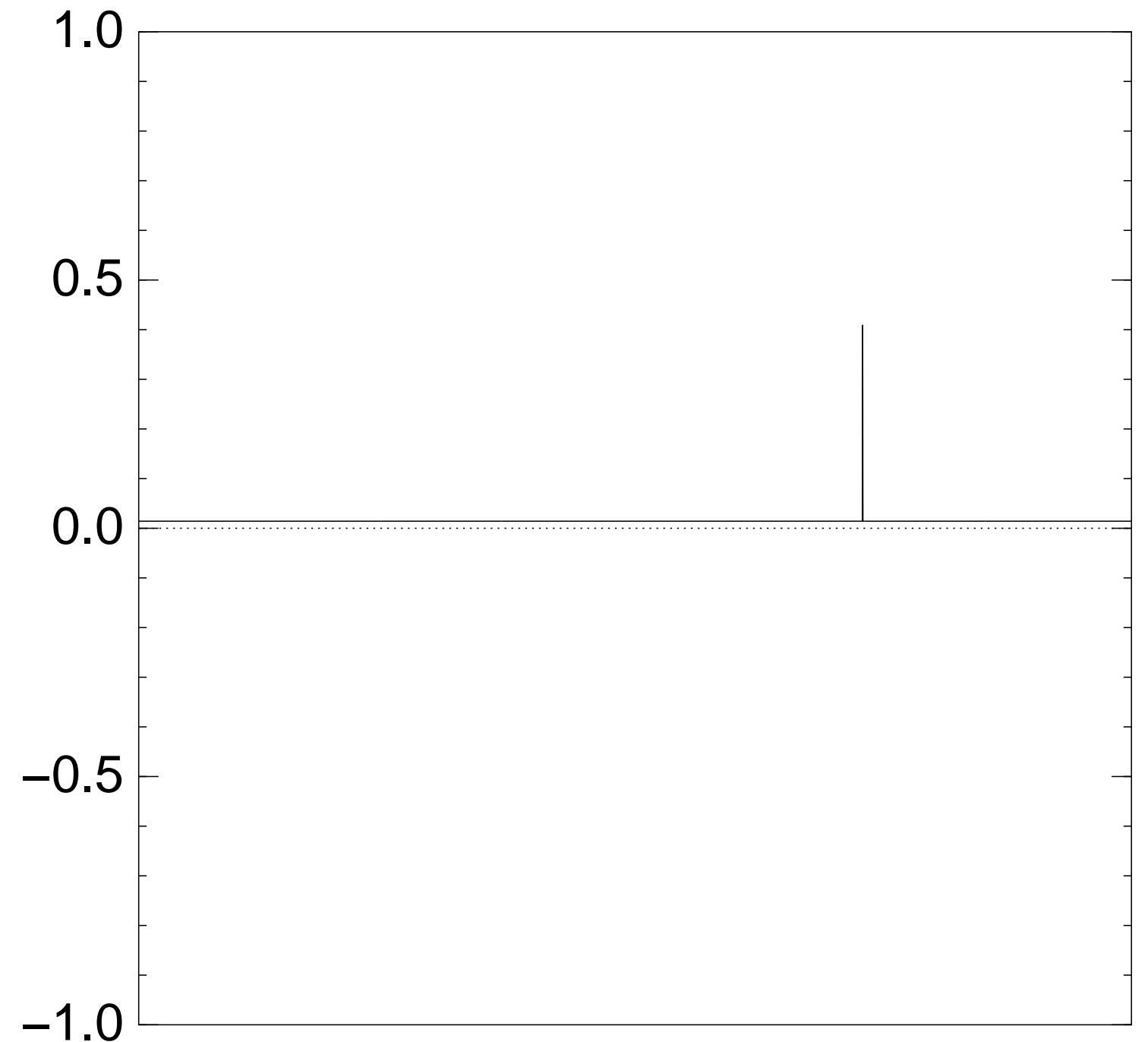
This is also fast.

Repeat Step 1 + Step 2 about  $0.58 \cdot 2^{0.5n}$  times.

Measure the  $n$  qubits.

With high probability this finds  $s$ .

Normalized graph of  $q \mapsto a_q$  for an example with  $n = 12$  after  $13 \times (\text{Step 1} + \text{Step 2})$ :



Start from uniform superposition over all  $n$ -bit strings  $q$ .

Step 1: Set  $a \leftarrow b$  where

$$b_q = -a_q \text{ if } f(q) = 0,$$

$$b_q = a_q \text{ otherwise.}$$

This is fast.

Step 2: “Grover diffusion”.

Negate  $a$  around its average.

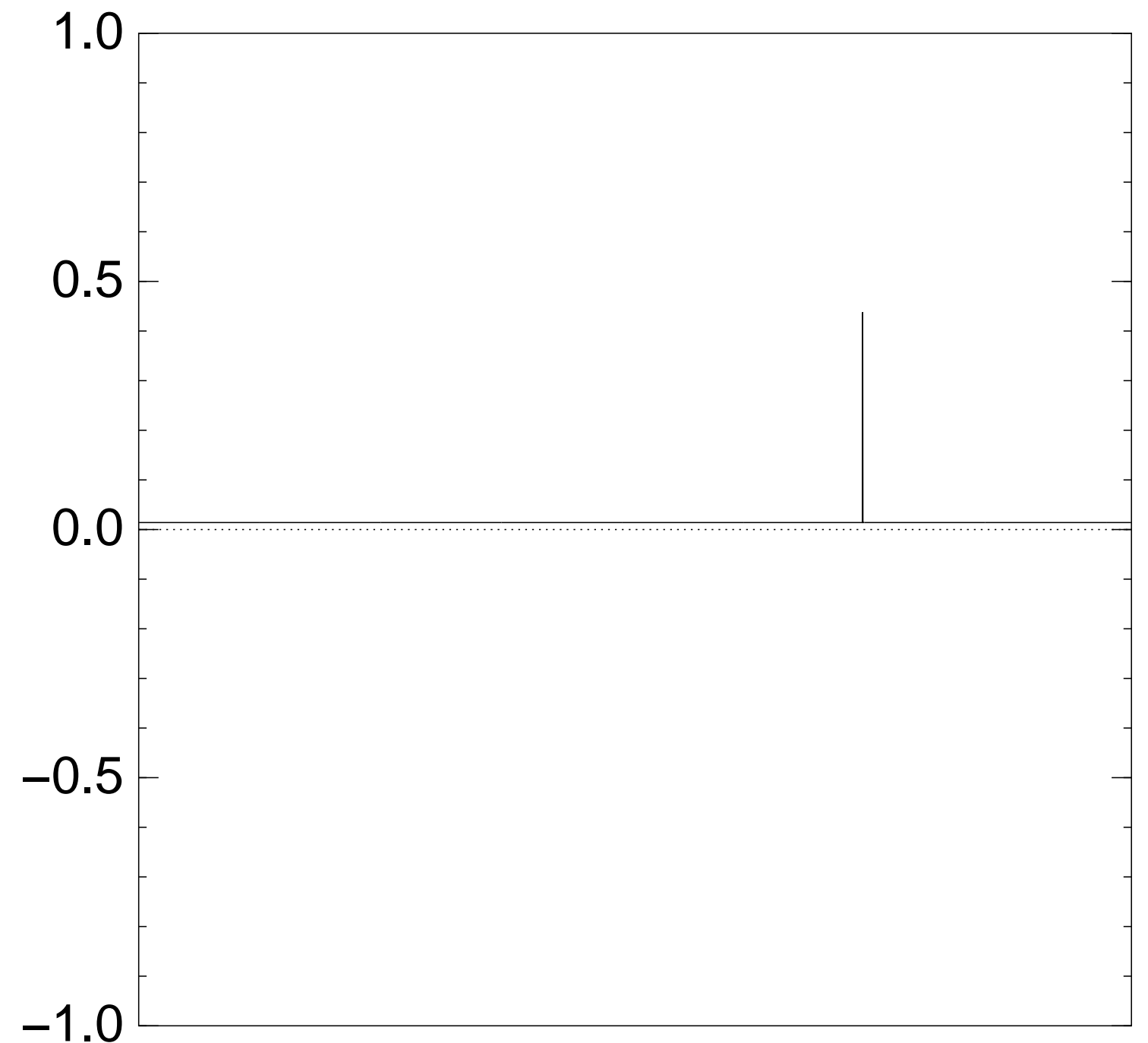
This is also fast.

Repeat Step 1 + Step 2 about  $0.58 \cdot 2^{0.5n}$  times.

Measure the  $n$  qubits.

With high probability this finds  $s$ .

Normalized graph of  $q \mapsto a_q$  for an example with  $n = 12$  after  $14 \times (\text{Step 1} + \text{Step 2})$ :



Start from uniform superposition over all  $n$ -bit strings  $q$ .

Step 1: Set  $a \leftarrow b$  where

$$b_q = -a_q \text{ if } f(q) = 0,$$

$$b_q = a_q \text{ otherwise.}$$

This is fast.

Step 2: “Grover diffusion”.

Negate  $a$  around its average.

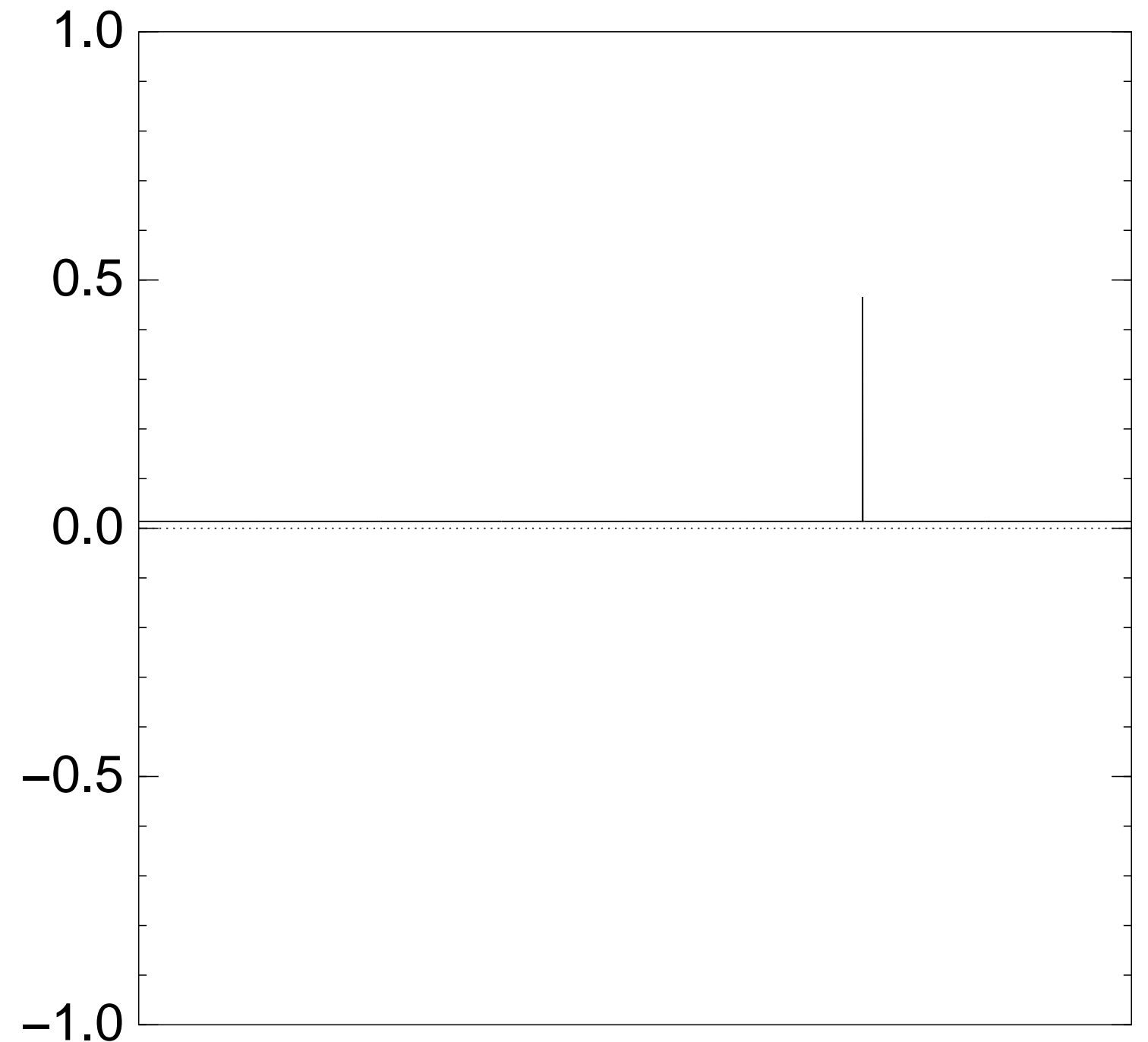
This is also fast.

Repeat Step 1 + Step 2 about  $0.58 \cdot 2^{0.5n}$  times.

Measure the  $n$  qubits.

With high probability this finds  $s$ .

Normalized graph of  $q \mapsto a_q$  for an example with  $n = 12$  after  $15 \times (\text{Step 1} + \text{Step 2})$ :



Start from uniform superposition over all  $n$ -bit strings  $q$ .

Step 1: Set  $a \leftarrow b$  where

$$b_q = -a_q \text{ if } f(q) = 0,$$

$$b_q = a_q \text{ otherwise.}$$

This is fast.

Step 2: “Grover diffusion”.

Negate  $a$  around its average.

