

An unknotter

Or: Local versus global

Accept ~~Change~~ what you cannot ~~change~~ accept

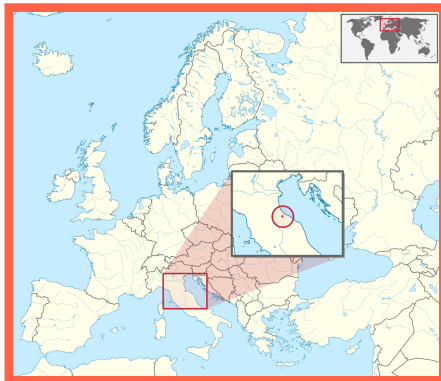


I report on work of many people (Brittenham, Hermiller, Applebaum et al. (sorry too many), Dranowski, Kabkov, ...)

An unknotter

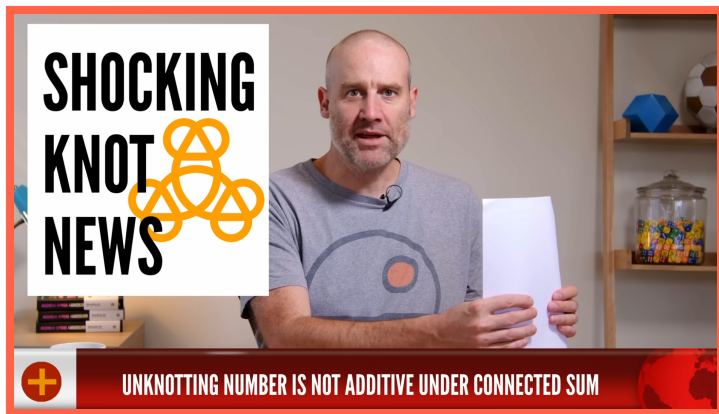
Or: Local versus global

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Unknotting is hard

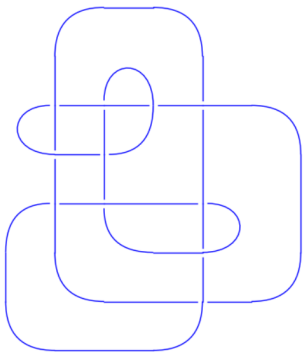


- ▶ **Amazing** Standupmaths <https://www.youtube.com/watch?v=Dx7f-nGohVc>
- ▶ **My take away** You really do not want to do this by hand :-)
- ▶ **Idea** Let a machine do it

Unknotting is hard

It actually works quite well (details later)

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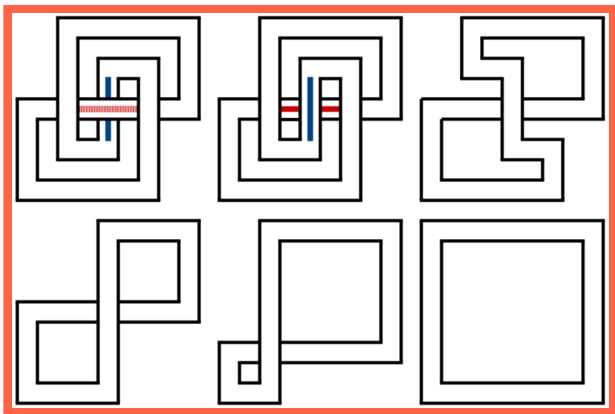
The smallest knot for which it found a nontrivial reduction sequence is the 11 crossing knot above

▶ Amazing Stand

▶ My take away

▶ Idea Let a machine do it

Unknotting is hard

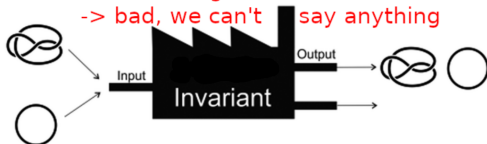


- ▶ Unknotting number u = minimal number of crossings flops needed to reach the unknot
- ▶ Dream You can find u on a minimal crossing diagram
- ▶ Problem (why this is so difficult, part a) That is not true

Unknotting is hard

Problem (why this is so difficult, part b)

The invariant gives the same answer
-> bad, we can't say anything



The invariant gives different answers
-> good, the knots are not the same



I didn't know any "good" invariant
that gives "interesting" information about u

Side quest With details to come, we can change that!

