

*twenty-five comparators is optimal  
when sorting nine inputs  
(and twenty-nine for ten)*

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november 10th, 2014

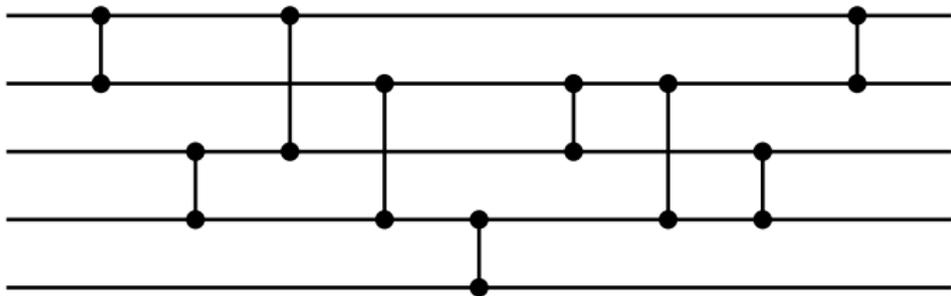
# *outline*

*sorting  
networks in a  
nutshell*

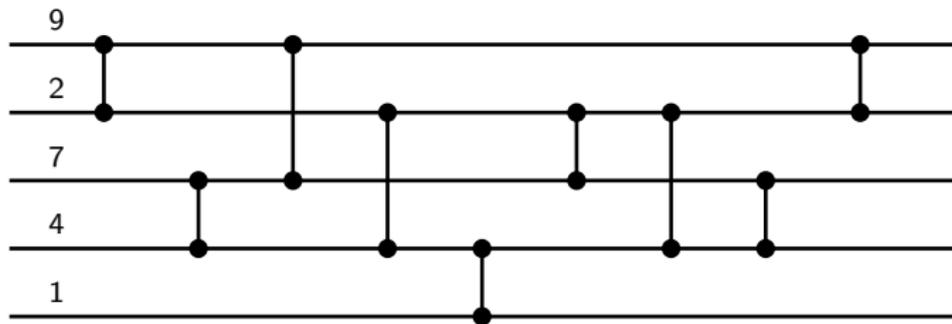
*encoding the  
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*conclusions &  
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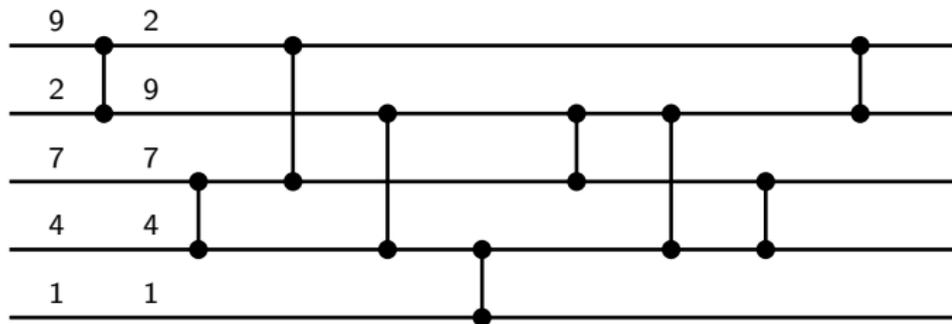
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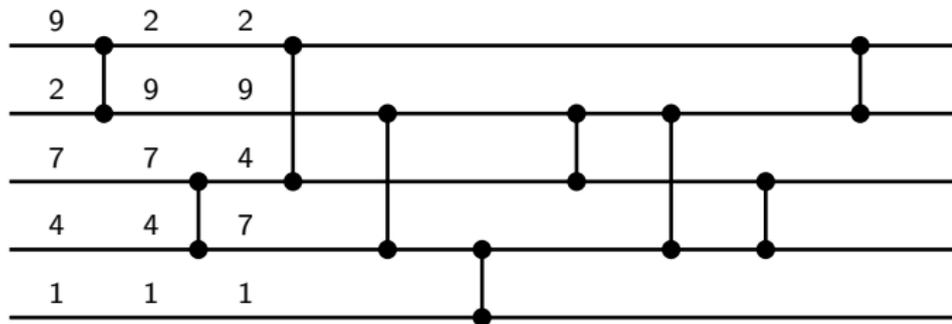