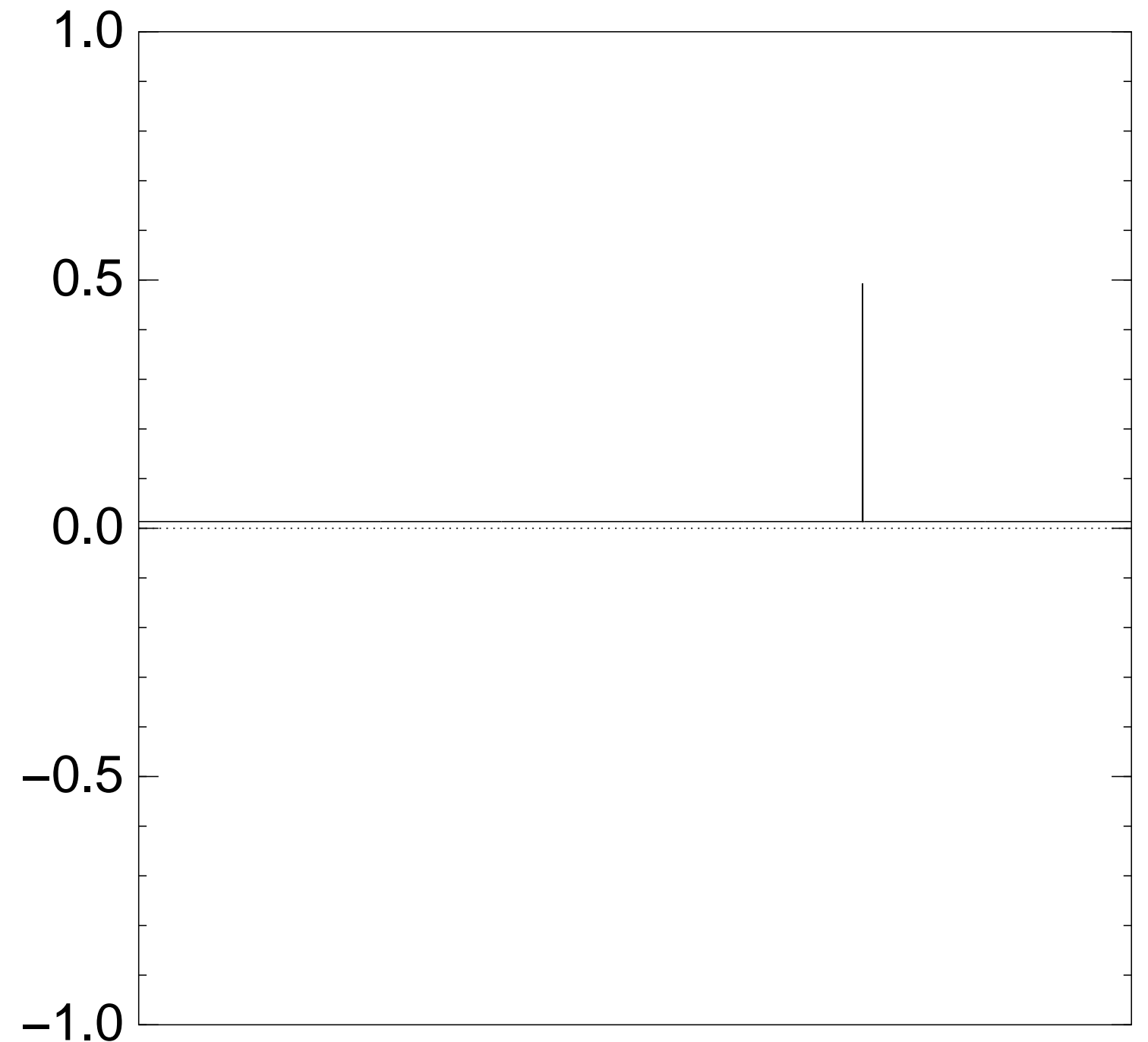
This is also fast.

Repeat Step 1 + Step 2 about  $0.58 \cdot 2^{0.5n}$  times.

Measure the  $n$  qubits.

With high probability this finds  $s$ .

Normalized graph of  $q \mapsto a_q$  for an example with  $n = 12$  after  $16 \times (\text{Step 1} + \text{Step 2})$ :



Start from uniform superposition over all  $n$ -bit strings  $q$ .

Step 1: Set  $a \leftarrow b$  where

$$b_q = -a_q \text{ if } f(q) = 0,$$

$$b_q = a_q \text{ otherwise.}$$

This is fast.

Step 2: “Grover diffusion”.

Negate  $a$  around its average.

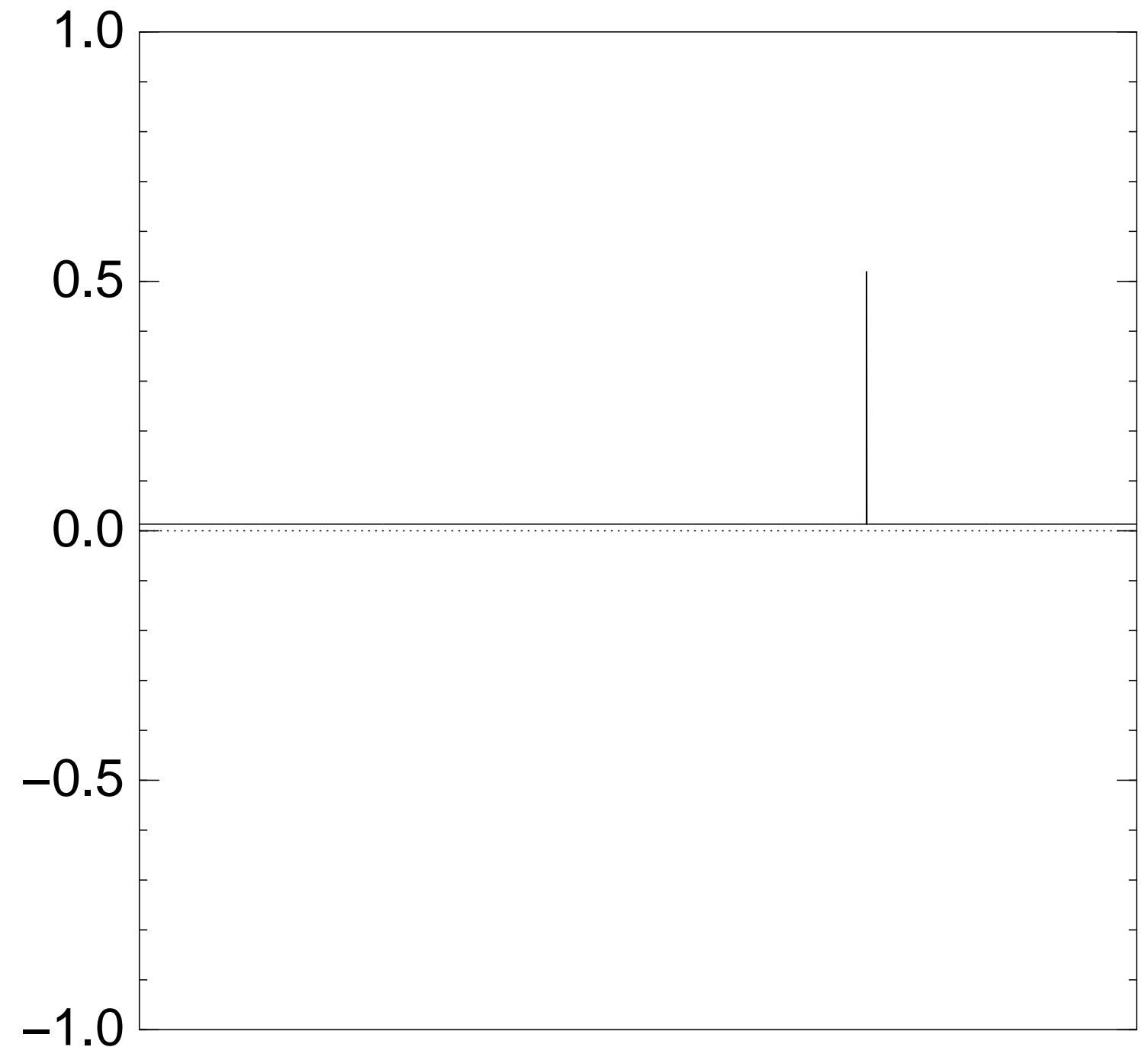
This is also fast.

Repeat Step 1 + Step 2 about  $0.58 \cdot 2^{0.5n}$  times.

Measure the  $n$  qubits.

With high probability this finds  $s$ .

Normalized graph of  $q \mapsto a_q$  for an example with  $n = 12$  after  $17 \times (\text{Step 1} + \text{Step 2})$ :



Start from uniform superposition over all  $n$ -bit strings  $q$ .

Step 1: Set  $a \leftarrow b$  where

$$b_q = -a_q \text{ if } f(q) = 0,$$

$$b_q = a_q \text{ otherwise.}$$

This is fast.

Step 2: “Grover diffusion”.

Negate  $a$  around its average.

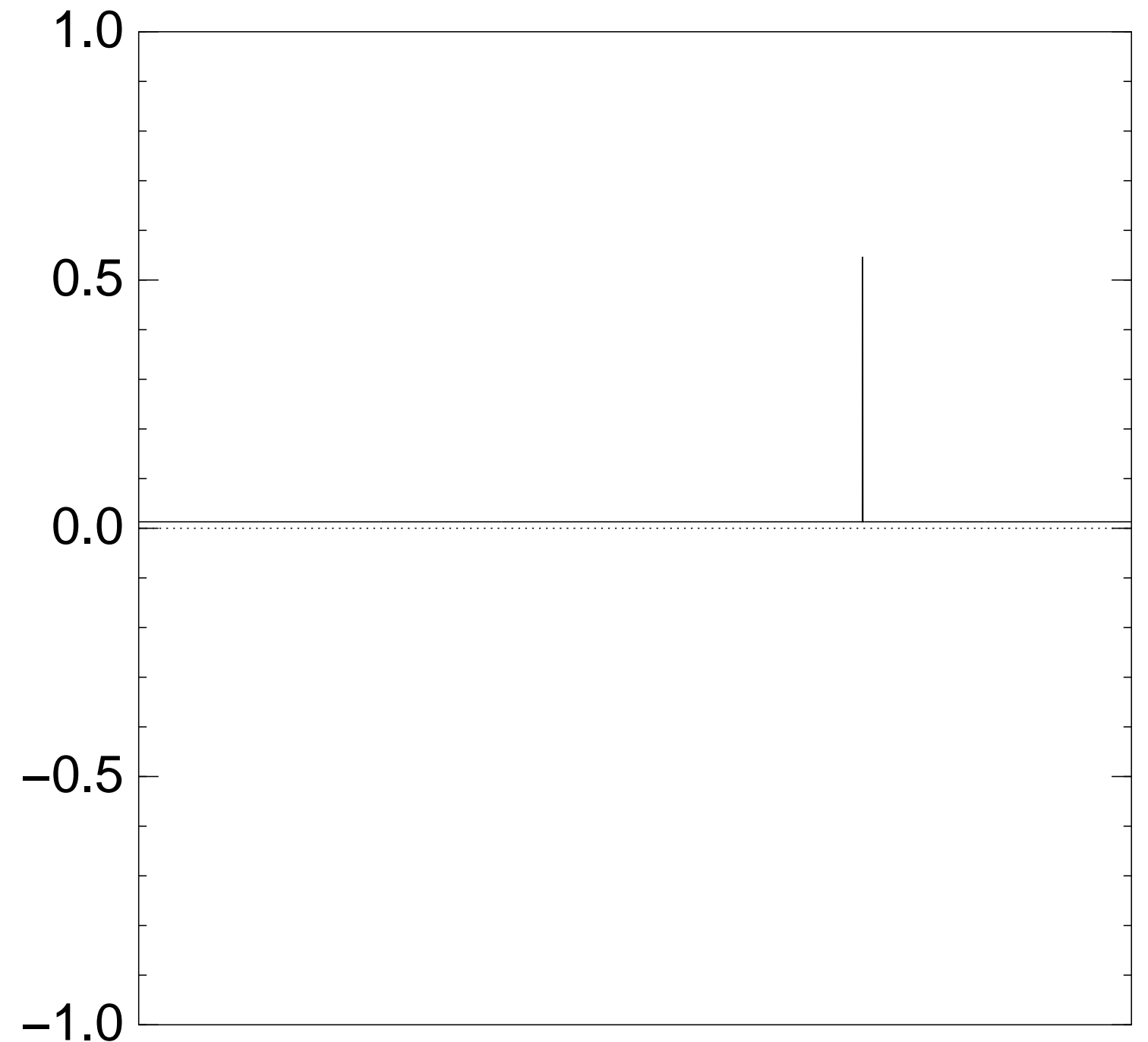
This is also fast.

Repeat Step 1 + Step 2 about  $0.58 \cdot 2^{0.5n}$  times.

Measure the  $n$  qubits.

With high probability this finds  $s$ .

Normalized graph of  $q \mapsto a_q$  for an example with  $n = 12$  after  $18 \times (\text{Step 1} + \text{Step 2})$ :



Start from uniform superposition over all  $n$ -bit strings  $q$ .

Step 1: Set  $a \leftarrow b$  where

$$b_q = -a_q \text{ if } f(q) = 0,$$

$$b_q = a_q \text{ otherwise.}$$

This is fast.

Step 2: “Grover diffusion”.

Negate  $a$  around its average.

