

# Quantum cryptanalysis

Daniel J. Bernstein

---

Main question in

quantum cryptanalysis:

What is the most efficient

quantum algorithm

to attack this cryptosystem?

(For comparison, main question

in non-quantum cryptanalysis:

What is the most efficient

non-quantum algorithm

to attack this cryptosystem?)

“Quantum algorithm”

means an algorithm that  
a quantum computer can run.

i.e. a sequence of instructions,  
where each instruction is  
in a quantum computer’s  
supported instruction set.

**How do we know which  
instructions a quantum  
computer will support?**

(Something to think about:

Do we really know the answer  
for non-quantum computers?)

Quantum computer type 1 (QC1):  
contains many “qubits” ;  
can efficiently perform  
“NOT gate”, “Hadamard gate”,  
“controlled NOT gate”, “ $T$  gate” .

**Making these instructions work  
is the main goal of quantum-  
computer engineering today.**

Combine these instructions  
to compute “Toffoli gate” ;  
... “Simon’s algorithm” ;  
... “Shor’s algorithm” ; etc.

General belief: Traditional CPU  
isn’t QC1; e.g. can’t factor quickly.

Quantum computer type 2 (QC2):  
stores a simulated universe;  
efficiently simulates the  
laws of quantum physics  
with as much accuracy as desired.

This is the original concept of  
quantum computers introduced  
by 1980 Manin (English version:  
[paper](#) appendix), [1982 Feynman](#).

General belief: any QC1 is a QC2.

Partial proof: see, e.g.,

[2011 Jordan–Lee–Preskill](#)

“Quantum algorithms for  
quantum field theories” .

Quantum computer type 3 (QC3):  
efficiently computes anything  
that any possible physical  
computer can compute efficiently.

General belief: any QC2 is a QC3.

Argument for belief:

any physical computer must  
follow the laws of quantum  
physics, so a QC2 can efficiently  
simulate any physical computer.

General belief: any QC3 is a QC1.

Argument for belief:

look, we're building a QC1.

## State of a non-quantum computer

Data (“state”) stored in 3 bits:

a list of 3 elements of  $\{0, 1\}$ .

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.:  $(1, 1, 1, 1, 1, 0, 0, 0, 1,$

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

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

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

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

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

## 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].

e.g.: [0, 0, 0, 0, 0, 1, 0, 0].

Data stored in 4 qubits: a list of 16 numbers, not all zero. e.g.:

[3, 1, 4, 1, 5, 9, 2, 6, 5, 3, 5, 8, 9, 7, 9, 3].

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.

## Measuring a quantum computer

Can simply look at a bit.

Cannot simply look at the list of numbers stored in  $n$  qubits.

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

- produces  $n$  bits and
- “collapses” the state.

If  $n$  qubits have state

$[a_0, a_1, \dots, a_{2^n-1}]$  then

measurement produces  $q$

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

“Collapse”: New state is all zeros except 1 at position  $q$ .



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;

111 = 7 with probability 1/8.

“Quantum RNG.”

Warning: Quantum RNGs sold  
today are **measurably 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 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:

$[3, 1, 4, 1, 5, 9, 2, 6] \mapsto$

$[5, 9, 2, 6, 3, 1, 4, 1].$

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.  $C_2\text{NOT}_0$ :

$[3, 1, 4, 1, 5, 9, 2, 6] \mapsto$

$[3, 1, 4, 1, 9, 5, 6, 2]$ .

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 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:

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

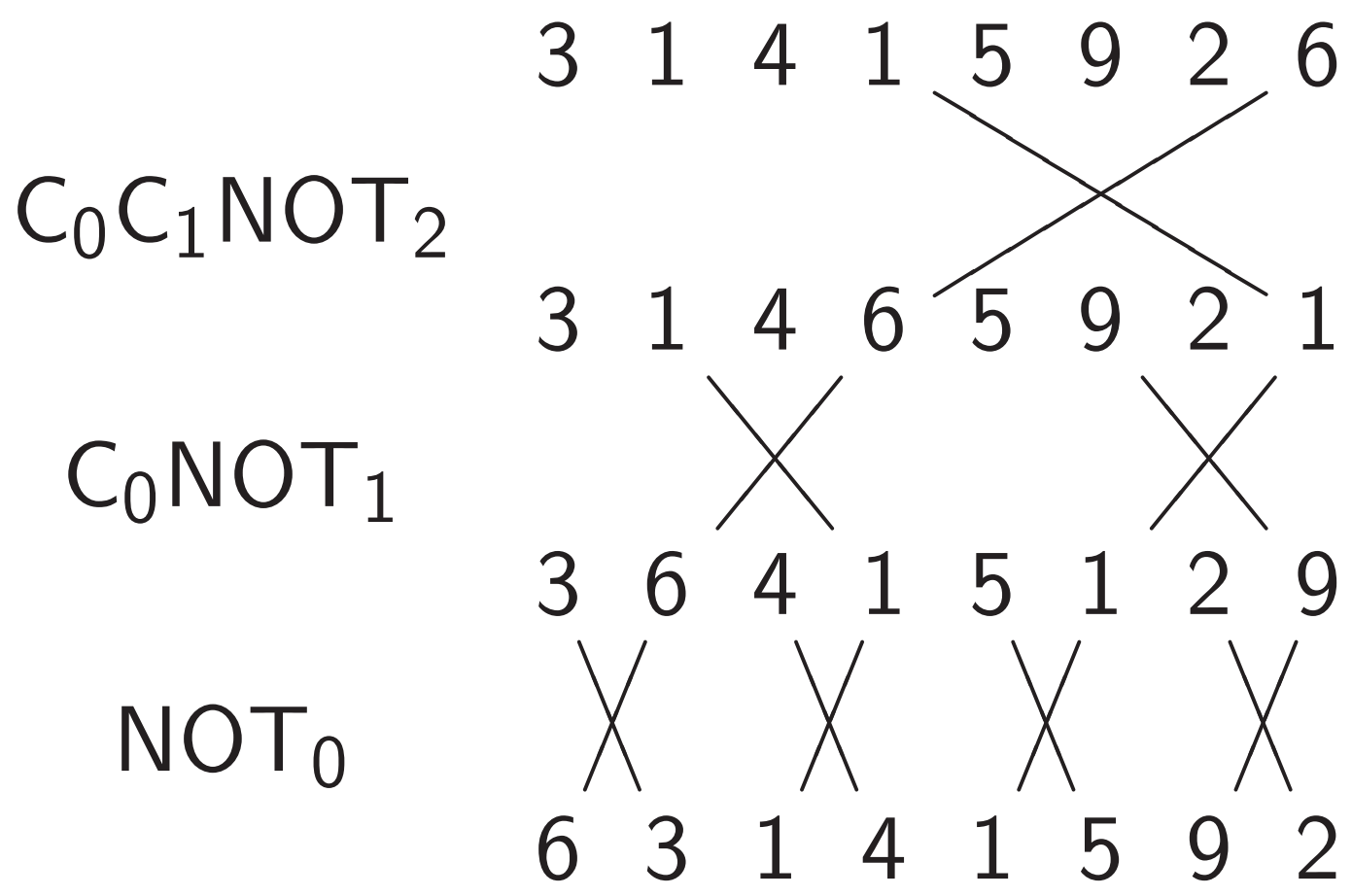
e.g.  $C_0C_1NOT_2$ :

$$[3, 1, 4, 1, 5, 9, 2, 6] \mapsto [3, 1, 4, 6, 5, 9, 2, 1].$$

# More shuffling

Combine NOT, CNOT, Toffoli to build other permutations.

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

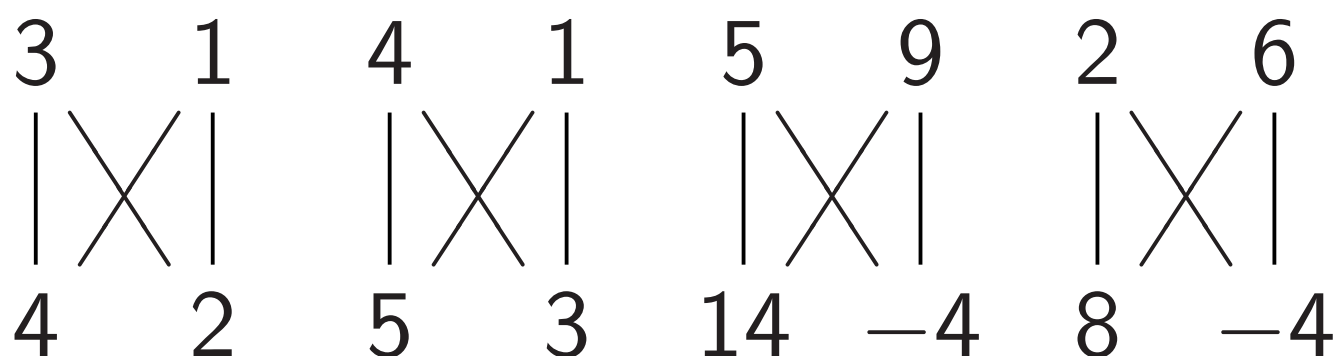




## Hadamard gates

Hadamard<sub>0</sub>:

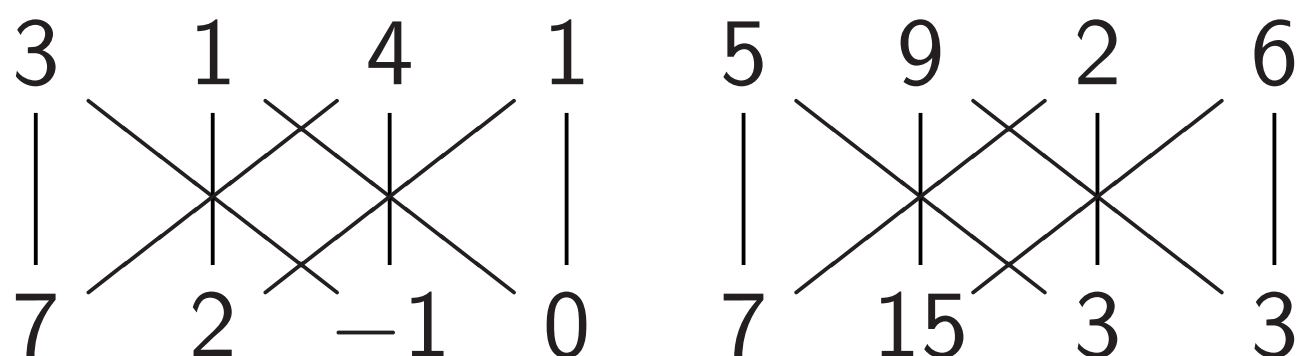
$$[a, b] \mapsto [a + b, a - b].$$



Hadamard<sub>1</sub>:

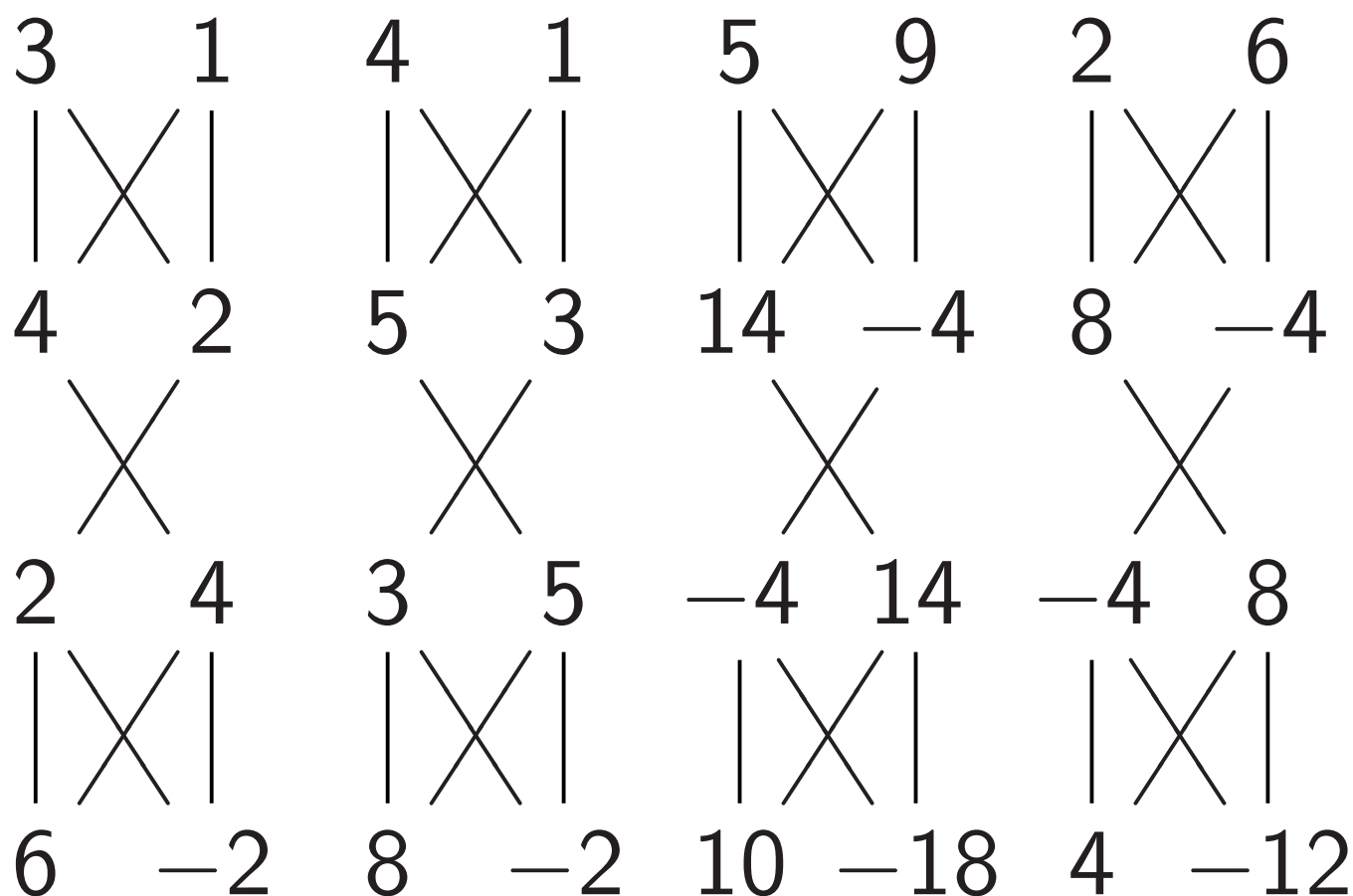
$$[a, b, c, d] \mapsto$$

$$[a + c, b + d, a - c, b - d].$$



## Some uses of Hadamard gates

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



“Multiplied each amplitude by 2.”

This is not physically observable.

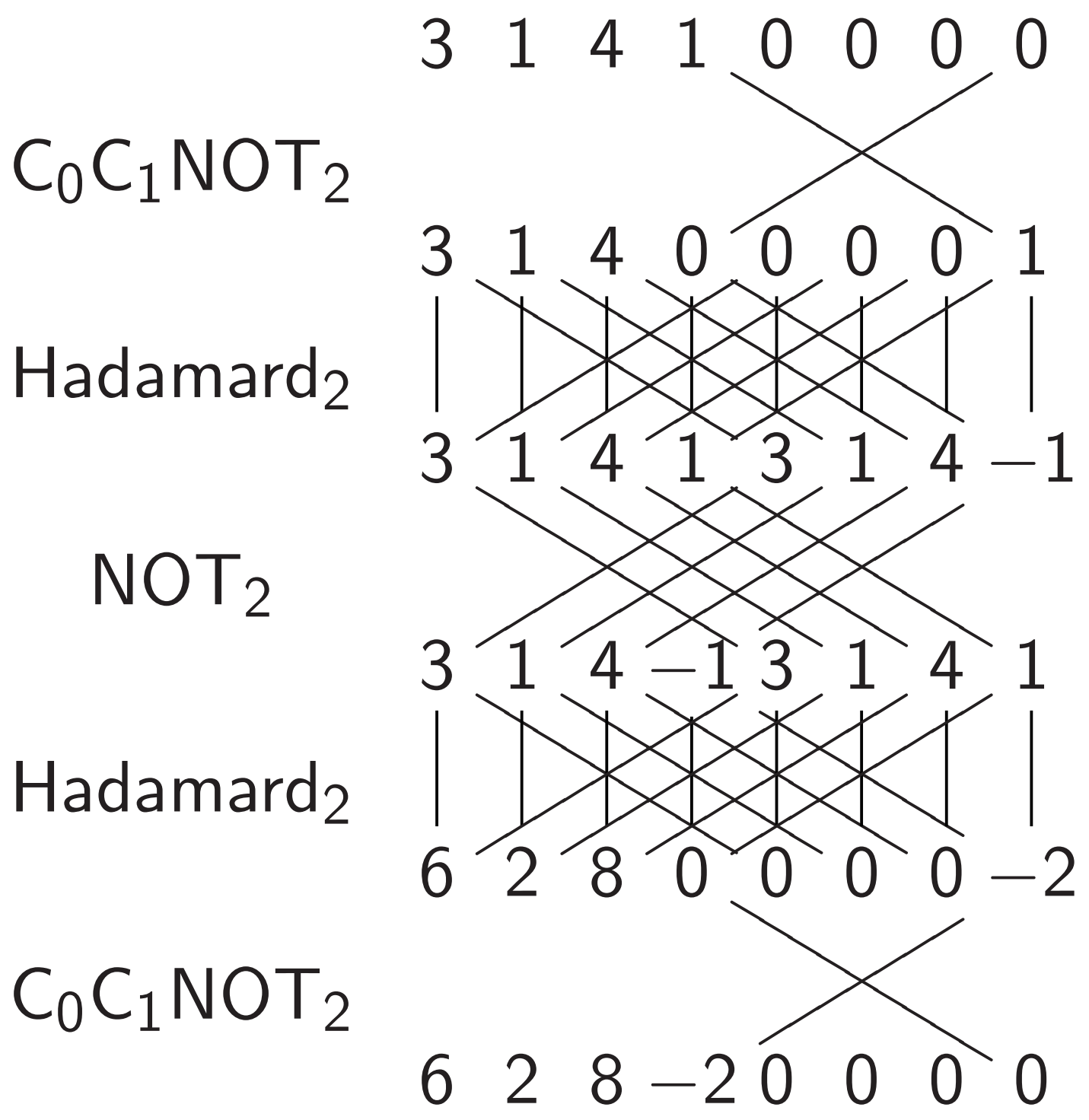
“Negated amplitude if  $q_0$  is set.”

No effect on measuring *now*.

Fancier example:

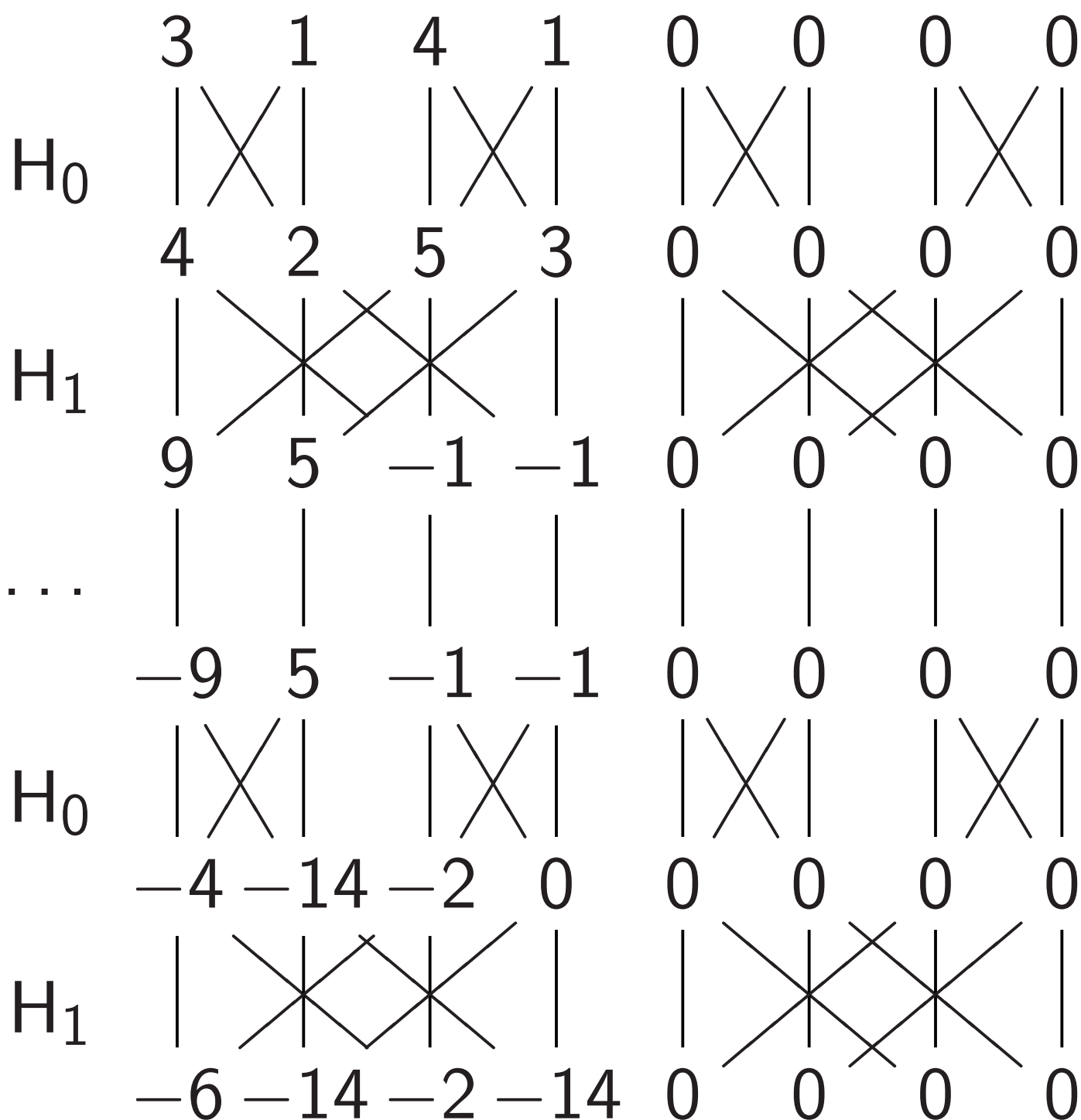
“Negate amplitude if  $q_0 q_1$  is set.”