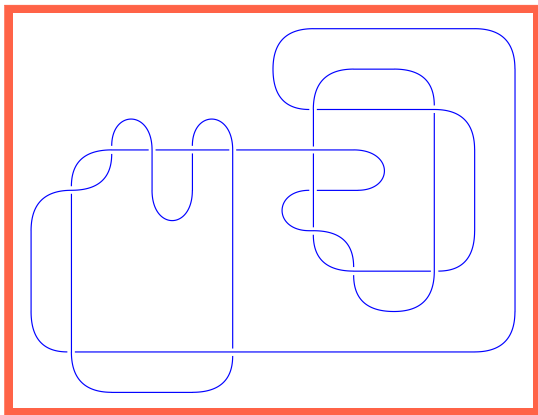
► Unknotting

► Dream Y

► Problem (why this is so difficult, part a) That is not true

ach the unknot

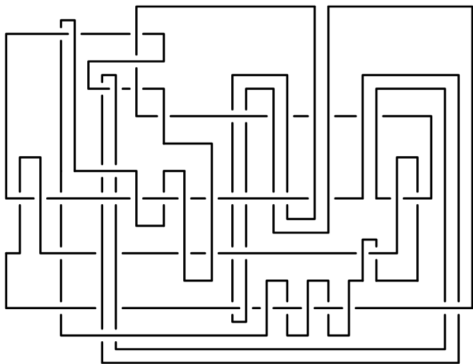
Unknotting is hard



-
- ▶ Open for a long time Unknotting number is additive under connected sum #
 - ▶ No! Brittenham–Hermiller 2025 (“obvious” diagrams hide the real unknotting)
 - ▶ Key What I am going to explain is a variation of their ideas

Unknotting is hard

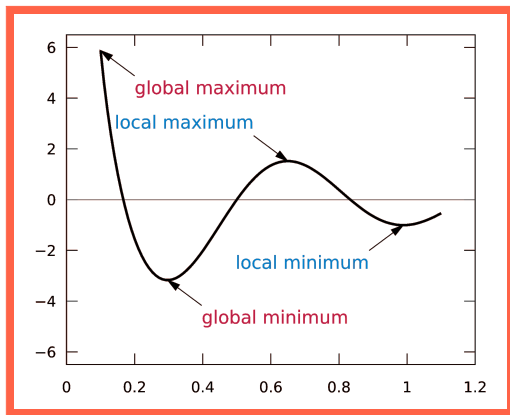
For $7_1 \# \text{mir}(7_1)$



The minimal known diagram with $u = 5$ has 56 crossings (**Wang–Zhang 2025**)

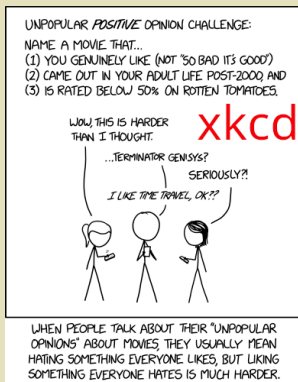
- ▶ Open for a long time (connected sum #)
- ▶ No! Brittenham–Hempel 2020 (obvious diagrams hide the real unknotting)
- ▶ Key What I am going to explain is a variation of their ideas

Unknotting is hard



- ▶ Problem a (rephrased) Human intuition is local; unknotting often is not
- ▶ Problem b Friendly invariants have not much to say about u
- ▶ Idea A perfect task for AI '=' reinforcement learning (RL)

Why is this perfect for AI?