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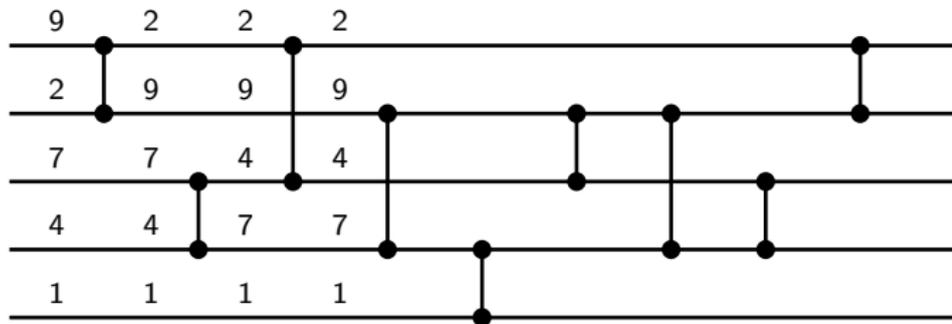
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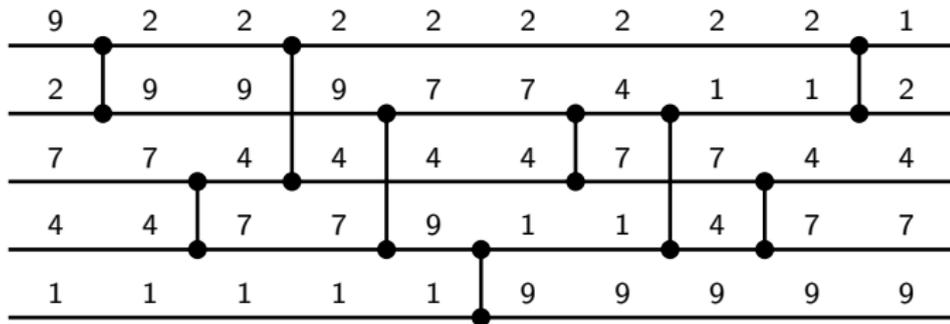
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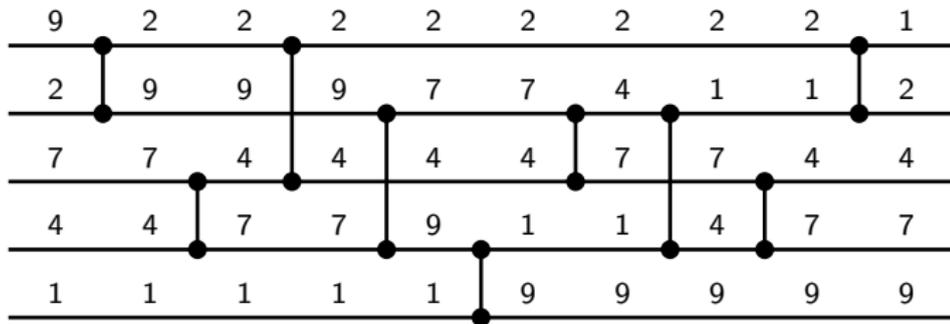
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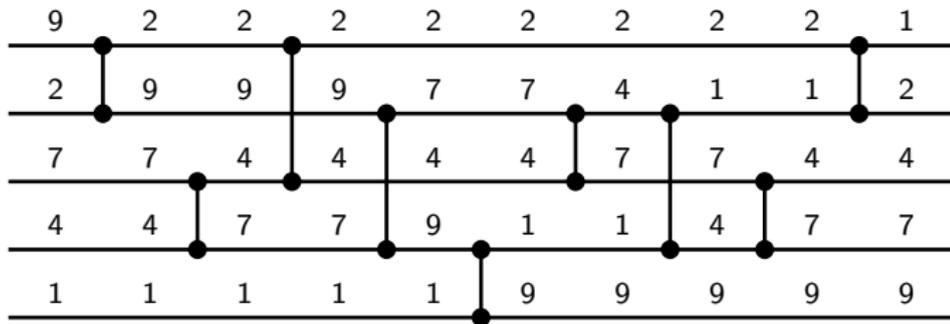


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*size* this net has 5 *channels* and 9 *comparators*

