# Black-Box Groups, Oracles and more

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Black-Box Groups

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Groups often arise as groups of symmetries in other areas. For example,

- symmetry groups of graphs
- symmetry groups of geometric structures
- crystallographic groups

Questions about a group G we would like to be able to answer using a computer algebra systems:

• |*G*|

- a composition series of G
- maximal subgroups of G
- normaliser of g for  $g \in G$
- $H \cap K$  for  $H, K \leq G$
- coset representatives for  $K \trianglelefteq G$
- automorphism group of G

- This summer school focusses on the geometric approach using Aschbacher's Theorem to design algorithms to answer the above questions.
- An alternative approach working with black box groups has been pursued with stunning results by Babai and collaborators since 1984 [3] culminating in [2].

G could be given as

- finitely presented group
- permutation group
- matrix group
- other descriptions

# **Finitely Presented groups**

*G* described by a finite presentation  $\{X \mid R\}$ .

- studied a lot
- e.g. polycyclic groups

Not the main focus of these lectures.

## **Permutation Groups**

Let  $S_n$  denote the group of all permutations on  $\Omega = \{1, ..., n\}$ . Usually  $G = \langle X \rangle$ , where  $X \subseteq S_n$ .

- studied extensively
- algorithms exist since 1950s
- very efficient

# Permutation Group Example

Let *G* be the symmetry group of the square, i.e. Let  $G = \langle (1,2,3,4), (1,2)(3,4) \rangle$ . Note: (1,2,3,4) \* (1,2)(3,4) = (2,4). More about permutation groups later.

## Matrix Groups

Matrix groups are the main focus of this summer school.

$$G = \langle X \rangle$$
, with  $X \subseteq GL(n, q)$ 

• practical algorithms designed over past 20 years

#### **Overview Articles**

For further reading please see the two very nice overview articles [4] and [5] by Eamonn O'Brien.

### Other representations

- factor groups
- homomorphic images of known groups
- kernels of homomorphisms

### summary

A group can come in many disguises. For example, as a

- permutation group
- matrix group
- finitely presented group
- a factor group of another group
- a homomorphic image of another group

### Black Box Groups

Black Box groups allow us to describe an arbitrary group to a computer without specifying anything more about the group.

Black Box groups were first introduced by Babai and Szemerédi in 1984 [3].

# Black Box Groups

*Q* a finite set, called *alphabet* and  $N \in \mathbb{N}$  and  $S \subseteq Q^N$ . Let

- = be an equivalence relation on S,
- $\ast$  a binary relation on S and
- -1 a unary relation on *S*.
- If the following 3 conditions hold
  - induces a binary operation \* on S/=
  - 2 G = (S/=, \*) is a group;
  - [-1] induces a unary operation  $^{-1}$  on S/[=] where, for  $s \in S$ ,  $[s]^{-1}$  is the inverse of [s] in G.

then (S/=,\*) is called a *Black-Box Group* and (S,=,\*,-1) a *Black-Box representation* of (S/=,\*).

#### Definition

Let  $Y \leq S$  and define  $X = \{[y] \mid y \in Y\}$ . If X is a generating set for G then (Y, =, \*, -1) is called a *generating tuple* for the black-box group G.

#### Example

Let  $Q = \{0, 1\}$  and N = 3. Put  $S = Q^N$  and define

•  $(i, j, k) \equiv (\ell, m, n)$  if and only if j = m and k = n.

• 
$$(i, j, k) [*](\ell, m, n) :=$$
  
 $(i * \ell, (j + m + k * n) \pmod{2}, (k + n) \pmod{2}$ 

• 
$$(i, j, k)$$
 :=  $(i, (j + k) \pmod{2}, k)$ 

What group is this?

#### Example

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• 
$$(i, j, k)$$
 :=  $(i, (j + k) \pmod{2}, k)$ 

The group is isomorphic to  $C_4$ .

# Complexity of algorithms for groups

We make some assumptions to simplify describing the complexity of algorithms for groups:

- all group operations cost the same, i.e.  $\mu$ .
- we have some bound in the size of the input

### Definition

The (worst case) complexity of an algorithm for a group is the maximum number of (group) operations required.