1) There could be a subtle pattern (not detectable by humans) that AI could exploit

2) Repeating the search a million times is likely helpful

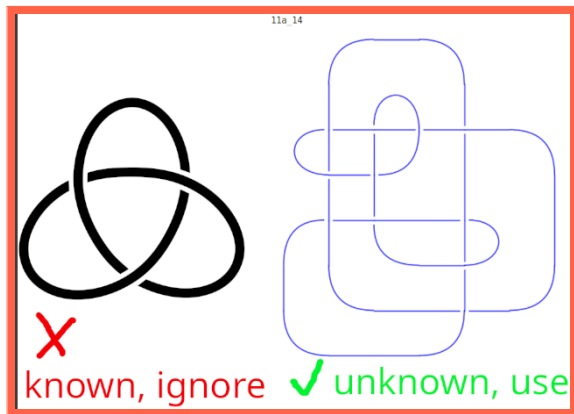
3) This is really not something you want to do by hand :-)

► Problem

► Problem

► Idea

The pipeline



- ▶ **Input** Take a knot diagram from a list (when u is not known)
- ▶ **Goal** Improve the upper bound by finding a better route to the unknot
- ▶ **Key** Do not insist on working with the smallest visible diagram

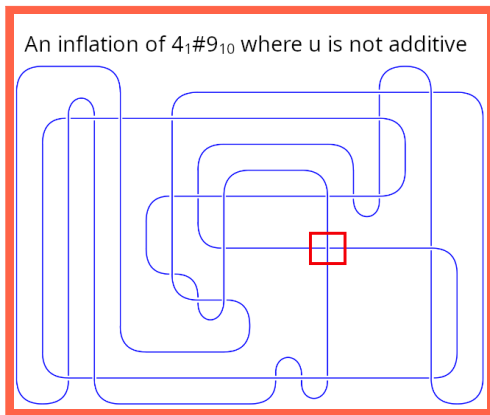
The pipeline

Why start from a list?

10_1	[-2, 8, 1, -1, 2, -2, ↗[[2,11,3,12],[4,20,↗	1
10_2	[1, 11, 1, -1, 2, -2, ↗[[1,13,2,12],[3,14,↗	3
10_3	[-4, 6, 1, -1, 2, -3, ↗[[2,10,3,9],[4,17,5↗	2
10_4	[-5, 5, 1, -2, 3, -3, ↗[[2,13,3,14],[4,11,↗	2
10_5	[-9, 1, -1, 2, -3, 4, ↗[[2,13,3,14],[4,17,↗	2
10_6	[0, 10, 1, -1, 3, -4, ↗[[2,11,3,12],[4,16,↗[2,3]	
10_7	[-1, 9, 1, -2, 4, -5, ↗[[1,12,2,13],[3,15,↗	1
10_8	[-2, 8, 1, -1, 2, -3, ↗[[1,12,2,13],[3,10,↗	2
10_9	[-3, 7, 1, -2, 3, -4, ↗[[1,13,2,12],[3,16,↗	1
10_10	[-7, 3, -1, 2, -3, 5, ↗[[1,11,2,10],[3,18,↗	1
10_11	[-3, 7, 1, -1, 3, -5, ↗[[2,15,3,16],[4,13,↗[2,3]	
10_12	[-8, 2, -1, 2, -4, 6, ↗[[2,11,3,12],[4,18,↗	2
10_13	[-4, 6, 1, -2, 5, -7, ↗[[2,17,3,18],[4,8,5↗	2
10_14	[0, 10, 1, -2, 4, -6, ↗[[1,14,2,15],[3,17,↗	2
10_15	[-6, 4, -1, 2, -4, 6, ↗[[2,6,3,5],[4,16,5,1↗	2
10_16	[-3, 7, 1, -2, 4, -5, ↗[[1,10,2,11],[3,15,↗	2
10_17	[-5, 5, -1, 2, -3, 5, ↗[[6,2,7,1],[12,4,13↗	1