Assumes  $q_2 = 0$ : “ancilla” qubit.



Affects measurements: “Negate amplitude around its average.”

$$[3, 1, 4, 1] \mapsto [1.5, 3.5, 0.5, 3.5].$$



## Simon's algorithm

Assumptions:

- Given any  $u \in \{0, 1\}^n$ ,  
can efficiently compute  $f(u)$ .
- Nonzero  $s \in \{0, 1\}^n$ .
- $f(u) = f(u \oplus s)$  for all  $u$ .
- $f$  has no other collisions.

Goal: Figure out  $s$ .

Non-quantum algorithm to find  $s$ :  
compute  $f$  for many inputs,  
hope to find collision.

Simon's algorithm finds  $s$  with  
 $\approx n$  quantum evaluations of  $f$ .

## Example of Simon's algorithm

Step 1. Set up pure zero state:

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, 0, 0, 0, 0, 0, 0, 0.

This example is for a function  $f$  with 3-bit input and 3-bit output.

# Example of Simon's algorithm

Step 2.0. Hadamard<sub>0</sub>:

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, 0, 0, 0, 0, 0, 0.





## Example of Simon's algorithm

Step 2.2. 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.

Step 3 will apply the function  $f$  (a specific function in this example), computing  $f(u)$  in universe  $u$ .

## Example of Simon's algorithm

Step 3a.  $C_0NOT_3$ :

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.

## Example of Simon's algorithm

Step 3b. More entry 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.

## Example of Simon's algorithm

Step 3c. More entry 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.

## Example of Simon's algorithm

Step 3d. More entry 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.

## Example of Simon's algorithm

Step 3e. More entry 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.

## Example of Simon's algorithm

Step 3f. More entry 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.

## Example of Simon's algorithm

Step 3g. More entry 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.



## Example of Simon's algorithm

Step 3h. More entry 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.

## Example of Simon's algorithm

Step 3i. More entry 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.

## Example of Simon's algorithm

Step 3j. Final entry 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.

## Example of Simon's algorithm

Step 3j. Final entry 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.

## Example of Simon's algorithm

Step 4.0. 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  $-1$ .

# Example of Simon's algorithm

Step 4.1. 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,  $\bar{1}$ .

# Example of Simon's algorithm

Step 4.2. 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.

## Example of Simon's algorithm

Step 4.2. 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.

Step 5: Measure. Obtain some information about the surprise: a random vector orthogonal to 101.



Repeat to figure out 101.

Generalize Step 3 to any function  $u \mapsto f(u)$  with  $f(u) = f(u \oplus s)$ .

“Usually” algorithm figures out  $s$ .

Shor’s algorithm replaces  $\oplus$  with more general  $+$  operation.

Many spectacular applications.

e.g. Shor finds “random”  $s$  with  $2^u \bmod N = 2^{u+s} \bmod N$ .

Easy to factor  $N$  using this.

e.g. Shor finds “random”  $s, t$  with  $4^u 9^v \bmod p = 4^{u+s} 9^{v+t} \bmod p$ .

Easy to compute discrete logs.

## Grover's algorithm

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

Goal: Figure out  $s$ .

Non-quantum algorithm to find  $s$ :  
compute  $f$  for many inputs,  
hope to find output 0.

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

Grover's algorithm takes only  $2^{n/2}$   
quantum evaluations of  $f$ .  
e.g.  $2^{64}$  instead of  $2^{128}$ .

Start from uniform superposition over  $n$ -bit strings  $u$ : each  $a_u = 1$ .

Step 1: Set  $a \leftarrow b$  where

$$b_u = -a_u \text{ if } f(u) = 0,$$

$$b_u = a_u \text{ otherwise.}$$

This is fast if  $f$  is fast.

Step 2: “Grover diffusion”.

Negate  $a$  around its average.

This is also fast.

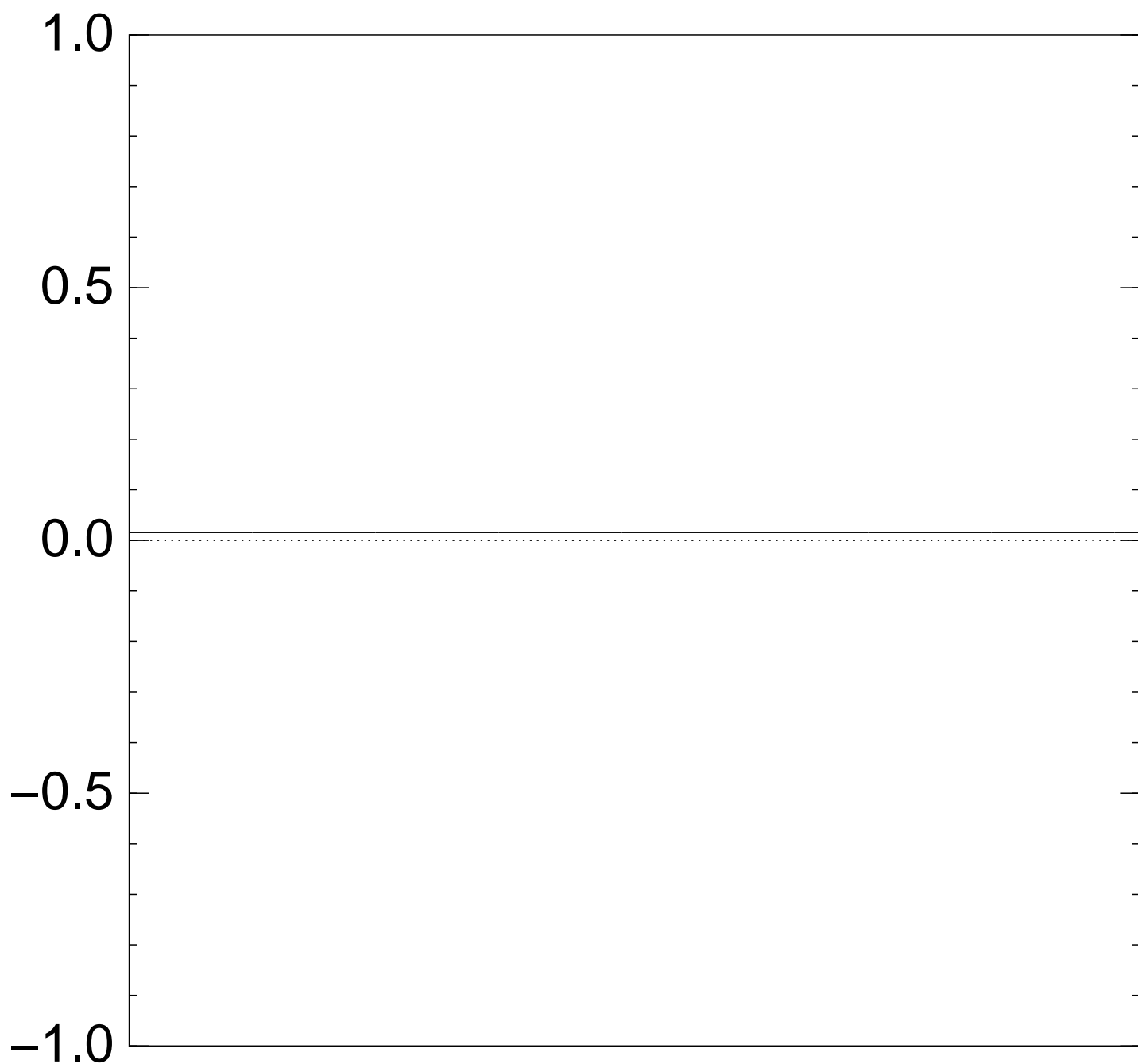
Repeat Step 1 + Step 2

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

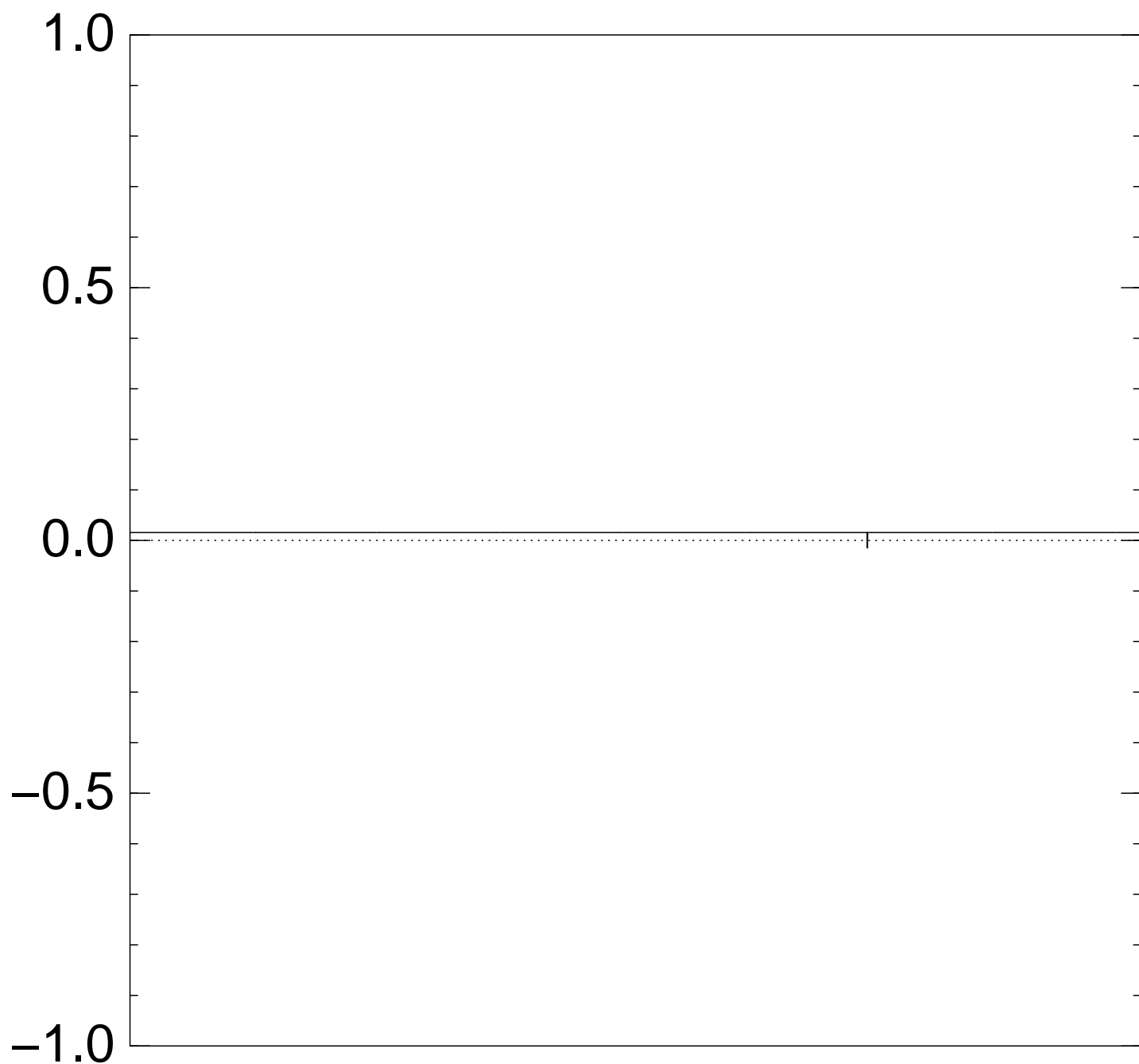
Measure the  $n$  qubits.

With high probability this finds  $s$ .

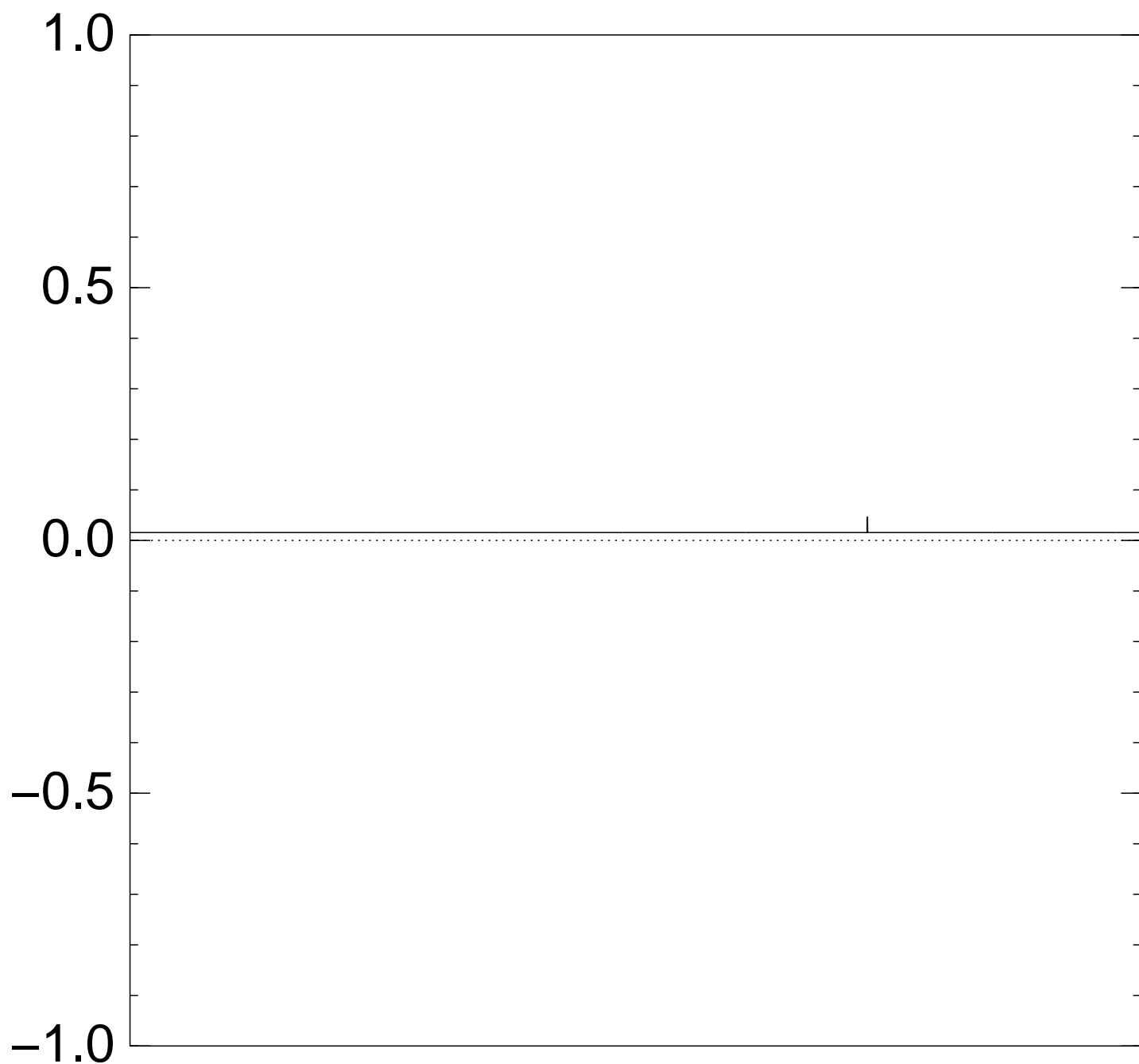
Normalized graph of  $u \mapsto a_u$   
for an example with  $n = 12$   
after 0 steps:



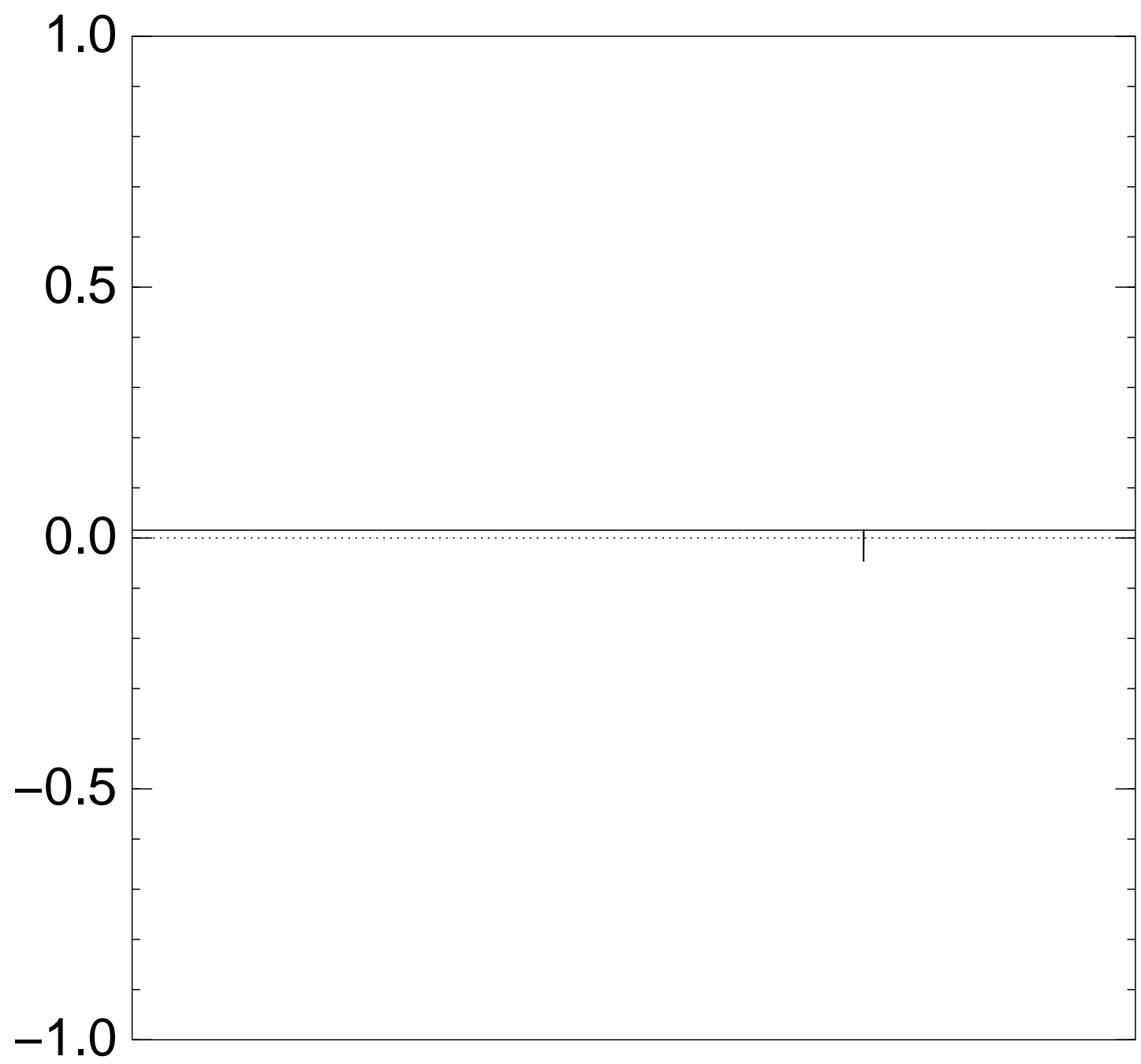
Normalized graph of  $u \mapsto a_u$   
for an example with  $n = 12$   
after Step 1:



