



Automated Symmetric Constructions in Discrete Geometry

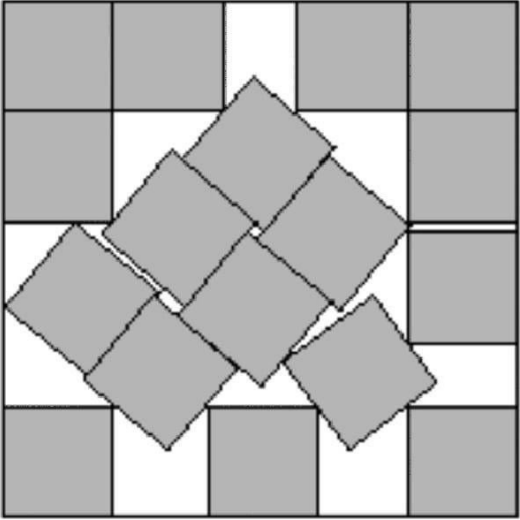
Bernardo Subercaseaux, Ethan Mackey, Long Qian, Marijn J. H. Heule
Carnegie Mellon University

The smallest square we know
that can fit 17 unit-squares
has side ~ 4.675 .

It's very ugly, and not known
to be optimal.

Nathan 🔍 @NathanpmYoung · 23h
god is dead and the most efficient way to pack 17 squares into a square killed him

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$s = 4.675+$
Found by John Bidwell
in 1997.

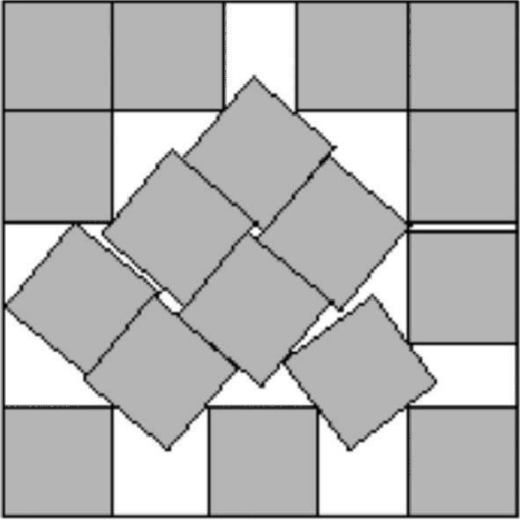
74 1.330 12,3K 500K

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how can we look for

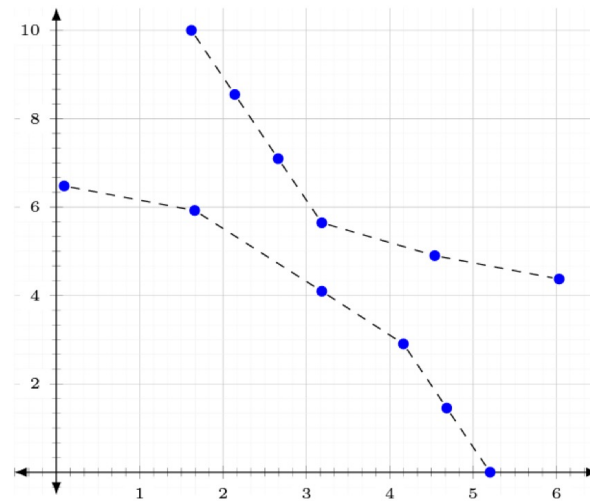
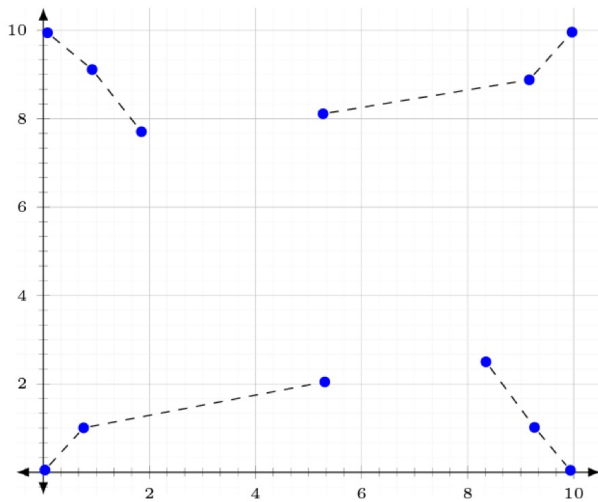
- beautiful solutions?
- clear patterns?
- generalizability?

Story from C1CM 2024

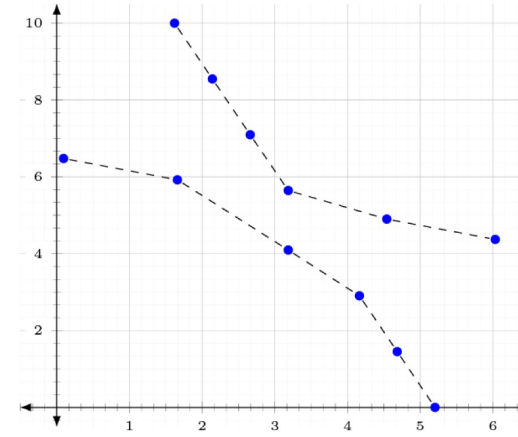
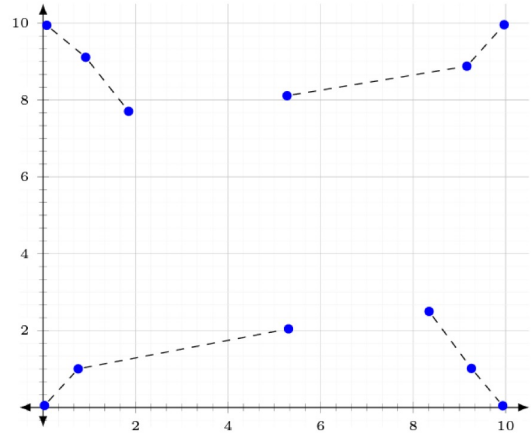
We were trying to understand the minimum number of convex pentagons for N points in the plane without three of them on a line

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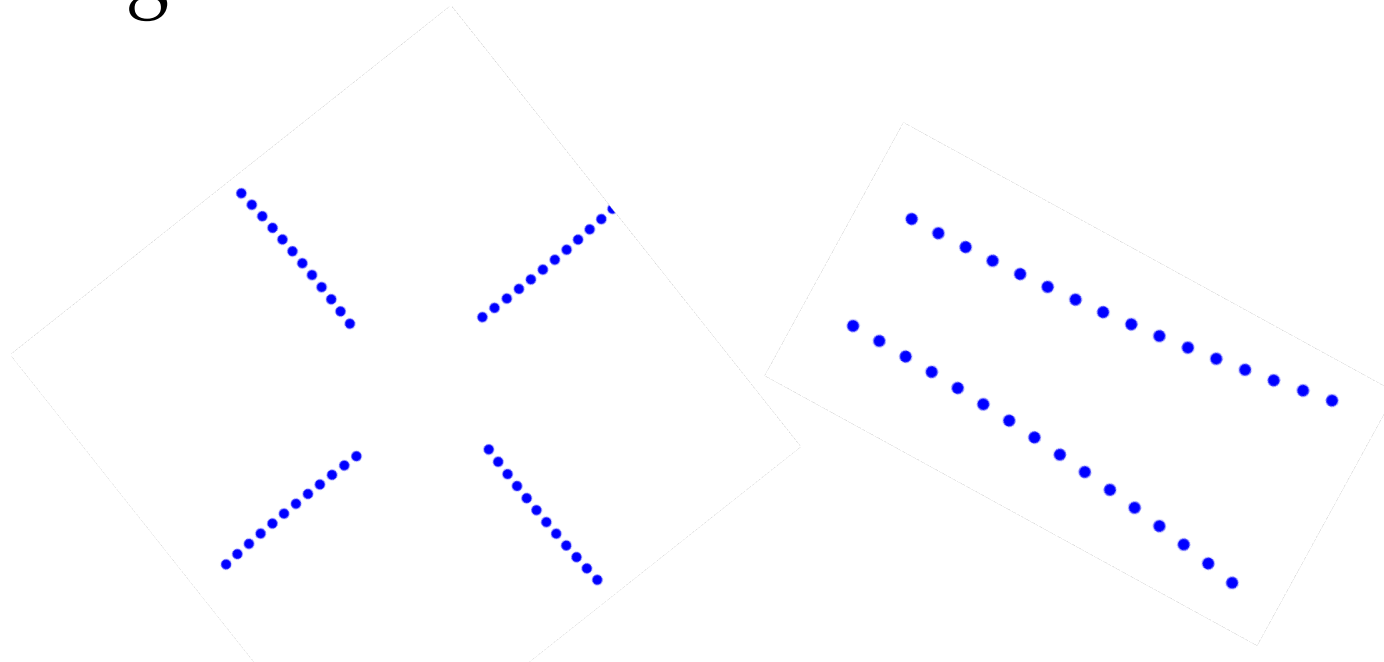
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We used SAT solvers and other computational tools to find optimal solutions for small N



The solutions generalized into nice constructions for all N



Story from CICM 2024

We were lucky!

In general, even if nice patterned solutions exist

- 1) computer searches might find the ugly ones first
- 2) usually too many to enumerate them all and check for a nice one

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Question: how do we factor in “niceness” into our constraints/objective?

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Question: how do we factor in “niceness” into our constraints/objective?

Answer is context dependent,
but we will show that **symmetry** is a great place to start!