# Examples

group	size of input	Cost of *
Sn	n	<i>O</i> ( <i>n</i> )
GL(n,q) BB $G$	$n^2 \log(q)$	$O(n^3 \log(q))$
BB G	N	$\mu$
$\{0,1\}^N$		

# Example: Cost of Element Order

### Algorithm 1: ORDER(g)

Input:  $g \in G$ Output: o(g)t := 1;while  $g^t \neq 1$  do t := t+1;end return t;

#### Cost

The cost of Algorithm ORDER on input g is  $O(o(g)\mu)$ .

### This can be pretty bad ...

Black-Box Groups

# Order of elements in the symmetric group $S_n$

#### Landau 1909

$$\lim_{n\to\infty}\frac{\max_{g\in S_n}(o(g))}{n^{\sqrt{n}}}=1.$$

Hence computing the order of an element in  $S_n$  like this has worst case complexity  $O(n^{\sqrt{n}}\mu)$ .

# Complexity of BB algorithms

#### Definition

- Suppose the cost of =, \* and -1 is  $\mu$ .
- Count the number of calls to =, \* and -1.
- We write this cost as a function of the size of the input.
- Input: words in *Q* of length *N*.
- Size of the input:  $N \log(Q) \sim N$  (for fixed |Q|).

# Computing orders of elements in groups

- In permutation groups: cheap (cycle structure)
- In matrix groups: (Eamonn's lectures)
- In BB groups: expensive!

Let (S, [=], [\*], [-1]) be a black box representation of the group G = (S/[=],\*) and  $s \in S$ . To determine the order of [s], we need to find the smallest positive integer k with  $s^k := s | * | s | * | \cdots | * | s | = | [1_q].$ 

This requires o([s]) calls of [\*].

### Example

### $G = \mathbb{Z}_{2^N}$ , the cyclic group of order $2^N$ . An element can have order $2^N$ . The size of the input is *N*, thus

computing the order can be an exponential algorithm!

### **Order Tests**

### Does g have order dividing m?

Let  $k = \lfloor \log_2(m) \rfloor$ . Compute •  $B_m = b_0 + b_1 2 + ... + b_k 2^k$ , the binary representation of m. •  $g, g^2, ..., g^{2^k}$ •  $g^m = g^{b_0} \underbrace{*}(g^2)^{b_1} \underbrace{*} \cdots \underbrace{*}(g^{2^k})^{b_k}$ Cost:  $O(k\mu)$  to compute  $g, g^2, ..., g^{2^k}$  and  $O(k\mu)$  multiplications for  $g^m$ , thus  $O(\log(m)\mu)$ .

# **Deterministic Algorithms**

#### deterministic algorithm

- computes an output for all (allowable) inputs
- same input yields same output
- output correct

# **Randomised Algorithms**

#### randomised algorithm

Uses sequence random bits

- computes an output for most (allowable) inputs
- output depends on random bits and input
- output maybe incorrect

### Monte-Carlo Algorithm

Let  $0 < \varepsilon < 1$ . A randomised algorithm is a Monte-Carlo algorithm with error probability  $\varepsilon$  if the algorithm returns an output for an allowable input and the probability that the output for an allowable input is correct is at least  $1 - \varepsilon$ .

### Algorithm 2: HEARTCARD( $\varepsilon$ )

Input: Standard deck of 32 playing cards, real  $\varepsilon$ Output: a card

for i in [1...M] do
 pick a random card in the deck;
 if the card is a heart card then
 return the card;
 else

put card back into deck;

end

end

return a random card in the deck;

## HeartCard

The algorithm is a Monte-Carlo algorithm since it

- always returns an output
- Output depends on random bits
- The output is correct if a heart card is returned
- The output is incorrect if a card of suit other than heart is returned

# Analysis of example algorithm

- At any stage there are 32 cards in the deck.
- The probability that a random card is heart is 1/4.
- How large must *M* be such that the probability of an incorrect answer is at most ε?