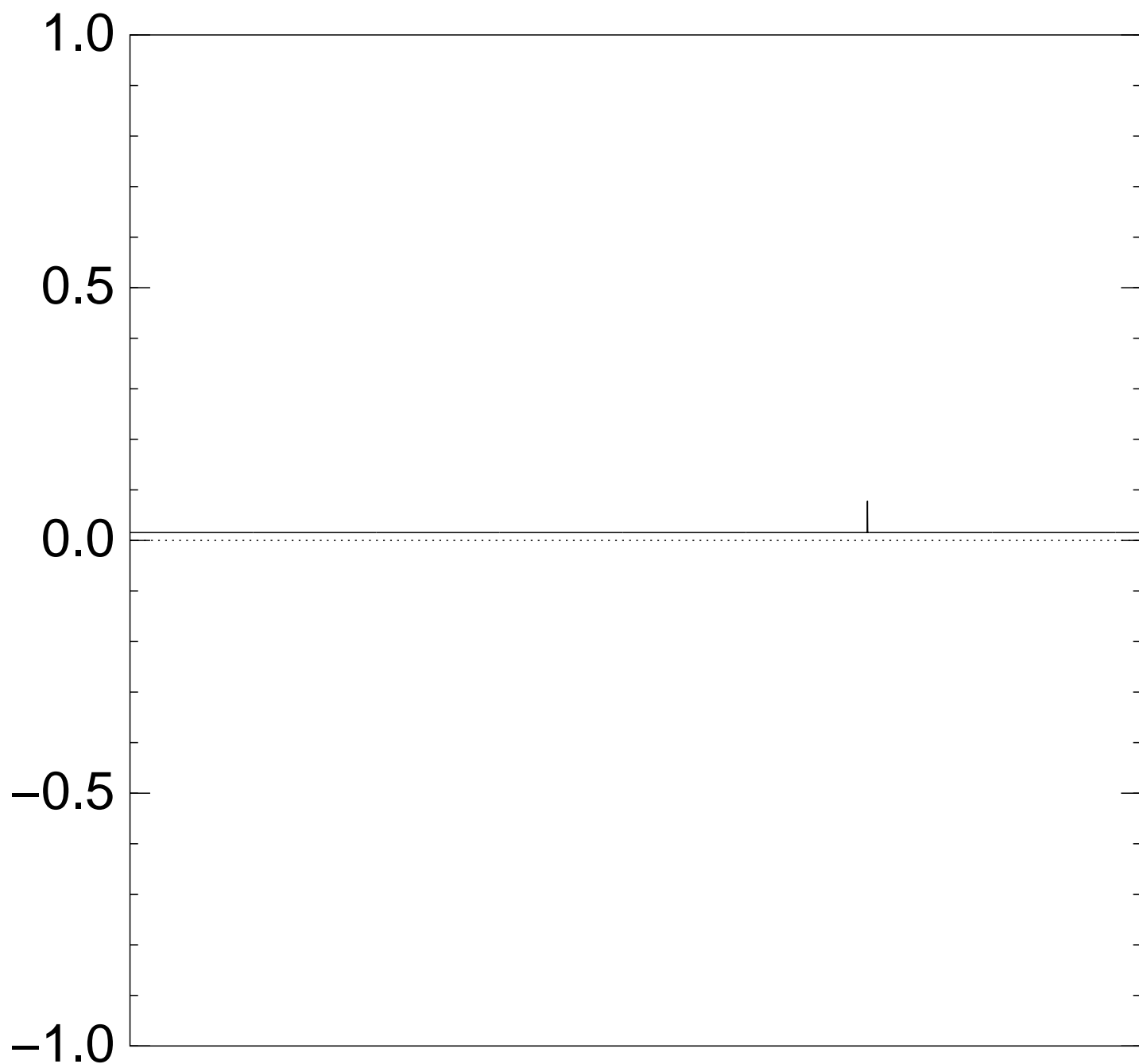
Normalized graph of  $u \mapsto a_u$   
for an example with  $n = 12$   
after Step 1 + Step 2:



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

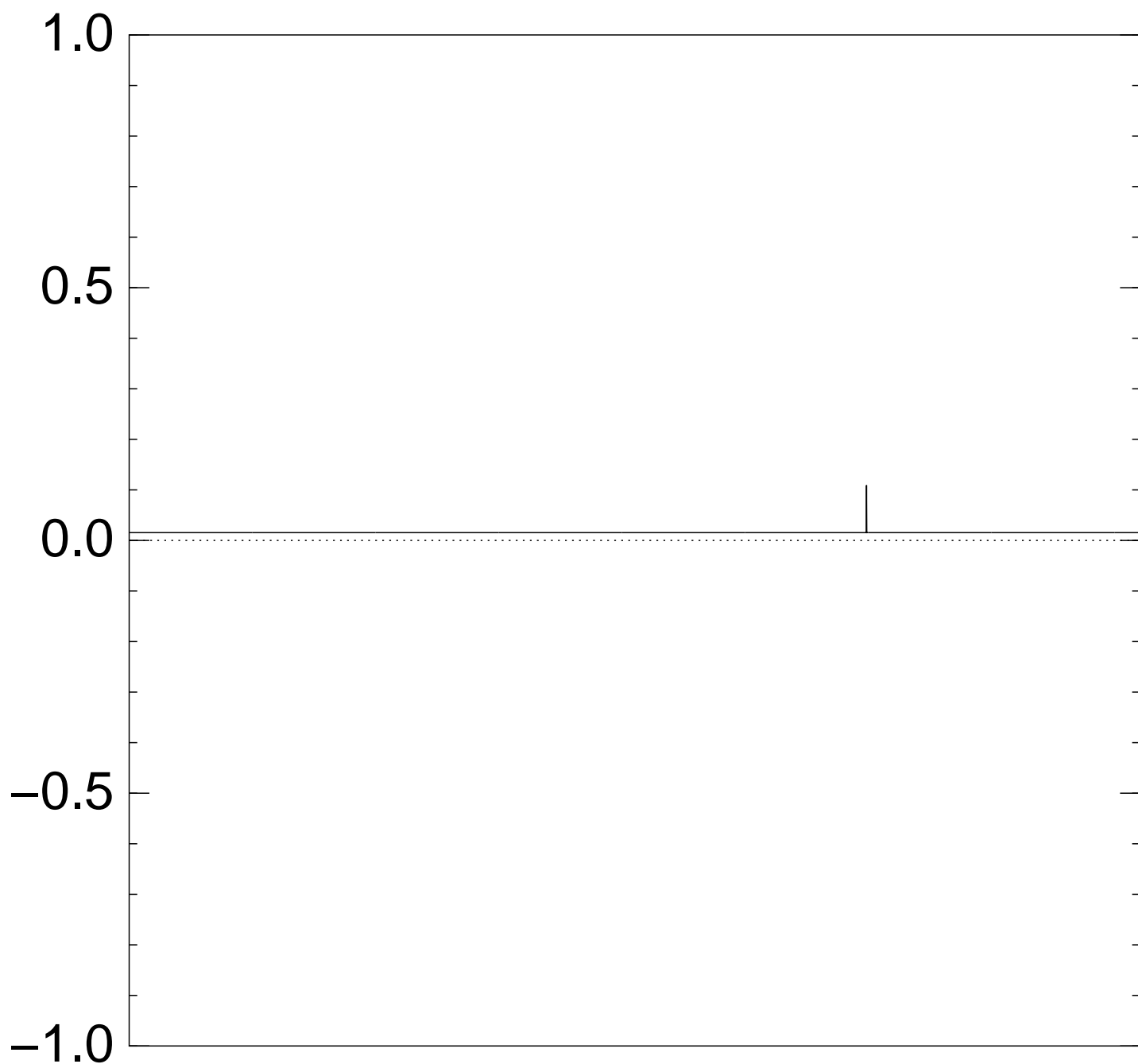


Normalized graph of  $u \mapsto a_u$   
for an example with  $n = 12$   
after  $2 \times$  (Step 1 + Step 2):

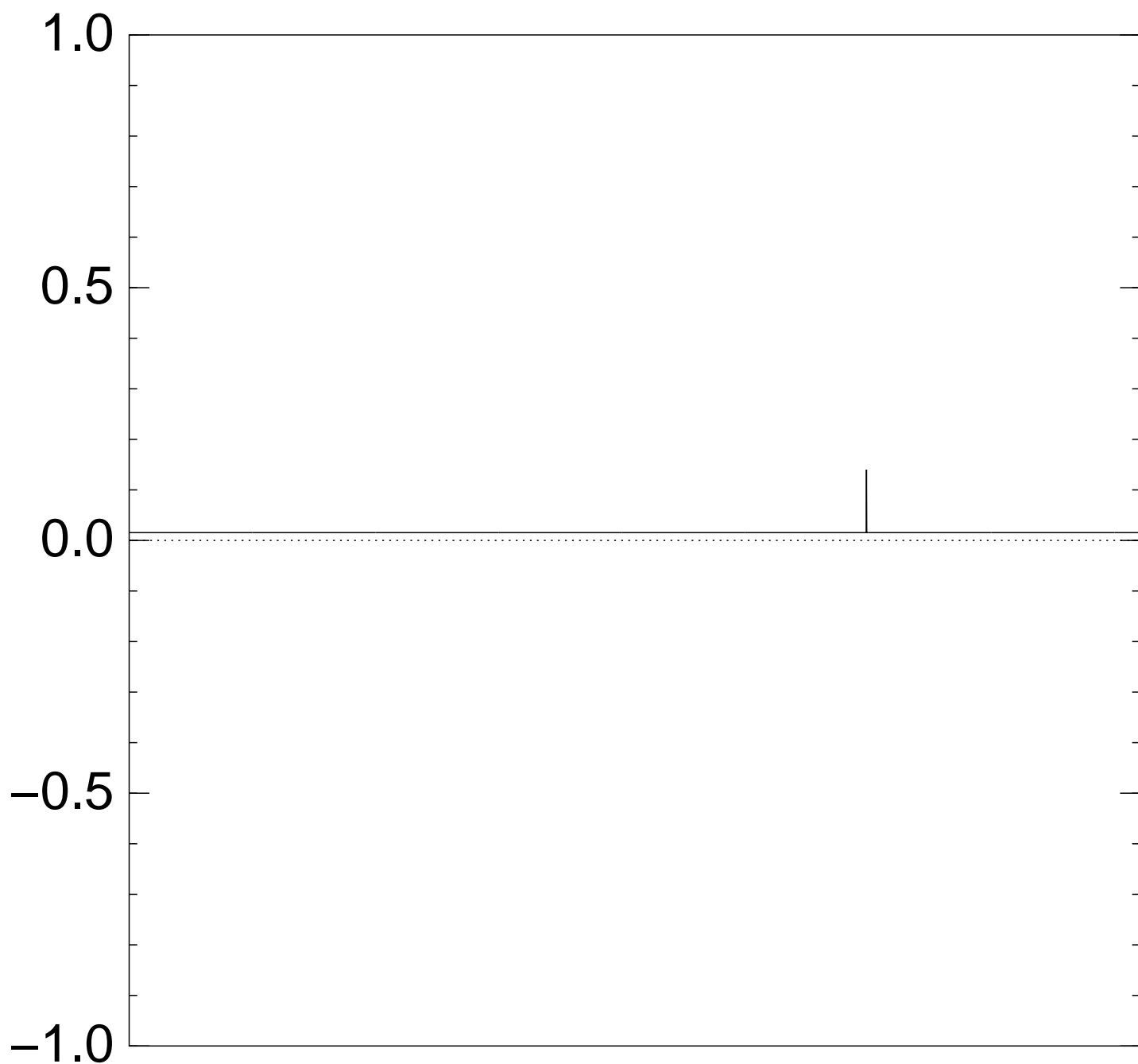




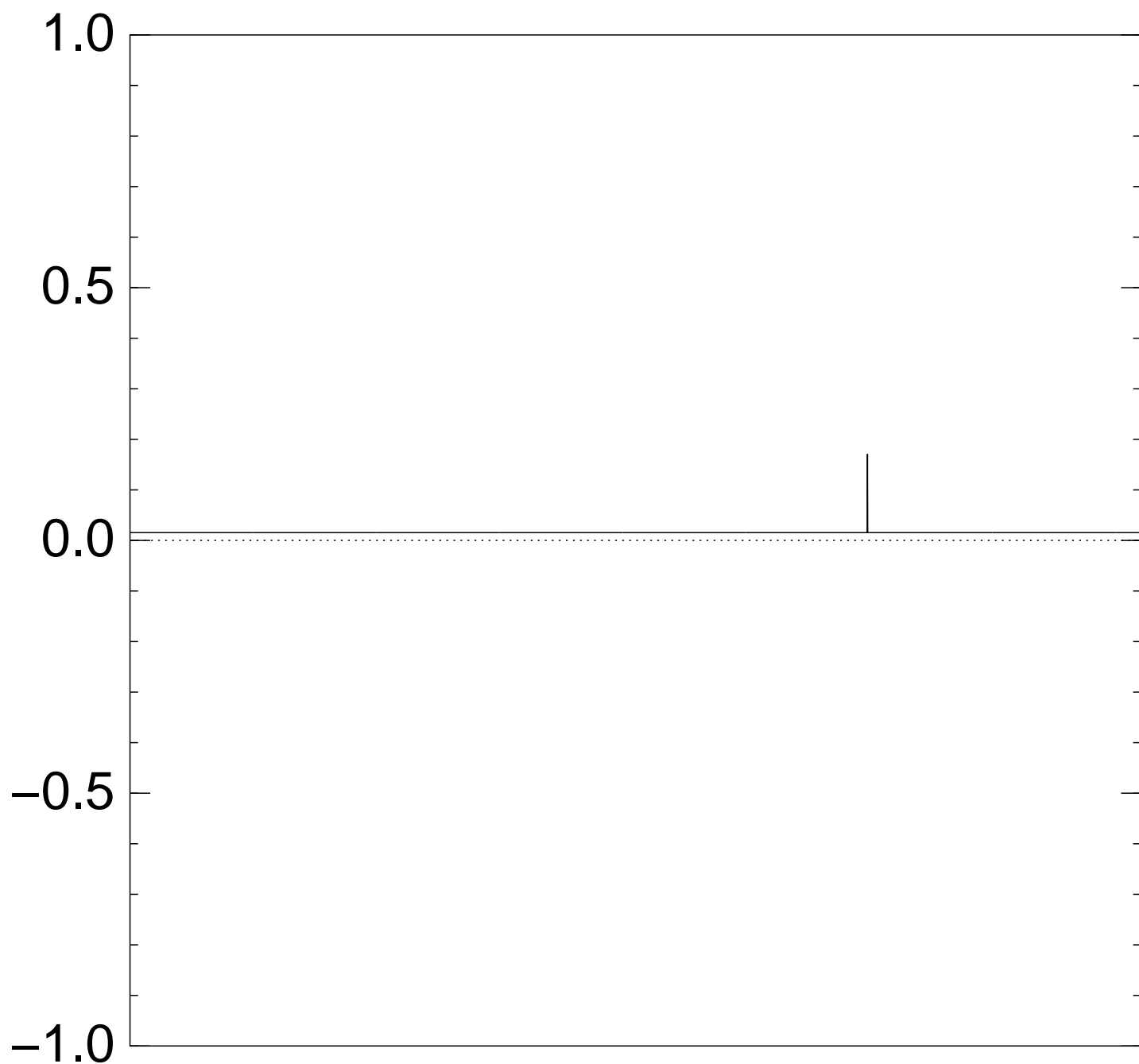
Normalized graph of  $u \mapsto a_u$   
for an example with  $n = 12$   
after  $3 \times$  (Step 1 + Step 2):



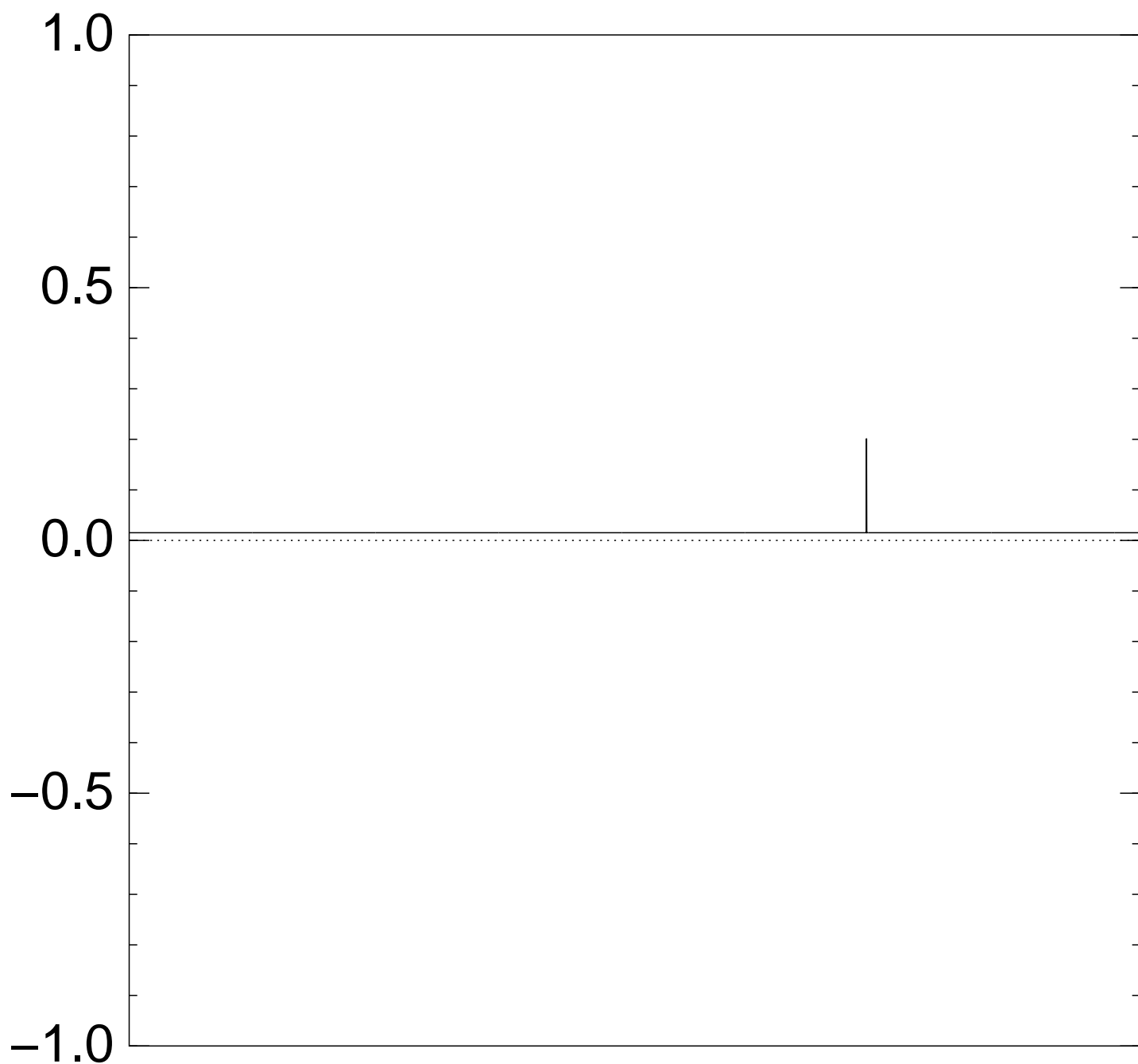
Normalized graph of  $u \mapsto a_u$   
for an example with  $n = 12$   
after  $4 \times$  (Step 1 + Step 2):



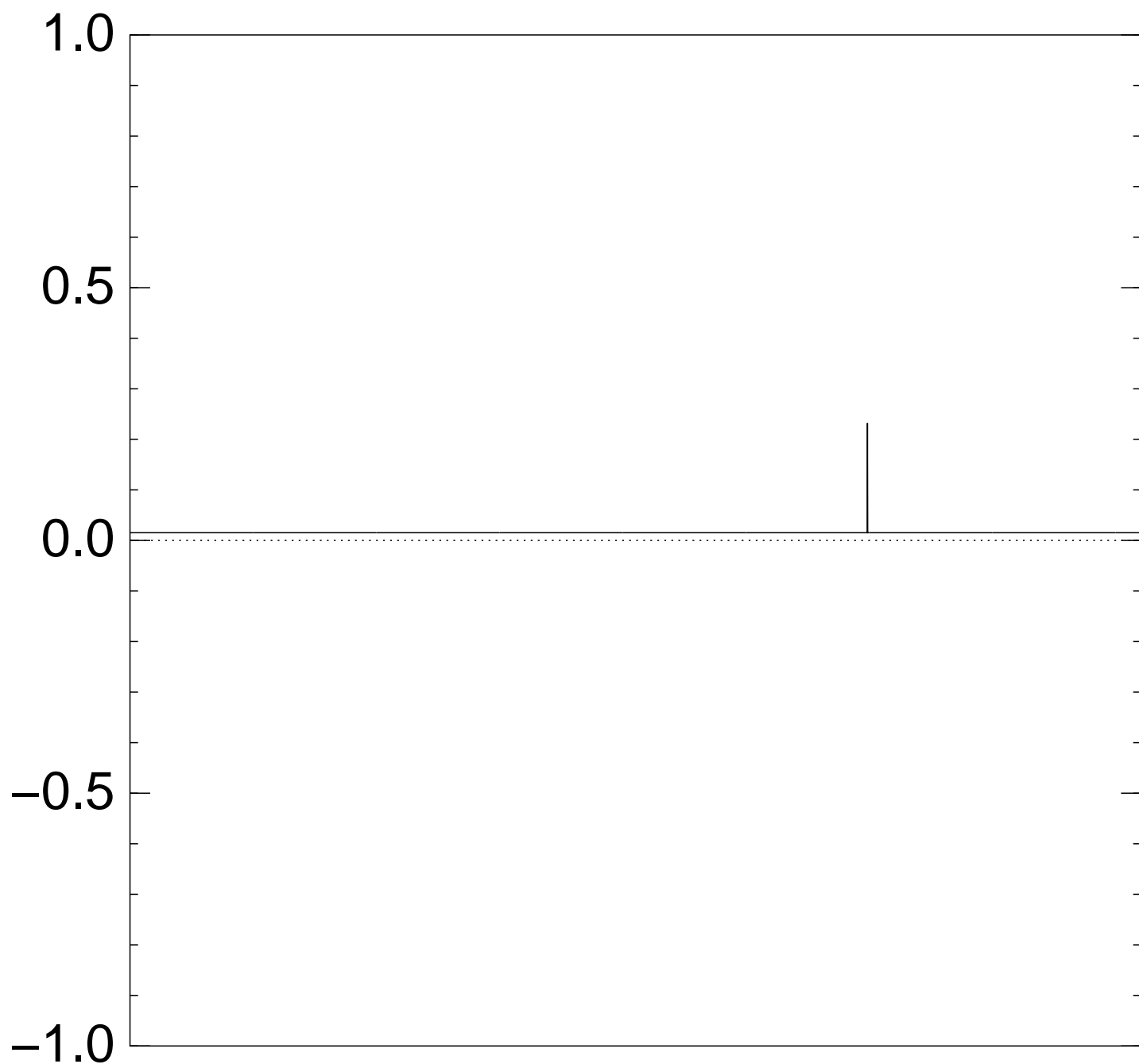
Normalized graph of  $u \mapsto a_u$   
for an example with  $n = 12$   
after  $5 \times$  (Step 1 + Step 2):