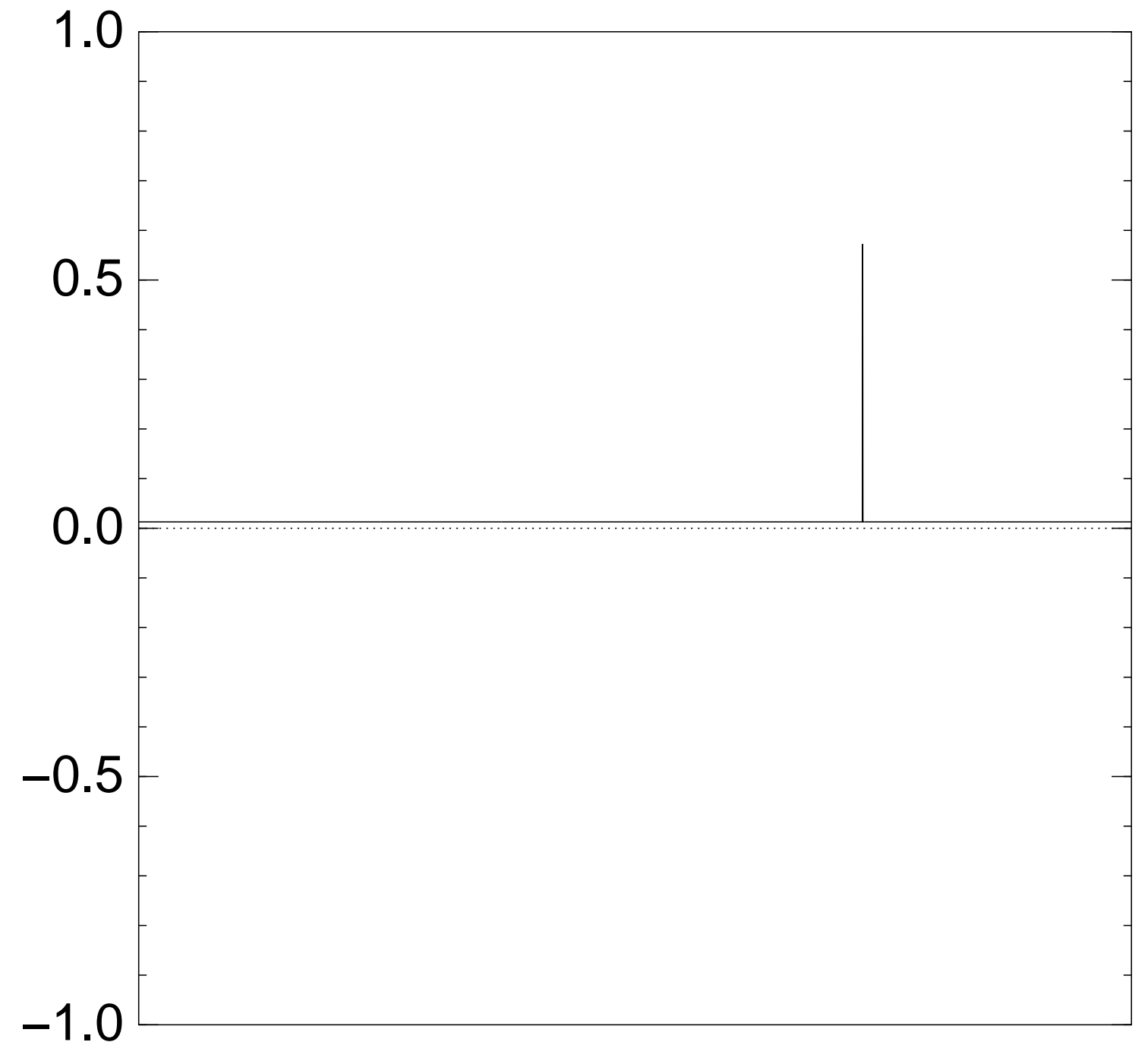
This is also fast.

Repeat Step 1 + Step 2 about  $0.58 \cdot 2^{0.5n}$  times.

Measure the  $n$  qubits.

With high probability this finds  $s$ .

Normalized graph of  $q \mapsto a_q$  for an example with  $n = 12$  after  $19 \times (\text{Step 1} + \text{Step 2})$ :





Start from uniform superposition over all  $n$ -bit strings  $q$ .

Step 1: Set  $a \leftarrow b$  where

$$b_q = -a_q \text{ if } f(q) = 0,$$

$$b_q = a_q \text{ otherwise.}$$

This is fast.

Step 2: “Grover diffusion”.

Negate  $a$  around its average.

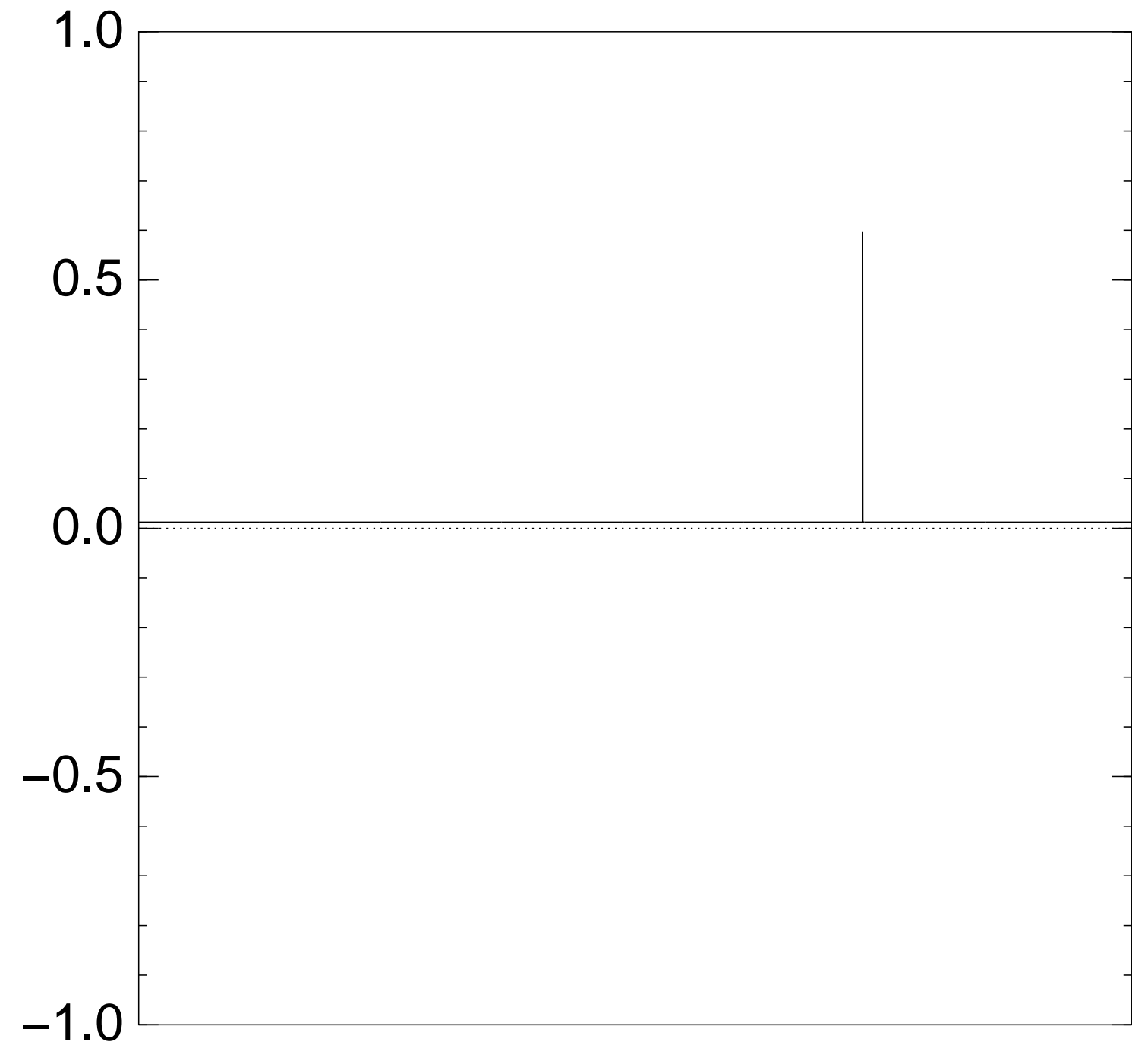
This is also fast.

Repeat Step 1 + Step 2 about  $0.58 \cdot 2^{0.5n}$  times.

Measure the  $n$  qubits.

With high probability this finds  $s$ .

Normalized graph of  $q \mapsto a_q$  for an example with  $n = 12$  after  $20 \times (\text{Step 1} + \text{Step 2})$ :



Start from uniform superposition over all  $n$ -bit strings  $q$ .

Step 1: Set  $a \leftarrow b$  where

$$b_q = -a_q \text{ if } f(q) = 0,$$

$$b_q = a_q \text{ otherwise.}$$

This is fast.

Step 2: “Grover diffusion”.

Negate  $a$  around its average.

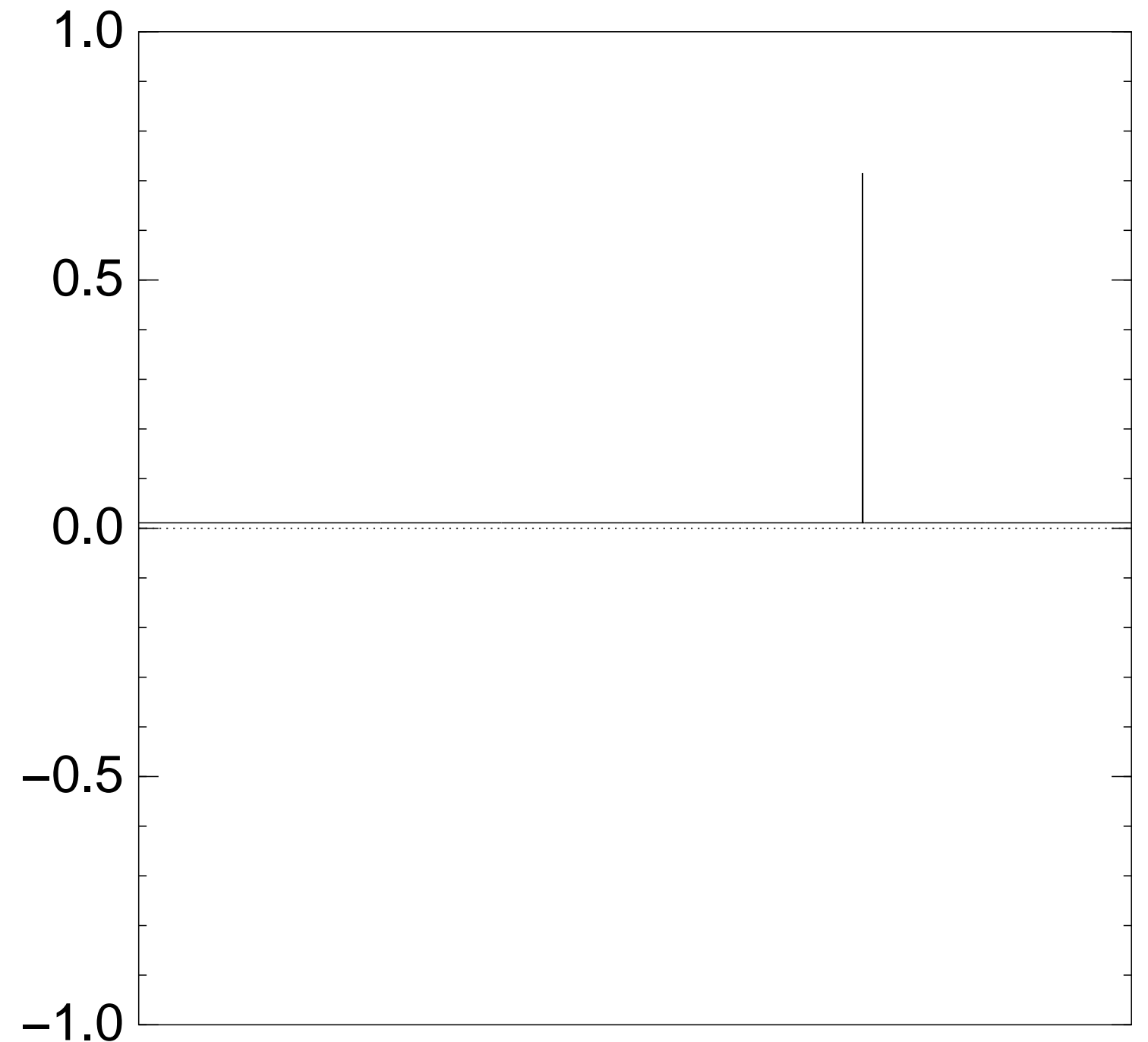
This is also fast.

Repeat Step 1 + Step 2 about  $0.58 \cdot 2^{0.5n}$  times.

Measure the  $n$  qubits.

With high probability this finds  $s$ .

Normalized graph of  $q \mapsto a_q$  for an example with  $n = 12$  after  $25 \times (\text{Step 1} + \text{Step 2})$ :



Start from uniform superposition over all  $n$ -bit strings  $q$ .

Step 1: Set  $a \leftarrow b$  where

$$b_q = -a_q \text{ if } f(q) = 0,$$

$$b_q = a_q \text{ otherwise.}$$

This is fast.

Step 2: “Grover diffusion”.

Negate  $a$  around its average.

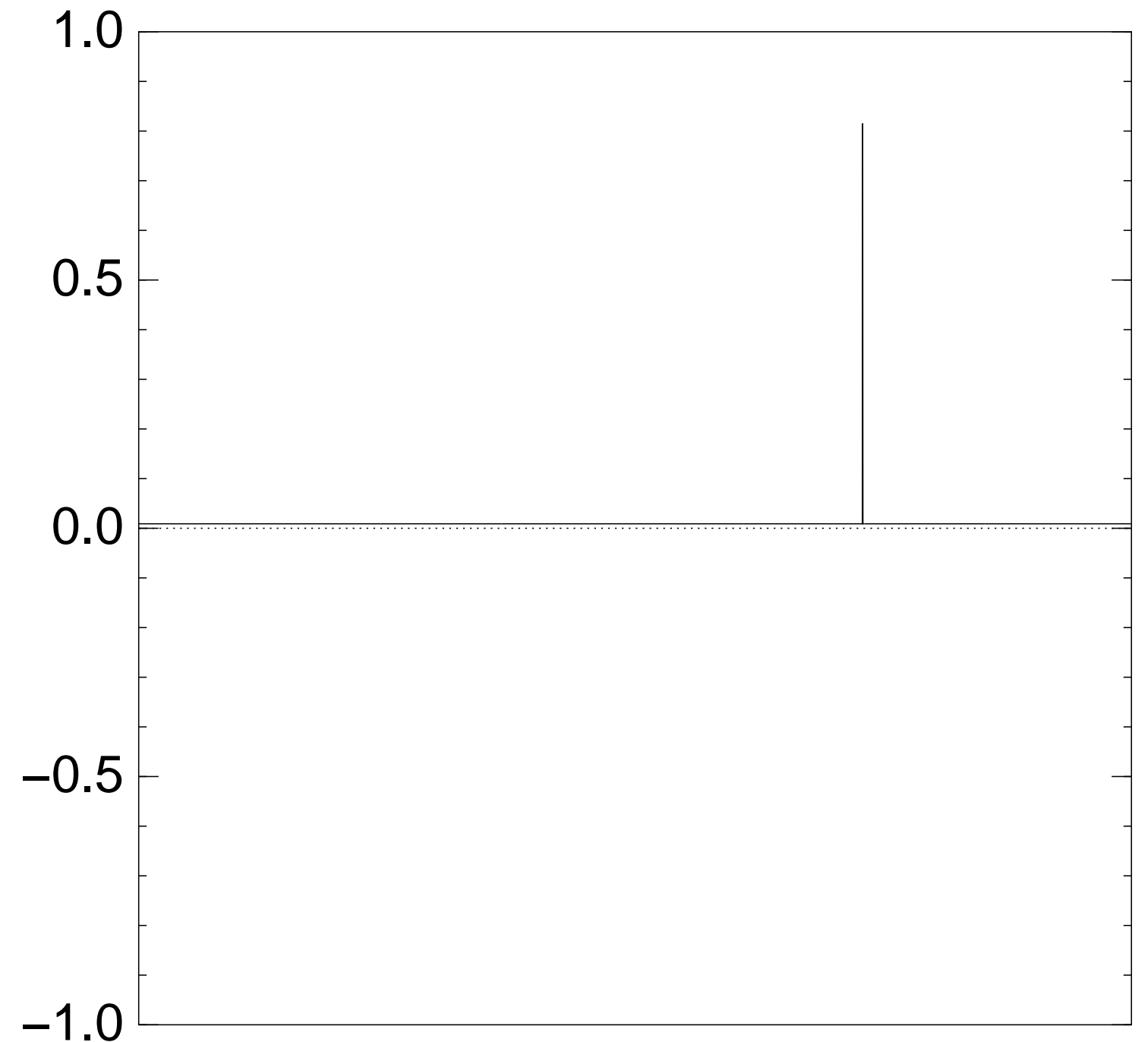
This is also fast.