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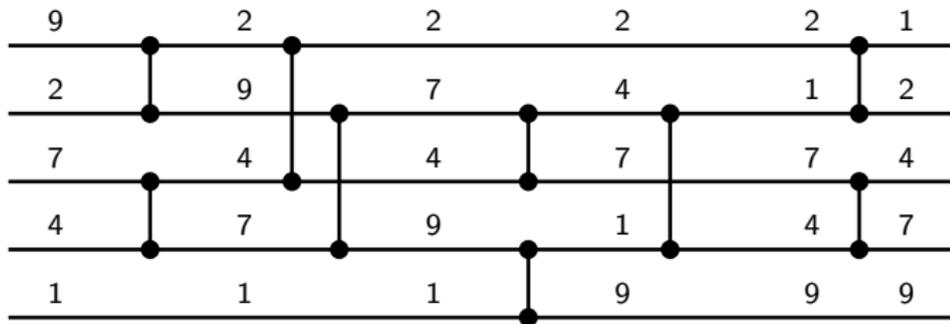


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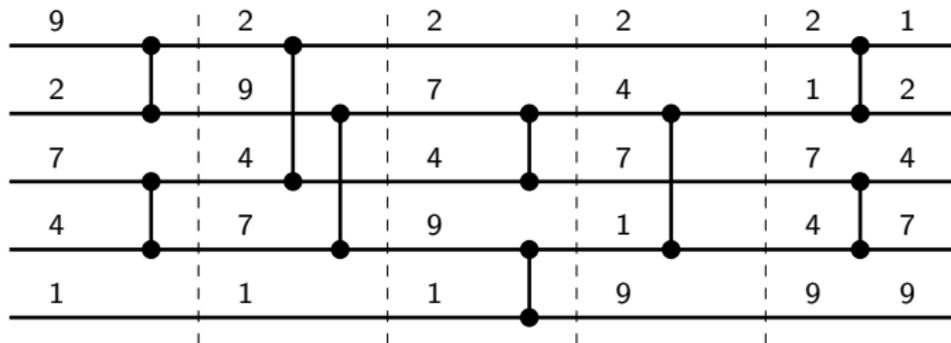


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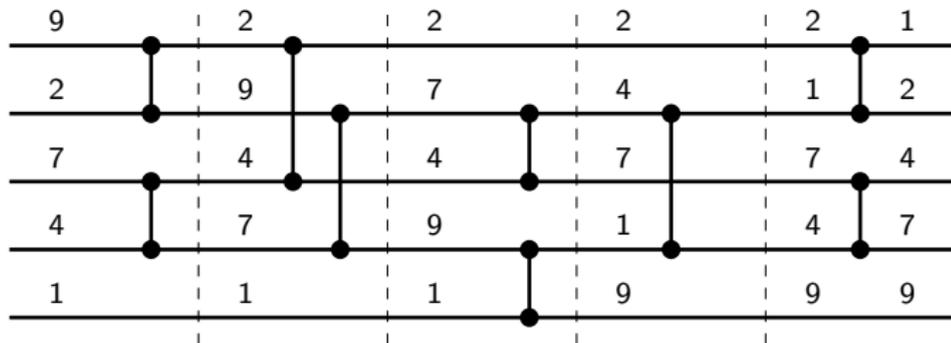


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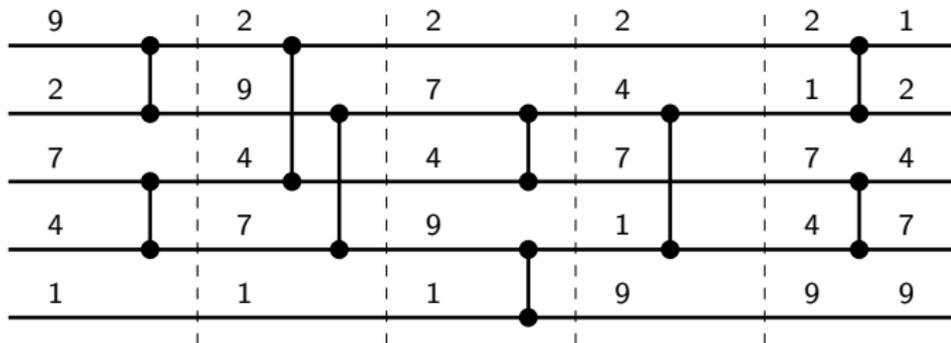
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*depth* this net has 5 *layers*

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*more info* see d.e. knuth, *the art of computer programming*, vol. 3

## *the optimization problems*

*the optimal size  
problem*

what is the minimal number of *comparators* on a sorting network on  $n$  channels ( $s_n$ )?