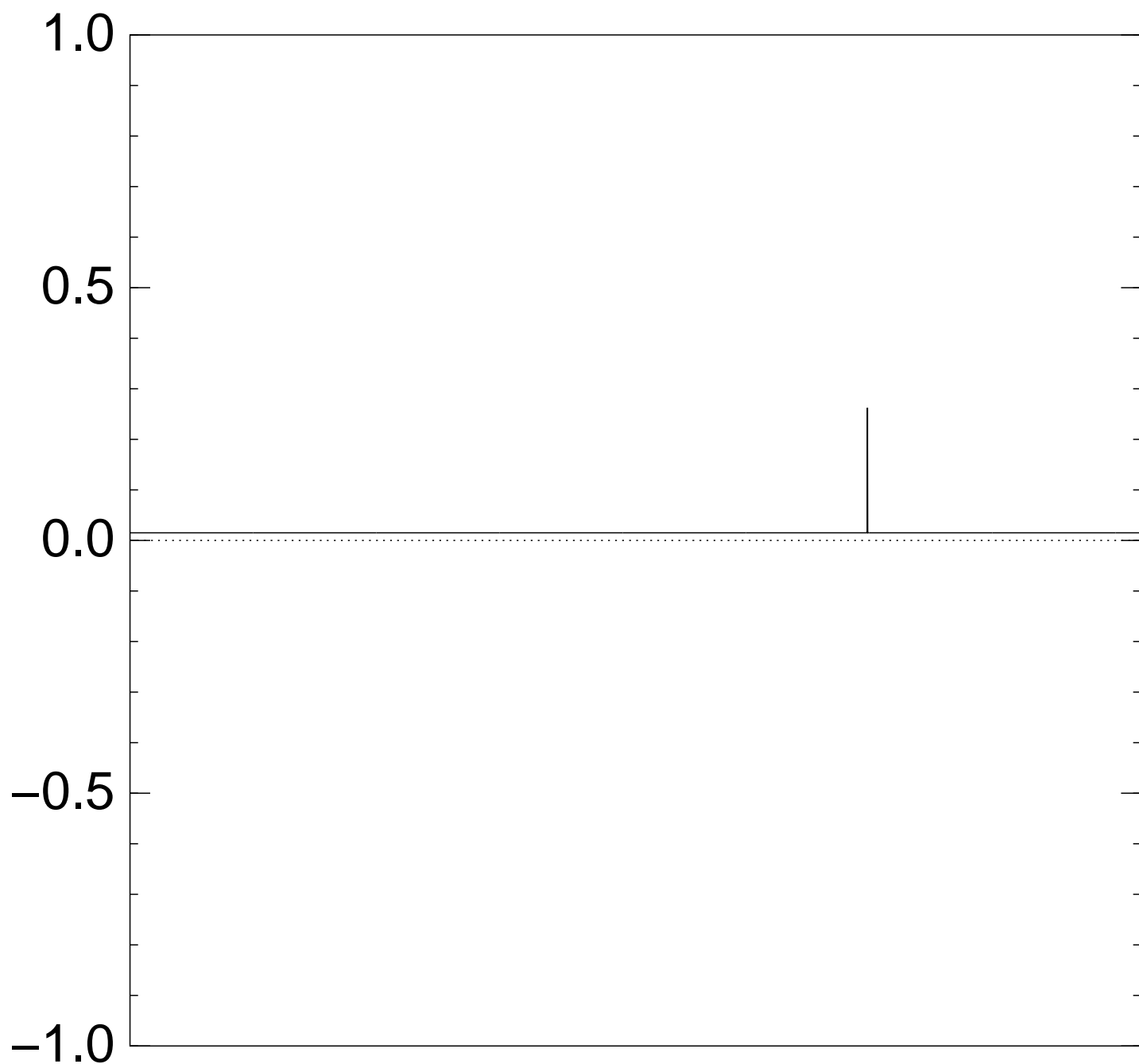
Normalized graph of  $u \mapsto a_u$   
for an example with  $n = 12$   
after  $6 \times$  (Step 1 + Step 2):



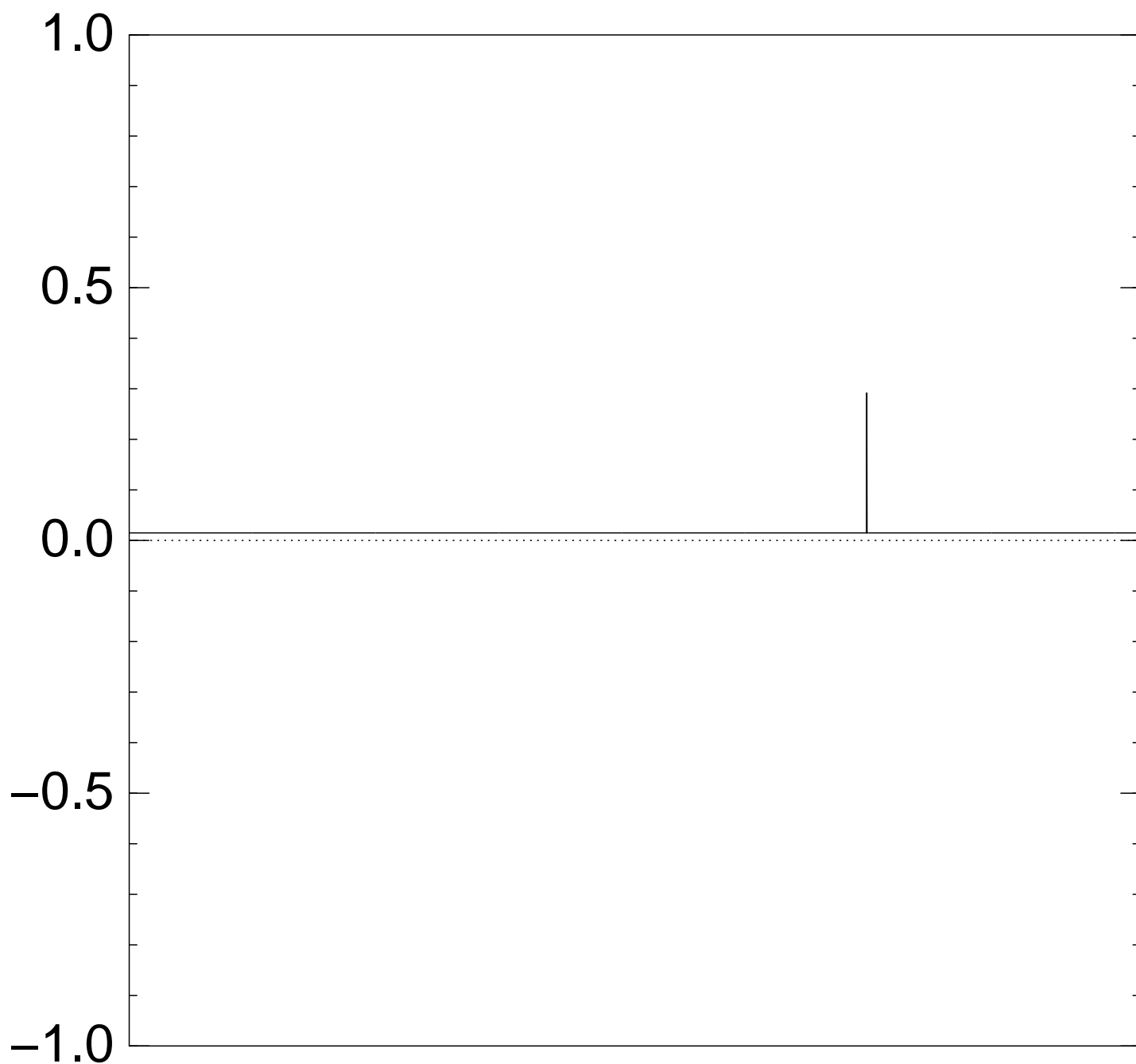
Normalized graph of  $u \mapsto a_u$   
for an example with  $n = 12$   
after  $7 \times$  (Step 1 + Step 2):



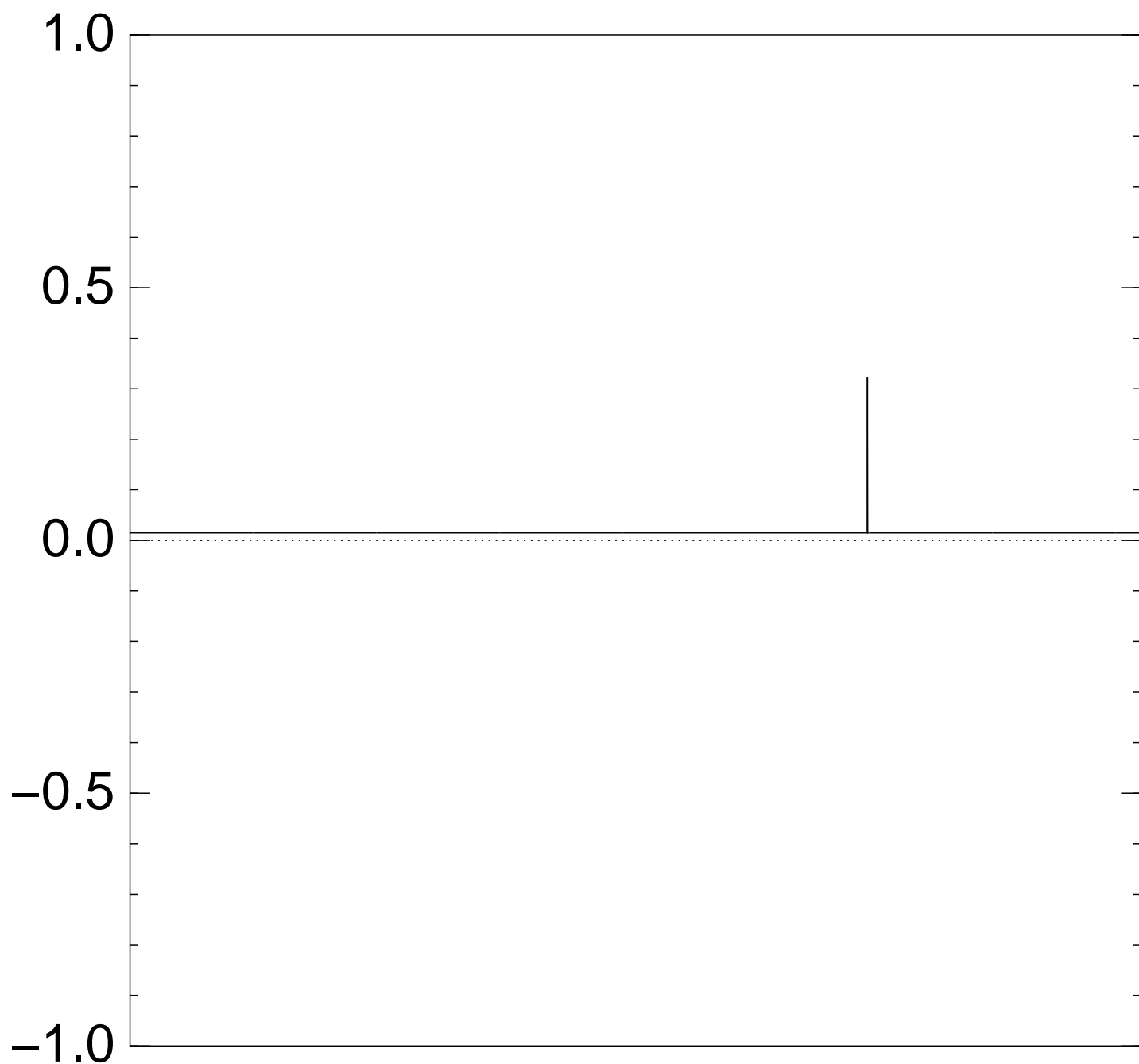
Normalized graph of  $u \mapsto a_u$   
for an example with  $n = 12$   
after  $8 \times$  (Step 1 + Step 2):



Normalized graph of  $u \mapsto a_u$   
for an example with  $n = 12$   
after  $9 \times$  (Step 1 + Step 2):

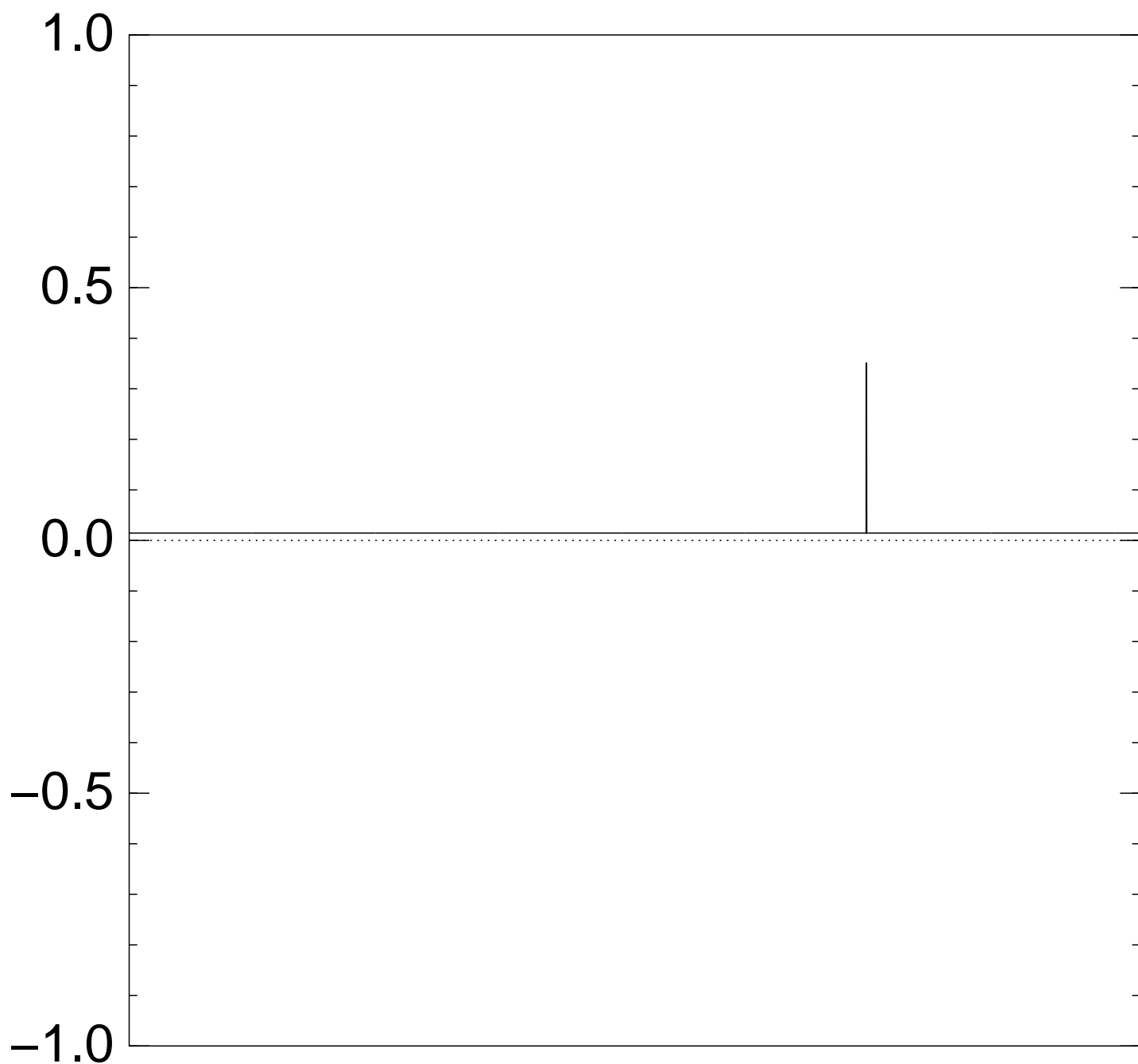


Normalized graph of  $u \mapsto a_u$   
for an example with  $n = 12$   
after  $10 \times$  (Step 1 + Step 2):

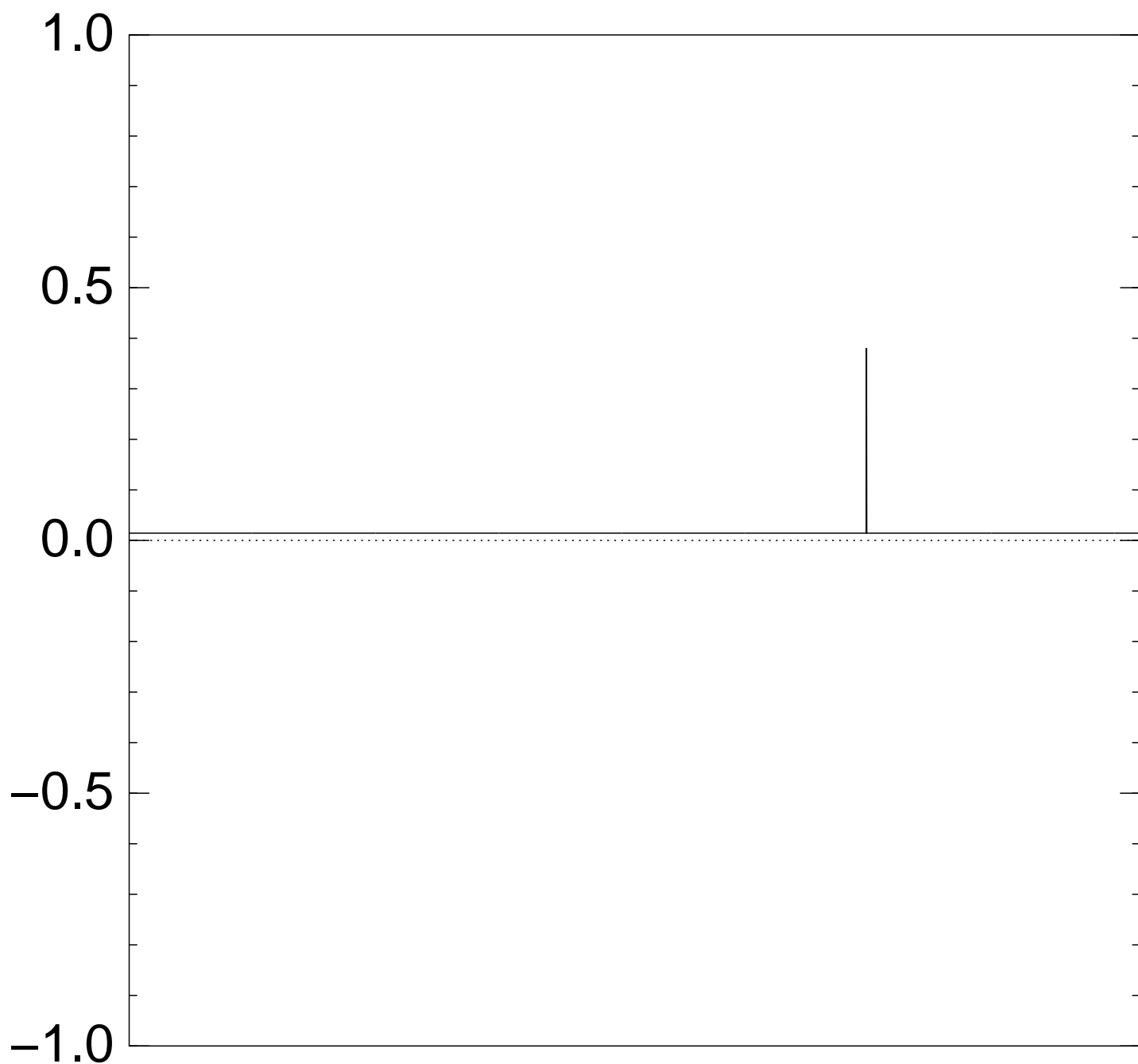




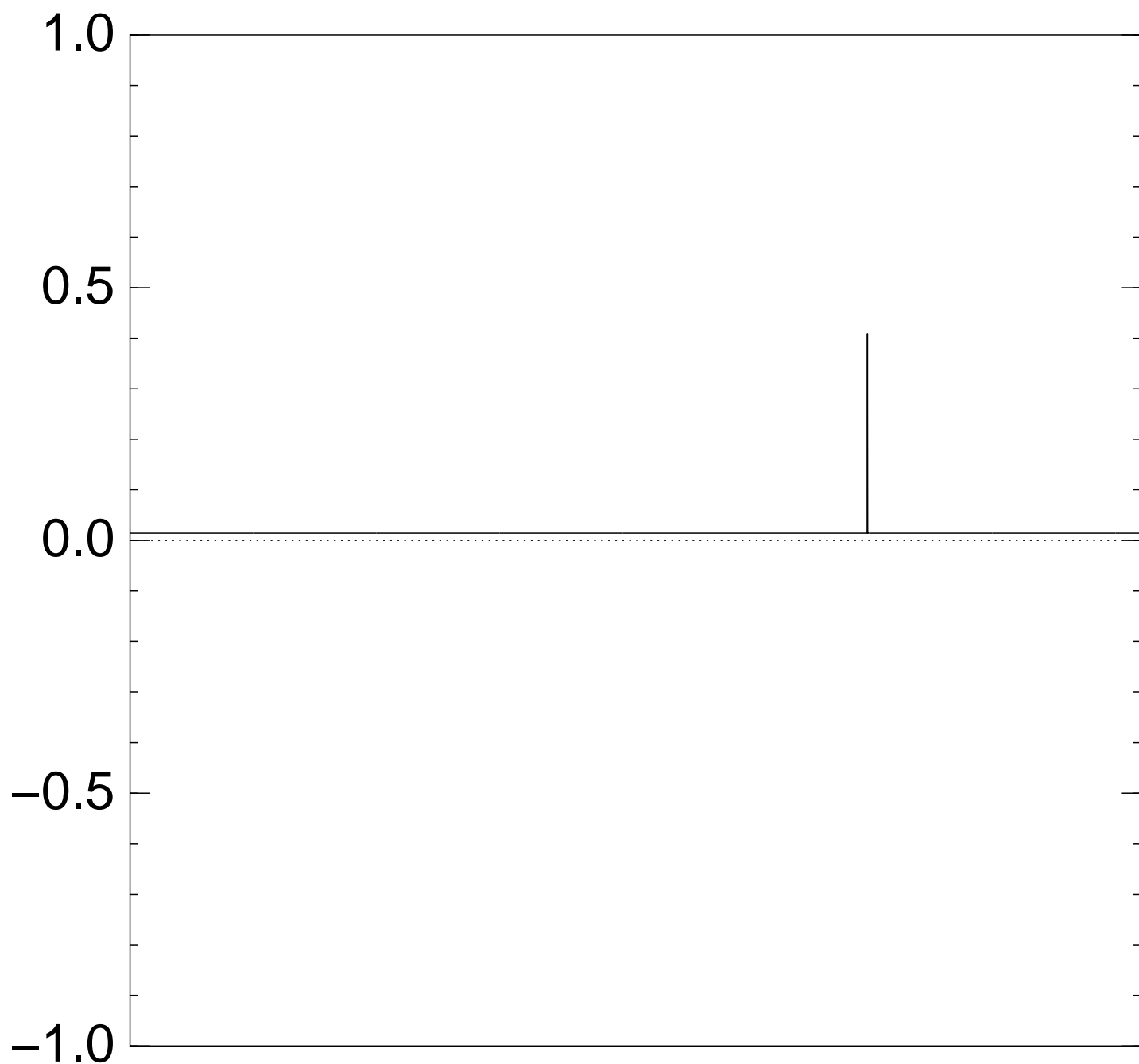
Normalized graph of  $u \mapsto a_u$   
for an example with  $n = 12$   
after  $11 \times$  (Step 1 + Step 2):