Happy Ending Problem (Erdős-Szekeres conjecture)

Given $k \geq 3$, how many points in the plane,
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Esther Klein



Paul Erdős

George Szekeres

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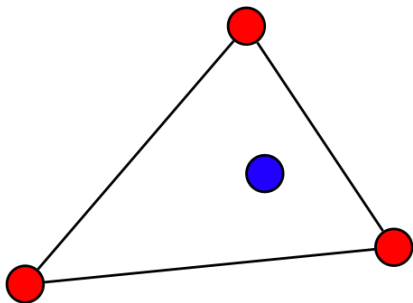
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Trivial: $g(3) = 3$

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Interesting: $g(4) = 5$

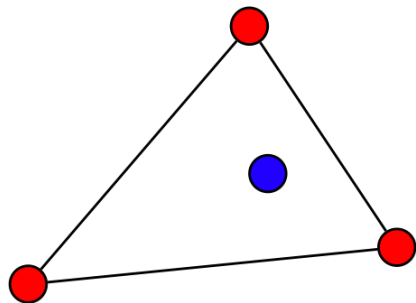


4 points is not enough

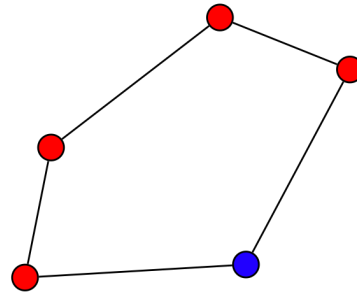
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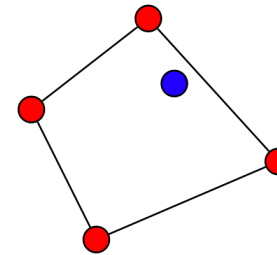
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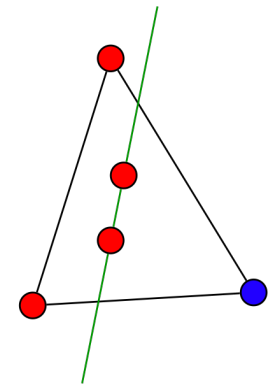
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(a)



(b)



(c)

5 points are enough (Klein, ~1933)

Happy Ending Problem (Erdős-Szekeres conjecture)

$$g(5) = 9, g(6) = 17, g(7) = ???$$

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The bound $g(6) = 17$ was published in 2006 by Lindsay Peters and George Szekeres.

Szekeres died at 95, months before the paper appeared in print, and within one hour of the passing of his wife Esther Szekeres-Klein.

You can read a nice obituary at tinyurl.com/SK-Obituary

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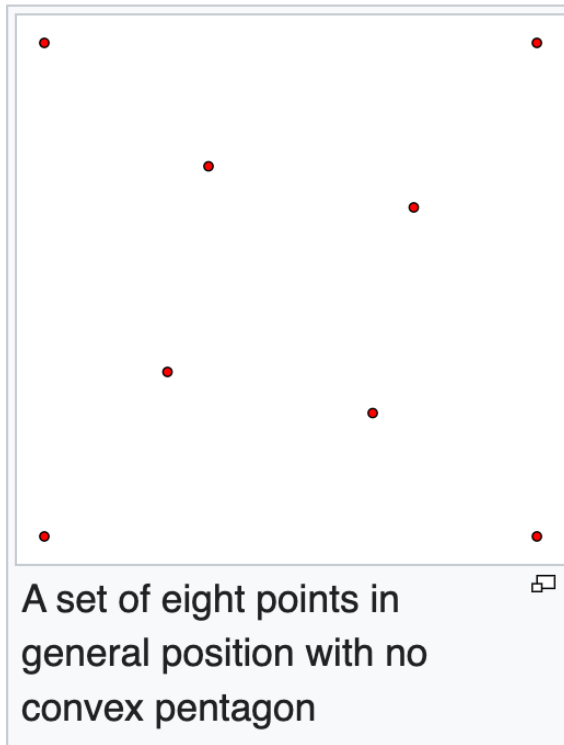
$$g(k) = 2^{k-2} + 1 ?!$$

Happy Ending Problem (Erdős-Szekeres conjecture)

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Wikipedia page right now:

Nice!

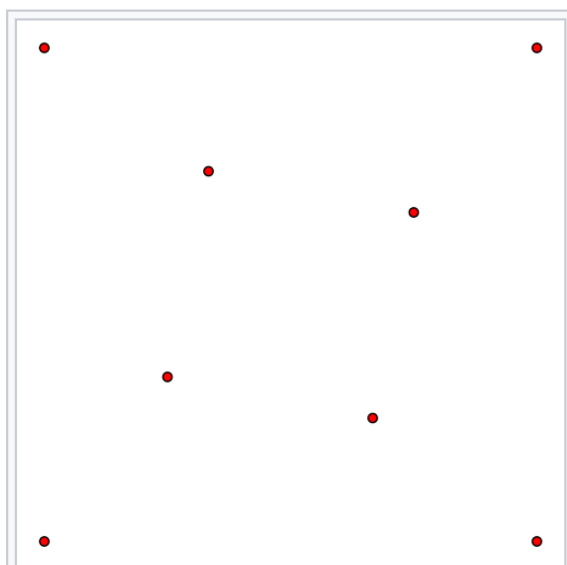


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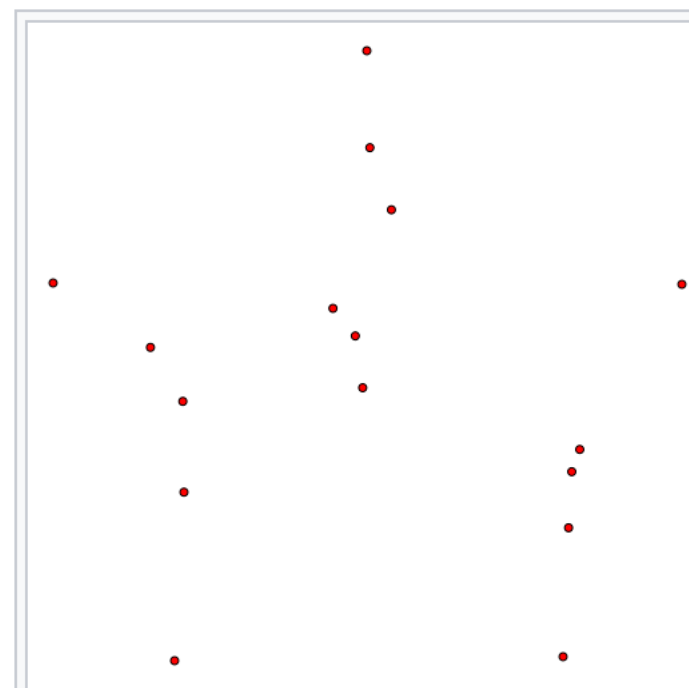
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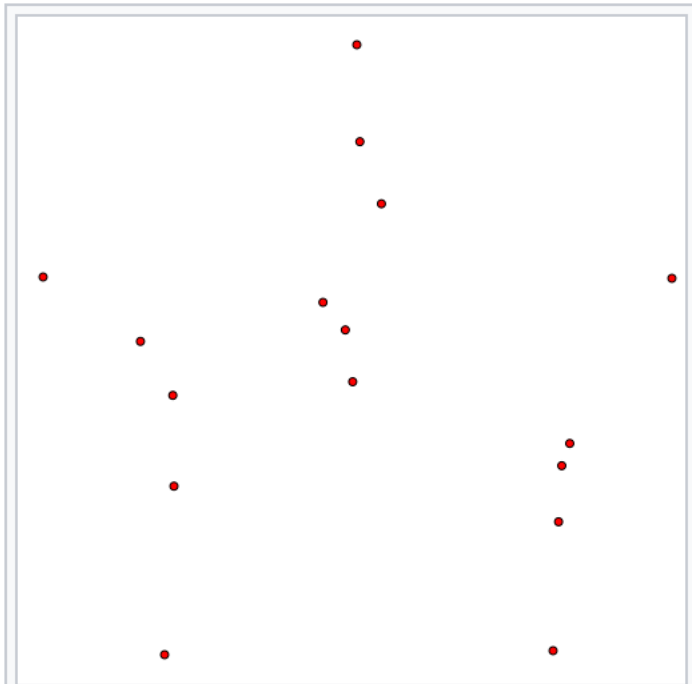
A set of eight points in
general position with no
convex pentagon



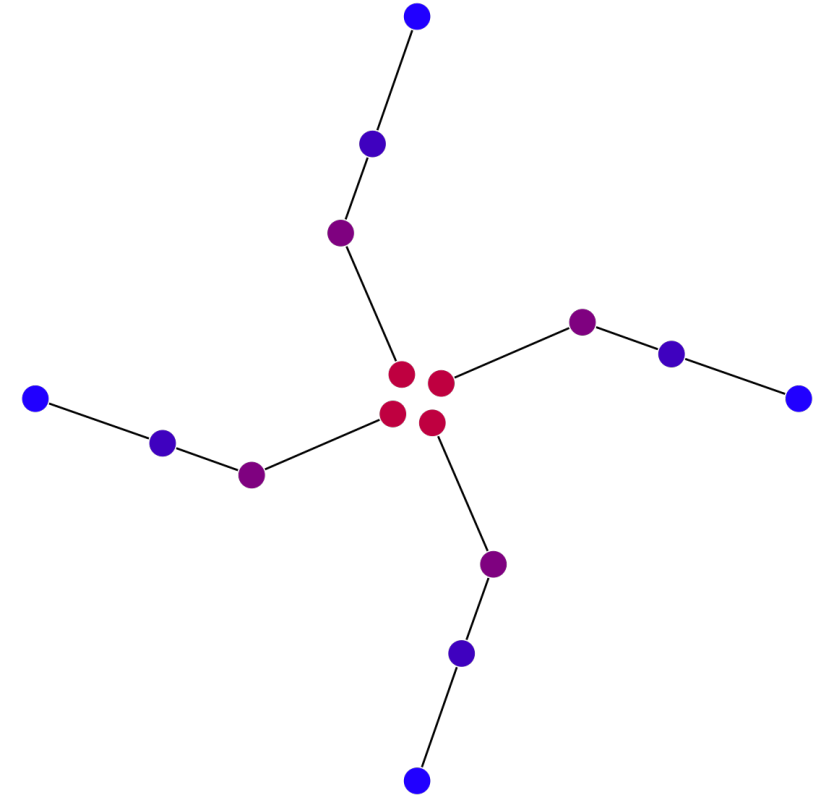
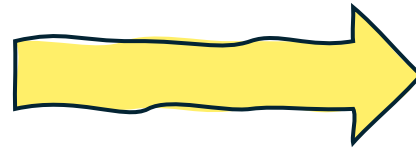
A set of sixteen points in
general position with no
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Not so
nice ☹️