# Analysis of example algorithm

Probability of not returning a heart in *M* random selections is

$$(1-\frac{1}{4})^M = \left(\frac{3}{4}\right)^M$$

Now

$$\left(\frac{3}{4}\right)^M \leq \varepsilon$$

if and only if

$$M \geq rac{\log(arepsilon^{-1})}{\log(4/3)} \sim 3.5 * \log(arepsilon^{-1}).$$

For example, if  $\varepsilon = 1/100$  then we need M > 16, while if  $\varepsilon = 1/10$  then we need M > 8.

# **Complexity of Algorithms**

- worst case time complexity: maximum number of basic operations as function of size of the input.
- worst case space complexity: maximum number of storage units required as function of size of the input.
- the number of random bits used is usually a parameter.

Complexity of example:

 $\log(\varepsilon^{-1})/\log(4/3)$  random selections and  $\log(\varepsilon^{-1})/\log(4/3)$  card checks.

# Complexity of algorithms

Let *N* be an upper bound on the size of the input. Depending on the behaviour of the function f(N) yielding the complexity, we classify algorithms as follows:

### Name and behaviour of f(N)

exponential	exponential in N
polynomial	polynomial in N
linear	constant multiple of N
nearly linear	constant multiple of $\log^{c}(N)N$

# Complexity of algorithms

We use the Big-O notation:

- *f*(*n*) ∈ *O*(*g*(*n*)) if there is a *C* such that *f*(*n*) ≤ *Cg*(*n*) for all sufficiently large *n*.
- $f(n) \in o(g(n))$  if  $\lim_{n\to\infty} \frac{f(n)}{g(n)} = 0$ .
- $f(n) \in \Theta(g(n))$  if there are  $C_1, C_2$  such that  $C_1g(n) \le f(n) \le C_2g(n)$  for all sufficiently large *n*.

# 1-Sided Monte-Carlo Algorithms

Given  $\varepsilon$  with  $0 < \varepsilon < 1$ .

- answer true or false questions
- answer true is provably correct
- answer false might be incorrect
- the probability that the answer is false and should have been true is less than  $\varepsilon$

### Example

### **Algorithm 3**: ISABELIANSUBGROUP( $G, H, \varepsilon$ )

**Input**: Black Box Group *G* with  $H \le G$ , real  $\varepsilon$  and a method to construct random elements in *H* **Output**: true or false

```
M :=?;

for i in [1... M] do

g := PSEUDORANDOM(H);

h := PSEUDORANDOM(H);

if [g, h] \neq 1 then

return false;

end

end

return true;
```

### Example

What is M?

- If *H* is abelian the algorithm returns true and the answer is correct.
- If *H* is non-abelian then the probability that two random elements *g* and *h* in *H* do not commute is at least 3/8 (Gustafson, 1973).
- Thus each repetition of the for-loop will fail to find a pair to witness that *H* is non-abelian with probability at most 5/8.
- The probability that in *M* repetitions we failed to find a non-commuting pair is  $(5/8)^M$ .
- If we require  $(5/8)^M < \varepsilon$ , we choose  $M \ge \frac{\log(\varepsilon^{-1})}{\log(8/5)} \ge 2.13 \log(\varepsilon^{-1})$ .

# 1-Sided Monte-Carlo Algorithms

**Algorithm 4**: ISABELIANSUBGROUP( $G, H, \varepsilon$ )

**Input**: Black Box Group *G* with  $H \le G$ , real  $\varepsilon$  and a method to construct random elements in *H* **Output**: true or false

```
M := \frac{\log(\varepsilon^{-1})}{\log(8/5)};
for i in [1 ... M] do
g := PSEUDORANDOM(H);
h := PSEUDORANDOM(H);
if [g, h] \neq 1 then
return false;
end
end
return true;
```

#### Complexity of Algorithms

# Las Vegas Algorithms

Introduced in 1979 by Lásló Babai.