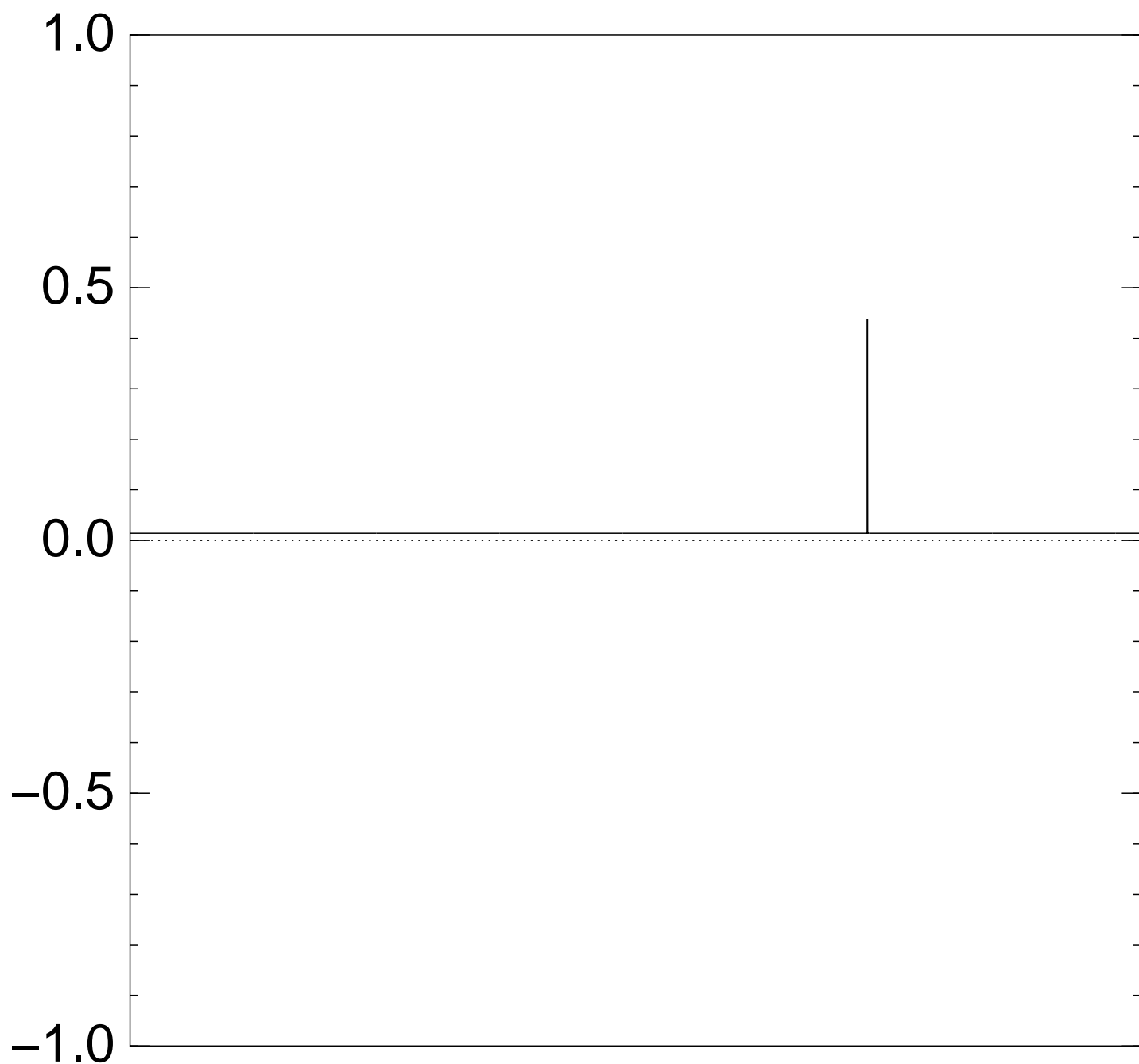
Normalized graph of  $u \mapsto a_u$   
for an example with  $n = 12$   
after  $12 \times$  (Step 1 + Step 2):



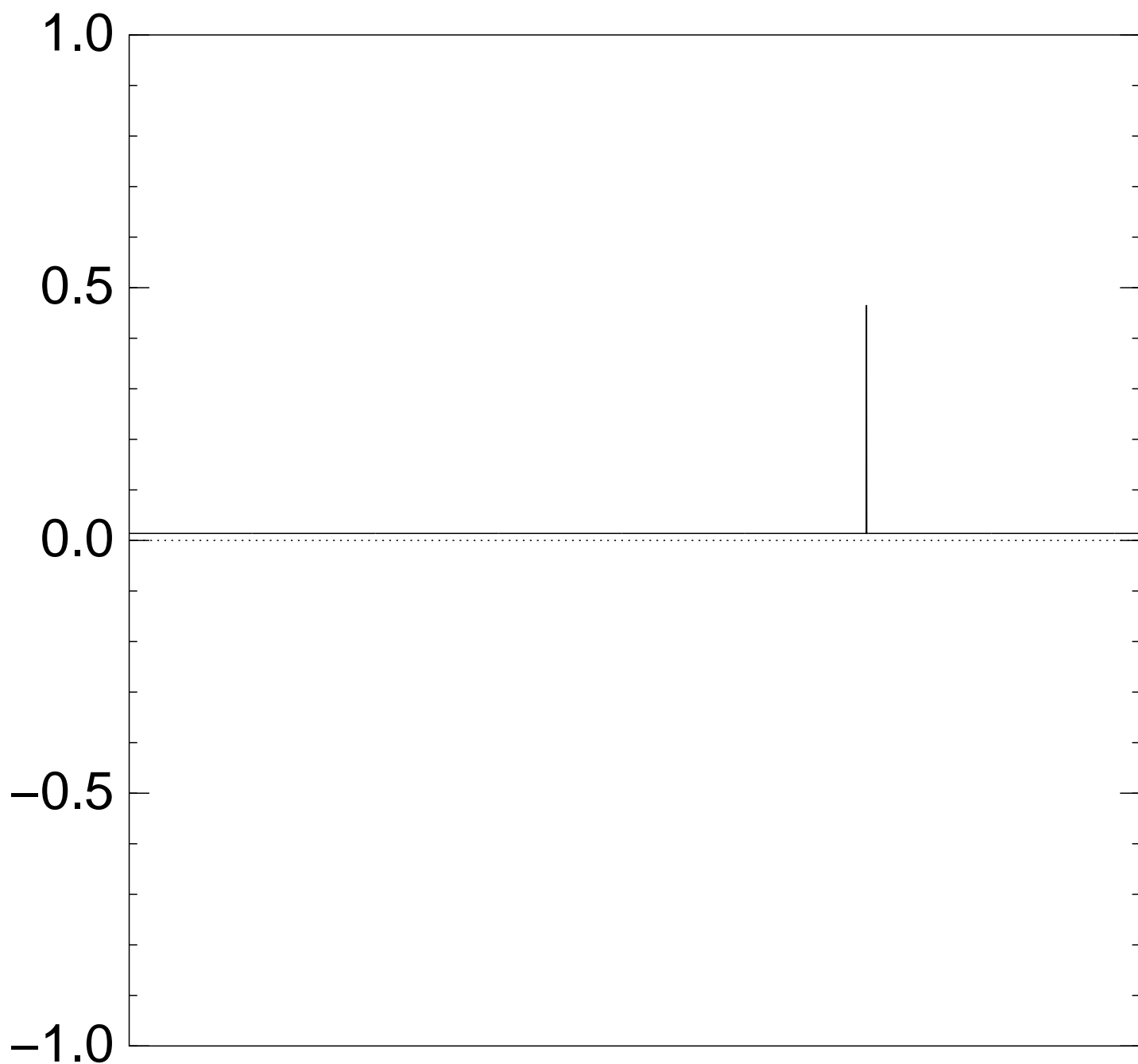
Normalized graph of  $u \mapsto a_u$   
for an example with  $n = 12$   
after  $13 \times$  (Step 1 + Step 2):



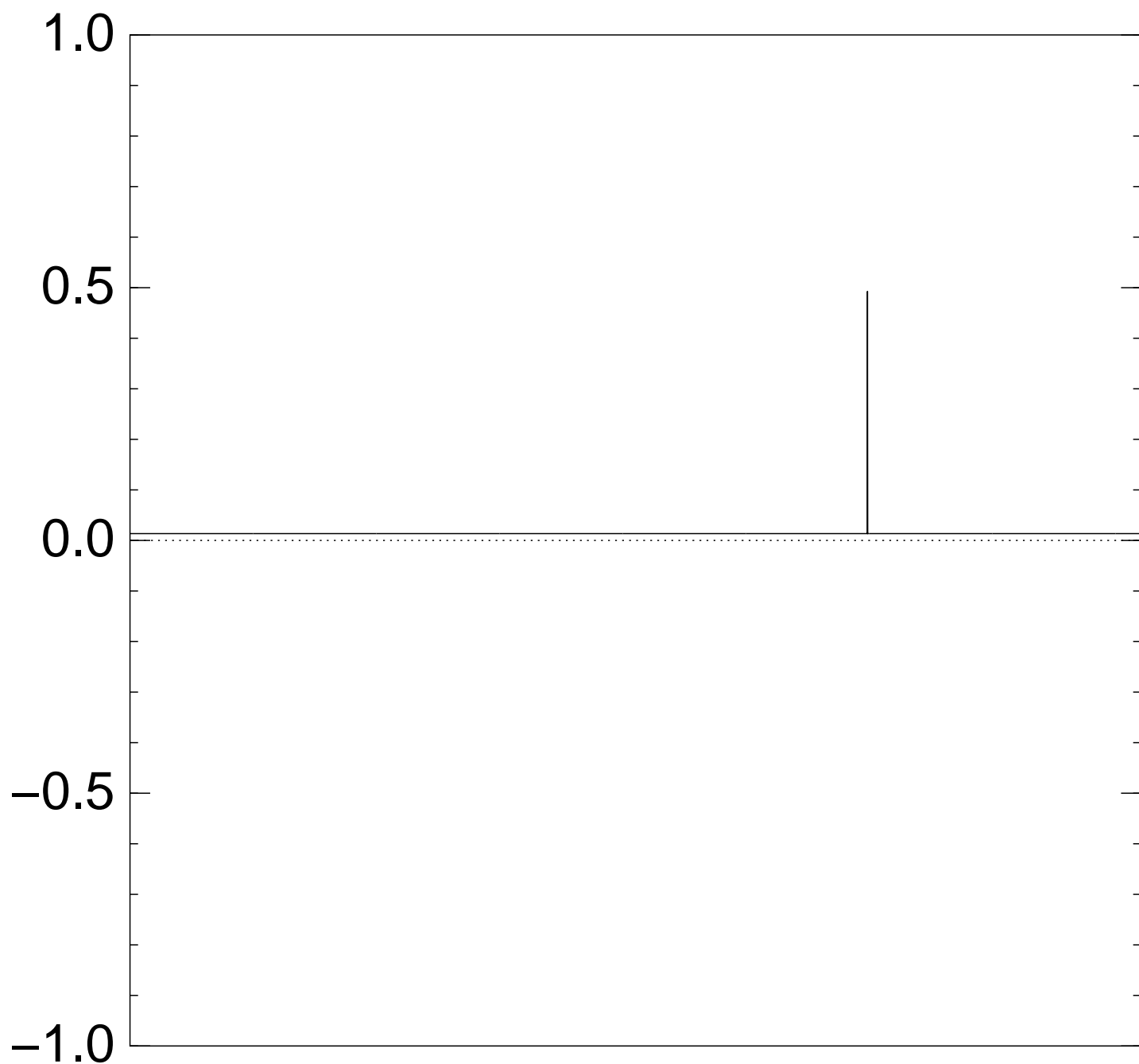
Normalized graph of  $u \mapsto a_u$   
for an example with  $n = 12$   
after  $14 \times$  (Step 1 + Step 2):



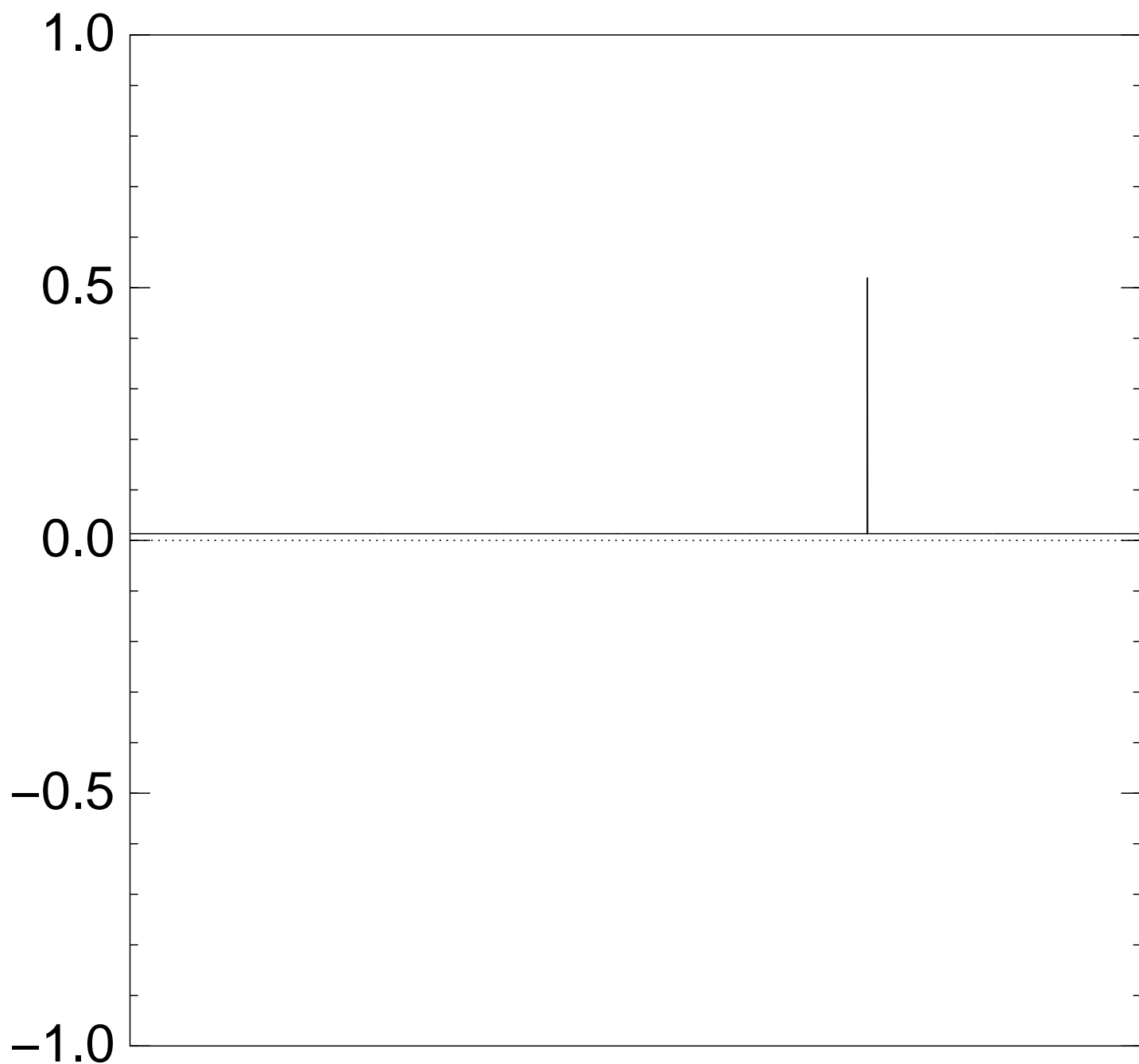
Normalized graph of  $u \mapsto a_u$   
for an example with  $n = 12$   
after  $15 \times$  (Step 1 + Step 2):



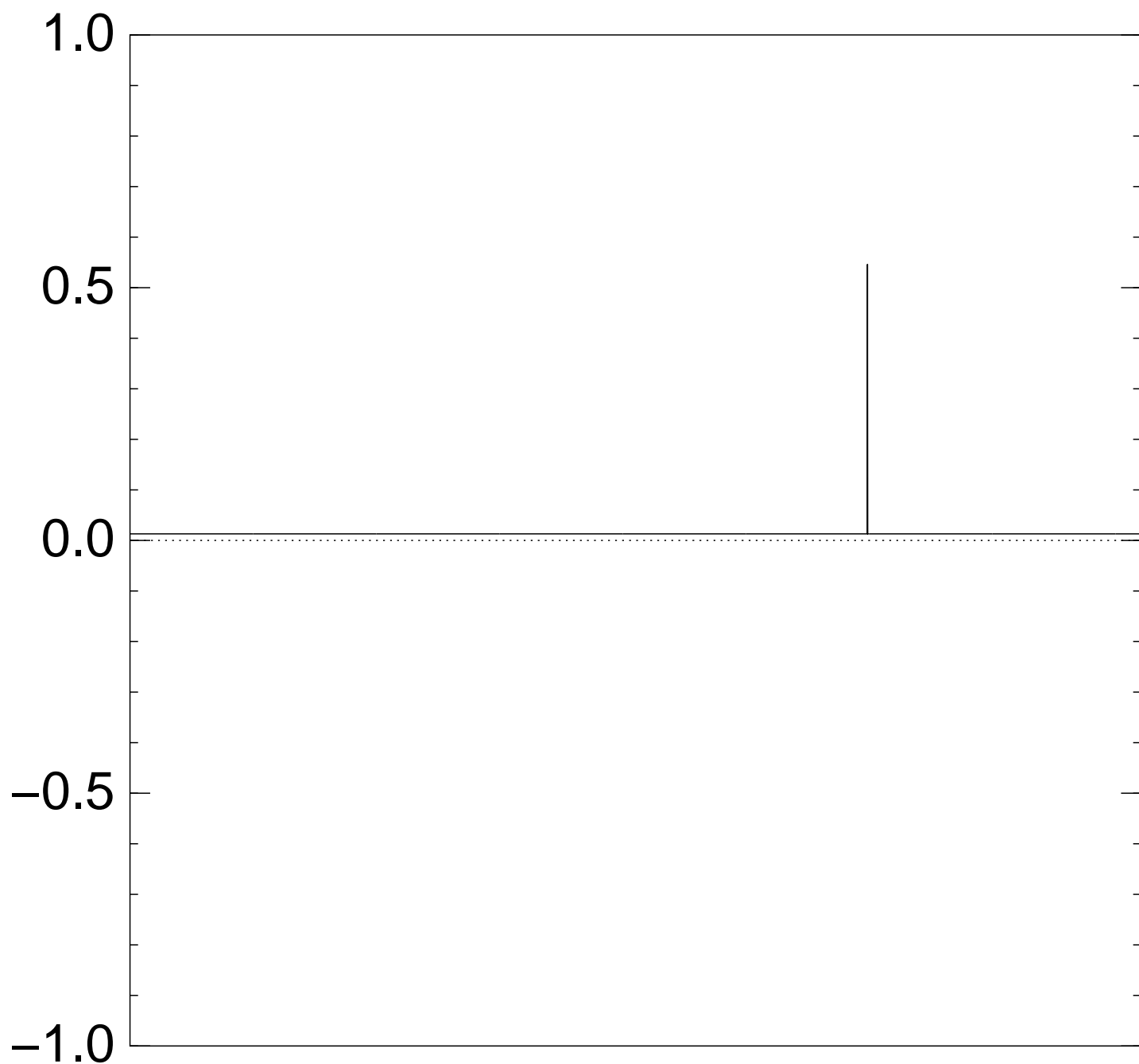
Normalized graph of  $u \mapsto a_u$   
for an example with  $n = 12$   
after  $16 \times$  (Step 1 + Step 2):