*the optimal  
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what is the minimal number of *layers* on a sorting network on  $n$  channels ( $t_n$ )?

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knuth 1973

$n$	1	2	3	4	5	6	7	8	9	10
$s_n$	0	1	3	5	9	12	16	19	25	29
$t_n$	0	1	3	3	5	5	6	6	7	7

$n$	11	12	13	14	15	16
$s_n$	35	39	45	51	56	60
$t_n$	8	8	9	9	9	9

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*parberry 1991*

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$s_n$	0	1	3	5	9	12	16	19	25	29
									23	27
$t_n$	0	1	3	3	5	5	6	6	<b>7</b>	<b>7</b>

$n$	11	12	13	14	15	16
$s_n$	35	39	45	51	56	60
	31	35	39	43	47	51
$t_n$	8	8	9	9	9	9
	<b>7</b>	<b>7</b>	<b>7</b>	<b>7</b>	<b>7</b>	<b>7</b>

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bundala &  
závodný 2013

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*our  
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## *an exponential explosion*

*parberry 1991*

- exploration of symmetries  $\rightsquigarrow$  fixed first layer
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- intensive sat-solving

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techniques not directly applicable to the size problem

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*9 channels  
(depth)*

2620 possibilities for each layer  
 $\rightsquigarrow 2620^6 \approx 3.2 \times 10^{20}$  6-layer networks

*9 channels  
(size)*

36 possibilities for each comparator  
 $\rightsquigarrow 36^{24} \approx 2.2 \times 10^{37}$  24-comparator networks

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## comparator networks

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network*

a *comparator network*  $C$  on  $n$  channels is a sequence of *comparators*  $(i, j)$  with  $1 \leq i < j \leq n$

*output*

$C(\vec{x})$  denotes the *output* of  $C$  on  $\vec{x} = x_1 \dots x_n$

*binary outputs*

the set of binary outputs of  $C$  is

$$\text{outputs}(C) = \{C(\vec{x}) \mid x \in \{0, 1\}^n\}$$

*sorting network*

a comparator network  $C$  is a *sorting network* if  $C(\vec{x})$  is sorted for every input  $\vec{x}$

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“ $C$  is a sorting network on  $n$  channels” is co-NP (complete)

## *sat encoding*

$\text{Network} = \langle c(I_1, J_1), \dots, c(I_k, J_k) \rangle$

$$\text{valid}_{n,k}(\text{Network}) = \bigwedge_{i=1}^k \text{new\_int}(I_i, 1, n) \wedge \text{new\_int}(J_i, 1, n) \wedge \text{int\_lt}(I_i, J_i)$$

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$$\psi_{I,J}(\vec{x}, \vec{y}) = \bigwedge_{i=1}^n \neg \text{int\_eq}(I, i) \wedge \neg \text{int\_eq}(J, i) \rightarrow (y_i \leftrightarrow x_i)$$

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$$\text{sorts}_{n,k}(\text{Network}, \vec{b}) = \bigwedge_{i=1}^k \varphi_{I_i, J_i}(\vec{x}_{i-1}, x_i) \wedge \psi_{I_i, J_i}(\vec{x}_{i-1}, \vec{x}_i)$$
$$\vec{x}_0 = \vec{b}, \quad \vec{x}_k = \text{sort}(\vec{b})$$

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this is compiled with the bee constraint compiler into a cnf formula  $\Psi(n, k)$

*theorem*

$\Psi(n, k)$  is satisfiable iff there is a sorting network on  $n$  channels with  $k$  comparators

## *practical evaluation*

optimal sorting networks (sat)					
$n$	$k$	bee	#clauses	#vars	sat
4	5	0.18	1916	486	0.01
5	9	1.03	10159	2550	0.03
6	12	4.55	35035	8433	2.45
7	16	21.68	114579	26803	16.70
8	19	82.93	321445	73331	$\infty$
9	25	452.55	977559	219950	$\infty$

smaller networks (unsat)					
$n$	$k$	bee	#clauses	#vars	sat
4	4	0.15	1480	356	0.01
5	8	0.90	8963	2221	1.27
6	11	3.99	32007	7657	242.02
7	15	19.04	107227	25000	$\infty$
8	18	73.34	304145	69221	$\infty$
9	24	406.67	937773	210715	$\infty$