- ▶ **Input** Take One targets knots with a known lower bound (known) but still no sharp upper bound in hand
- ▶ **Goal** Improve so every success gives genuinely new information the unknot
- ▶ **Key** Do not insist on working with the smallest visible diagram

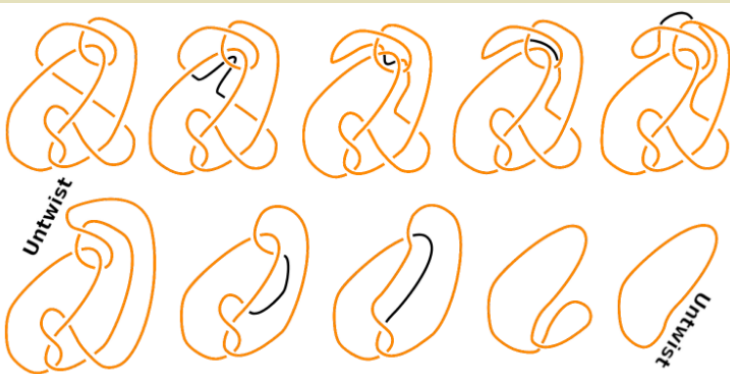
The pipeline



- ▶ **Step 1** First inflate the diagram in a controlled but nontrivial way
- ▶ **Essentially** A first RL machine guesses promising diagrams and crossings
- ▶ **Hope** A larger diagram may reveal a much better crossing to flop

The pipeline

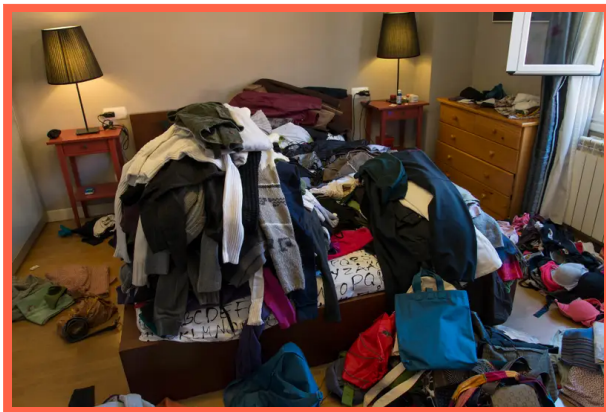
Why inflate first?



In general: sometimes diagrams are “just stuck”
The decisive crossing change may simply not appear on small diagrams
so inflation creates better chances for a good flop

- ▶ **Stuck** Sometimes diagrams are “just stuck”
- ▶ **Essentially** A first RE machine guesses promising diagrams and crossings
- ▶ **Hope** A larger diagram may reveal a much better crossing to flop

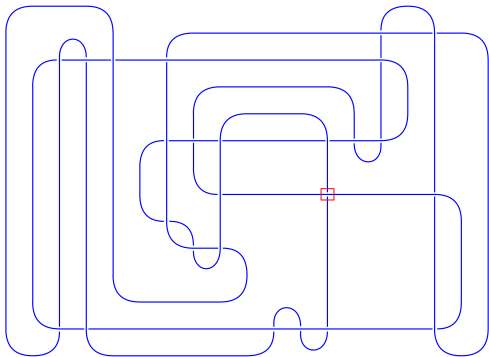
The pipeline



- ▶ **Step 2** Flop one chosen (RL agent 1) crossing on that inflated diagram
- ▶ **Best case** This lands us dramatically closer to the unknot than before
- ▶ **Reality** The new diagram is usually still much too messy to read

The pipeline

After the flop



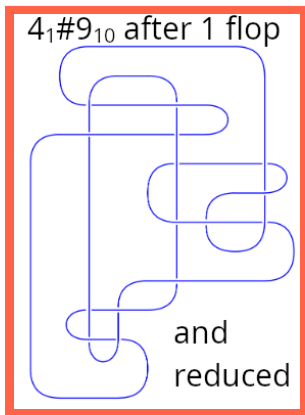
Even after the right crossing change
the diagram usually still looks quite hopeless
so we pass it on to a second RL agent