Normalized graph of  $u \mapsto a_u$   
for an example with  $n = 12$   
after  $17 \times$  (Step 1 + Step 2):

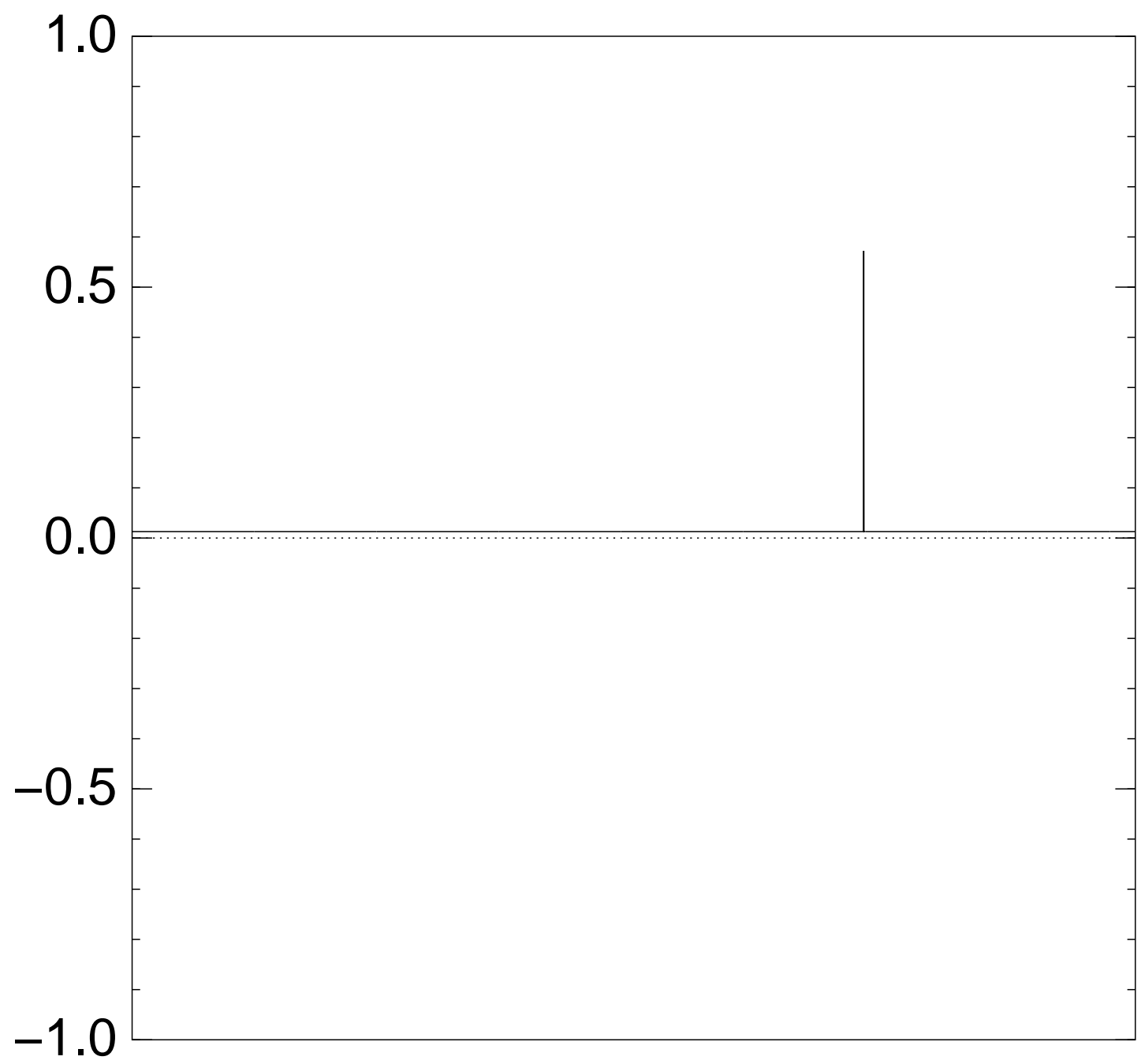


Normalized graph of  $u \mapsto a_u$   
for an example with  $n = 12$   
after  $18 \times$  (Step 1 + Step 2):

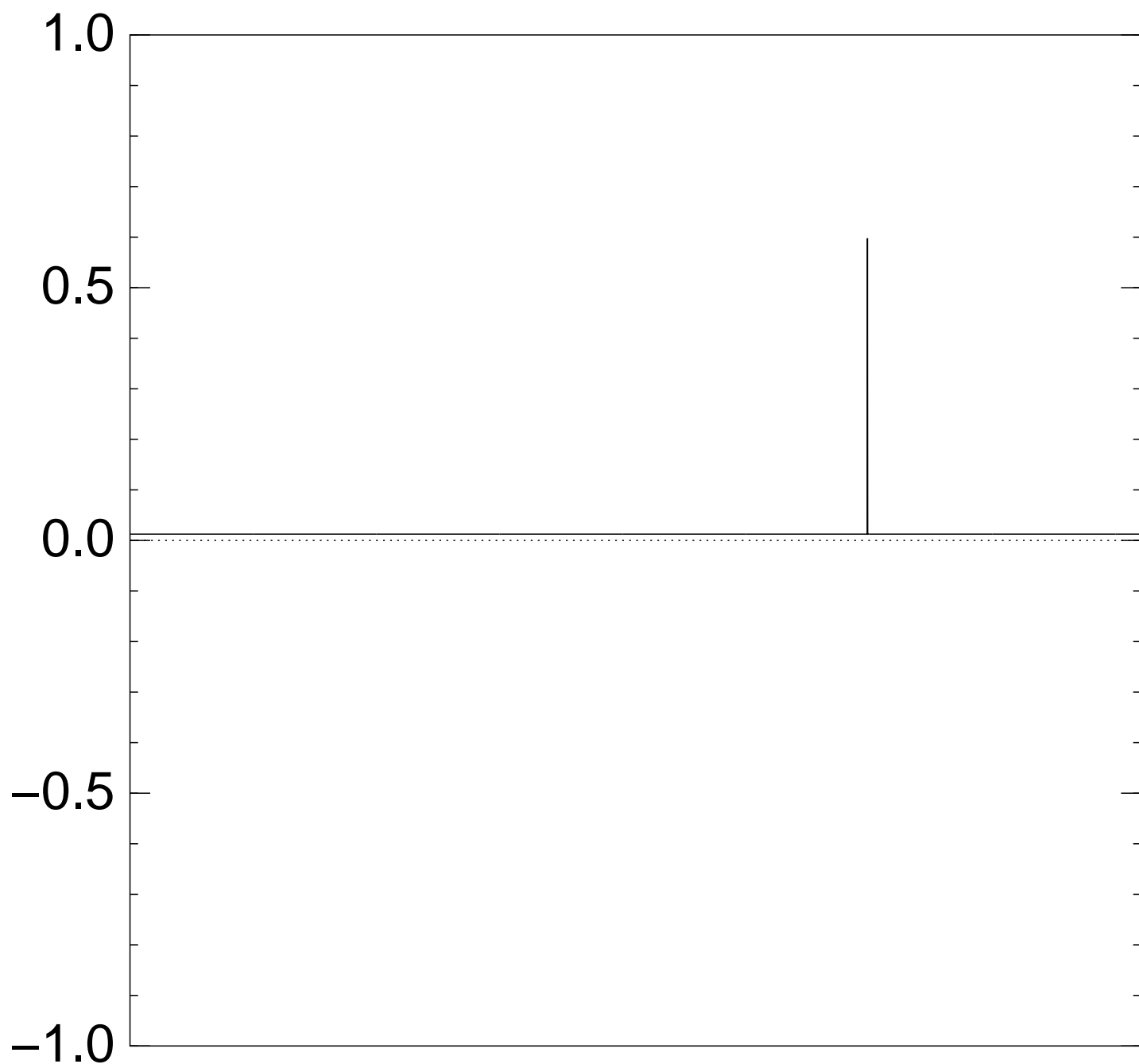




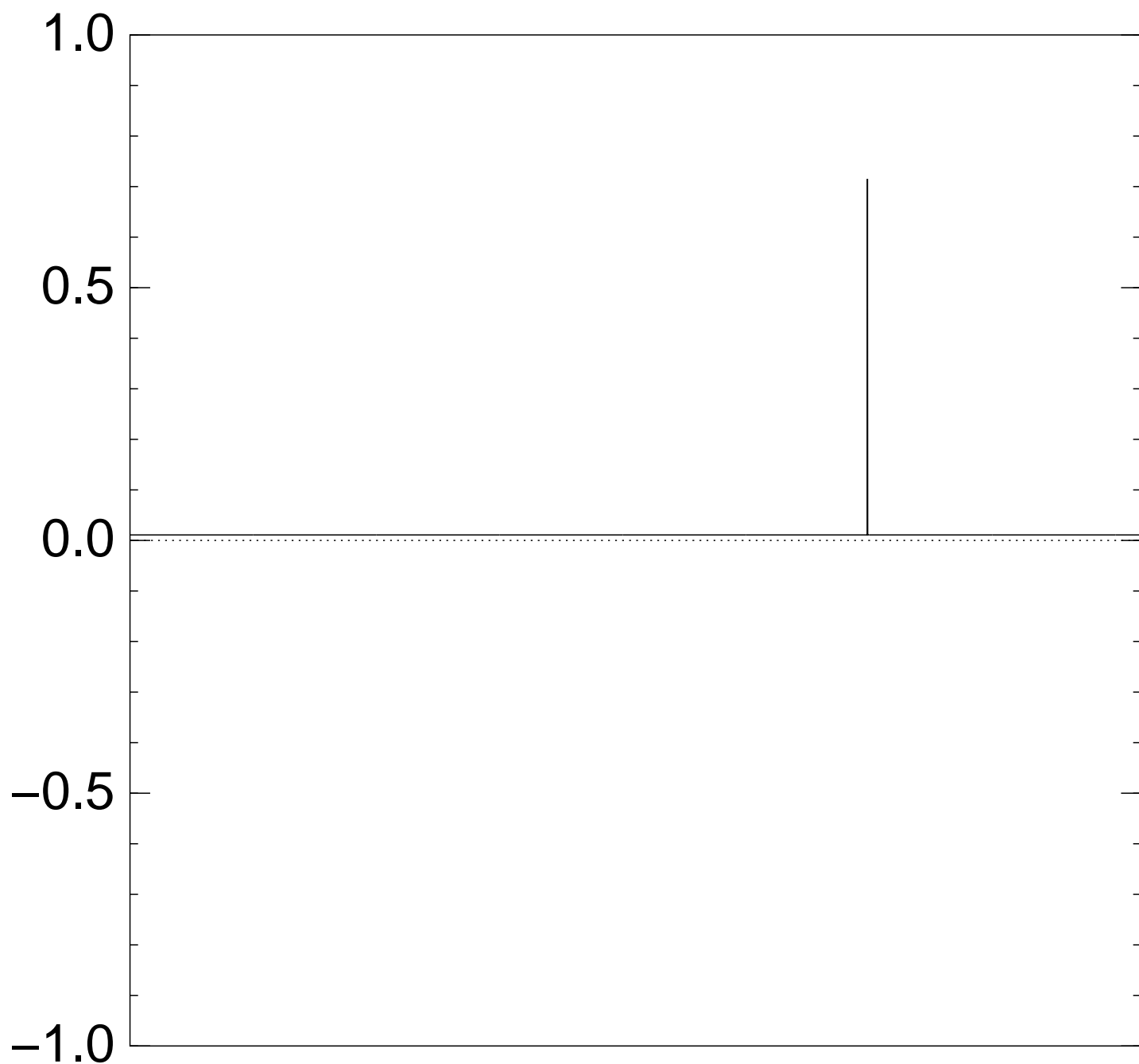
Normalized graph of  $u \mapsto a_u$   
for an example with  $n = 12$   
after  $19 \times$  (Step 1 + Step 2):



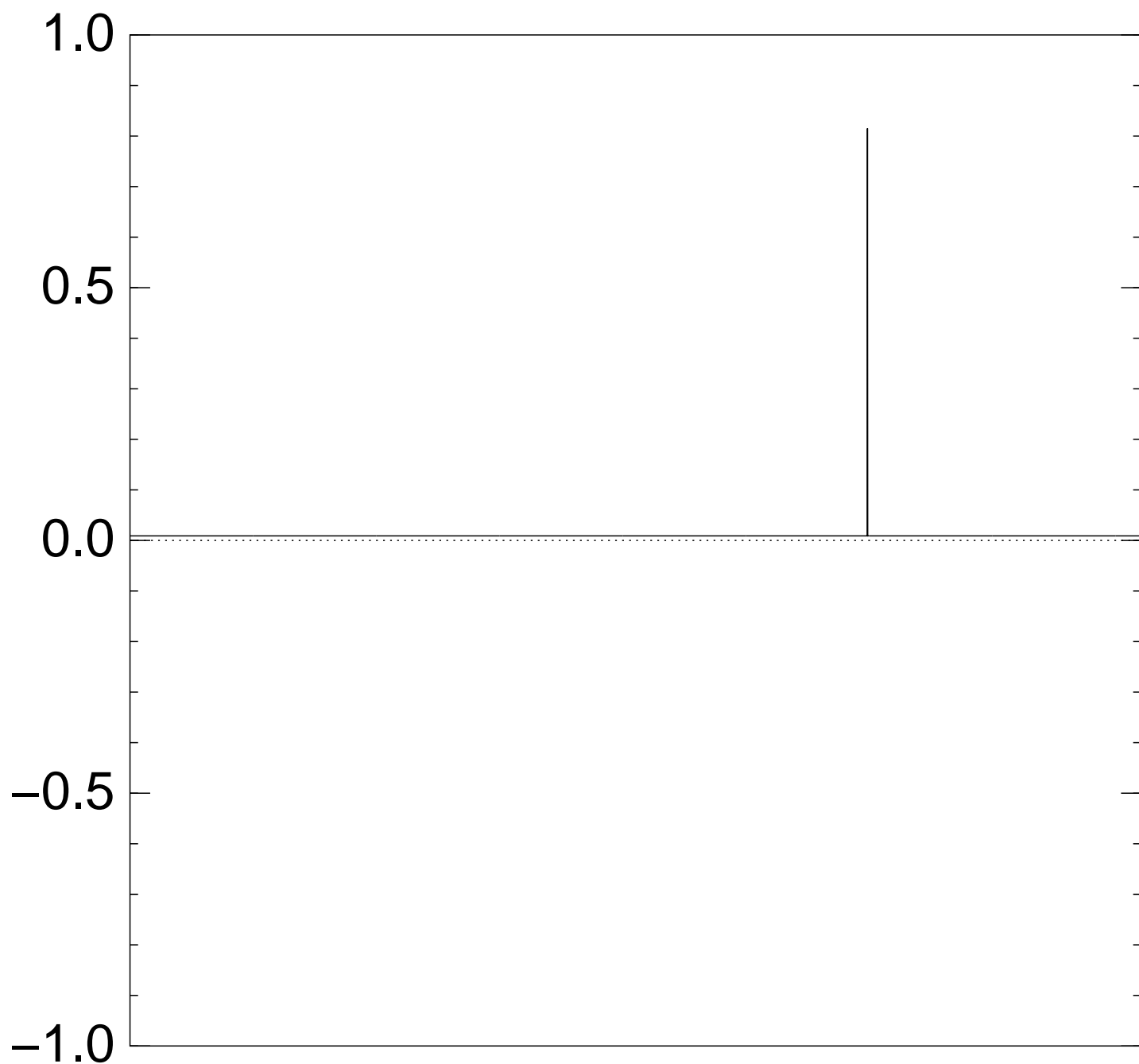
Normalized graph of  $u \mapsto a_u$   
for an example with  $n = 12$   
after  $20 \times$  (Step 1 + Step 2):



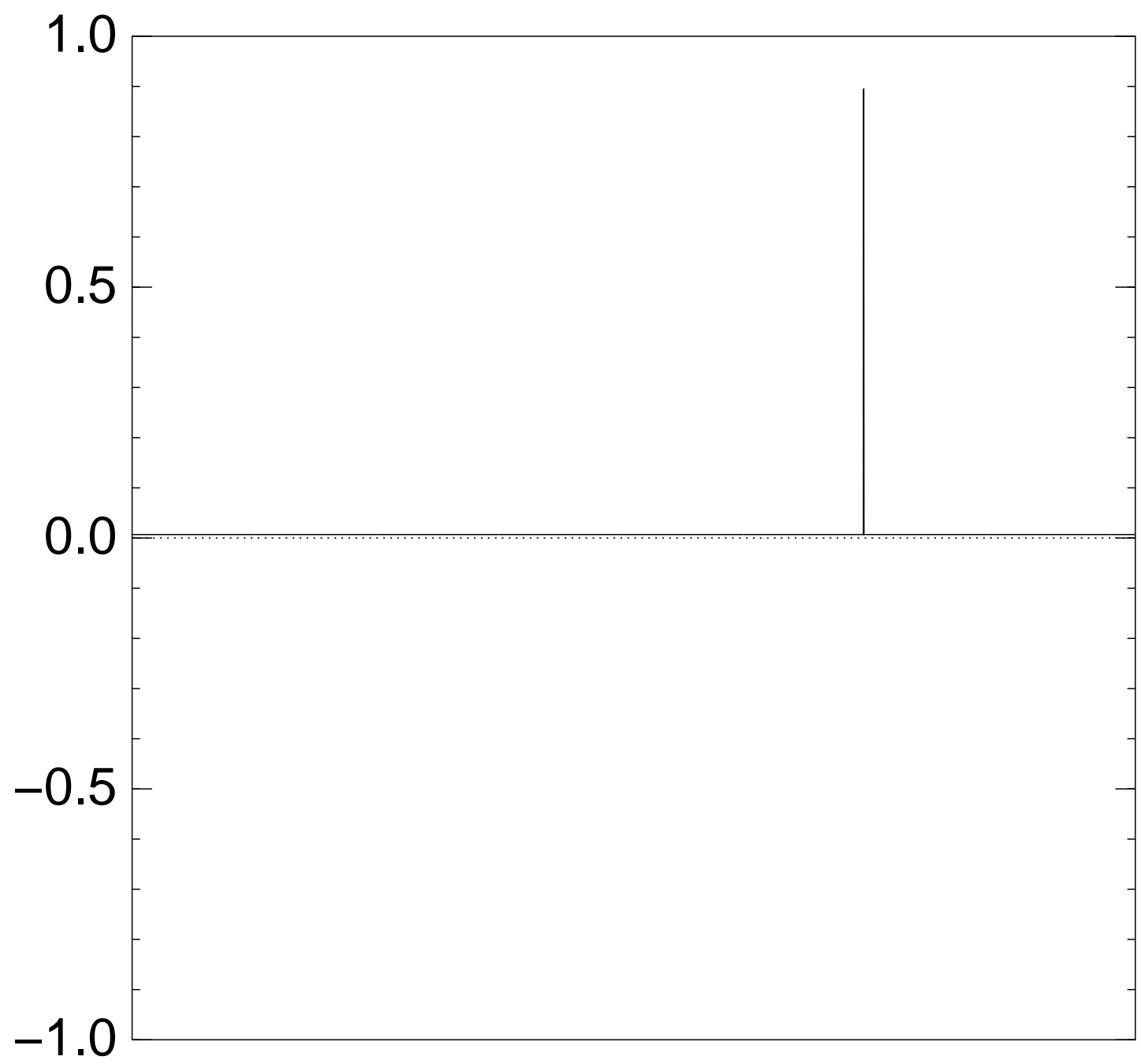
Normalized graph of  $u \mapsto a_u$   
for an example with  $n = 12$   
after  $25 \times$  (Step 1 + Step 2):



Normalized graph of  $u \mapsto a_u$   
for an example with  $n = 12$   
after  $30 \times$  (Step 1 + Step 2):