Repeat Step 1 + Step 2 about  $0.58 \cdot 2^{0.5n}$  times.

Measure the  $n$  qubits.

With high probability this finds  $s$ .

Normalized graph of  $q \mapsto a_q$  for an example with  $n = 12$  after  $30 \times (\text{Step 1} + \text{Step 2})$ :



Start from uniform superposition over all  $n$ -bit strings  $q$ .

Step 1: Set  $a \leftarrow b$  where

$$b_q = -a_q \text{ if } f(q) = 0,$$

$$b_q = a_q \text{ otherwise.}$$

This is fast.

Step 2: “Grover diffusion”.

Negate  $a$  around its average.

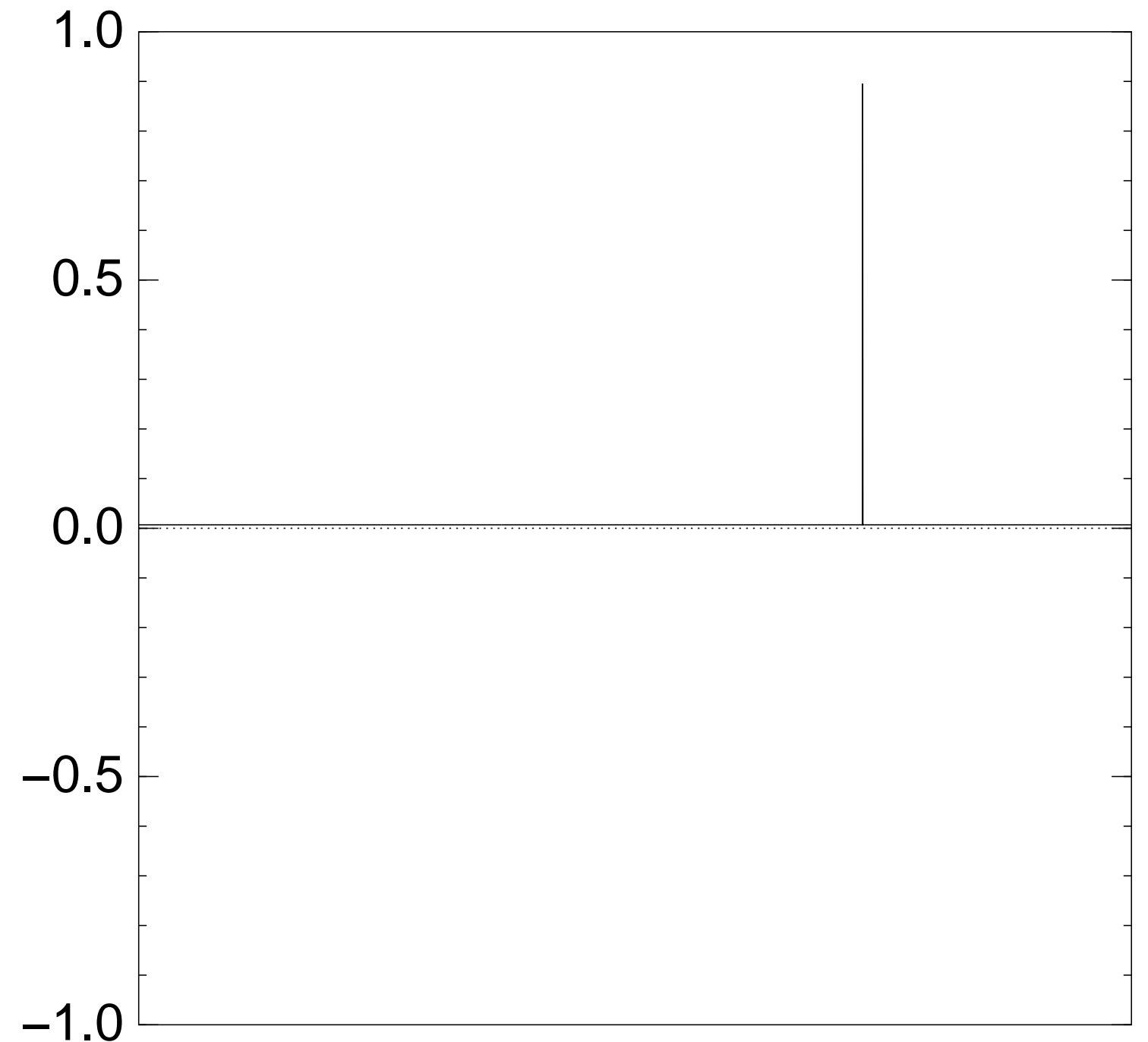
This is also fast.

Repeat Step 1 + Step 2 about  $0.58 \cdot 2^{0.5n}$  times.

Measure the  $n$  qubits.

With high probability this finds  $s$ .

Normalized graph of  $q \mapsto a_q$  for an example with  $n = 12$  after  $35 \times$  (Step 1 + Step 2):



Good moment to stop, measure.

Start from uniform superposition over all  $n$ -bit strings  $q$ .

Step 1: Set  $a \leftarrow b$  where

$$b_q = -a_q \text{ if } f(q) = 0,$$

$$b_q = a_q \text{ otherwise.}$$

This is fast.

Step 2: “Grover diffusion”.

Negate  $a$  around its average.

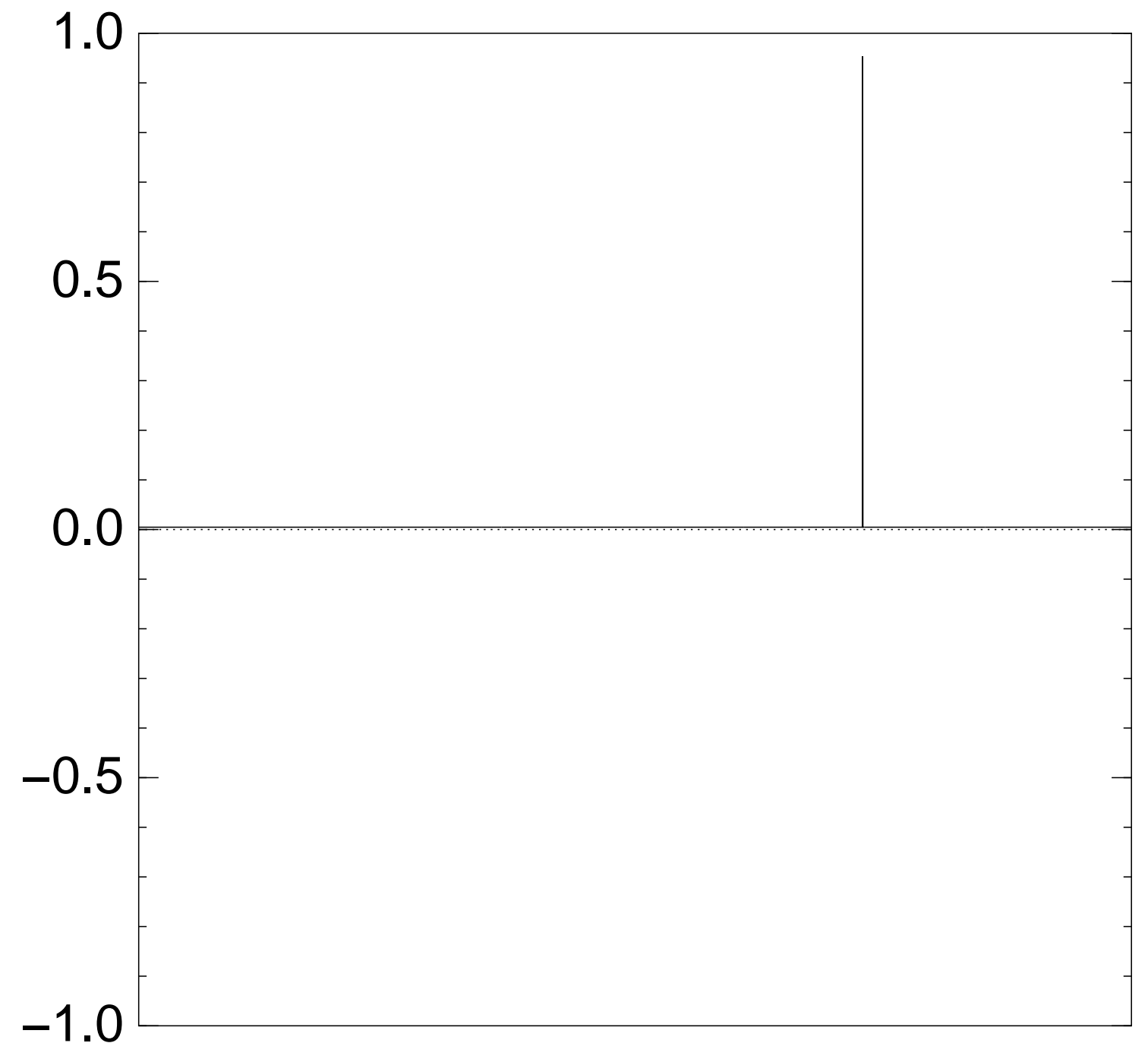
This is also fast.

Repeat Step 1 + Step 2 about  $0.58 \cdot 2^{0.5n}$  times.

Measure the  $n$  qubits.

With high probability this finds  $s$ .

Normalized graph of  $q \mapsto a_q$  for an example with  $n = 12$  after  $40 \times (\text{Step 1} + \text{Step 2})$ :



Start from uniform superposition over all  $n$ -bit strings  $q$ .

Step 1: Set  $a \leftarrow b$  where

$$b_q = -a_q \text{ if } f(q) = 0,$$

$$b_q = a_q \text{ otherwise.}$$

This is fast.

Step 2: “Grover diffusion”.

Negate  $a$  around its average.

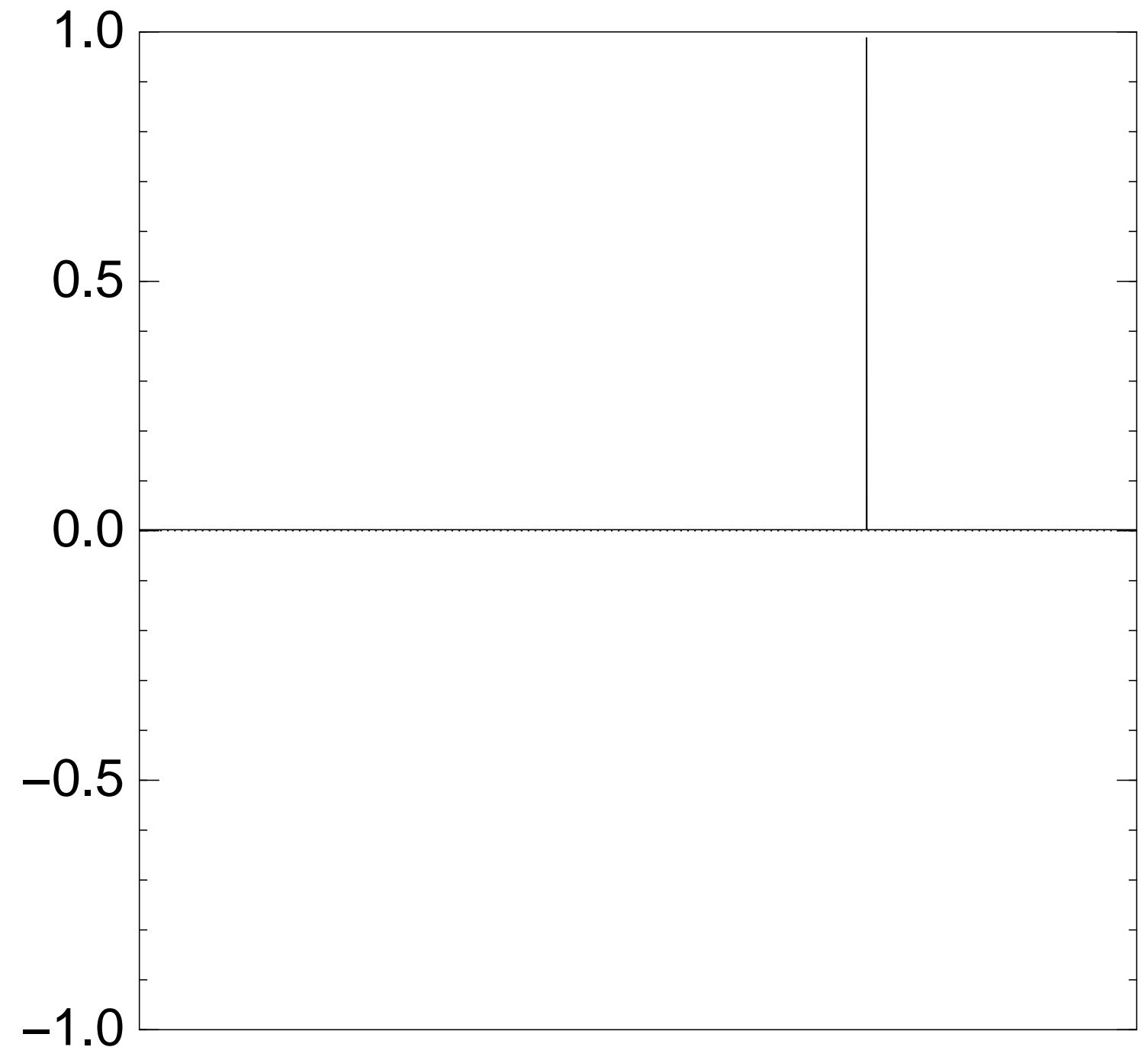
This is also fast.