▶ Step 2 Flop

▶ Best case

▶ Reality The new diagram is usually still much too messy to read

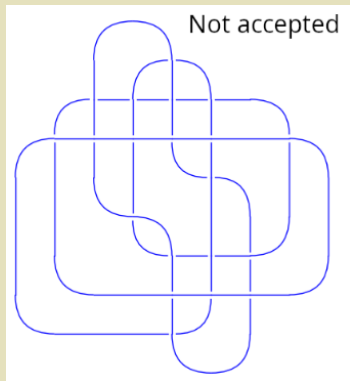
The pipeline



- ▶ Step 3 Reduce the result with the unknotter (RL agent 2; more details later)
- ▶ Step 4 Then we identify the reduced knot using cheap invariants
- ▶ Self-healing Every success produces new training data: the system improves

The pipeline

Crossing reduction is key



After crossing flop the diagram often still has > 20 crossings
So one can't find it on the list in general (invariants are not perfect!)
so the number of crossings needs to be reduced first

▶ Step

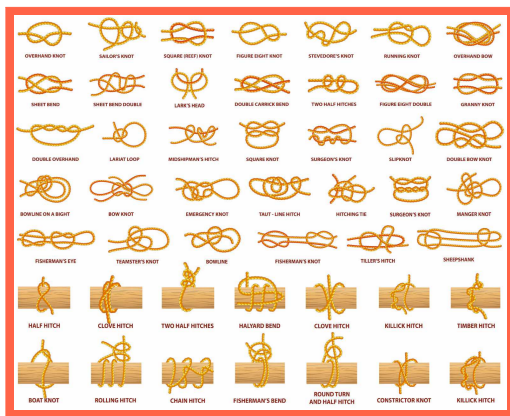
▶ Step

▶ Self-healing Every success produces new training data: the system improves



- ▶ **New 1** Improved bounds for u for prime knots (sometimes identifying u)
- ▶ **New 2** Verified/new examples of unexpected u under $\#$
- ▶ **Update soon?** It is still running

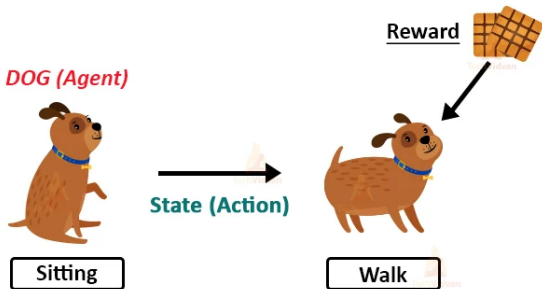
The unknotter (“find the crossing number”)



- ▶ **RL** = learning a strategy from rewards (and penalties)
- ▶ **Here** The environment is the space of knot diagrams and local moves
- ▶ **Goal** Learn which move is most promising from the current diagram

What is RL?

Reinforcement Learning in ML



One tries actions in a given environment and gets feedback on whether they were helpful so over time one learns a good strategy

▶ RL = le

▶ Here The environment is the space of knot diagrams and local moves

▶ Goal Learn which move is most promising from the current diagram

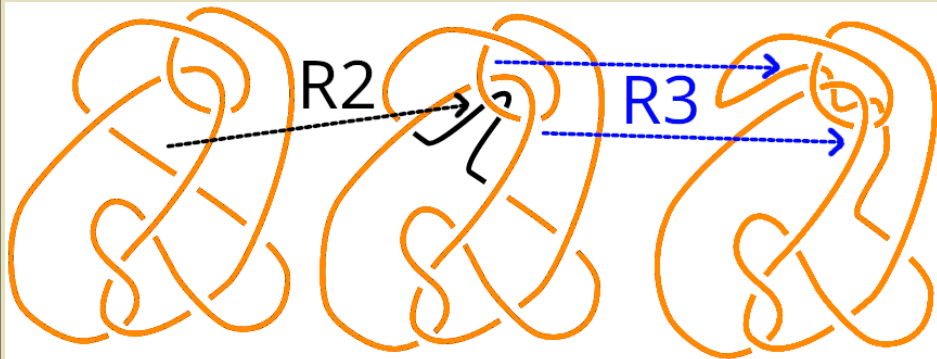
The unknotter (“find the crossing number”)