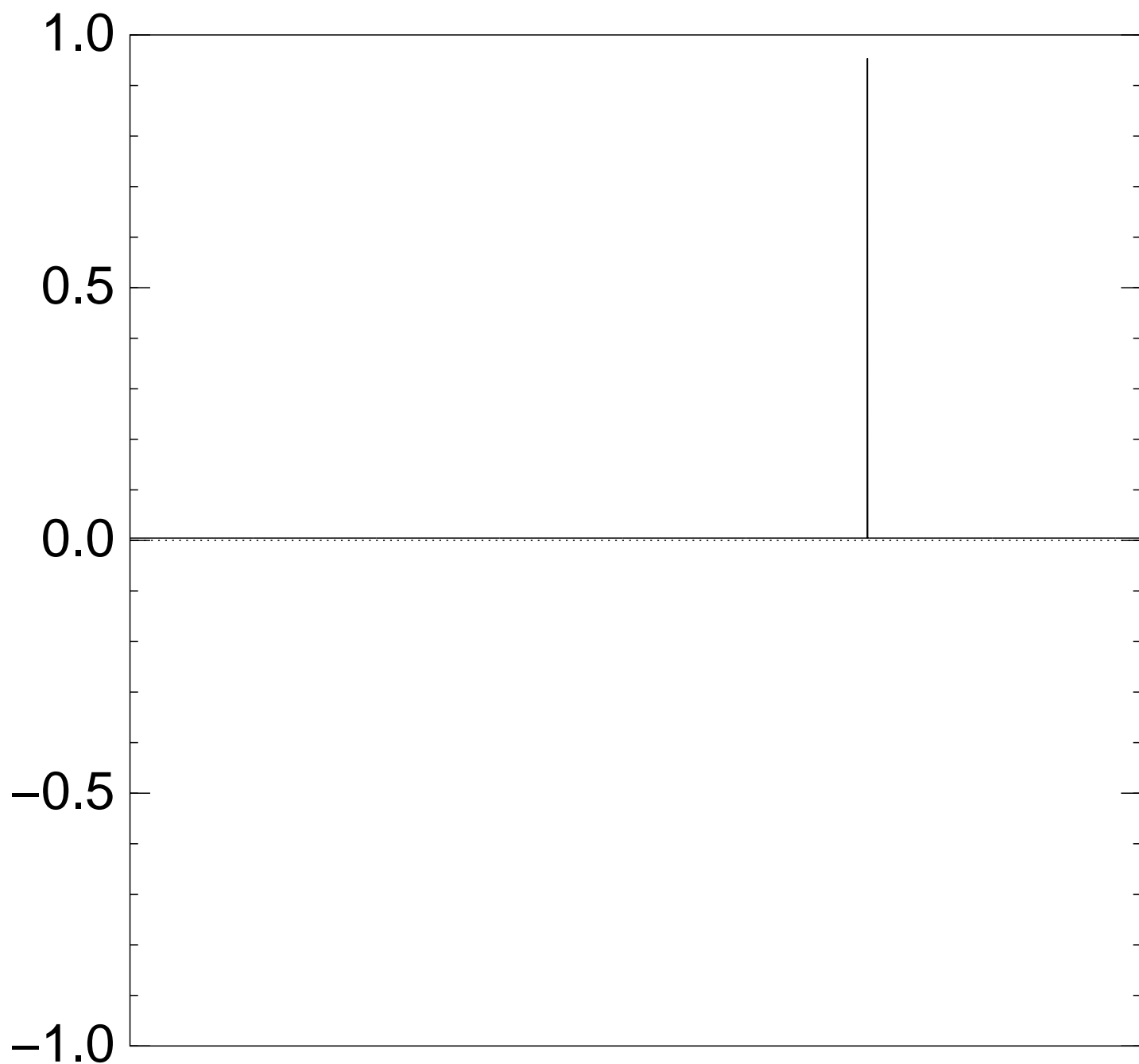


Normalized graph of  $u \mapsto a_u$   
for an example with  $n = 12$   
after  $35 \times$  (Step 1 + Step 2):

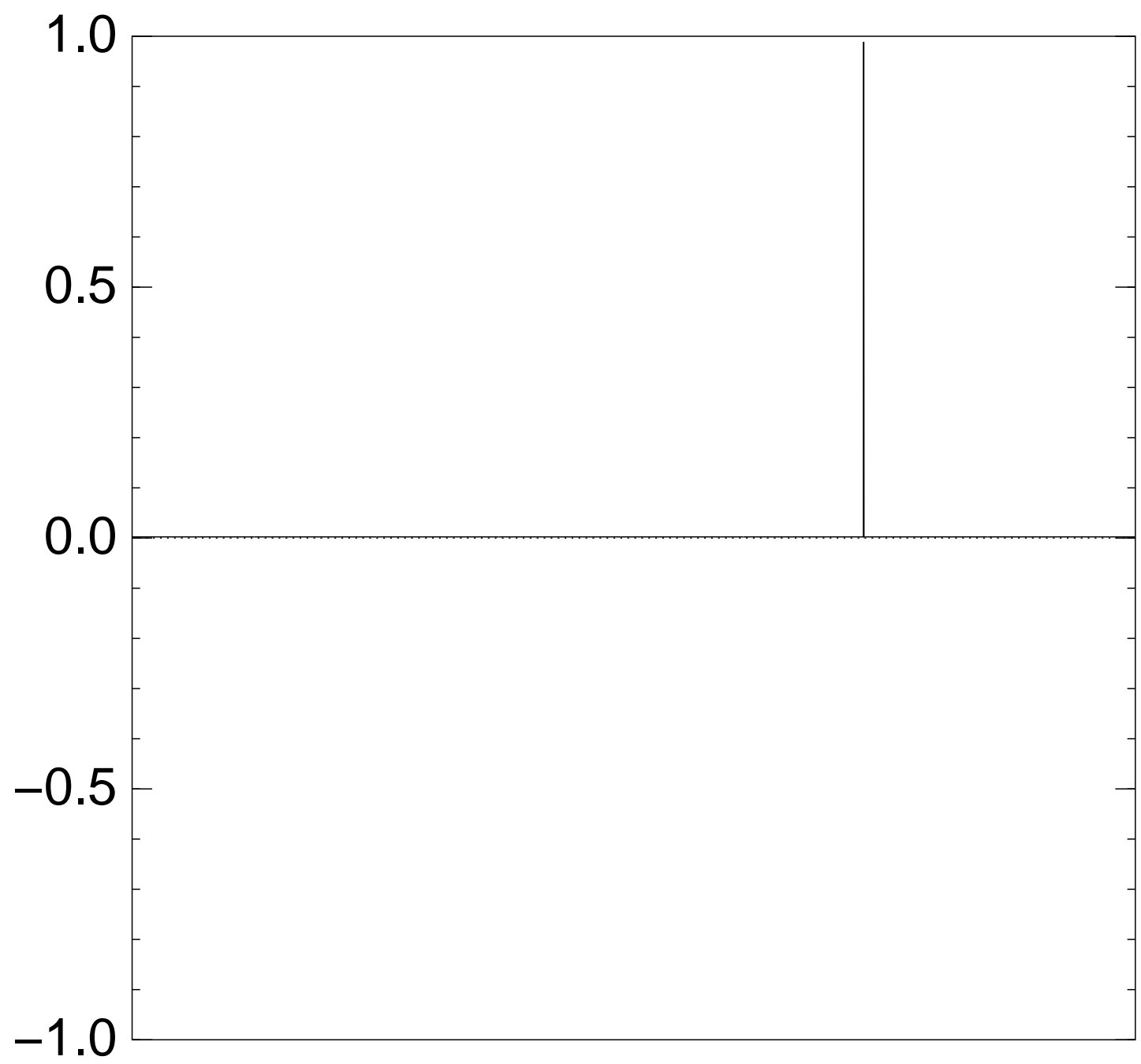


Good moment to stop, measure.

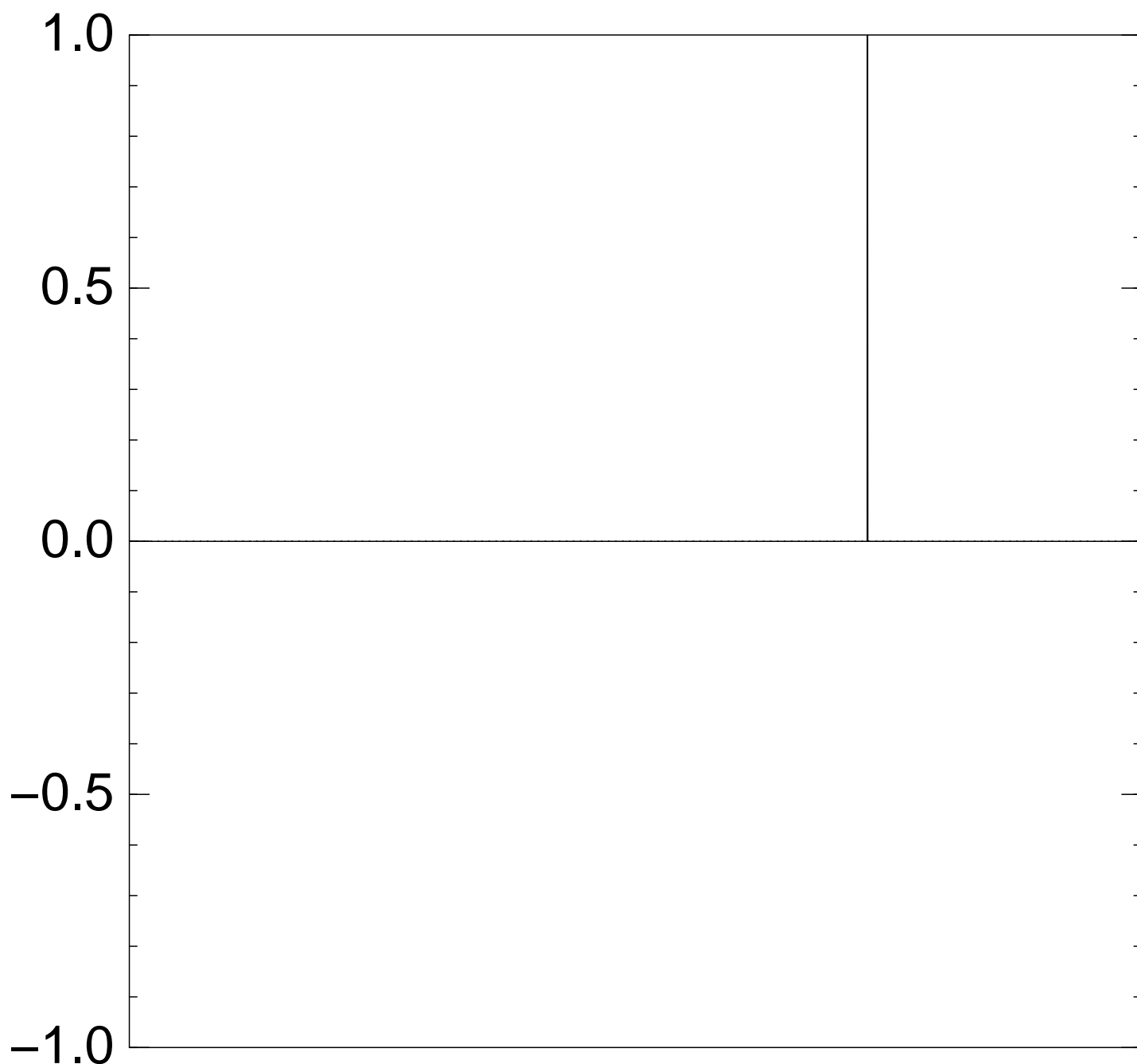
Normalized graph of  $u \mapsto a_u$   
for an example with  $n = 12$   
after  $40 \times$  (Step 1 + Step 2):



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



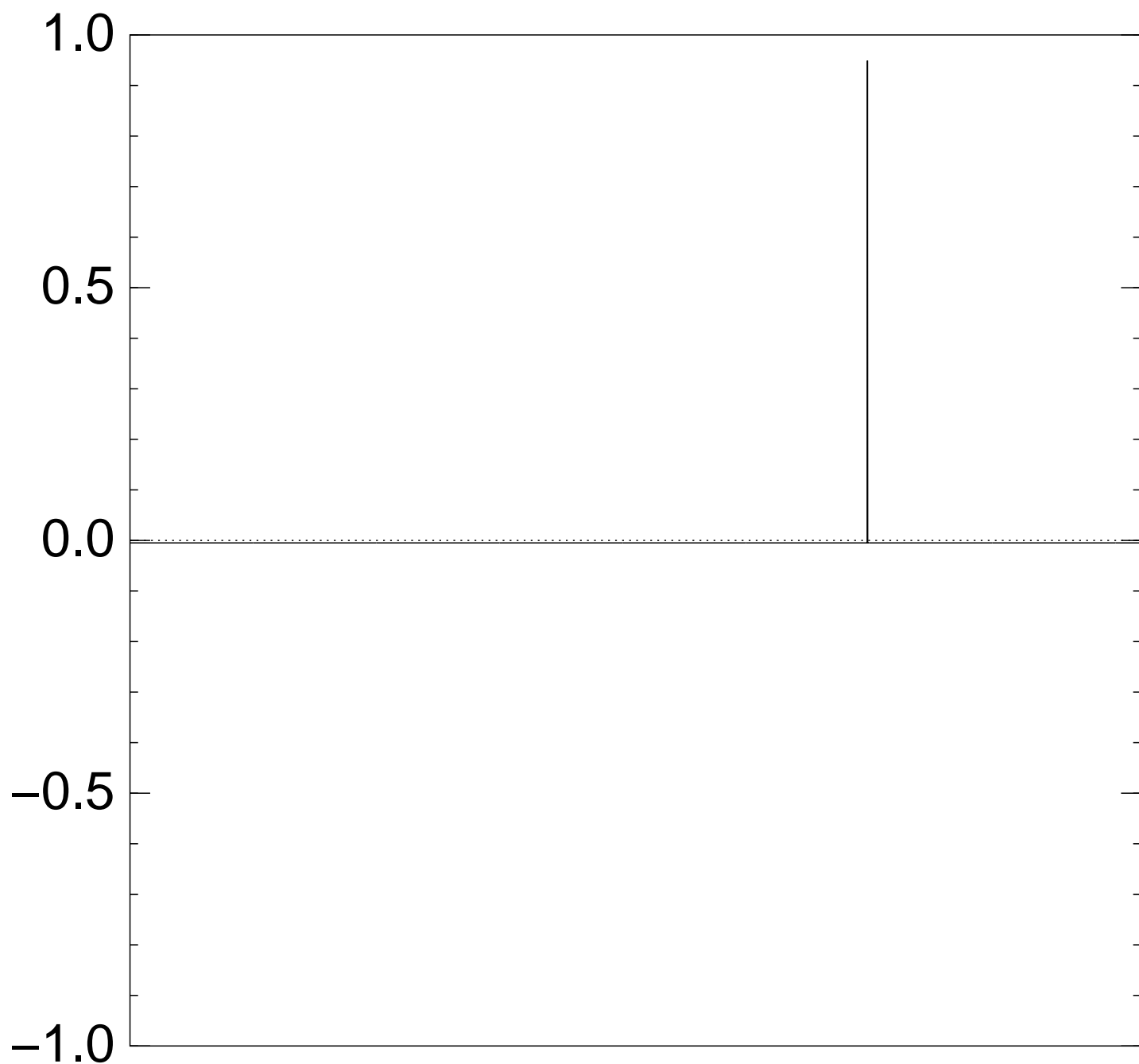
Normalized graph of  $u \mapsto a_u$   
for an example with  $n = 12$   
after  $50 \times$  (Step 1 + Step 2):



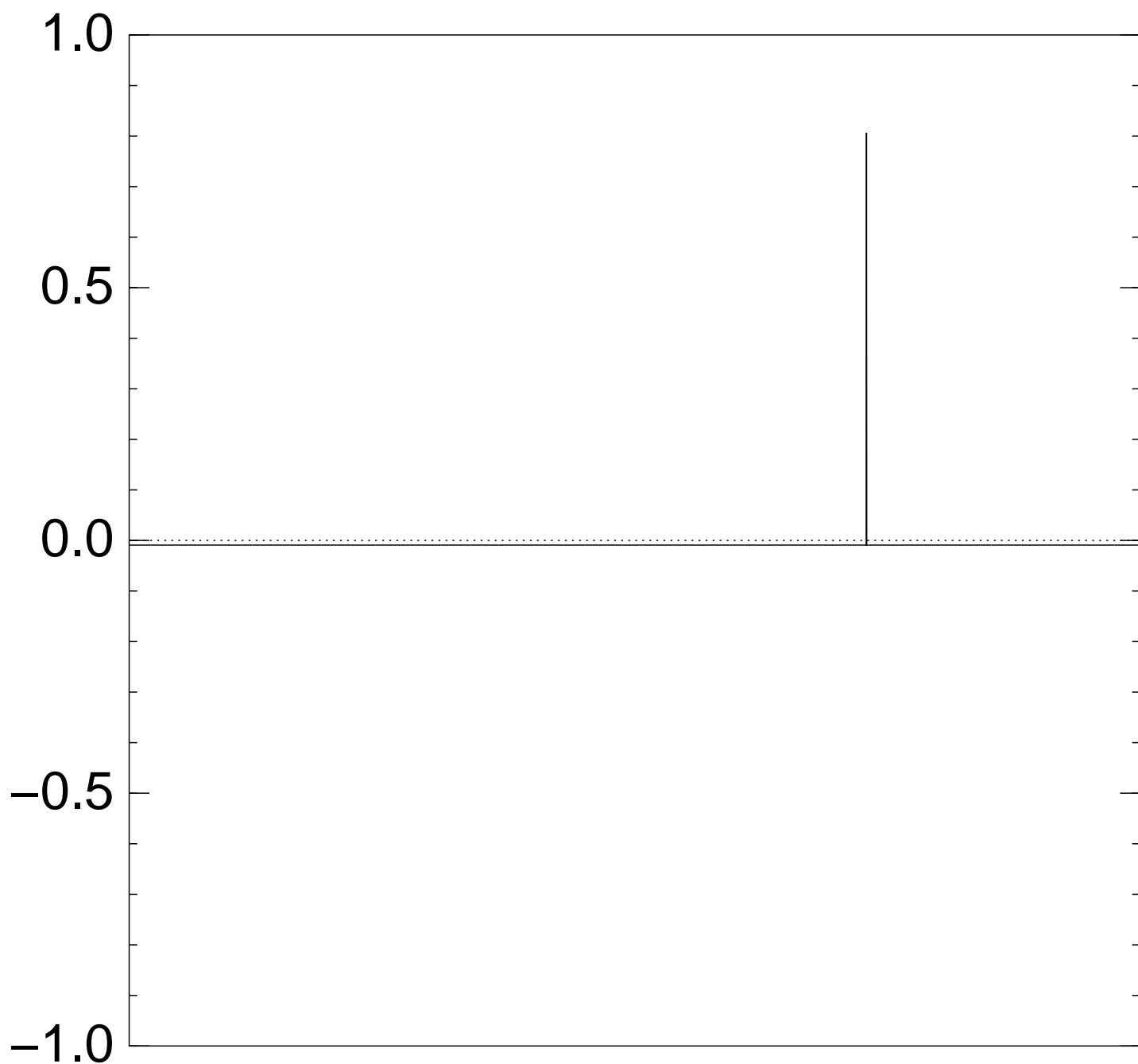
Traditional stopping point.