Repeat Step 1 + Step 2 about  $0.58 \cdot 2^{0.5n}$  times.

Measure the  $n$  qubits.

With high probability this finds  $s$ .

Normalized graph of  $q \mapsto a_q$  for an example with  $n = 12$  after  $45 \times$  (Step 1 + Step 2):



Start from uniform superposition over all  $n$ -bit strings  $q$ .

Step 1: Set  $a \leftarrow b$  where

$$b_q = -a_q \text{ if } f(q) = 0,$$

$$b_q = a_q \text{ otherwise.}$$

This is fast.

Step 2: “Grover diffusion”.

Negate  $a$  around its average.

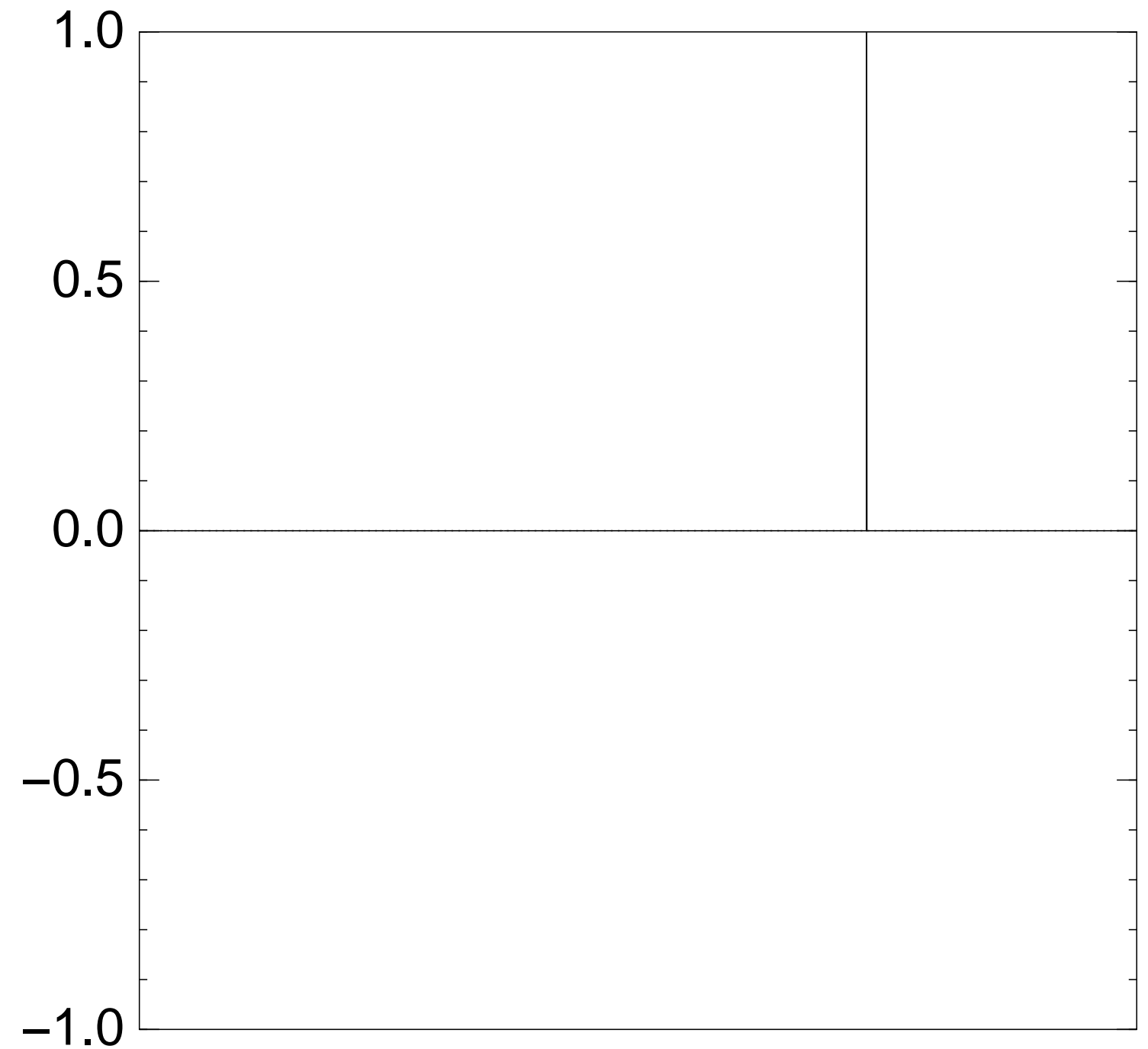
This is also fast.

Repeat Step 1 + Step 2 about  $0.58 \cdot 2^{0.5n}$  times.

Measure the  $n$  qubits.

With high probability this finds  $s$ .

Normalized graph of  $q \mapsto a_q$  for an example with  $n = 12$  after  $50 \times (\text{Step 1} + \text{Step 2})$ :



Traditional stopping point.

Start from uniform superposition over all  $n$ -bit strings  $q$ .

Step 1: Set  $a \leftarrow b$  where

$$b_q = -a_q \text{ if } f(q) = 0,$$

$$b_q = a_q \text{ otherwise.}$$

This is fast.

Step 2: “Grover diffusion”.

Negate  $a$  around its average.

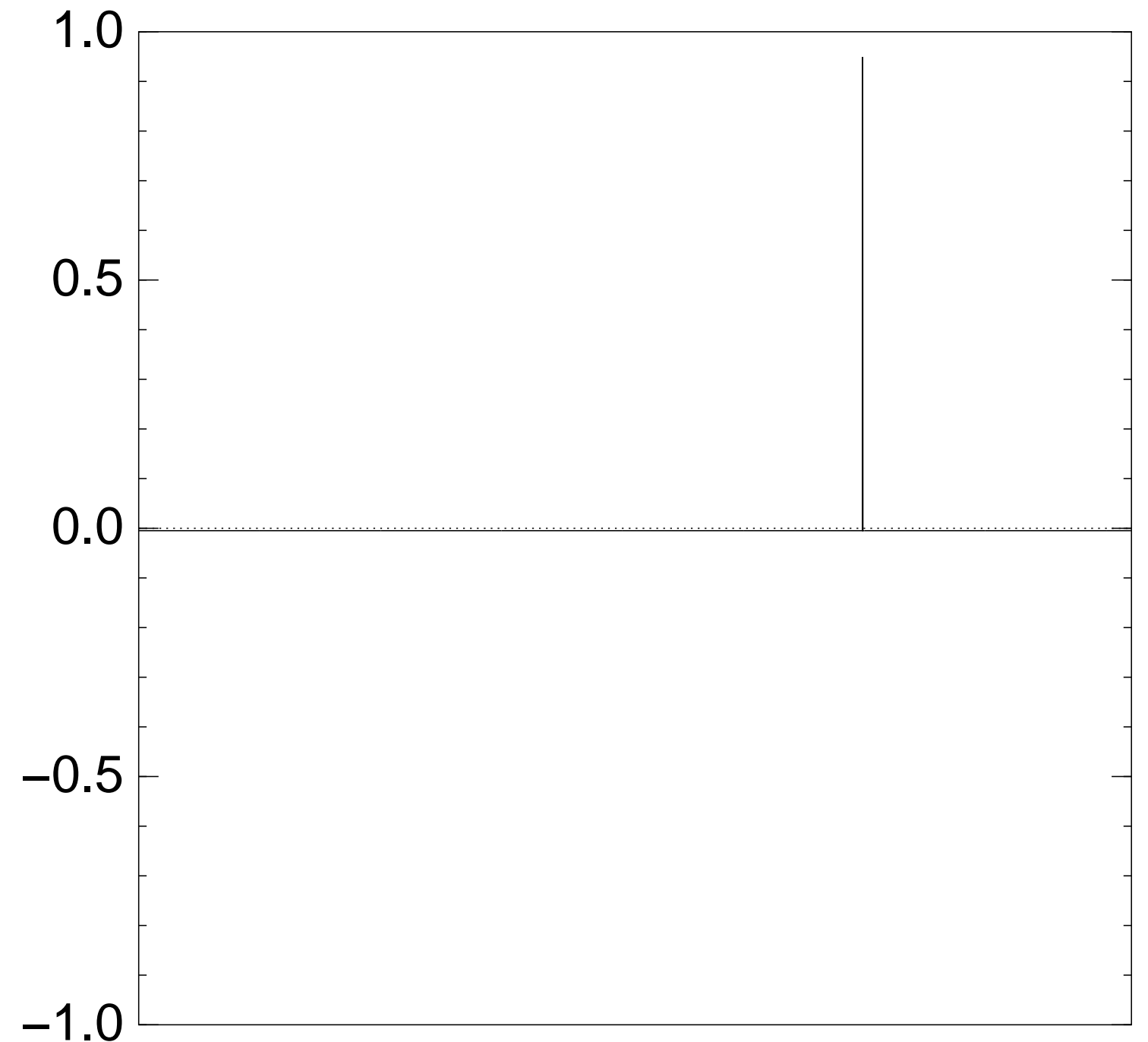
This is also fast.

Repeat Step 1 + Step 2 about  $0.58 \cdot 2^{0.5n}$  times.

Measure the  $n$  qubits.

With high probability this finds  $s$ .

Normalized graph of  $q \mapsto a_q$  for an example with  $n = 12$  after  $60 \times (\text{Step 1} + \text{Step 2})$ :





Start from uniform superposition over all  $n$ -bit strings  $q$ .

Step 1: Set  $a \leftarrow b$  where

$$b_q = -a_q \text{ if } f(q) = 0,$$

$$b_q = a_q \text{ otherwise.}$$

This is fast.

Step 2: “Grover diffusion”.

Negate  $a$  around its average.

This is also fast.

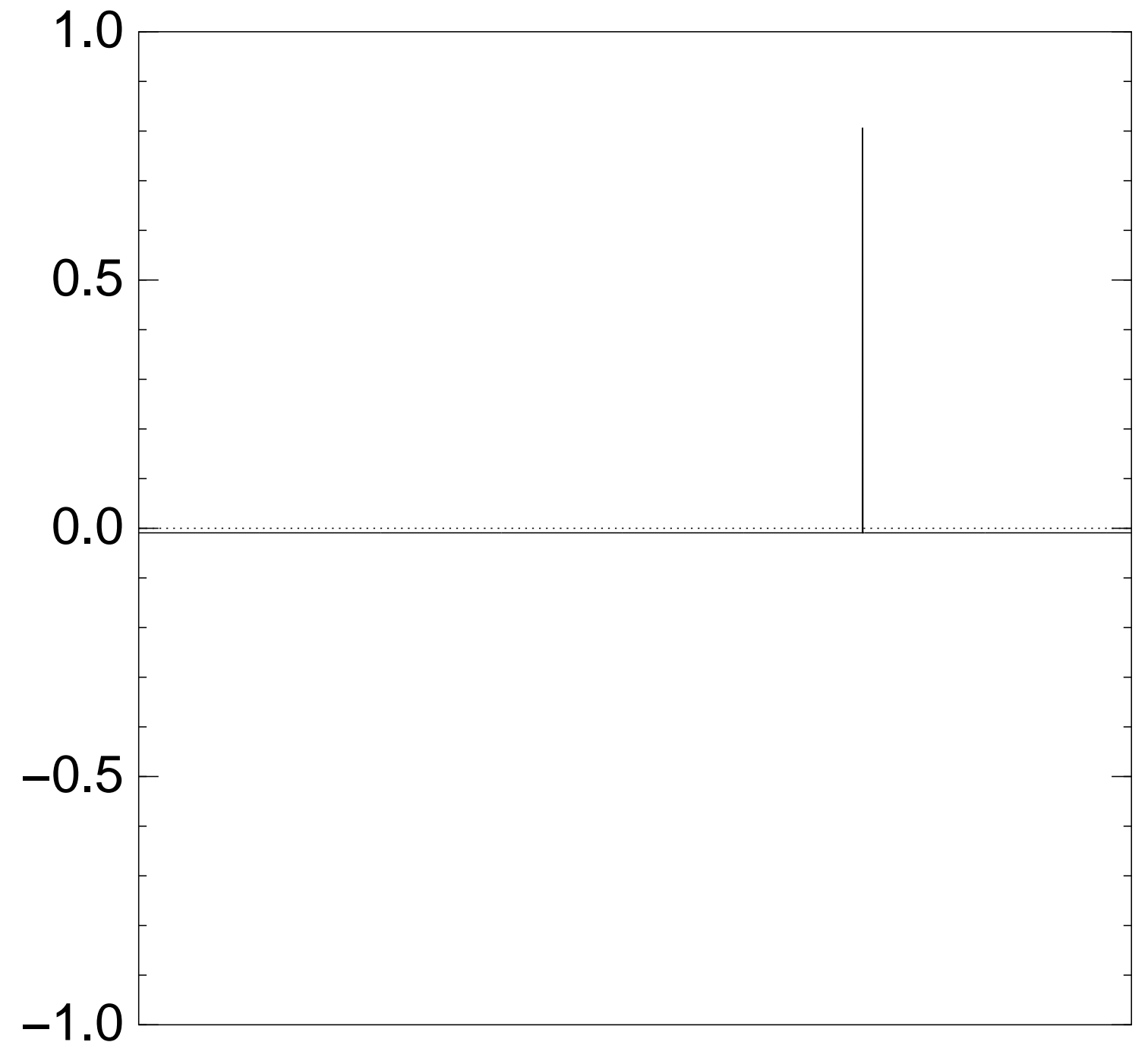
Repeat Step 1 + Step 2

about  $0.58 \cdot 2^{0.5n}$  times.