- ▶ **Moves** Simplify (R1, R2), shuffle (R3), and also add crossings (R1, R2)
- ▶ **Reward** Simpler diagrams get rewarded but getting stuck is penalized
- ▶ **Subtlety** Good moves do not always look good immediately

The unknotter (“find the crossing number”)

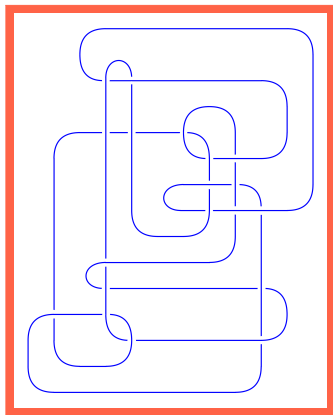
The crucial surprise



If one only simplifies greedily
one may miss the real route completely
so first adding crossings can be the winning move

- ▶ **Reward** Simpler diagrams get rewarded but getting stuck is penalized
- ▶ **Subtlety** Good moves do not always look good immediately

The unknotter (“find the crossing number”)



-
- ▶ **Test set** There are families of hard and very hard unknot diagrams
 - ▶ **Meaning** “No obvious simplification” (see Applebaum et al.)
 - ▶ **Point** These are the right benchmarks if one wants to be taken seriously

The unknotter (“find the crossing number”)

Hard vs. very hard

From Applebaum et al.:

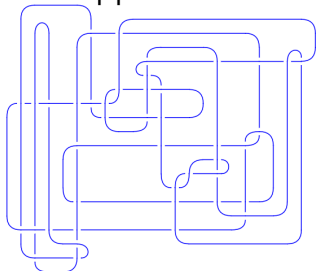


FIGURE 17. A 42-crossing hard unknot diagram with 6225 R3-equivalent diagrams that we have not been able to simplify by calling SnapPy's 'global' heuristic 10000 times.

- ▶ **Test set** There are many examples of hard unknot diagrams. These examples are designed to defeat short-sighted simplification strategies (Applebaum et al.)
- ▶ **Meaning** “No one has been able to simplify them so they are a very meaningful stress test (Applebaum et al.)”
- ▶ **Point** These are the right benchmarks if one wants to be taken seriously

The unknotter (“find the crossing number”)

	run	resolved_count	resolved_pct	median_min_crossings	max_min_crossings
0	1	367	95.324675	0.0	46
1	2	367	95.324675	0.0	38
2	3	365	94.805195	0.0	39
3	4	357	92.727273	0.0	46
4	5	357	92.727273	0.0	41
5	6	360	93.506494	0.0	46
6	7	367	95.324675	0.0	39
7	8	368	95.584416	0.0	46
8	9	365	94.805195	0.0	46
9	10	368	95.584416	0.0	40

- ▶ **Result** On these hard examples the unknotter performs surprisingly well
- ▶ **Strong point** It finds genuinely nontrivial reduction sequences in practice
- ▶ **Message** Letting the machine first add crossings was exactly the right idea

The unknotter (“find the crossing number”)

run r

0	1
1	2
2	3
3	4
4	5
5	6
6	7
7	8
8	9
9	10

Hard knot punchline



ssings

46
38
39
46
41
46
39
46
46
40

- ▶ Result C The unknotter is not merely a toy (no, it is and I like it) singly well
- ▶ Strong p it succeeds on serious benchmark families in practice
and finds genuinely useful reduction paths
- ▶ Message Letting the machine first add crossings was exactly the right idea