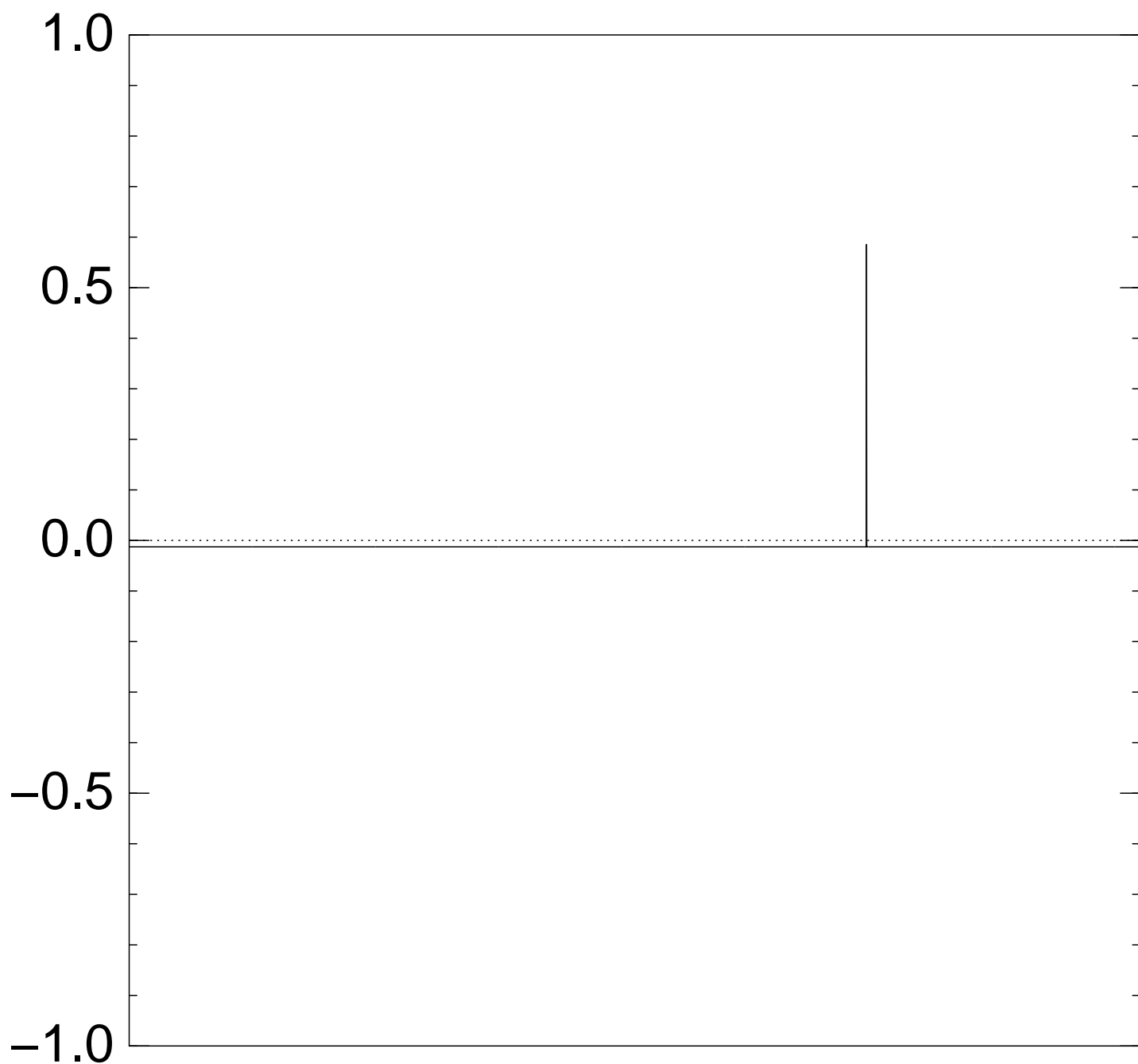
Normalized graph of  $u \mapsto a_u$   
for an example with  $n = 12$   
after  $60 \times$  (Step 1 + Step 2):



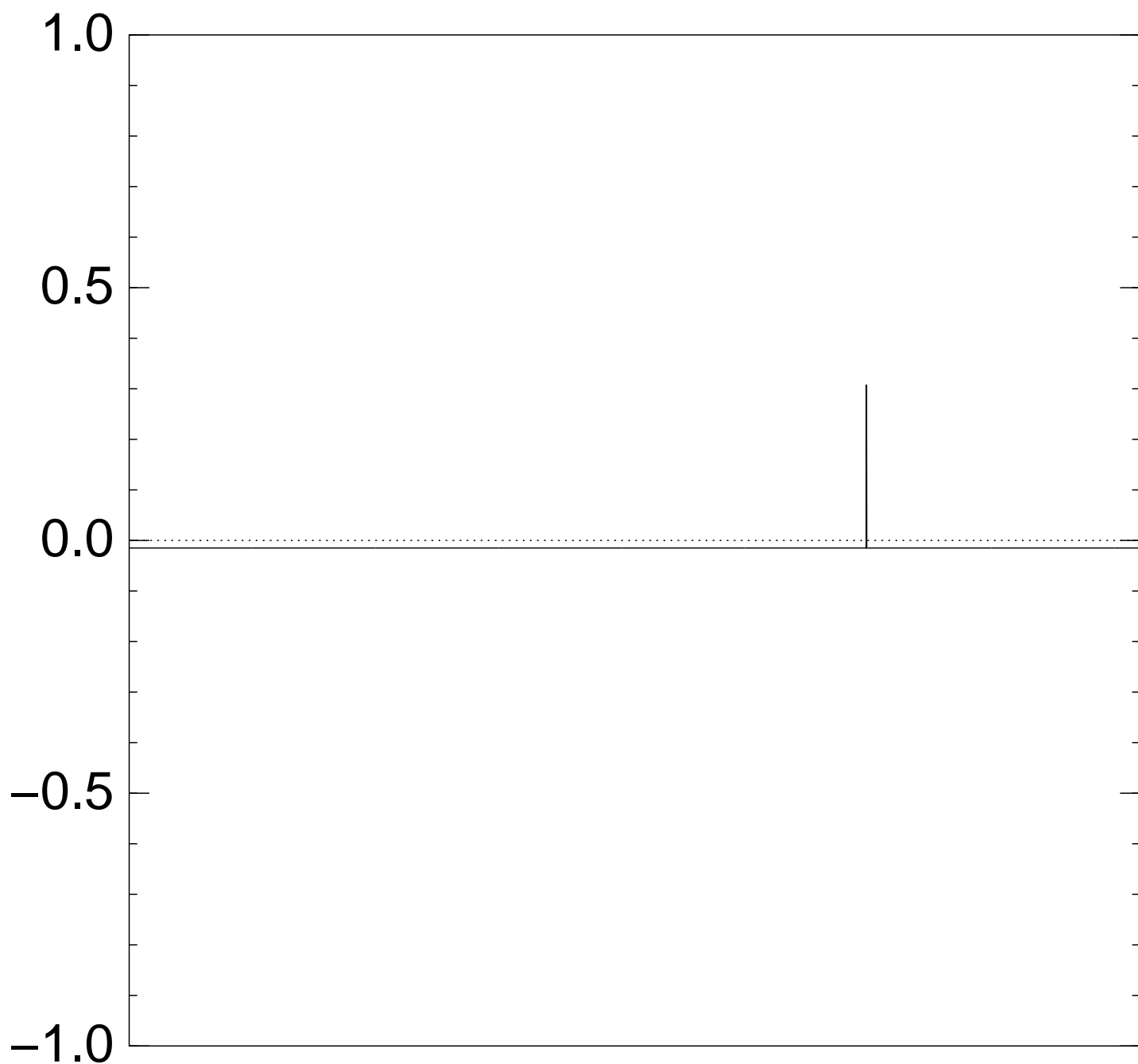
Normalized graph of  $u \mapsto a_u$   
for an example with  $n = 12$   
after  $70 \times$  (Step 1 + Step 2):



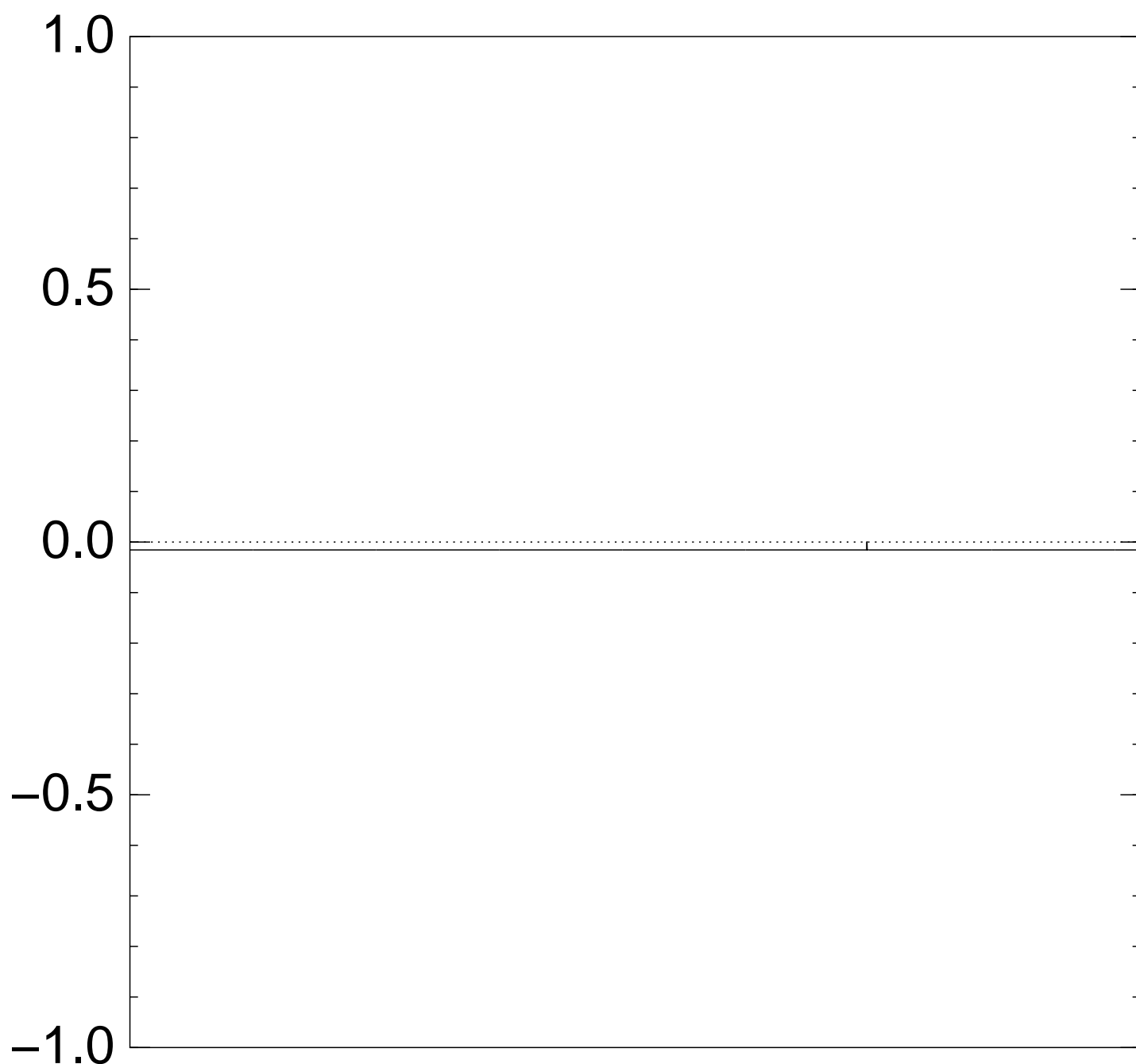
Normalized graph of  $u \mapsto a_u$   
for an example with  $n = 12$   
after  $80 \times$  (Step 1 + Step 2):



Normalized graph of  $u \mapsto a_u$   
for an example with  $n = 12$   
after  $90 \times$  (Step 1 + Step 2):



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



Very bad stopping point.

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

- (1)  $a_u$  for roots  $u$ ;
- (2)  $a_u$  for non-roots  $u$ .

Step 1 + Step 2

act linearly on this vector.

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.

## Many more quantum algorithms

2021: Your CPU consists of transistors performing bit ops.

Can think of any algorithm running on that CPU as a sequence of bit operations.

Can simulate these bit operations and output using NOT, CNOT, CCNOT, and measurement on a quantum computer.

So {non-quantum algorithms} can be viewed as a subset of {quantum algorithms}.

This subset includes the fastest algorithms known for many computations. Learn how to design non-quantum algorithms!

Assuming quantum computers: Fastest known quantum-physics simulators, fastest algorithms to factor “hard” integers, etc. are outside this subset. Learn how to design quantum algorithms!

Often techniques for designing non-quantum algorithms are combined with techniques specific to quantum algorithms.



2001 Shor survey regarding 1994 Shor and 1996 Grover: “These techniques for constructing faster algorithms for classical problems on quantum computers are the only two significant ones which have been discovered so far.”

2021: Shor’s algorithm and Grover’s algorithm continue to play critical roles. There are also several useful generalizations and further ideas adding to the landscape of quantum speedups.

## Some common Grover variants

What if  $f$  has many roots?

Can try same algorithm.

Analysis and optimization depend on  $R = \#\{\text{roots of } f\}$ .

Non-quantum search:  $\approx 2^n / R$  evaluations of  $f$ .

Quantum search:  $\approx (2^n / R)^{1/2}$  quantum evaluations of  $f$ .

Alternative approach, instead of redoing analysis and optimization: restrict  $f$  to a (pseudo)random input set; use unique-root Grover.

What if there are many “good” values of  $f$ , not just value 0?

Can modify algorithm: instead of negating when  $f(q) = 0$ , negate when  $g(f(q)) = 0$ .

Or simply apply original Grover to the composition  $q \mapsto g(f(q))$ .

What if one doesn't know  $R$ ?

Can modify algorithm. Or repeat original algorithm with sequence of guesses for  $R$ , starting with  $2^n$  and decreasing exponentially. Approximation of  $R$  suffices.

More interesting generalization:  
quantum walks. Seems more  
powerful than original Grover.

Say  $u \mapsto f(u)$  isn't very fast  
but have a very fast algorithm  
 $u, u', f(u) \mapsto f(u')$  for  $u'$  in a  
specified set of "neighbors" of  $u$ .  
Want to find "good"  $f(u)$ .

Non-quantum random walk:  
Start with one  $u$ ; compute  $f(u)$ .  
Replace  $u$  by random neighbor;  
repeat enough times for mixing;  
check if good; keep repeating.

Quantum walk: (repetitions)<sup>1/2</sup>.

Extreme example of walk:

Completely unrestricted neighbors.

Recompute  $f$  at each step.

Mixes instantly. Same as Grover.

More interesting example:

Ambainis distinctness algorithm.

Say  $f$  has  $2^n$  inputs,

exactly one collision  $\{p, q\}$ .

“Collision”:  $p \neq q; f(p) = f(q)$ .

Problem: find this collision.

Generic non-quantum algorithm:

nearly  $2^n$  calls to  $f$ .

Ambainis, using quantum walk:

$\approx 2^{2n/3}$  calls to  $f$ .

Sketch of Ambainis details:

For  $S \subseteq \{\text{inputs}\}$  with  $\#S = \sigma$ ,

define  $\varphi(S) = (\tau, T)$  where

$\tau = \#\{f(i) : i \in S\}$  and

$T$  is the *multiset* of  $f(i)$  for  $i \in S$ .

Define “good” to mean  $\tau < \sigma$ .

Chance of good:  $(\sigma/2^n)^2$ .

To walk from  $S$  to neighbor  $S'$ :

delete one elt, insert one elt.

Non-quantum setup cost  $\sigma$ ;

then inner · outer loops  $\sigma \cdot (2^n/\sigma)^2$ .

Quantum:  $\sigma$ ; then  $\sigma^{1/2} \cdot (2^n/\sigma)$ .

Take  $\sigma$  to minimize  $\sigma + 2^n/\sigma^{1/2}$ .

## Some common Shor variants

Simon used addition in  $(\mathbf{Z}/2)^n$ .

Shor used addition in  $\mathbf{Z}$  or  $\mathbf{Z}^2$   
for factorization or discrete logs.

Can use addition in  $\mathbf{Z}^n$ .

“Continuous” version:  $\mathbf{R}^n$ ,  
with careful precision handling.

In all of these algorithms,  
naturally find “random”  $s$   
satisfying  $f(u) = f(u + s)$ .

Watch out for hypotheses on  $f$   
and exact meaning of “random”.

What if the function  $f$  is defined on a more general group, not necessarily commutative?

Terminology:  $\{\text{periods of } f\}$  is also called the “stabilizer group” of  $f$  under the natural group action. In quantum algorithms, the “hidden-subgroup problem” (HSP) is to find this group.

Shor’s idea + extra work handles arbitrary finite groups with  $O(n)$  evaluations of  $f$ .

Massive caveat here: also need huge  $f$ -independent computation!



Kuperberg: For dihedral group, reduce the extra computation at some cost in  $f$  evaluations. Total cost is superpolynomial but subexponential:  $2^{O(\sqrt{n})}$  evaluations of  $f$  + overhead.

Shor already handles some easy subgroups of the dihedral group. For hard cases, Kuperberg solves the “hidden-*shift* problem”:  
find  $s$  in a commutative group given *two* functions  $f_0, f_1$  satisfying  $f_1(u) = f_0(u + s)$ .

## The impact on cryptography

2021.12: A Firefox connection to `https://google.com` is encrypted and authenticated by AES-128-GCM, using a key exchanged by the X25519 ECDH system, with the key exchange signed by an ECDSA-NIST-P-256 key, with the signing key certified by an RSA-2048 key, which in turn is certified by an RSA-4096 key, which is trusted by Firefox. SHA-256, SHA-384 also appear.

Shor's algorithm breaks RSA and ECC in polynomial time. Panic!

“But nobody has a big enough quantum computer yet!”

— Will large-scale attackers *tell us* that they've built a big enough quantum computer?

Also, leaks show that they're already recording ciphertexts that they'll try to decrypt later.

Also, upgrading everything to post-quantum cryptography won't happen instantaneously.

“Is the polynomial small enough to be a real threat?”

— 2019 [Gidney–Ekerå](#)

“How to factor 2048 bit RSA integers in 8 hours using 20 million noisy qubits” combines an improved version of Shor’s algorithm with “plausible physical assumptions for large-scale superconducting qubit platforms” .

Most of the cost is for error correction, using many imperfect qubits to simulate the perfect qubits inside Shor’s algorithm.

Same paper says 7 hours with 26 million qubits for a big discrete log in  $\mathbf{F}_p^*$  if  $p$  is a 2048-bit prime and  $(p - 1)/2$  is also prime.

([Other papers](#): lower costs for 256-bit elliptic-curve discrete log.)

Useful comparison: *non-quantum* modular exponentiation on an Intel CPU core is  $>2^{20} \times$  faster.

Reasonable estimates: quantum computer will cost  $2^{20} \times$  more; overall cost of qubit operation will be  $2^{40} \times$  cost of bit operation.

Some reasons  $2^{40}$  can improve:

- Lower-cost qubits.
- Less noise in qubits.
- Better qubit connectivity.
- Better error-correction methods.
- Better reversibility conversions.

Beyond modular exponentiation:  
each algorithm needs analysis of  
quantum/non-quantum cost ratio.

CCNOT costs  $>100\times$  CNOT;

reversibility conversions interact  
with high-level algorithm details;  
error-correction cost depends on  
size of computation; and so on.

## Further impact: Grover

Typical symmetric-crypto attack:

Guess the secret AES-128 key,

see if guess decrypts ciphertext

to `<HTML><HEAD><met...`

If not, try further guesses.

Non-quantum attack succeeds

in  $2^{127}$  guesses on average,

which sounds too expensive

for most people to worry about.

(Bitcoin:  $\approx 2^{92}$  hashes/year.)

Grover takes only  $2^{64}$  quantum

evaluations of AES. Panic!

Quantum AES evaluation:

$\approx 2^{15}$  qubit operations.

Similar cost to  $2^{55}$  bit operations.

Attack costs  $\approx 2^{119}$  bit operations.

Also, Grover speedup

comes from *serial* iterations.

$2^{64}$  nanoseconds = 585 years,

and 1ns iterations won't be easy.

To run  $2^{10} \times$  faster,

need  $2^{20}$  quantum computers:

$\approx 2^{129}$  bit operations.

To run  $2^{20} \times$  faster,

need  $2^{40}$  quantum computers:

$\approx 2^{139}$  bit operations.



Can still be lower cost than non-quantum AES attack under reasonable assumptions re quantum-computer progress, but much more expensive than Shor RSA-2048 attack.

Many commentators conclude that AES-128 is safe.

However, AES-128 exposes many protocols to **multi-target attacks** that are already feasible today. So use AES-256. (Or ChaCha20: bigger security margin, no timing attacks, no block-size attacks.)

Every Grover application runs into the same questions.

How many years is the user willing to wait for results?

How many *serial* iterations can be carried out in that time for the target function  $f$ ?

Does this outweigh ratio between qubit-op cost and bit-op cost?

For cryptographic risk management, should presume some Grover speedup depending on quantum-computer progress, but have to account for costs of quantum evaluation of  $f$ .

For many applications of Grover (and quantum-walk algorithms), claims of quantum speedups in the literature rely critically on underestimating the cost of  $f$ .

Example: Non-quantum algorithm finds SHA-256 collision in  $2^{128}$  evaluations. Quantum algorithm finds SHA-256 collision in  $2^{85}$  evaluations *plus*  $2^{85}$  random accesses to  $2^{85}$  memory locations. The literature does not state a physically plausible cost model where quantum algorithm wins.

## Post-quantum cryptography

What do cryptographers do against quantum computers?

2003: **Coined** the term “post-quantum cryptography” .

2006, 2008, 2010, 2011, 2013, 2014, 2016, 2017, 2018, 2019, 2020, 2021, . . . :

**PQCrypto** conferences.

2015: NSA issued statement.

2016: NIST announced

Post-Quantum Cryptography Standardization Project.

2017: NIST received and posted 69 complete submissions.

Almost all submissions have faster attacks known today *even without quantum computers.*

About half have been shown to not meet their security claims.

New attack advances are continuing to appear in 2021.

Much worse status than previous cryptographic competitions.

Cryptanalysts are overloaded.

Presumably many attacks haven't been found yet.

Major directions so far in  
*quantum* cryptanalysis of PQC:

1. “Small” Grover applications.

Main design strategy:

Try to find attack components that can be viewed as huge searches for “good” objects.

Some state-of-the-art attacks built this way: quantum AES key search; quantum preimages for SPHINCS+; [quantum ISD](#) for Classic McEliece; [quantum XL](#) for MQ systems; [quantum enumeration](#) for lattice systems.

2. “Big” Grover applications, quantum walks, etc.

Main design strategy: Try to find attack components that can be viewed as collision searches.

Some state-of-the-art attacks built this way, *assuming memory costs magically disappear*:

quantum collisions for SHA-256;

quantum claw-finding for SIKE;

quantum sieving (and another approach) for lattice systems;

quantum combinatorial attacks for lattice systems.

3. Kuperberg applications and optimizations. Interesting example: [attacking](#) CRS/CSIDH, isogeny-based systems for small non-interactive key exchange.

(This subexponential CRS attack prompted the development of SIKE. SIKE is smaller than CRS/CSIDH for *sufficiently large* security levels against known attacks, but cutoff is unclear. SIKE also opens up [new attack avenues](#) and doesn't provide non-interactive key exchange.)



## 4. Shor applications.

Interesting example: discrete logarithms in groups related to number fields, combined with **further techniques**, led to a **polynomial-time attack** breaking usual “cyclotomic  $h^+ = 1$ ” case of Gentry STOC 2009 “Fully homomorphic encryption using ideal lattices” and some newer lattice-based cryptosystems.

Latest developments:

see recent talk on  **$S$ -unit attacks** against Ideal-SVP.

5. New ideas for quantum attacks. Recent example:

“Quantum algorithms for variants of average-case lattice problems via filtering” .

6. Analyzing and optimizing costs of all of these algorithms in much more detail.

7. Changing cryptosystems to enable attacks: e.g. “Please use your secret key on a quantum computer to decrypt the following superposition of ciphertexts.”