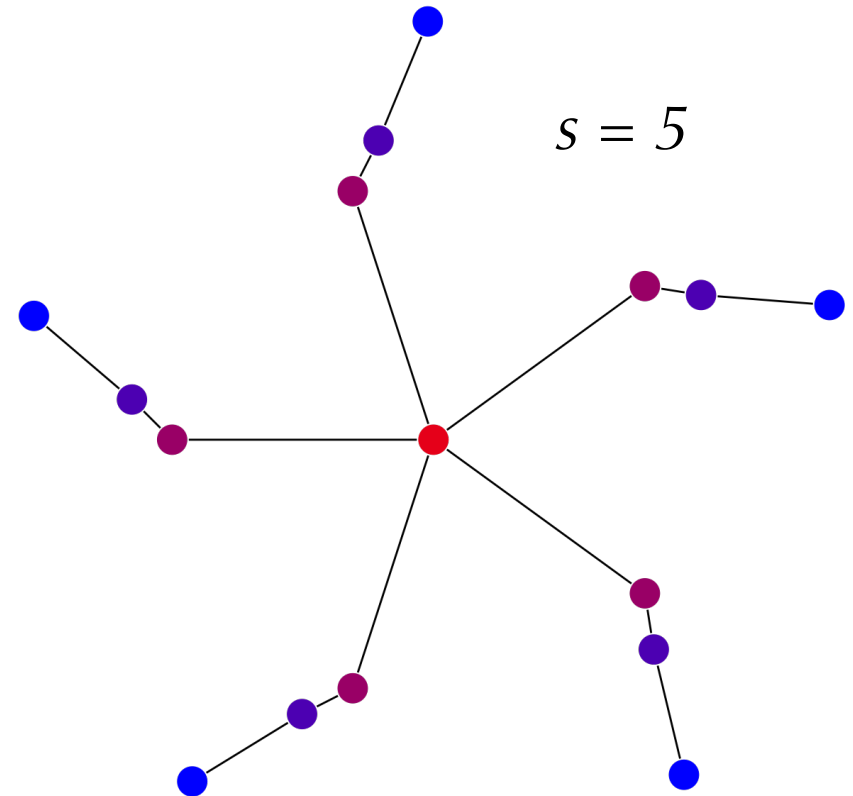
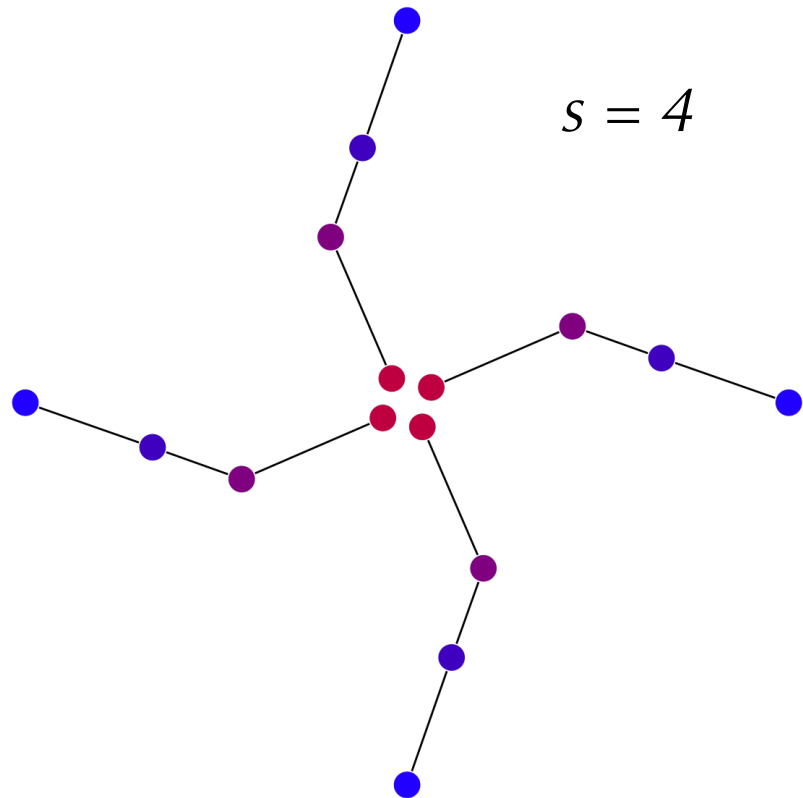
An example of what we achieve in our paper:



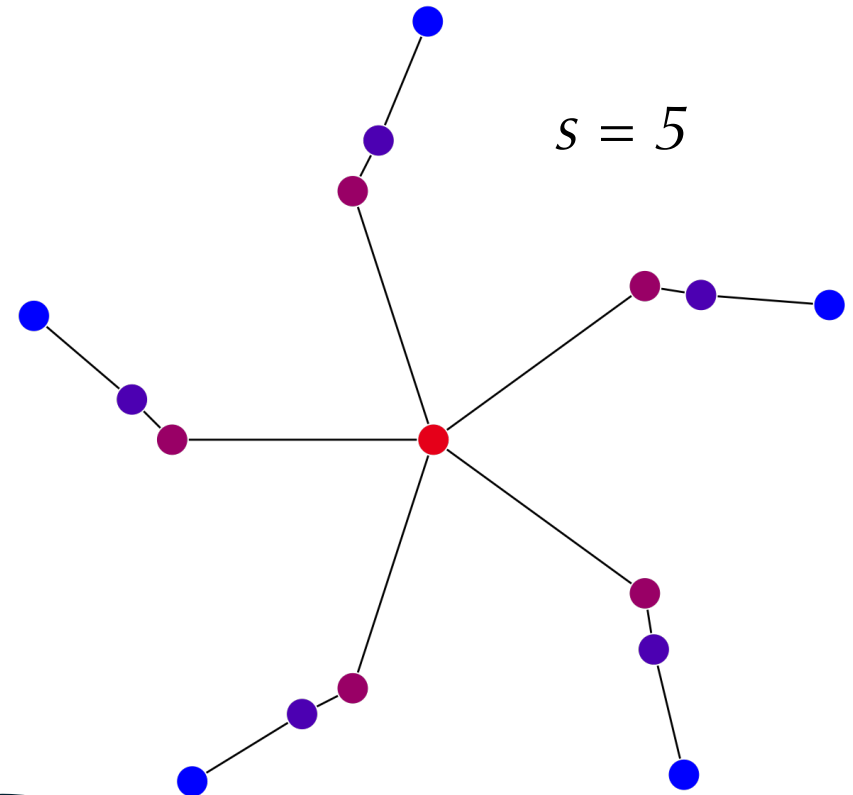
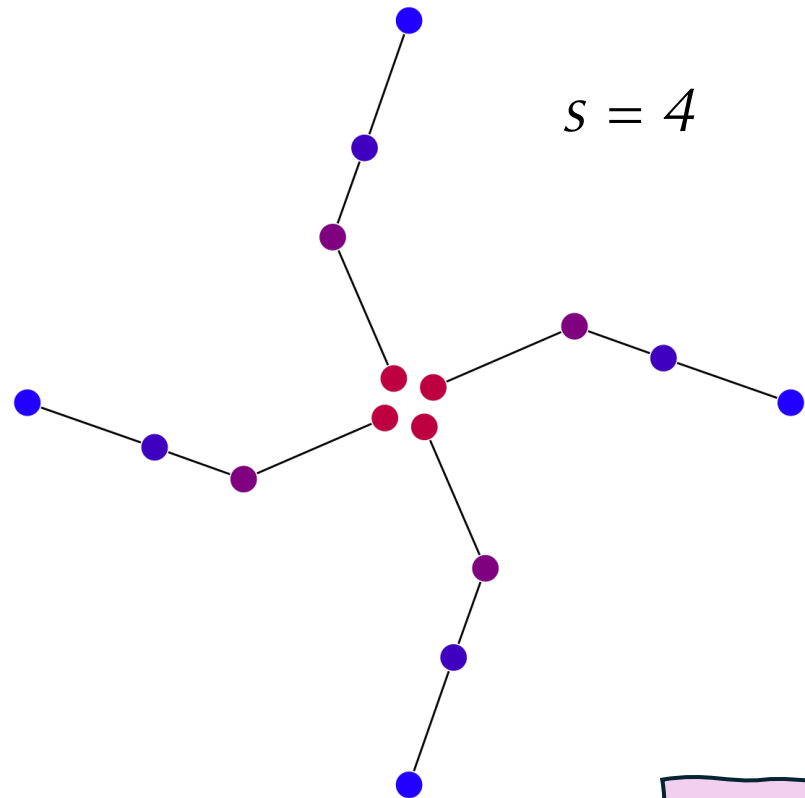
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No 3-fold symmetric
solution exists!



Automated Symmetric Constructions in Discrete Geometry

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All you need to know about SAT solving

Q: What is a SAT solver?