#### Definition

A Las Vegas algorithm with error probability  $\varepsilon$  with  $0 < \varepsilon < 1$  is a randomised algorithm which either returns the correct answer or reports failure. The probability that it reports failure on an allowable input is at most  $\varepsilon$ .

#### Lessons

- randomised algorithms often draw conclusions from particular elements
- the probability of failing to find the element often features in complexity
- often need lower bounds for the probability of the elements
- the better the lower bound the smaller the complexity

#### Oracles

Sometimes the Black Box model is too restrictive.

- Some computations too hard, e.g. Discrete Log, see below
- Some computations more efficient, e.g. ElementOrder in S<sub>n</sub>

#### Example:

The order of (1,2)(3,4,5)(6,7,8,9,10) is just the lcm of the cycle lengths, i.e.  $2 \cdot 3 \cdot 5 = 30$ .

### historical solution

In these situations we consult the oracle.



#### Oracles

We add new black boxes to the description of a black box group and call them oracles. For example, an Order Oracle o could be used to compute the order of a group element.

# Complexity of BB Algorithms with Oracles

We treat the Oracle as an unknown quantity and give the complexity as a function of

- the size of the input
- the number of random elements required
- the number of calls to black boxes
- the number of calls to an oracle

### Discrete Logarithm

Let  $G = \langle a \rangle$  be a cyclic group with *n* elements.

**Discrete Log Problem** 

Given  $b \in G$ , find *t* such such that  $a^t = b$ .

*t* is called a discrete logarithm of *b* with respect to *a*. Note that two discrete logarithms of *b* with respect to *a* are congruent modulo |G|.

$$t = \log_a(b).$$

### Example

Let p = 1009.

#### Problem:

```
Find x with 11^x = 135 in GF(p).
```

#### How can we do this?

- $11^1 = 11$
- 11<sup>2</sup> = 121
- 11<sup>3</sup> = 322
- 11<sup>5</sup> = 620
- 11<sup>6</sup> = 766
- 11<sup>7</sup> = 354
- :
- $11^{1000} = 135$

### Example

- The size of the input is log(p) as every element in GF(p) can be stored in log(p) = N bits.
- Thus computing all powers of 11 modulo p cost  $p = \log(p)^{\log(p)/\log\log(p)} = N^{N/\log(N)}$  basic operations.
- Thus exponential in the size of the input.

### Discrete Logarithm Problem

#### **Open Problem**

Is there a polynomial time algorithm that computes the discrete logarithm of  $\log_a(b)$  for a, b in a cyclic group ?

#### **Existing Algorithms**

- Faster than naive algorithm, but still exponential.
- Often trading time for space, e.g. Baby-step Giant-step or Pollard's Rho Algorithm (both  $O(\sqrt{n})$ ), where *n* is the order of the cyclic group, i.e.  $n = N^{N/\log(N)}$ .

We usually treat the Discrete Log Problem as an Oracle.

Given a group *G* by a generating set *X*, we would like to be able to represent a given  $g \in G$  as a word in *X*. However, this word should not be too long.

Suppose we know for *h* in some group *G* that we can compute an element *g* that we need for some purpose as  $g = h^{32}$ .

Example

$$g=h^{2^5}=h*h*\cdots*h$$

requires 31 group operations.

We need a better way to record that  $g = h^{32}$ .

#### Example

A faster way for obtaining *g* from *h*:

$$[w_1 = h, w_2 = (w_1, w_1), w_3 = (w_2, w_2), w_4 = (w_3, w_3), w_5 = (w_4, w_4), w_6 = (w_5, w_5)].$$

Evaluate this list in *G*, by computing first  $w_1$ , then  $w_2$ , where (a, b) means multiply *a* and *b*. We find:

$$w_1 = h, w_2 = w_1 * w_1 = h^2, w_3 = w_2 * w_2 = h^4,$$
  
 $w_4 = w_3 * w_3 = h^8, w_5 = w_4 * w_4 = h^{16}, w_6 = w_5 * w_5 = h^{32} = g.$ 

This requires 5 multiplications.

