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# Deep Cooperation of CDCL and Local Search for SAT

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# Outline

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- Introduction
- Main Ideas
  - A Novel Framework of hybrid solvers
  - Phase Resetting with Local Search Assignments
  - Branching with Conflict Frequency in Local Search
- Experiments

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# Introduction - SAT

SAT: Given a propositional formula  $\varphi$ , test whether there is an assignment to the variables that makes  $\varphi$  true.

e.g., a CNF formula

$$\varphi = (x_1 \vee \neg x_2) \wedge (x_2 \vee x_3) \wedge (x_2 \vee \neg x_4) \wedge (\neg x_1 \vee \neg x_3 \vee x_4)$$

Hard

- The first problem that is proved to be NP-Complete [S. Cook, 1971]
- ETH says 3-SAT cannot be solved in  $2^{o(n)}$  time, and SETH says  $k$ -SAT needs roughly  $2^n$  time for large  $k$ .



Important Applications

- EDA
- Software verification
- Automatic Theorem Proving
- cryptography
- ...



Improve the efficiency of SAT Solving

# Introduction - CDCL

- The most popular approach: CDCL, since 1996 (evolved from DPLL)

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**Algorithm 1:** Typical CDCL algorithm:  $\text{CDCL}(F, \alpha)$

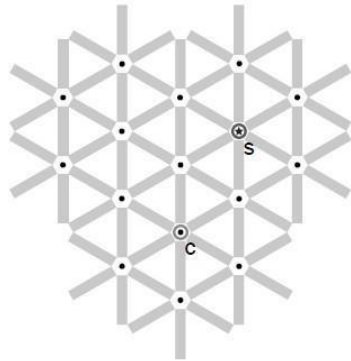
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```
1  $dl \leftarrow 0;$            //decision level
2 if  $\text{UnitPropagation}(F, \alpha) == \text{CONFLICT}$  then return UNSAT
3 while  $\exists$  unassigned variables do
    |   /* PickBranchVar picks a variable to assign and
    |     picks the respective value                               */
4    $(x, v) \leftarrow \text{PickBranchVar}(F, \alpha);$ 
5    $dl \leftarrow dl + 1;$ 
6    $\alpha \leftarrow \alpha \cup \{(x, v)\};$ 
7   if  $\text{UnitPropagation}(F, \alpha) == \text{CONFLICT}$  then
8     |    $bl \leftarrow \text{ConflictAnalysis}(F, \alpha);$ 
9     |   if  $bl < 0$  then
10    |   |   return UNSAT;
11    |   |   else
12    |   |   |    $\text{BackTrack}(F, \alpha, bl);$ 
13    |   |   |    $dl \leftarrow bl;$ 
14 return SAT;
```

- clause learning
- Lazy data structures
- Restarting
- branching heuristics
  - Pick a variable
  - Pick the respective phase
- ...

# Introduction - SLS

- The other important paradigm: stochastic local search (SLS), since 1992
  - a main incomplete method biased towards the satisfiable side.
  - Begin with a complete assignment and iteratively modify the assignment



# Introduction – Challenge on hybrid solving

## Ten Challenges in Propositional Reasoning and Search

Bart Selman, Henry Kautz, and David McAllester

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[http://www. research, att.com/~selman/challenge](http://www.research.att.com/~selman/challenge)

**Challenge 7:** Demonstrate the successful combination of stochastic search and systematic search techniques, by the creation of a new algorithm that outperforms the best previous examples of both approaches.

---**AAAI 1997**, Bart Selman, Henry Kautz and David McAllester

# Introduction – Related works

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- Use a local search solver as the main body solver.