A: A highly optimized program for deciding if a Boolean formula is satisfiable

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A: $(x_1 \vee \neg x_2 \vee x_3) \wedge (\neg x_1 \vee x_2)$

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Q: And how do I write this on my computer ?

A:

 test.cnf

```
p cnf 3 2
1 -2 3 0
-1 2 0
```

All you need to know about SAT solving

Q: And then what?

A:

terminal

```
> kissat test.cnf  
[...]  
s SATIISFIABLE  
v -1 2 3 0
```

or

terminal

```
> kissat test2.cnf  
[...]  
s UNSATISFIABLE
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or



All you need to know about SAT solving

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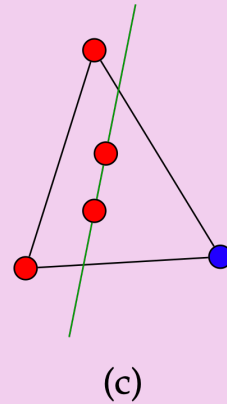
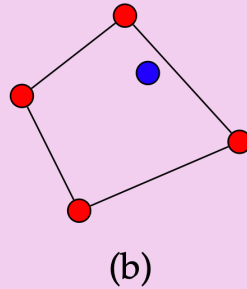
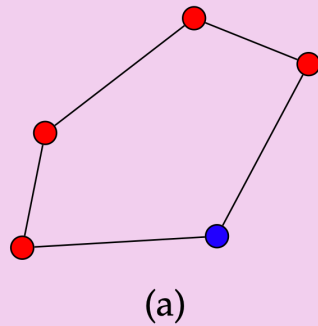
or



You can also get a
reproducible proof!

Now I'll talk about discrete geometry,
and using SAT for it.

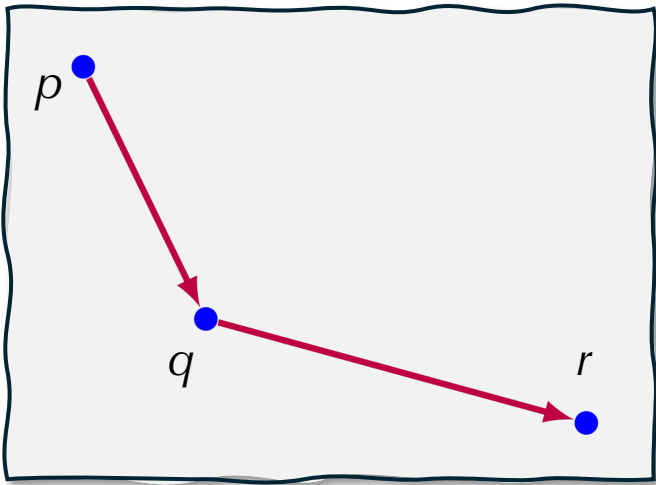
I want you to keep in mind the insight of Esther Klein:



$\mathbb{R}^2$ might be uncountably infinite,

But a finite number of well-chosen
cases might answer our questions.

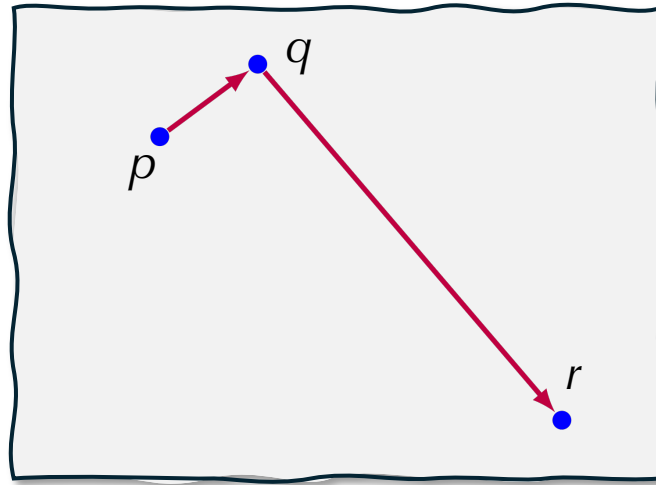
Discrete Orientations (Knuth CC-Axioms)



Counterclockwise (CC)

$$A_{p,q,r}$$

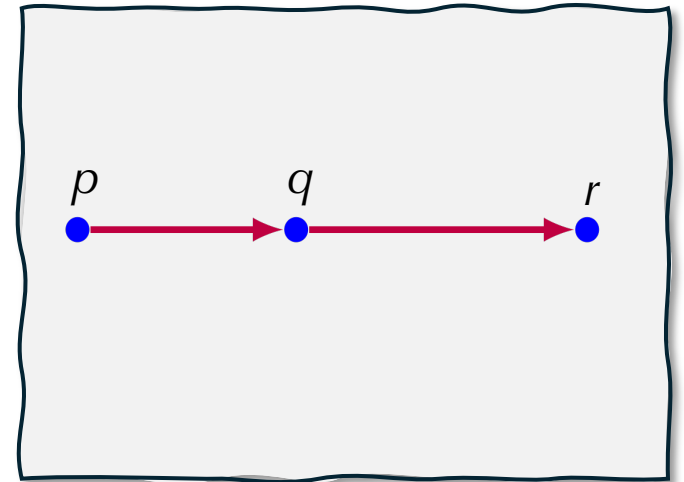
p is **Above** directed line qr



Clockwise

$$B_{p,q,r}$$

p is **Below** directed line qr



Collinear

$$C_{p,q,r}$$

Discrete Orientations ($\sim$ Knuth CC-Axioms)

The orientations of (q, r, p) , or (q, p, r) or (p, r, q) , etc... can all be deduced from (p, q, r) .

Instead of writing e.g., $A_{p, r, q}$ we can write $\neg A_{p, q, r}$

We thus have $3\binom{n}{3}$ variables for n points.

Discrete Orientations (~Knuth CC-Axioms)

Some combinations of variables don't make geometric sense!

$$C_{p,q,r} \wedge C_{q,r,s} \wedge A_{p,q,s}$$

Axiom 1. $C_{p,q,r} \wedge C_{q,r,s} \rightarrow C_{p,q,s}$

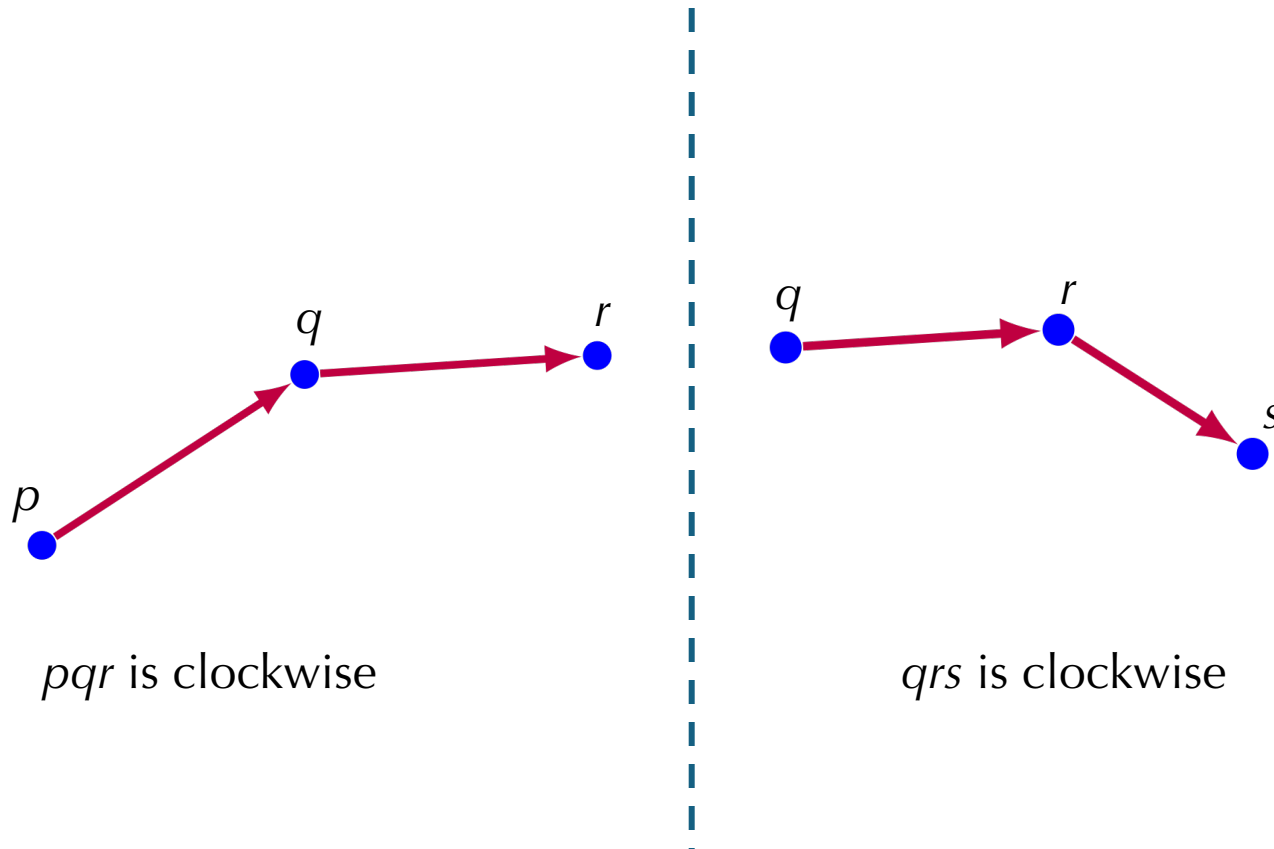
Axiom 2. $C_{p,q,r} \wedge C_{q,r,s} \rightarrow C_{p,r,s}$

Axiom 3. $C_{p,q,r} \wedge A_{q,r,s} \rightarrow A_{p,r,s}$

(and so on...)

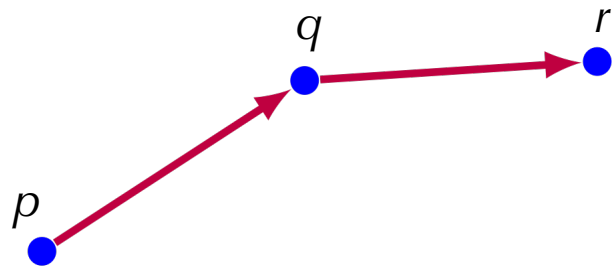
Discrete Orientations (~Knuth CC-Axioms)

Some axioms discard a significant portion of “ungeometrical configurations”,