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#### Example

Suppose h, b are generators for G and

$$g = h^4 * b^{-1} * h^8$$

lies in a particular subgroup. Then we record SLP for *g*:

$$[w_1 = h, w_2 = (w_1, w_1), w_3 = (w_2, w_2), w_4 = (w_3, w_3), w_5 = b, w_6 = (w_5, -1), w_7 = (w_3, w_6), w_8 = (w_7, w_4)].$$

The entries of the SLP are called cells. Cells contain pointers to previous cells and operations.

We see:  $w_1 = h$ ,  $w_2 = h^2$ ,  $w_3 = h^4$ ,  $w_4 = h^8$ ,  $w_5 = b$ ,  $w_6 = b^{-1}$ ,  $w_7 = h^4 * b^{-1}$  and  $w_8 = g$ .

Let G = S/= be a Black Box group given by a Black Box generating tuple (Y, =, \*, -1). Let  $g \in G$ .

#### Definition

A Straight-Line Program (SLP) for g is a list  $L = [w_1, \ldots, w_n]$ , for which each cell satisfies one of:

• 
$$w_i \in Y$$
,

• 
$$w_i = (w_j, -1)$$
 with  $j < i$ ,

• 
$$w_i = (w_j, w_k, [*])$$
 with  $k, j < i$ ,

such that the evaluation of  $w_n$  is g.

#### The evaluation of an *SLP* of length *m* requires *m* BB operations.

## **Better Straight-Line Programs**

We might allow more operations than \* and -1. For example, in a permutation group *G* the following operations can be computed quickly:

- $g^m$  for positive m
- $g^{-1} \cdot h$

#### Example

$$g = (1, 2, 3, 4, 5)(6, 7, 8, 9, 10, 11).$$
 Then  
 $g^4 = (1, 5, 4, 3, 2)(6, 10, 8)(7, 11, 9).$   
 $g^{-1} \cdot h$  can be computed as fast as  $g \cdot h$ .

# **Better Straight-Line Programs**

#### Definition

A Straight-Line Program (SLP) for *g* is a list  $L = [w_1, ..., w_n]$ , where every cell satisfies one of:

• 
$$w_i \in Y$$
,

• 
$$w_i = (w_j, -1)$$
 with  $j < i$ ,

• 
$$w_i = (w_j, w_k, *)$$
 with  $k, j < i$ ,

• 
$$w_i = (w_j, w_k, \frown^m)$$
 with  $k, j < i$ ,

• 
$$w_i = (w_j, w_k, -1)$$
 with  $k, j < i$ ,

such that the evaluation  $w_n$  is g.

The evaluation of an SLP of length m may now require more than m BB operations.

### **Better Straight-Line Programs**

A straight line program can encode several words, e.g. in our example

$$g = h^4 * b^{-1} * h^8$$
.

The SLP for g

$$[w_1 = h, w_2 = (w_1, w_1), w_3 = (w_2, w_2), w_4 = (w_3, w_3), w_5 = b, w_6 = (w_5, -1), w_7 = (w_3, w_6), w_8 = (w_7, w_4)].$$

with  $w_1 = h$ ,  $w_2 = h^2$ ,  $w_3 = h^4$ ,  $w_4 = h^8$ ,  $w_5 = b$ ,  $w_6 = b^{-1}$ ,  $w_7 = h^4 * b^{-1}$  and  $w_8 = g$ . It also encodes  $h^4$ ,  $h^4 \cdot b^{-1}$ , etc.

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# **Better Straight-Line Programs**

A straight line program can encode several words,

#### Encode several elements

By storing pointers to the elements in the SLP.

#### Example:

 $G = \langle a, b \rangle$  and  $M \leq G$  is generated by g, h which we found as SLP in a and b by random search. Most likely g and h contain many common subwords and share large chunks in their SLPs.

# **Evaluating Straight-Line Programs**

A straight line program can be evaluated linearly, e.g.