Measure the  $n$  qubits.

With high probability this finds  $s$ .

Normalized graph of  $q \mapsto a_q$  for an example with  $n = 12$  after  $70 \times (\text{Step 1} + \text{Step 2})$ :



Start from uniform superposition over all  $n$ -bit strings  $q$ .

Step 1: Set  $a \leftarrow b$  where

$$b_q = -a_q \text{ if } f(q) = 0,$$

$$b_q = a_q \text{ otherwise.}$$

This is fast.

Step 2: “Grover diffusion”.

Negate  $a$  around its average.

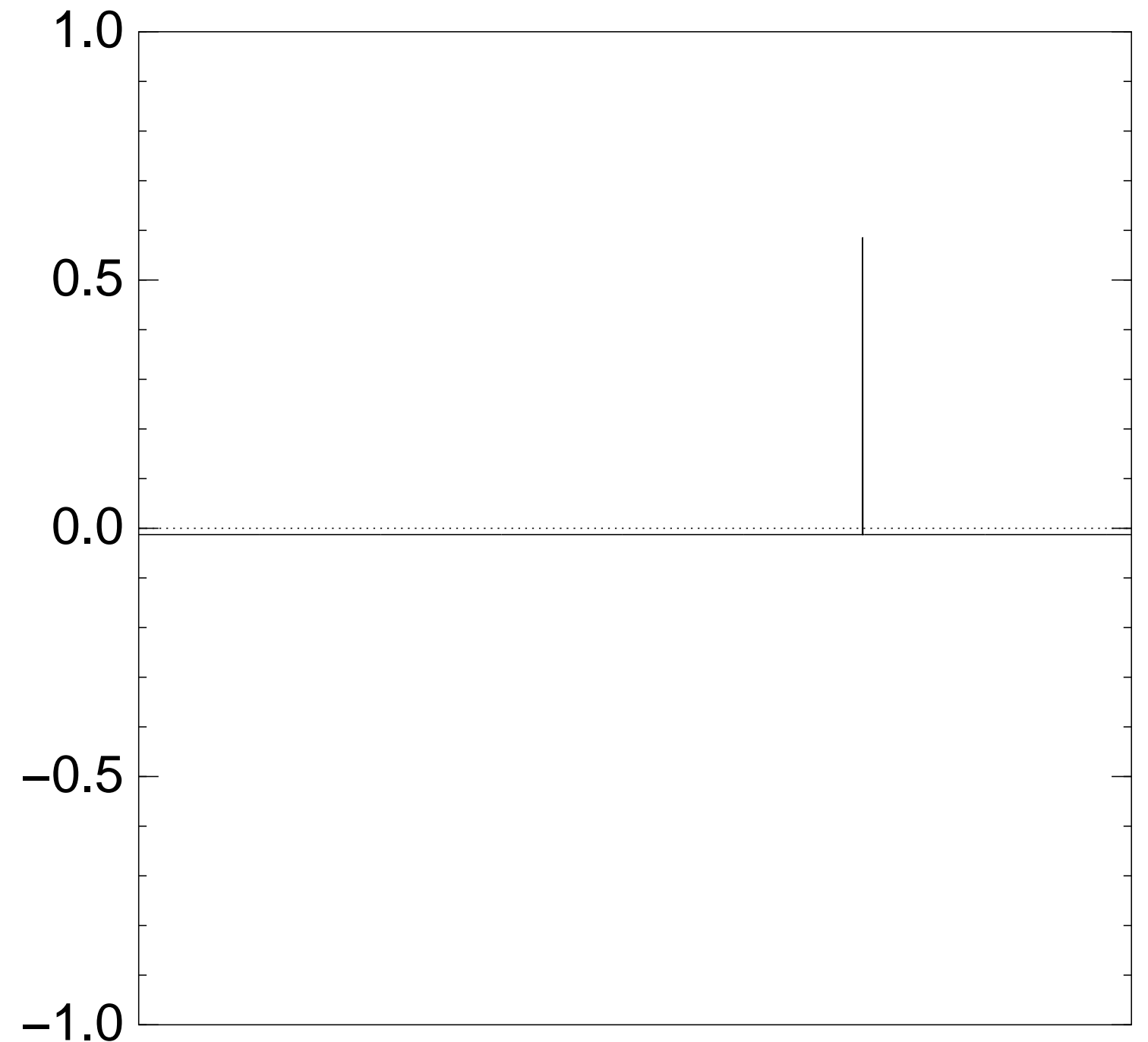
This is also fast.

Repeat Step 1 + Step 2 about  $0.58 \cdot 2^{0.5n}$  times.

Measure the  $n$  qubits.

With high probability this finds  $s$ .

Normalized graph of  $q \mapsto a_q$  for an example with  $n = 12$  after  $80 \times (\text{Step 1} + \text{Step 2})$ :



Start from uniform superposition over all  $n$ -bit strings  $q$ .

Step 1: Set  $a \leftarrow b$  where

$$b_q = -a_q \text{ if } f(q) = 0,$$

$$b_q = a_q \text{ otherwise.}$$

This is fast.

Step 2: “Grover diffusion”.

Negate  $a$  around its average.

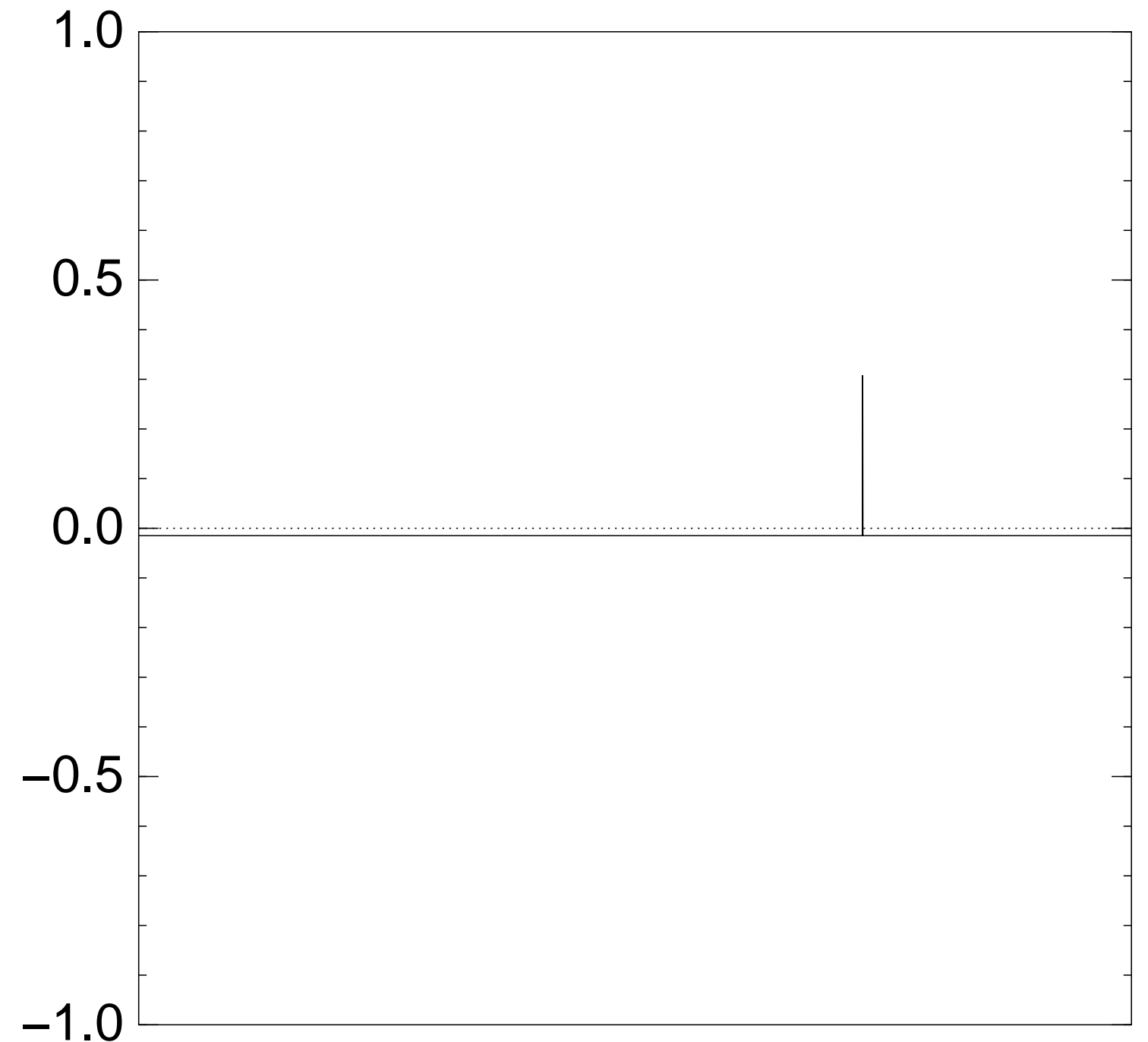
This is also fast.

Repeat Step 1 + Step 2 about  $0.58 \cdot 2^{0.5n}$  times.

Measure the  $n$  qubits.

With high probability this finds  $s$ .

Normalized graph of  $q \mapsto a_q$  for an example with  $n = 12$  after  $90 \times (\text{Step 1} + \text{Step 2})$ :



Start from uniform superposition over all  $n$ -bit strings  $q$ .

Step 1: Set  $a \leftarrow b$  where

$$b_q = -a_q \text{ if } f(q) = 0,$$

$$b_q = a_q \text{ otherwise.}$$

This is fast.

Step 2: “Grover diffusion”.

Negate  $a$  around its average.

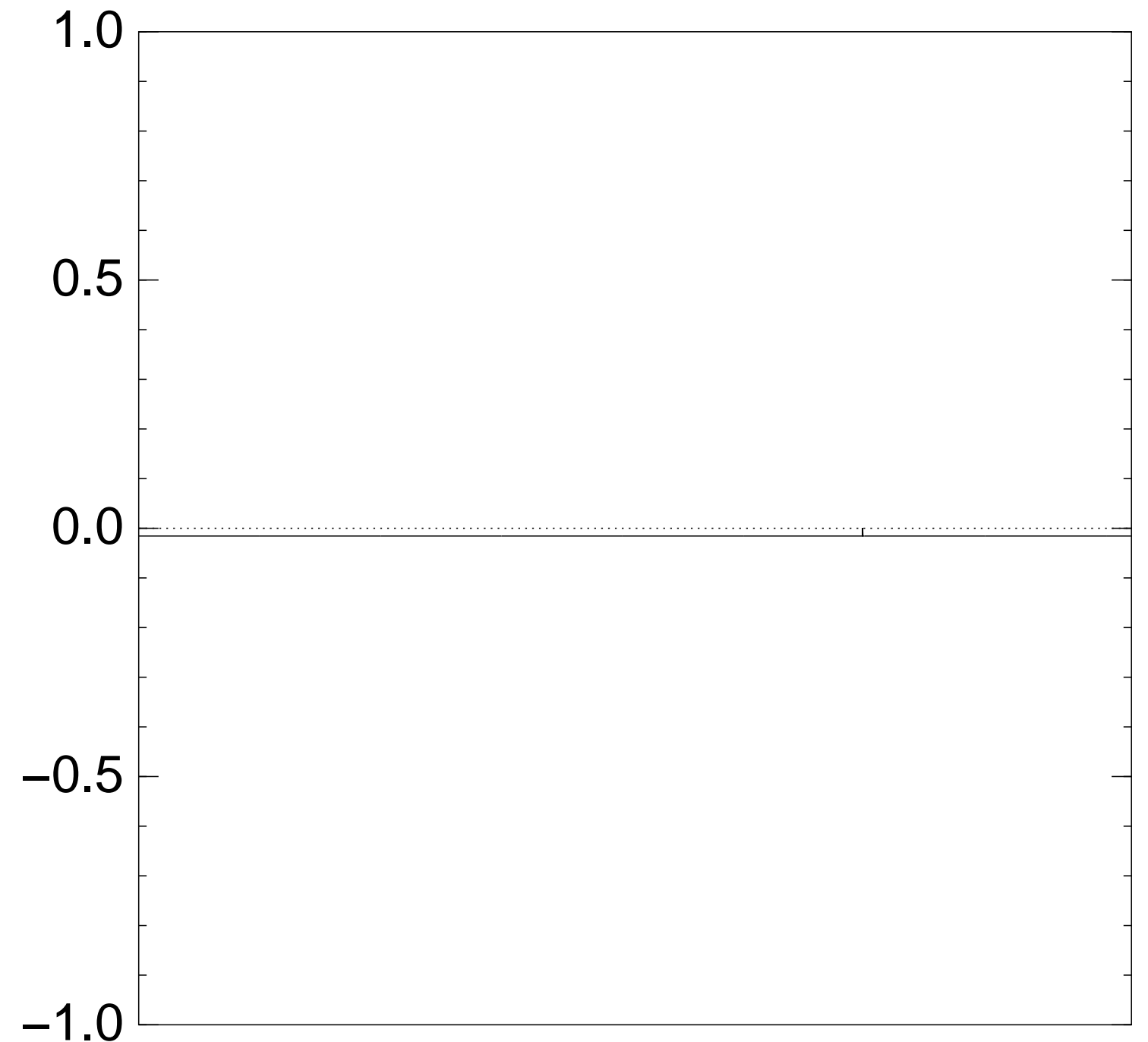
This is also fast.

Repeat Step 1 + Step 2 about  $0.58 \cdot 2^{0.5n}$  times.

Measure the  $n$  qubits.

With high probability this finds  $s$ .