But require points to be sorted left-to-right.

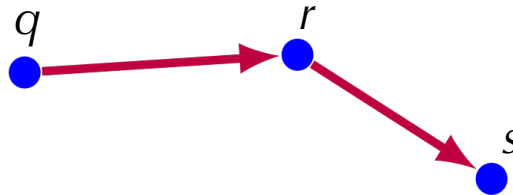


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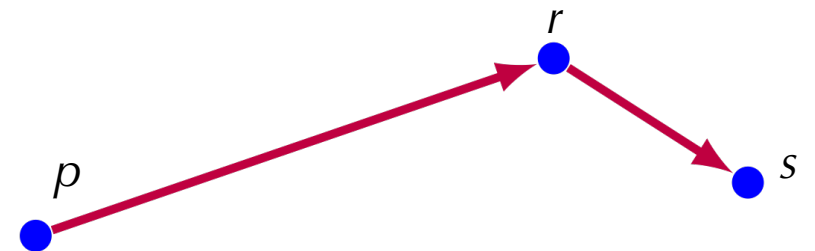
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pqr is clockwise



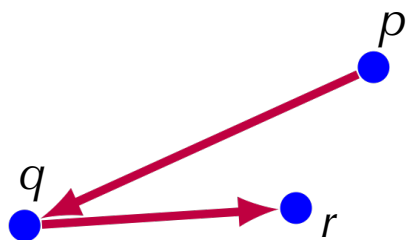
qrs is clockwise



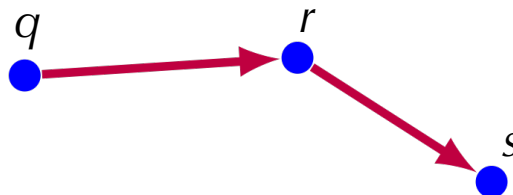
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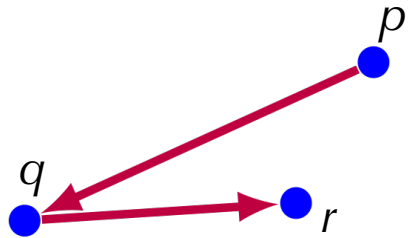
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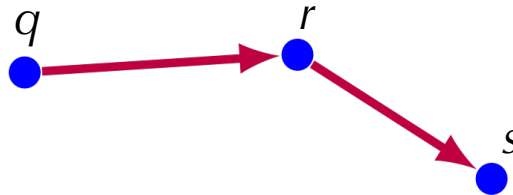
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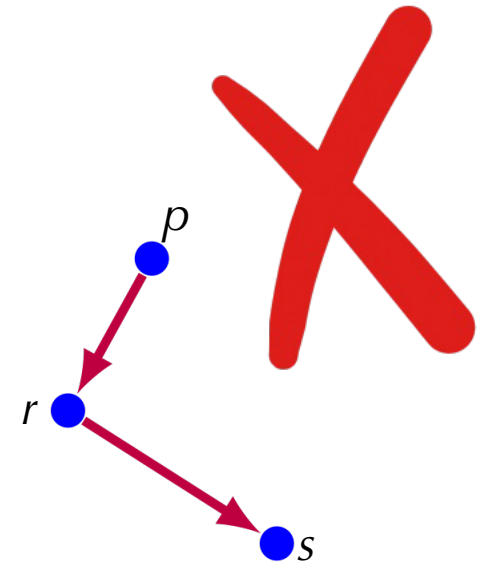
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qrs is clockwise



prs is **NOT** clockwise !

Erdős-Szekeres Encoding

Goal: n points, without 3 on a line, that do not contain a convex k -gon.

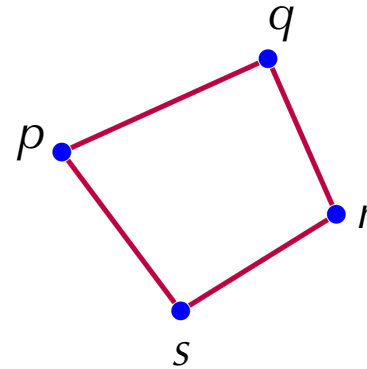
Base variables: $A_{p,q,r}$ (since no collinearities)

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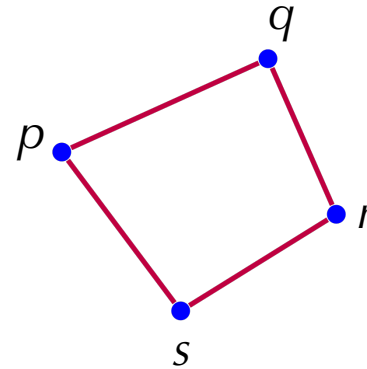


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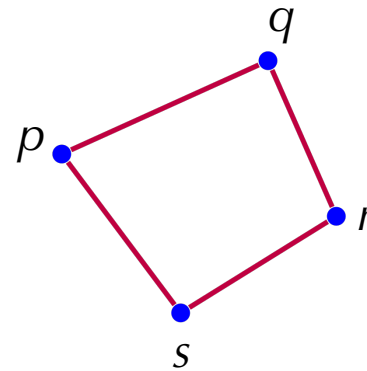
$$\text{Conv}_{p,q,r,s} \leftrightarrow ((A_{p,q,r} \leftrightarrow A_{p,r,s}) \leftrightarrow (A_{p,r,s} \leftrightarrow A_{q,r,s}))$$

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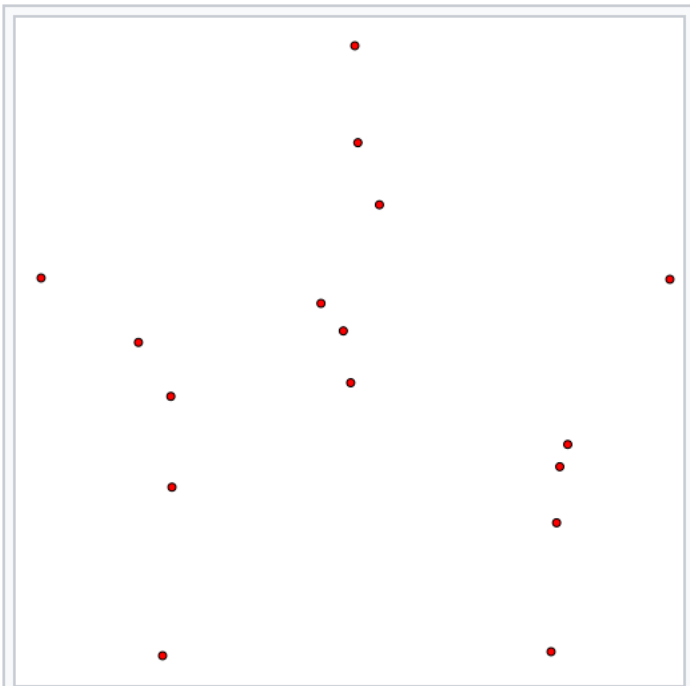
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Main constraint: for every subset of k -points, at least one of its 4-subsets is not convex.

Symmetry Constraints

We want s -fold rotational symmetries



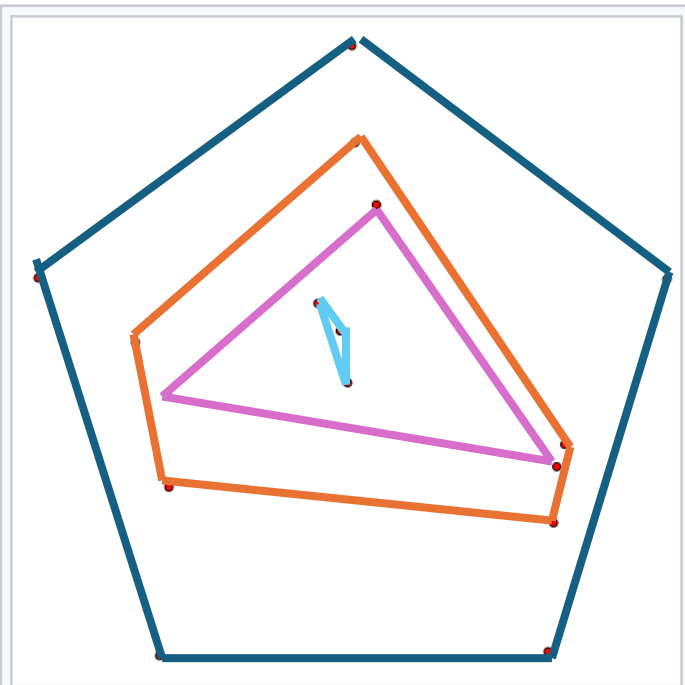
A set of sixteen points in
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To illustrate by example,
remember the Wikipedia picture.

Symmetry Constraints

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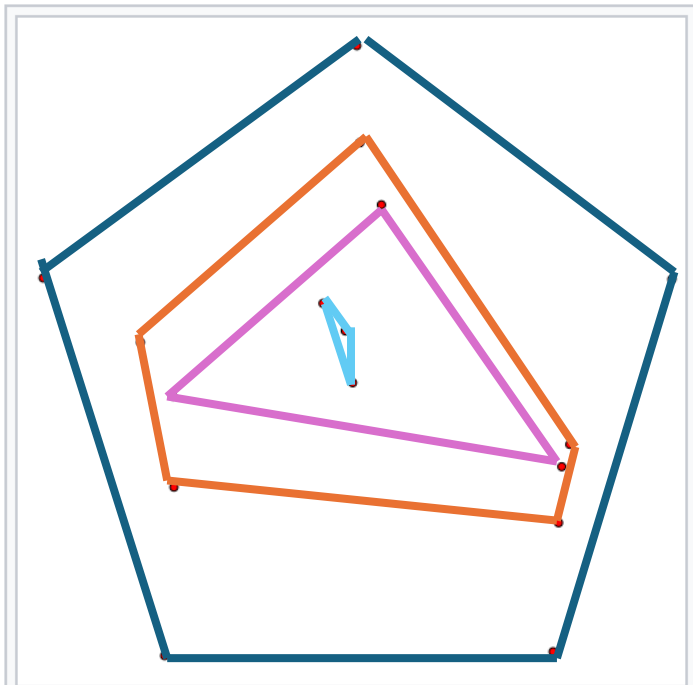
Its *convex-hull onion* is 5-5-3-3



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Symmetry Constraints

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A set of sixteen points in general position with no convex hexagon

Its *convex-hull onion* is 5-5-3-3

This cannot be made 5-fold symmetric nor 3-fold symmetric