#### Example

Consider the SLP

$$[w_1 = h, w_2 = (w_1, w_1), w_3 = (w_2, w_2), w_4 = (w_3, w_3), w_5 = b, w_6 = (w_5, -1), w_7 = (w_3, w_6), w_8 = (w_7, w_2)].$$

with 
$$w_1 = h$$
,  $w_2 = h^2$ ,  $w_3 = h^4$ ,  $w_4 = h^8$ ,  $w_5 = b$ ,  $w_6 = b^{-1}$ ,  $w_7 = h^4 * b^{-1}$  and  $w_8 = h^4 * b^{-1} * h^4$ .

This way we compute and store group elements  $w_1, \ldots, w_8$ .

#### However...

the element  $w_4$  was never used.

An SLP can be evaluated recursively, storing a counter how often a cell is visited.

Example: evaluate w<sub>8</sub>

 $[w_1 = h(0), w_2 = (w_1, w_1)(0), w_3 = (w_2, w_2)(0), w_4 = (w_3, w_3)(0)$  $w_5 = b(0), w_6 = (w_5, -1)(0), w_7 = (w_3, w_6)(0), w_8 = (w_7, w_2)(1)].$ 

- $w_8 = w_7 * w_2$ , so  $w_7$  and  $w_2$  visited and evaluated
- $w_2 = w_1 * w_1$  so  $w_1$  visited
- $w_7 = w_3 * w_6$ , so  $w_6$  and  $w_3$  visited and evaluated
- $w_6 = w_5^{-1}$ , so  $w_5$  visited
- $w_3 = w_2 * w_2$ , so  $w_2$  visited again, already evaluated.
- $w_4$  not visited, so not evaluated.

An SLP can be evaluated recursively, storing a counter how often a cell is visited.

Example: evaluate w<sub>8</sub>

 $[w_1 = h(0), w_2 = (w_1, w_1)(1), w_3 = (w_2, w_2)(0), w_4 = (w_3, w_3)(0)$  $w_5 = b(0), w_6 = (w_5, -1)(0), w_7 = (w_3, w_6)(1), w_8 = (w_7, w_2)(1)].$ 

- $w_8 = w_7 * w_2$ , so  $w_7$  and  $w_2$  visited and evaluated
- $w_2 = w_1 * w_1$  so  $w_1$  visited
- $w_7 = w_3 * w_6$ , so  $w_6$  and  $w_3$  visited and evaluated
- $w_6 = w_5^{-1}$ , so  $w_5$  visited
- $w_3 = w_2 * w_2$ , so  $w_2$  visited again, already evaluated.
- $w_4$  not visited, so not evaluated.

An SLP can be evaluated recursively, storing a counter how often a cell is visited.

Example: evaluate w<sub>8</sub>

 $[w_1 = h(2), w_2 = (w_1, w_1)(1), w_3 = (w_2, w_2)(0), w_4 = (w_3, w_3)(0)$  $w_5 = b(0), w_6 = (w_5, -1)(0), w_7 = (w_3, w_6)(1), w_8 = (w_7, w_2)(1)].$ 

- $w_8 = w_7 * w_2$ , so  $w_7$  and  $w_2$  visited and evaluated
- $w_2 = w_1 * w_1$  so  $w_1$  visited
- $w_7 = w_3 * w_6$ , so  $w_6$  and  $w_3$  visited and evaluated
- $w_6 = w_5^{-1}$ , so  $w_5$  visited
- $w_3 = w_2 * w_2$ , so  $w_2$  visited again, already evaluated.
- $w_4$  not visited, so not evaluated.

An SLP can be evaluated recursively, storing a counter how often a cell is visited.

Example: evaluate w<sub>8</sub>

 $[w_1 = h(2), w_2 = (w_1, w_1)(1), w_3 = (w_2, w_2)(1), w_4 = (w_3, w_3)(0)$  $w_5 = b(0), w_6 = (w_5, -1)(1), w_7 = (w_3, w_6)(1), w_8 = (w_7, w_2)(1)].$ 

- $w_8 = w_7 * w_2$ , so  $w_7$  and  $w_2$  visited and evaluated
- $w_2 = w_1 * w_1$  so  $w_1$  visited
- $w_7 = w_3 * w_6$ , so  $w_6$  and  $w_3$  visited and evaluated
- $w_6 = w_5^{-1}$ , so  $w_5$  visited
- $w_3 = w_2 * w_2$ , so  $w_2$  visited again, already evaluated.
- $w_4$  not visited, so not evaluated.