Normalized graph of  $q \mapsto a_q$  for an example with  $n = 12$  after  $100 \times$  (Step 1 + Step 2):



Very bad stopping point.

from uniform superposition  
 $n$ -bit strings  $q$ .

Set  $a \leftarrow b$  where

$a_q$  if  $f(q) = 0$ ,

otherwise.

fast.

“Grover diffusion”.

$a$  around its average.

also fast.

Step 1 + Step 2

$58 \cdot 2^{0.5n}$  times.

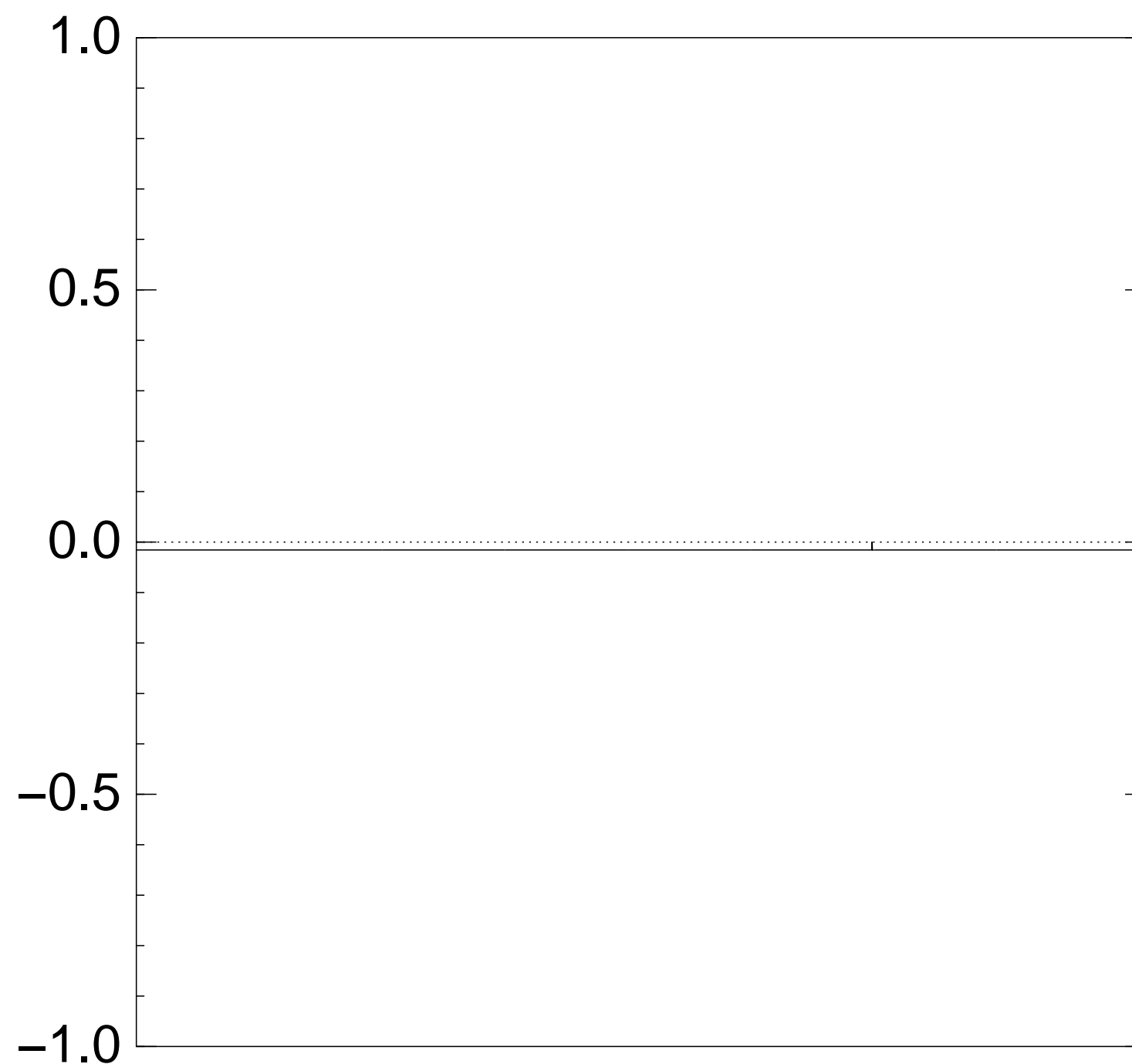
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Normalized graph of  $q \mapsto a_q$

for an example with  $n = 12$

after  $100 \times$  (Step 1 + Step 2):



Very bad stopping point.

$q \mapsto a_q$

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(2)  $a_q$  fo

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gs  $q$ .

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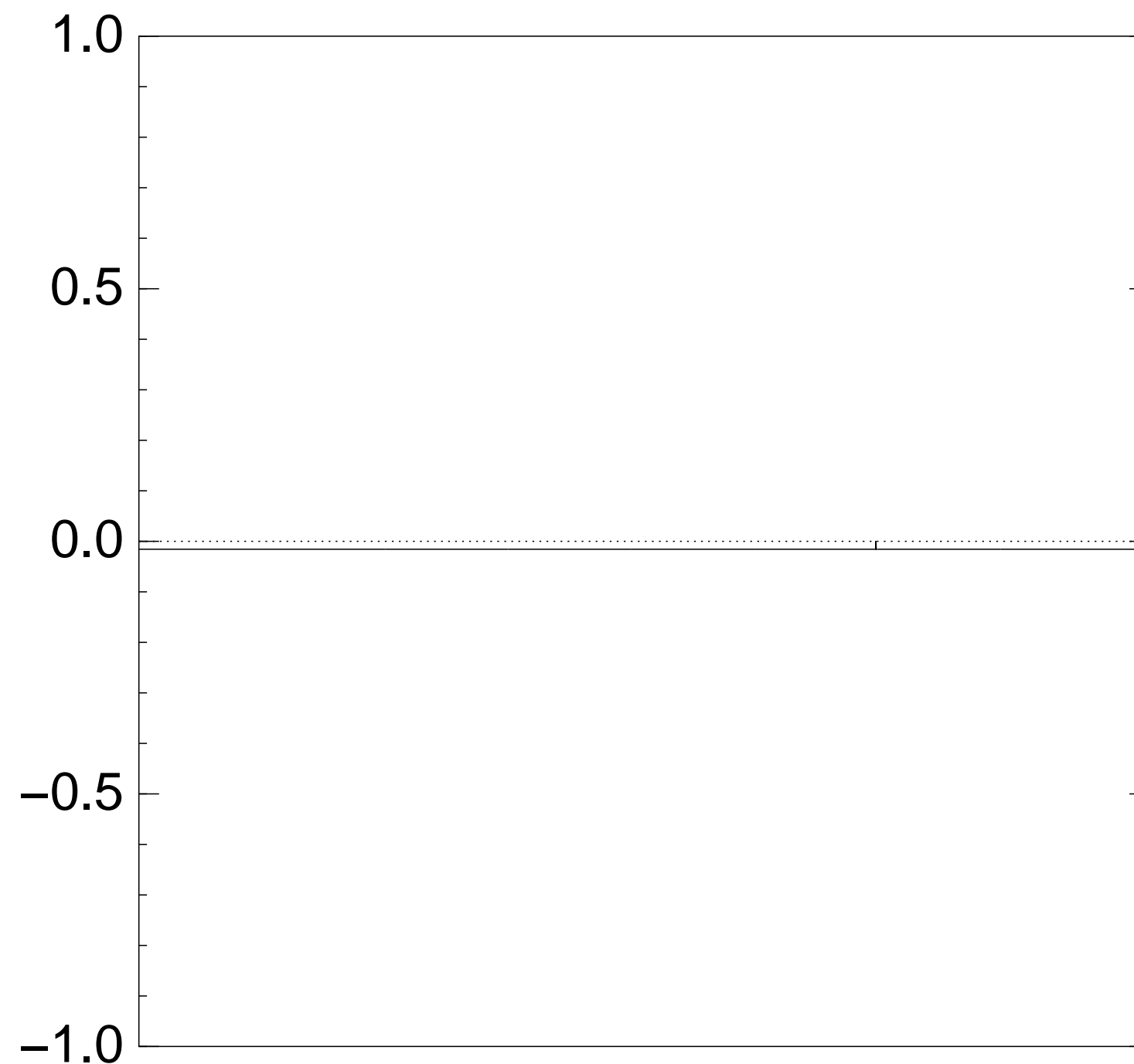
Step 2

times.

bits.

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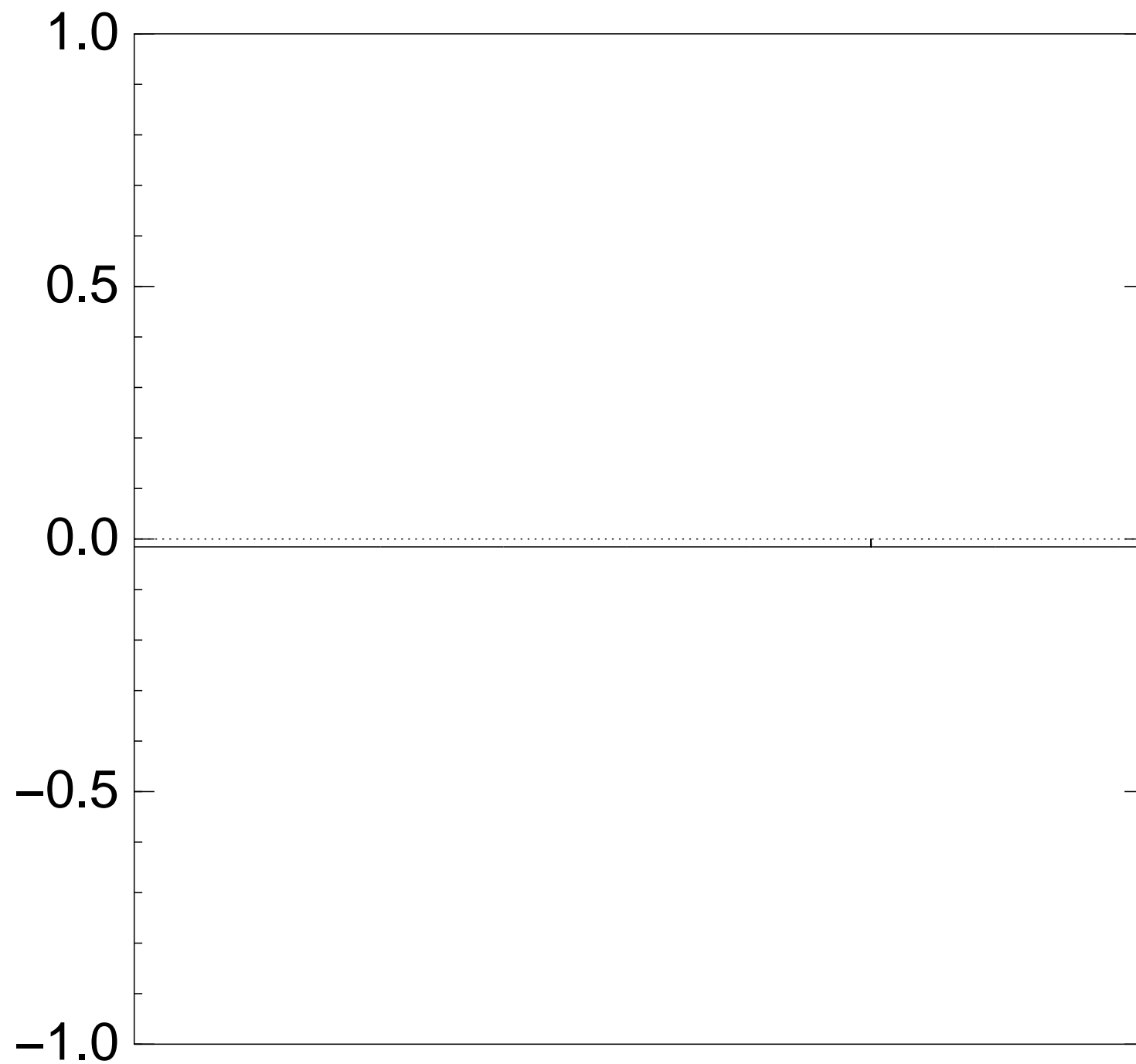
Normalized graph of  $q \mapsto a_q$   
for an example with  $n = 12$   
after  $100 \times$  (Step 1 + Step 2):



Very bad stopping point.