So let's aim for **5-5-5-1**, which can be made 5-fold symmetric.

Symmetry Constraints

We want s -fold rotational symmetries

Our symmetric 5-5-5-1 constructions will have orbits:

1 → 2 → 3 → 4 → 5
6 → 7 → 8 → 9 → 10
11 → 12 → 13 → 14 → 15
16

Symmetry Constraints

We want s -fold rotational symmetries

$1 \rightarrow 2 \rightarrow 3 \rightarrow 4 \rightarrow 5$

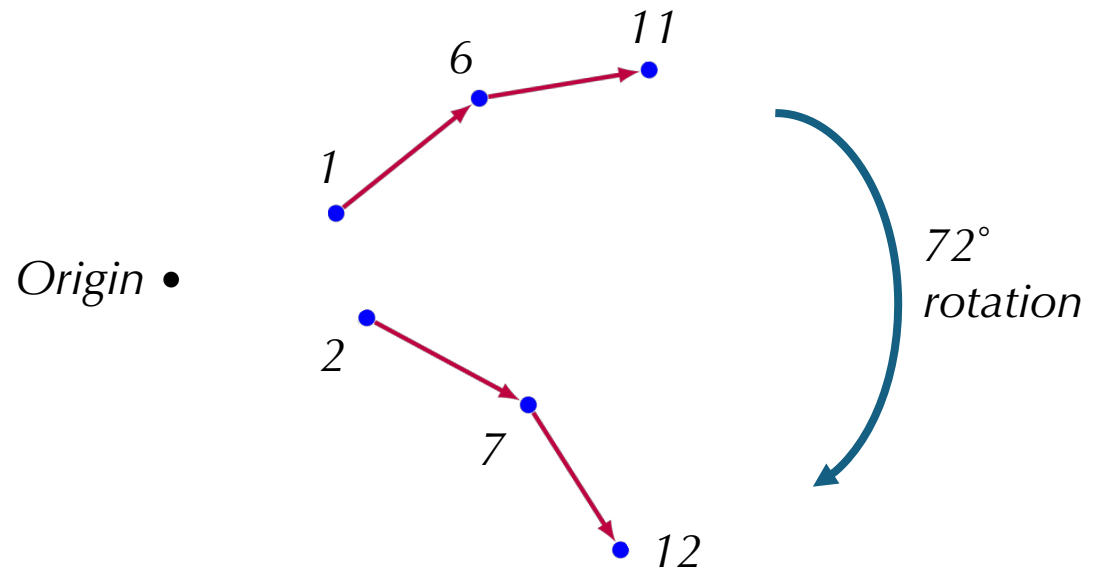
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This implies that

$$A_{1,6,11} \leftrightarrow A_{2,7,12} \leftrightarrow \dots$$



Symmetry Constraints

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View this as cycles of permutation π

$$A_{p,q,r} \leftrightarrow A_{\pi(p),\pi(q),\pi(r)}$$

Remember left-to-right ordering axioms?

Knuth (1992) showed that these reduce search space dramatically:

$$2^{O(n^3)} \rightarrow 2^{O(n^2)}$$

If we don't enforce symmetry, we can always relabel points so that they're sorted left-to-right (point 1 is leftmost, point n is rightmost)

But when enforcing a symmetry π , we commit to certain relationships between the labels.

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Example

point 16 is mapped to itself, which implies its at the origin, and thus cannot be the rightmost

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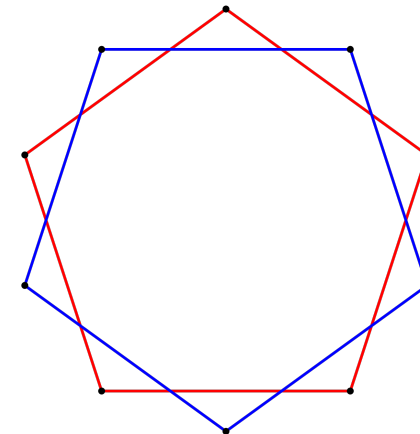
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Example 2

Orbits must remain in the same convex-hull layer; thus either (1,2,3,4,5) are all inside (11,12,13,14,15) or viceversa

Or there is a big convex layer of 10 points mixing two orbits.



Remember left-to-right ordering axioms?

But when enforcing a symmetry π , we commit to certain relationships between the labels.

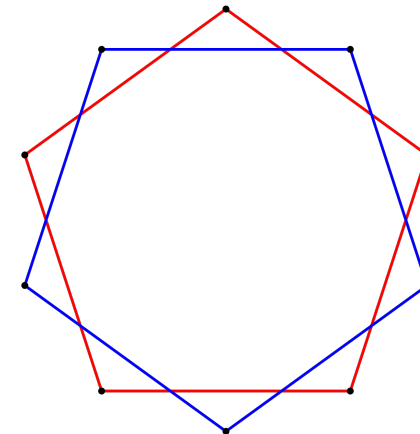
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16

Example 2

Orbits must remain in the same convex-hull layer; thus either (1,2,3,4,5) are all inside (11,12,13,14,15) or viceversa

Or there is a big convex layer of 10 points mixing two orbits.

So how can we enforce symmetries while still enforcing the left-to-right axioms that dramatically reduce search space?



Dynamic Ordering Axioms

(most interesting theoretical contribution of our paper, in my opinion)