(times in seconds, timeout = 1 week)

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*main idea*

divide the “big” sat problem into smaller problems

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“minimal” set  $\mathcal{F}$  of choices for  $I_1, J_1, \dots, I_\ell, J_\ell$  such that

$$\Psi_{n,k} \text{ is satisfiable iff } \bigvee_{f \in \mathcal{F}} \Psi_{n,k,f} \text{ is satisfiable}$$

where  $f = \langle I_1, J_1, \dots, I_\ell, J_\ell \rangle$

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*reduce* the size of  $\mathcal{F}$  using symmetry-breaking techniques

## *breaking symmetry i/ii*

*output lemma*  
*(parberry 1991)*

- $C$  and  $C'$  are comparator networks
  - $\text{outputs}(C) \subseteq \text{outputs}(C')$
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*permuted  
output lemma  
(bundala & Závodný 2013)*

- $C$  and  $C'$  comparator networks of depth 2
- $\pi(\text{outputs}(C)) \subseteq \text{outputs}(C')$  for some permutation  $\pi$
- $C'$  can be extended to a sorting network

then  $C$  can also be extended to a sorting network of depth 2

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## breaking symmetry ii/ii

### permuted output lemma (generalized)

- $C$  and  $C'$  comparator networks **of equal size**
- $\pi(\text{outputs}(C)) \subseteq \text{outputs}(C')$  for some permutation  $\pi$
- $C'$  can be extended to a sorting network

then  $C$  can also be extended to a sorting network of **the same size**

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### *subsumption*

$C \preceq C'$  when

$$\pi(\text{outputs}(C)) \subseteq \text{outputs}(C')$$

for some permutation  $\pi$

## *the generate-and-prune approach*

*init* set  $R_0^n = \{\emptyset\}$  and  $k = 0$

*repeat* until  $k > 1$  and  $|R_k^n| = 1$

*generate* construct  $N_{k+1}^n$  by extending each net in  $R_k^n$  by one comparator in all possible ways

*prune* construct  $R_{k+1}^n$  from  $N_{k+1}^n$  by keeping only one element of each minimal equivalence class w.r.t. the transitive closure of  $\preceq$

*step* increase  $k$

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*termination condition* if  $C$  is a sorting network on  $n$  channels of size  $k$ , then  $|R_k^n| = 1$

## *optimizations*

- only generate networks when the extra comparator does something
- prove and implement criteria for when subsumption will fail
- restrict the search space of possible permutations
- optimize data structures
- parallelize to 288 nodes

## *some numerology*

$n$	$s_n$	largest $ N_k^n $	largest $ R_k^n $	execution time
3	3	2	2	$\sim 0$
4	5	12	4	$\sim 0$
5	9	65	11	$\sim 0$
6	12	380	53	2 sec
7	16	7,438	678	2 min
8	19	253,243	16,095	6 hours
9	25	18,420,674	914,444	16 years

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parallel runtime for  $n = 9$ : 3 weeks

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4	5	12	4	$\sim 0$
5	9	65	11	$\sim 0$
6	12	380	53	2 sec
7	16	7,438	678	2 min
8	19	253,243	16,095	6 hours
9	25	18,420,674	914,444	16 years

parallel runtime for  $n = 9$ : 3 weeks

*the hard part*

going “over the peak” consumes most execution time



*collaboration is the key*

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*find* “minimal” set  $\mathcal{F}$  of choices for  $I_1, J_1, \dots, I_\ell, J_\ell$  such that

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different values of  $k$  give different total running times

# *outline*

*sorting  
networks in a  
nutshell*

*encoding the  
size problem in  
sat*

*conclusions &  
future work*

## *results & future work*

- exact values of  $s_9$  and  $s_{10}$
- technique may be adapted to settle higher values which are still unknown
- algorithms may be useful for *finding* smaller-than-currently-known networks
- further theoretical results may help proving optimality of best known upper bounds

thank you!