  - hybridGM, SATHYS
  - GapSAT: use CDCL as preprocessor before local search
- DPLL/CDCL as the main body solver
  - HINOTOS: local search finds subformulas for CDCL to solve
  - WalkSatz: calls WalkSAT at each node of a DPLL solver Satz.
  - CaDiCaL and Kissat: a local search solver is called when the solver resets the saved phases and is used only once immediately after the local search process
- Sequential call local search and CDCL
  - Sparrow2Riss, CCAnr+glucose, SGSeq



# Introduction – Related works

- Use a local search solver as the main body solver.
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  - GapSAT: use CDCL as preprocessor before local search
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Previous works did not lead to hybrid solvers essentially better than CDCL solvers on application instances.

the local search process

- Sequential call local search and CDCL
  - Sparrow2Riss, CCA<sub>nr</sub>+glucose, SGSeq

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## Idea 2: Phase Resetting with Local Search Assignments

- Phase selection is an important component of a CDCL solver.
- Most modern CDCL solvers utilize the phase saving heuristic (Pipatsrisawat & Darwiche, SAT 2007).
- Our idea:
  - After each time the CDCL is restarted, resets the saved phases of all variables with assignments produced by local search.

**Table 1.** Probability of different phases in our phase resetting mechanism

Phase Name	$\alpha_{max\_LS}[x]$	$\alpha_{latest\_LS}[x]$	$\alpha_{best\_LS}[x]$	no change
Probability	20%	65%	5%	10%

- $\alpha_{max\_LS}$  and  $\alpha_{best\_LS}$  serve for the aim to maximize the depth of the branch
- $\alpha_{latest\_LS}$  adds diversification

## Idea 3: Branching with Conflict Frequency in Local Search

- CDCL is a powerful framework owing largely to the utilization of the conflict information
- branching heuristics aim to promote conflicts.
- Can information from SLS be used to enhance branching heuristics to promote conflicts?

### Our idea:

$ls\_confl\_freq(x) = \#(\text{steps in which } x \text{ appears in unsatisfied clauses}) / \#total\_local\_search\_steps$

multiply  $ls\_confl\_freq(x)$  with 100, resulting  $ls\_confl\_num(x)$ .

**LS Enhanced VSIDS:** for each variable  $x$ , its activity is increased by  $ls\_confl\_num(x)$

**LS Enhanced LRB:** for each variable  $x$ , the number of learnt clause during its period  $I$  is creased by  $ls\_confl\_num(x)$ .

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# Experiments - Preliminaries

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- Base solver:
  - glucose (v4.2.1)
  - MapleLCMDistChronoBT-DL (v2.1)
  - Kissat\_sat (2414b6d)
  - CCAr
- Benchmarks:
  - The latest four SAT Competitions/Race (2017-2020)
  - US Federal Communication Commission(FCC) 10000 instances

# Experiments – Results on SC benchmarks

solver	#SAT	#UNSAT	#Solved	PAR2	#SAT	#UNSAT	#Solved	PAR2
	SC2017(351)				SC2018(400)			
glucose_4.2.1	83	101	184	5220.0	95	95	190	5745.9
glucose+rx	88	95	183	5237.1 20↑	113	95	208	5283.4 51↑
glucose+rx+rp	112	94	206	4618.2	141	87	228	4698.3
glucose+rx+rp+cf	110	94	204	4668.5	150	91	241	4438.2