Instead of enforcing a left-to-right order, we enforce
some linear ordering $<$ between the points

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Which one though?

You

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Which one though?

You

I don't know!
We just let the solver figure it out

Me

Dynamic Ordering Axioms

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Instead of enforcing a left-to-right order, we enforce *some* linear ordering $<$ between the points

We add variables $<_{p,q}$ which the solver will assign

Totality. $(<_{p,q} \vee <_{q,p})$, for all $1 \leq p \neq q \leq n$.

Asymmetry. $(<_{p,q} \leftrightarrow \overline{<_{p,q}})$, for all $1 \leq p \neq q \leq n$.

Transitivity. $(<_{p,q} \wedge <_{q,r} \rightarrow <_{p,r})$, for all $1 \leq p \neq q \neq r \leq n$.

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Instead of enforcing a left-to-right order, we enforce *some* linear ordering $<$ between the points

We add variables $<_{p,q}$ which the solver will assign

And then use this ordering to “guard” the axioms so that they only apply to ordered tuples

$$(<_{p,q} \wedge <_{q,r} \wedge <_{r,s}) \rightarrow (A_{p,q,r} \vee \overline{A_{p,r,s}} \vee A_{q,r,s})$$

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Why on earth would any linear ordering be correct?!

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Why on earth would any linear ordering be correct?!

You

I don't know!

I proved this system is equivalent to Knuth's axioms using a SAT solver!

Me

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There's no counterexample with 5 points (proved via SAT)

Realizability Problem

Suppose we solve the formula and obtain:

$A_{1,2,3} \mapsto \text{true}$
 $A_{1,2,4} \mapsto \text{false}$
 $A_{1,3,4} \mapsto \text{true}$
 $A_{1,2,5} \mapsto \text{false}$
 $\vdots$
 $A_{14,15,16} \mapsto \text{false}$

Now what?

It's complete for $\exists\mathbb{R}$ to see if a pointset can actually *realize* these orientations.

Realizability Problem

Previous work (CICM 2024):

Pose a non-linear program to a commercial solver (LocalSolver)