$q \mapsto a_q$  is complet  
by a vector of two  
(with fixed multipl  
(1)  $a_q$  for roots  $q$ ;  
(2)  $a_q$  for non-roo

Normalized graph of  $q \mapsto a_q$   
 for an example with  $n = 12$   
 after  $100 \times$  (Step 1 + Step 2):

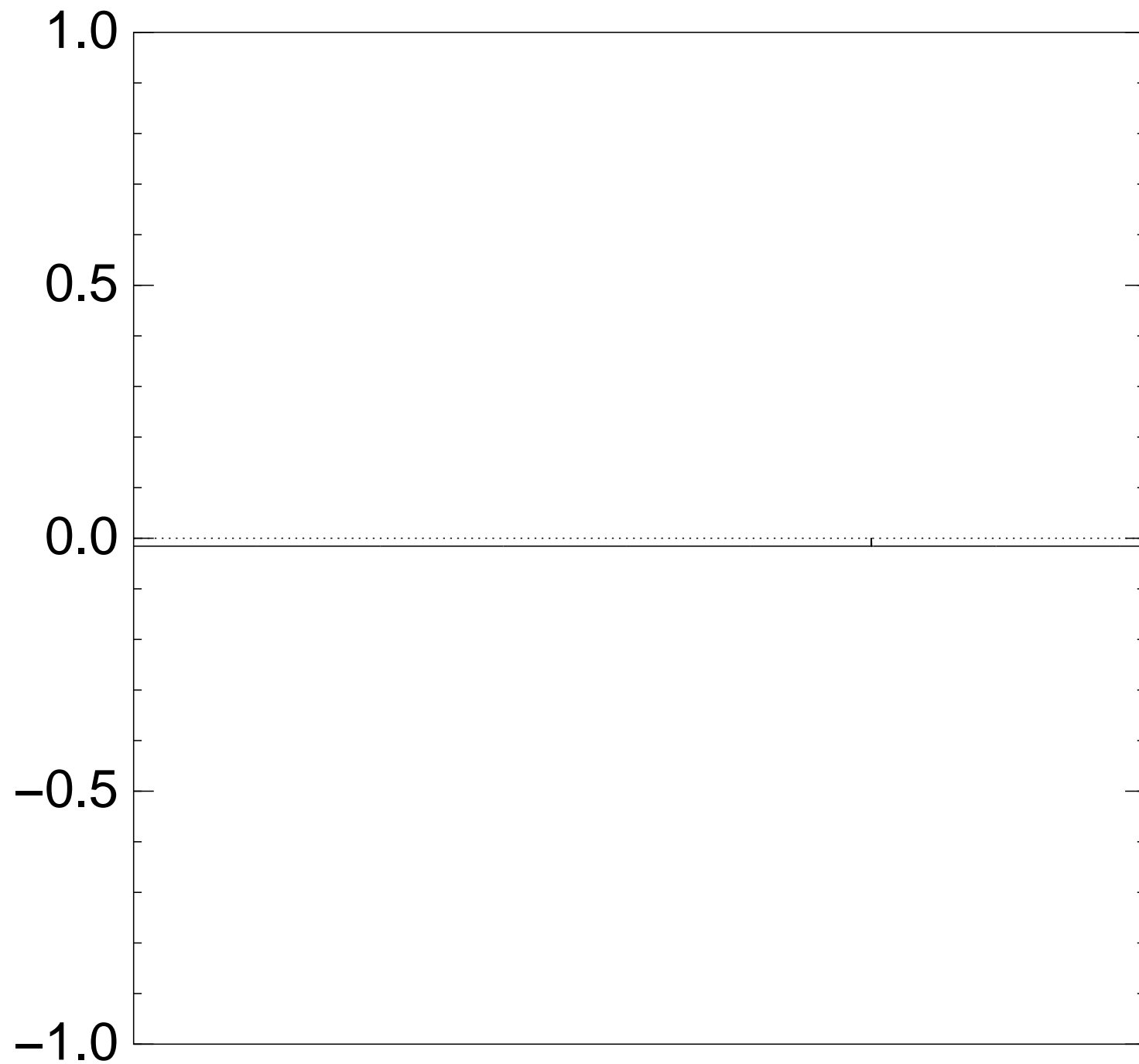


Very bad stopping point.

$q \mapsto a_q$  is completely described  
 by a vector of two numbers  
 (with fixed multiplicities):

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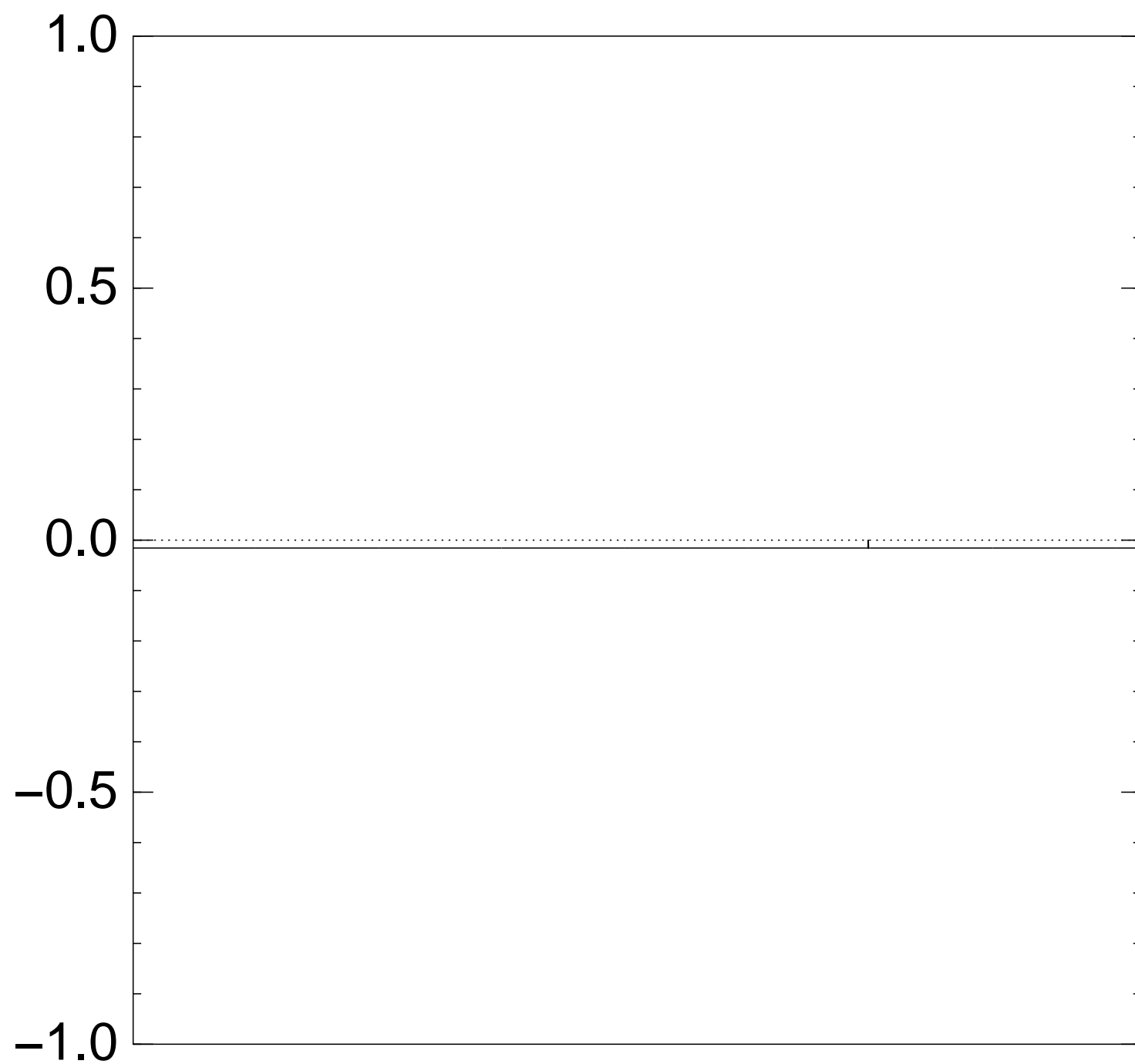
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after  $100 \times$  (Step 1 + Step 2):



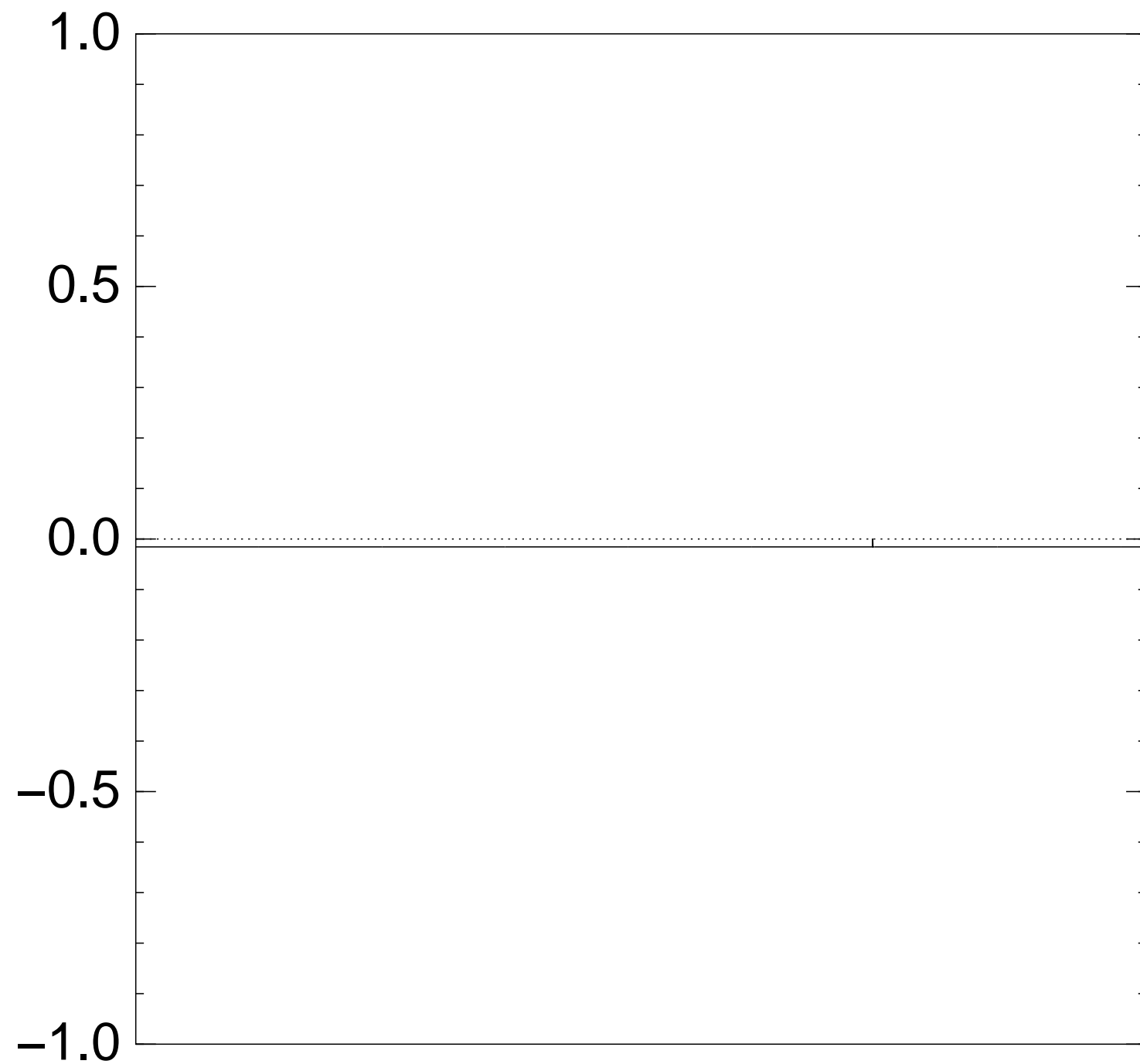
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Step 1 + Step 2  
act linearly on this vector.

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Easily compute eigenvalues  
and powers of this linear map  
to understand evolution  
of state of Grover's algorithm.

$\Rightarrow$  Probability is  $\approx 1$

after  $\approx (\pi/4)2^{0.5n}$  iterations.

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$0 \times (\text{Step 1} + \text{Step 2})$ :



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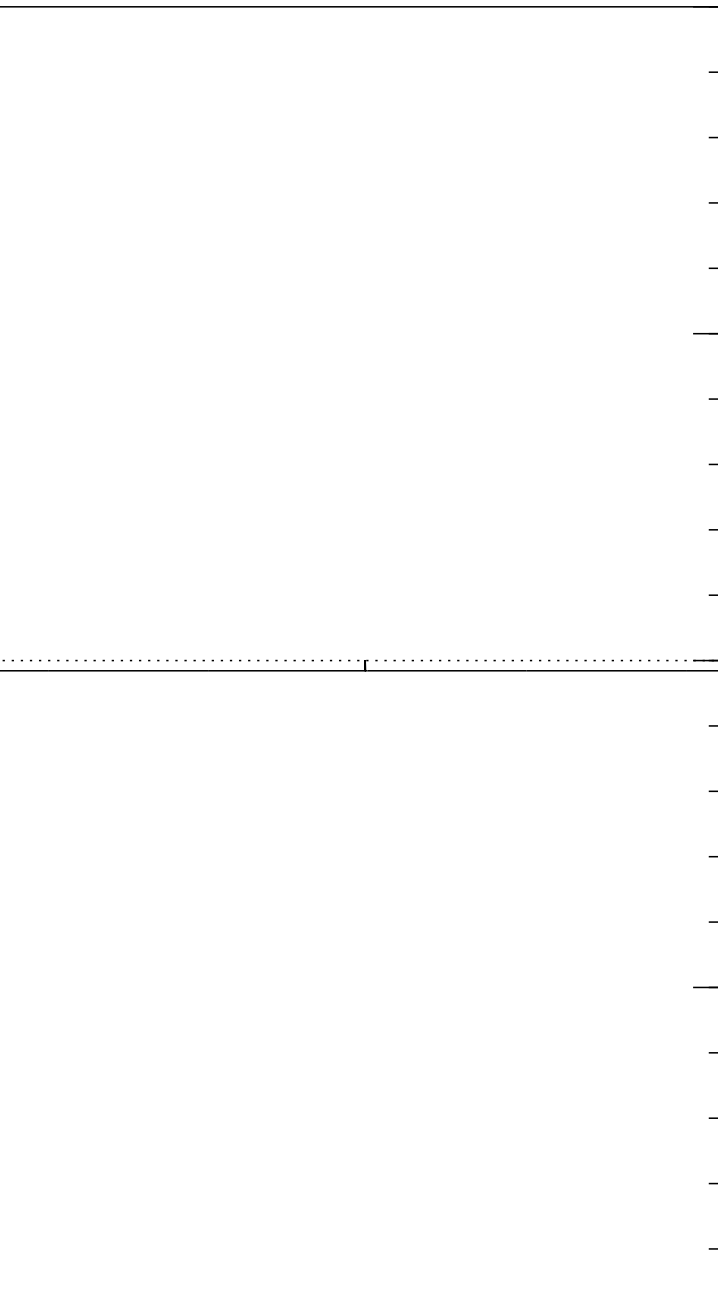
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of  $q \mapsto a_q$   
 with  $n = 12$   
 (Step 1 + Step 2):



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 "cyclotomic" case  
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 use "quantum wal

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## Many more applications

Shor generalizations:

e.g., poly-time attack breaking  
“cyclotomic” case of Gentry  
STOC 2009 “Fully homomorphic  
encryption using ideal lattices”

Grover generalizations:

e.g., fastest subset-sum attack  
use “quantum walks”.

Not just Shor and Grover:

e.g., subexponential-time  
CRS/CSIDH isogeny attack  
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