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Example: evaluate w<sub>8</sub>

 $[w_1 = h(2), w_2 = (w_1, w_1)(1), w_3 = (w_2, w_2)(1), w_4 = (w_3, w_3)(0)$  $w_5 = b(1), w_6 = (w_5, -1)(1), w_7 = (w_3, w_6)(1), w_8 = (w_7, w_2)(1)].$ 

- $w_8 = w_7 * w_2$ , so  $w_7$  and  $w_2$  visited and evaluated
- $w_2 = w_1 * w_1$  so  $w_1$  visited
- $w_7 = w_3 * w_6$ , so  $w_6$  and  $w_3$  visited and evaluated
- $w_6 = w_5^{-1}$ , so  $w_5$  visited

•  $w_3 = w_2 * w_2$ , so  $w_2$  visited again, already evaluated.

•  $w_4$  not visited, so not evaluated.

An SLP can be evaluated recursively, storing a counter how often a cell is visited.

Example: evaluate w<sub>8</sub>

 $[w_1 = h(2), w_2 = (w_1, w_1)(3), w_3 = (w_2, w_2)(1), w_4 = (w_3, w_3)(0)$  $w_5 = b(1), w_6 = (w_5, -1)(1), w_7 = (w_3, w_6)(1), w_8 = (w_7, w_2)(1)].$ 

- $w_8 = w_7 * w_2$ , so  $w_7$  and  $w_2$  visited and evaluated
- $w_2 = w_1 * w_1$  so  $w_1$  visited
- $w_7 = w_3 * w_6$ , so  $w_6$  and  $w_3$  visited and evaluated
- $w_6 = w_5^{-1}$ , so  $w_5$  visited
- $w_3 = w_2 * w_2$ , so  $w_2$  visited again, already evaluated.

• *w*<sub>4</sub> not visited, so not evaluated.

An SLP can be evaluated recursively, storing a counter how often a cell is visited.

Example: evaluate w<sub>8</sub>

 $[w_1 = h(2), w_2 = (w_1, w_1)(3), w_3 = (w_2, w_2)(1), w_4 = (w_3, w_3)(0)$  $w_5 = b(1), w_6 = (w_5, -1)(1), w_7 = (w_3, w_6)(1), w_8 = (w_7, w_2)(1)].$ 

- $w_8 = w_7 * w_2$ , so  $w_7$  and  $w_2$  visited and evaluated
- $w_2 = w_1 * w_1$  so  $w_1$  visited
- $w_7 = w_3 * w_6$ , so  $w_6$  and  $w_3$  visited and evaluated
- $w_6 = w_5^{-1}$ , so  $w_5$  visited
- $w_3 = w_2 * w_2$ , so  $w_2$  visited again, already evaluated.
- w<sub>4</sub> not visited, so not evaluated!

An SLP can be evaluated recursively, storing a counter how often a cell is visited.

- Cells with counter (0) can be deleted.
- During evaluation, store only group elements for visited cells.

More details in Bäärnhielm and Leedham-Green [1].

# For Further Reading I



Henrik Bäärnhielm and Charles Leedham-Green The product replacement prospector, preprint.



Lásló Babai, Robert Beals and Ákos Seress Polynomial-time Theory of Matrix Groups STOC'09, 2009.

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On the complexity of matrix group problems I FOCS'84, 229–240.

# For Further Reading II

#### Eamonn A. O'Brien

Towards effective algorithms for linear groups Finite Geometries, Groups and Computation, (Colorado), September 2004, 163–190, 2006.

Eamonn A. O'Brien Algorithms for matrix groups Groups St Andrews 2009 in Bath, LMS Lecture Notes 388, 297–323, 2011.