Maple-DL-v2.1	101	113	214	4531.0	133	102	235	4533.9
Maple-DL+rx	101	112	213	4591.3 9↑	149	101	250	4148.4 24↑
Maple-DL+rx+rp	111	103	214	4447.1	158	93	251	4147.1
Maple-DL+rx+rp+cf	116	107	223	4139.4	162	97	259	3927.6
Kissat_sat	115	114	229	3943.5	167	98	265	3786.4 17↑
Kissat_sat+cf	113	113	226	4001.0 3↓	178	104	282	3409.4
CCAnr	13	N/A	13	9629.9	56	N/A	56	8622.0
	SC2019(400)				SC2020(400)			
glucose_4.2.1	118	86	204	5437.6	68	91	159	6494.6
glucose+rx	120	84	204	5443.1 21↑	93	88	181	6018.3 62↑
glucose+rx+rp	134	85	219	5096.3	130	85	215	5125.7
glucose+rx+rp+cf	140	85	225	4923.6	134	87	221	4977.9
Maple-DL-v2.1	143	97	240	4601.8	86	104	190	5835.7
Maple-DL+rx	146	93	239	4602.4 9↑	121	105	226	4977.8 67↑
Maple-DL+rx+rp	155	94	249	4416.3	142	99	241	4589.2
Maple-DL+rx+rp+cf	154	95	249	4377.4	151	106	257	4171.1
Kissat_sat	159	88	247	4293.4 5↑	146	114	260	4048.8 10↑
Kissat_sat+cf	162	90	252	4211.7	157	113	270	3896.8
CCAnr	13	N/A	13	9678.3	45	N/A	45	8978.7



solver	Analysis for SAT				Analysis for UNSAT	
	#by LS	#SAT_bonus	#LS_call	LS_time(%)	#LS_call	LS_time(%)
SC2017(351)						
glucose+rx	20	11	24.28	21.66	16.36	5.52
glucose+rx+rp	10	33	17.77	18.46	14.33	4.86
glucose+rx+rp+cf	17	29	22.7	22.19	15.3	5.81
Maple+rx	16	9	13.86	7.52	11.18	2.03
Maple+rx+rp	11	15	9.63	10.43	6.54	2.36
Maple+rx+rp+cf	6	16	12.59	7.49	8.59	2.12
SC2018(400)						
glucose+rx	50	4	11.27	20.66	29.62	4.94
glucose+rx+rp	47	31	9.46	18.4	21.66	5.64
glucose+rx+rp+cf	53	36	11.43	20.28	20.62	6.64
Maple+rx	52	7	4.8	13.02	11.69	2.81
Maple+rx+rp	56	13	4.84	15.21	8.7	3.04
Maple+rx+rp+cf	51	18	6.52	12.53	15.62	2.94
SC2019(400)						
glucose+rx	14	8	26.46	10.79	17.42	6.39
glucose+rx+rp	10	26	22.68	8.67	14.59	5.14
glucose+rx+rp+cf	11	26	20.39	11.82	15.51	5.95
Maple+rx	14	7	12.66	2.67	12.94	1.98
Maple+rx+rp	9	14	8.6	3.17	16.59	2.53
Maple+rx+rp+cf	12	15	11.21	3.05	17.23	2.22
SC2020(400)						
glucose+rx	30	9	14.94	11.75	14.67	10.27
glucose+rx+rp	23	37	13.17	10.79	9.4	9.71
glucose+rx+rp+cf	23	37	12.78	11.67	10.52	10.28
Maple+rx	19	13	14.21	6.69	10.24	5.25
Maple+rx+rp	30	29	8.53	6.62	11.7	6.18
Maple+rx+rp+cf	23	36	10.95	6.05	14.17	5.42

# Experiments – Results on FCC benchmark

**Table 3.** Comparing with state-of-the-art solvers on FCC. glucose+ is short for glucose+rx+rp+cf, and malple+ is short for Maple-DL+rx+rp+cf.

Benchmark	glucose	glucose+	Maple	Maple+	kissat_sat	kissat_sat+cf	CCAnr
	#SAT	#SAT	#SAT	#SAT	#SAT	#SAT	#SAT
	#UNSAT	#UNSAT	#UNSAT	#UNSAT	#UNSAT	#UNSAT	#UNSAT
	#Solved	#Solved	#Solved	#Solved	#Solved	#Solved	#Solved
	PAR2	PAR2	PAR2	PAR2	PAR2	PAR2	PAR2
FCC (10000)	7330	8075	8084	8759	8192	8214	7853
	187	197	215	218	207	211	0
	7517	8272	8299	8977	8399	8425	7853
	2555.85	1850.58	1867.13	1243.66	1760.55	1734.61	2215.35

# Conclusion

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As far as we know, **this is the first work that meets the standard of the challenge 7** *"Demonstrate the successful combination of stochastic search and systematic search techniques, by the creation of a new algorithm that outperforms the best previous examples of both approaches."* **on standard application benchmarks.**

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**Thank you!**  
**Any question?**