$$\begin{array}{ll}\text{maximize} & z \\ \text{subject to} & z \leq \sqrt{(x_a - x_b)^2 + (y_a - y_b)^2}, \\ & \varepsilon < \sigma^*(a, b, c) \cdot [(y_c - y_a)(x_b - x_a) - (x_c - x_a)(y_b - y_a)], \\ & 0 \leq x_a \leq K, \\ & 0 \leq y_a \leq K,\end{array}$$

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Very slow, handled
dozens of points only

Localizer

We introduce a new open-source tool, **Localizer** (local search + realizer)

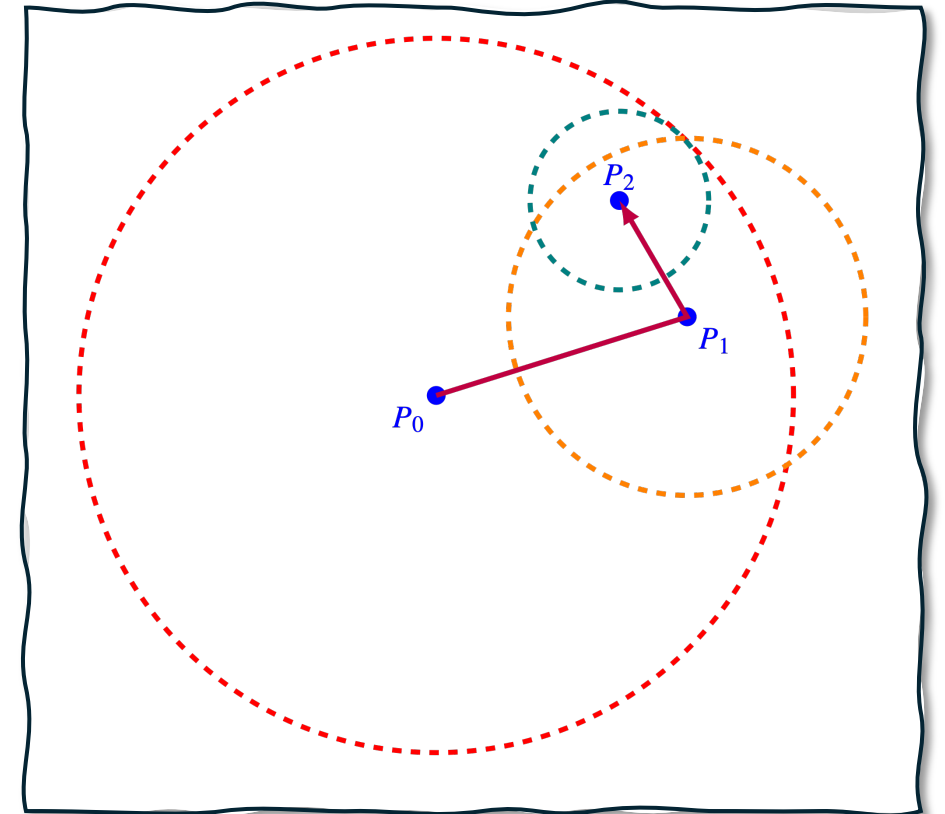
Count violated constraints, and move a point randomly to reduce this count

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Start with aggressive moves, and decrease exponentially
(Local version of simulated annealing)



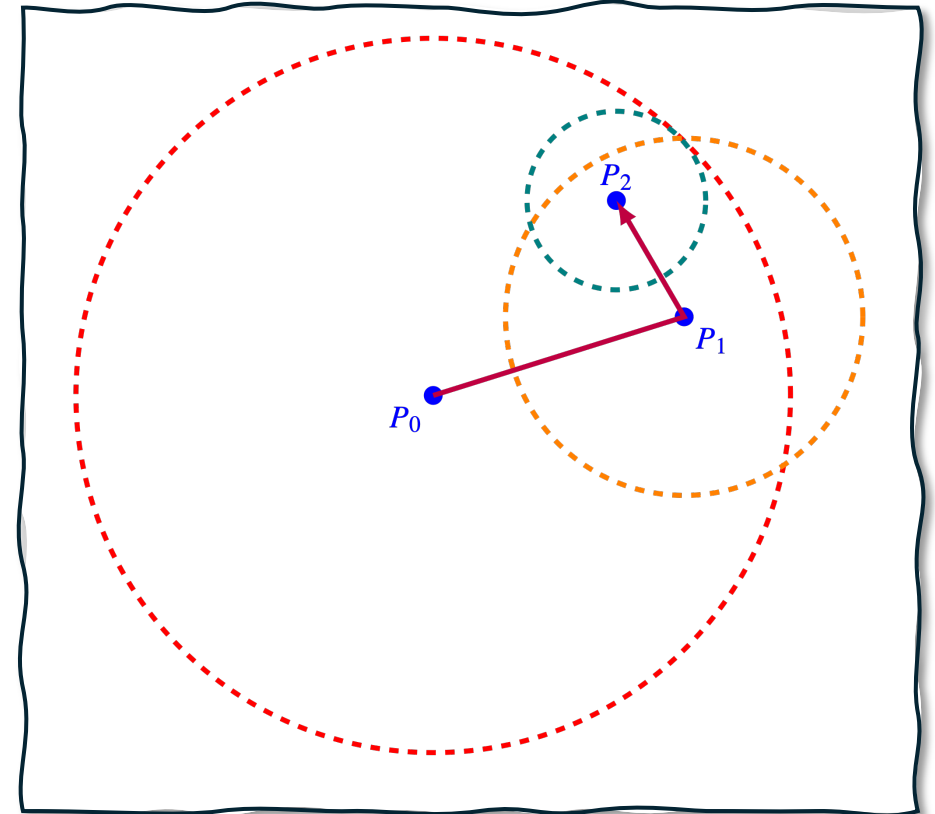
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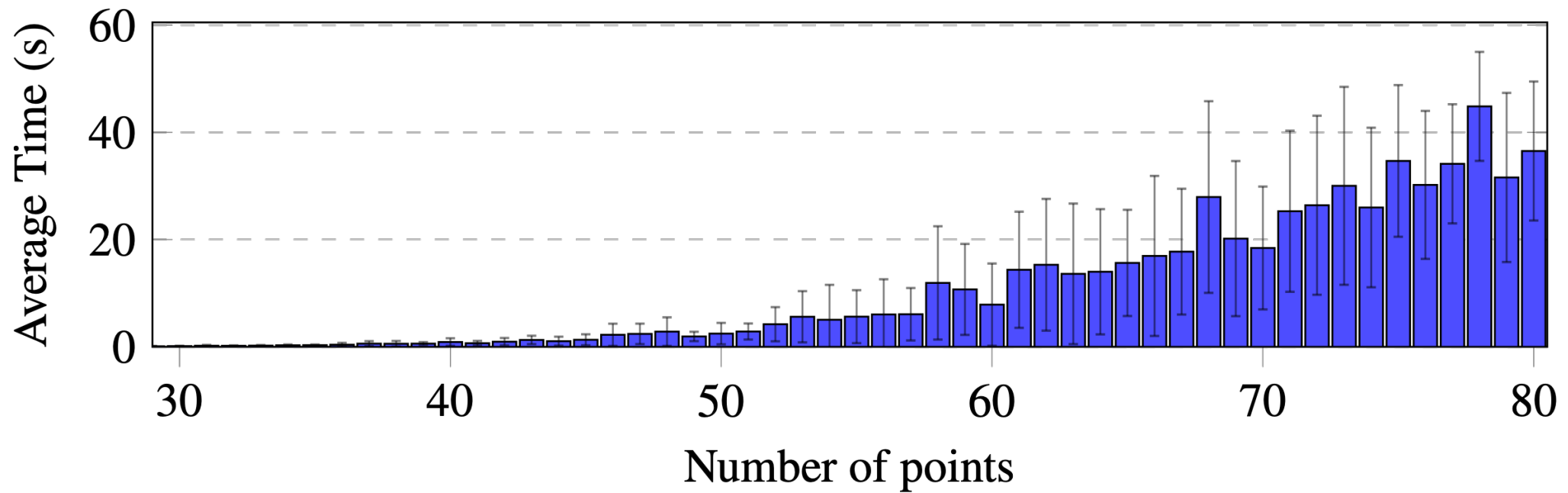
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Start with aggressive moves, and decrease exponentially
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Multi-threaded, written in C for extreme efficiency



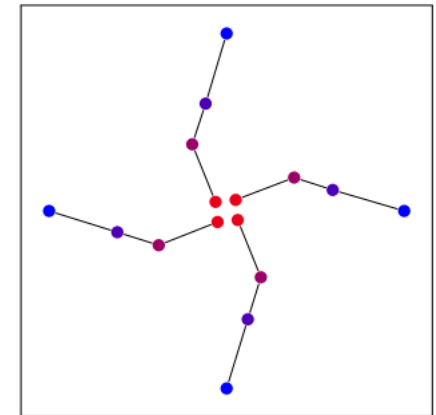
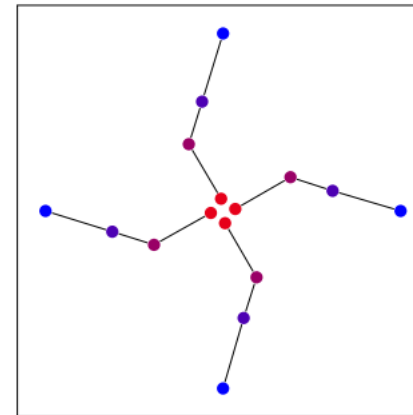
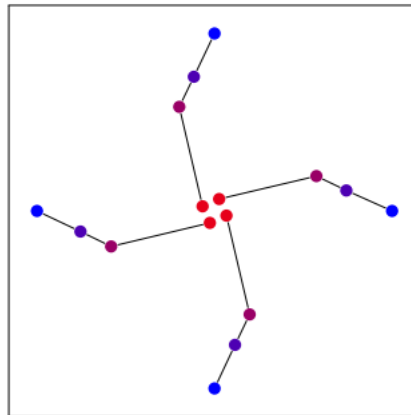
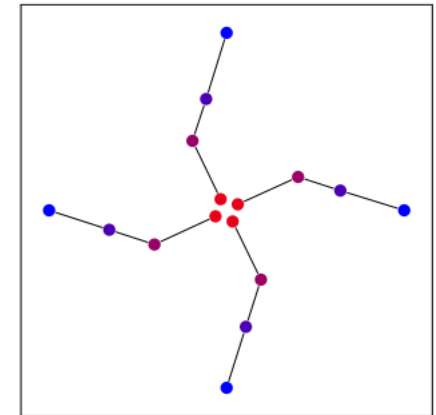
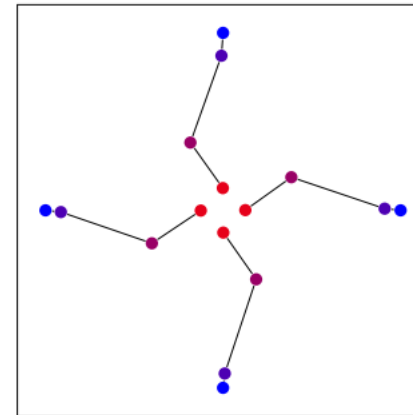
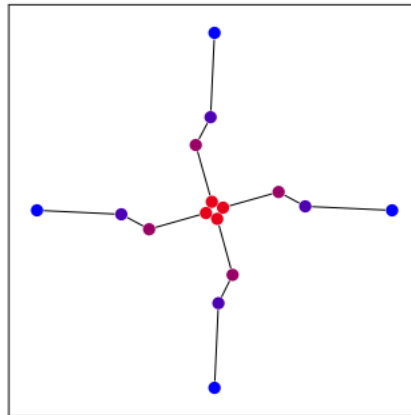
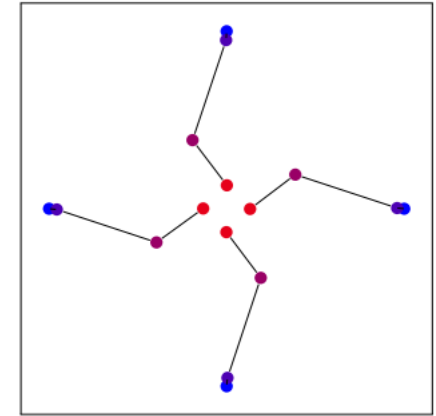
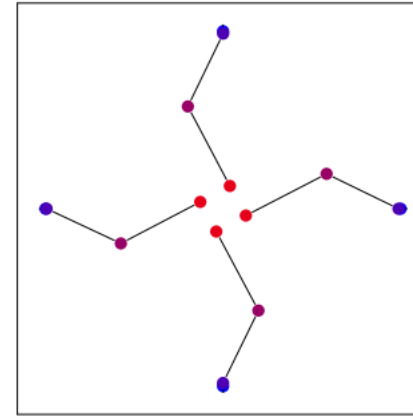
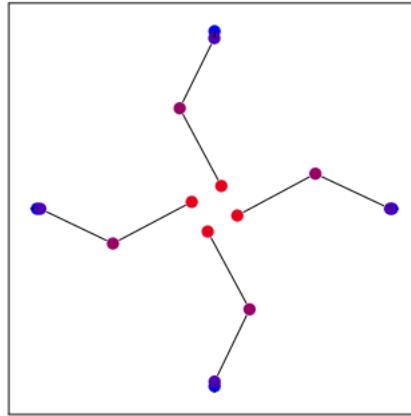
Localizer



Previous approach took hundreds of seconds for under < 20 points,
not even visible here!

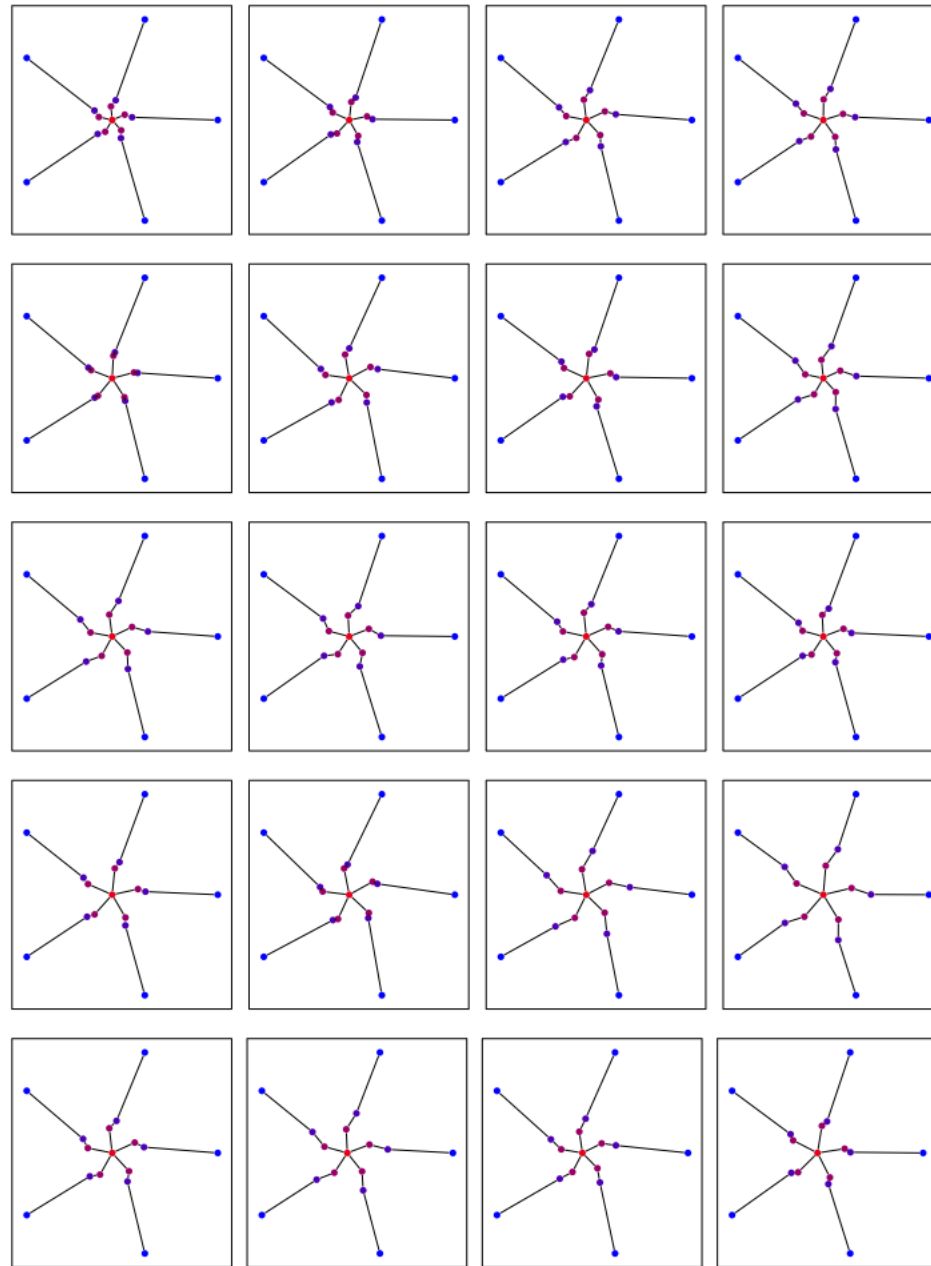
We get a complete enumeration:

9 non-isomorphic solutions with 4-fold symmetry.



We get a complete enumeration:

46 non-isomorphic solutions with 5-fold symmetry.



What about 32 points without a convex 7-gon?

Despite millions of solutions, even after filtering for basic isomorphisms, none of them is 4-fold symmetric, which is the expected symmetry for 32 points.

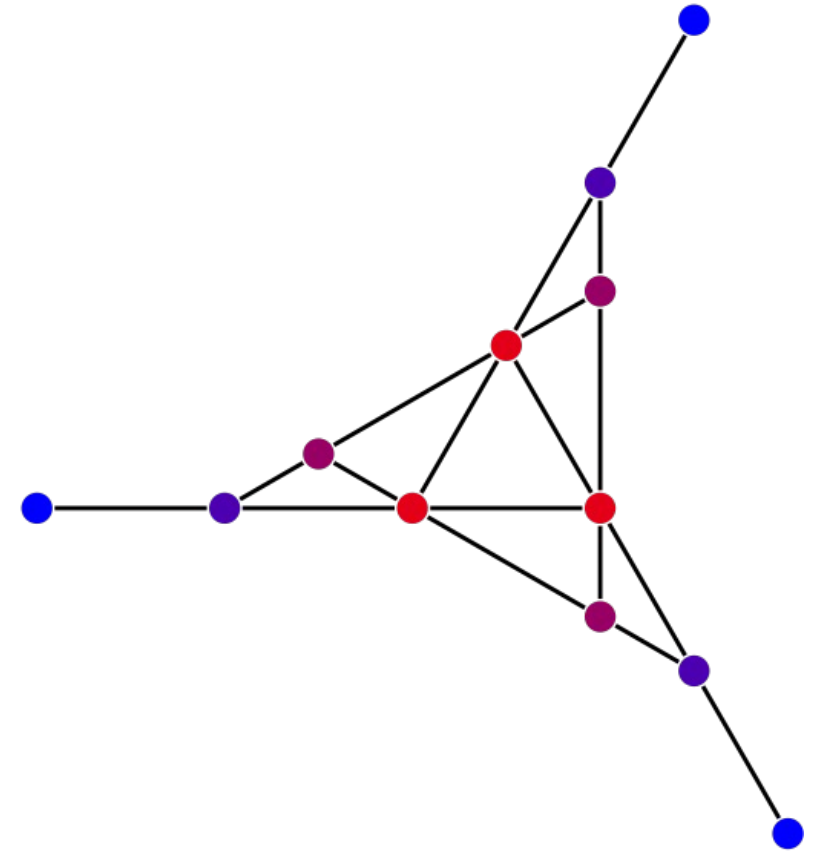
I don't have a good explanation for why this would be the case. But I'd like to know.

Everywhere Unbalanced Points

Alon's construction, circa 2000

If you think of any line L between two of these 12 points, then

$$| \# \text{points above } L - \# \text{points below } L | \geq 2$$

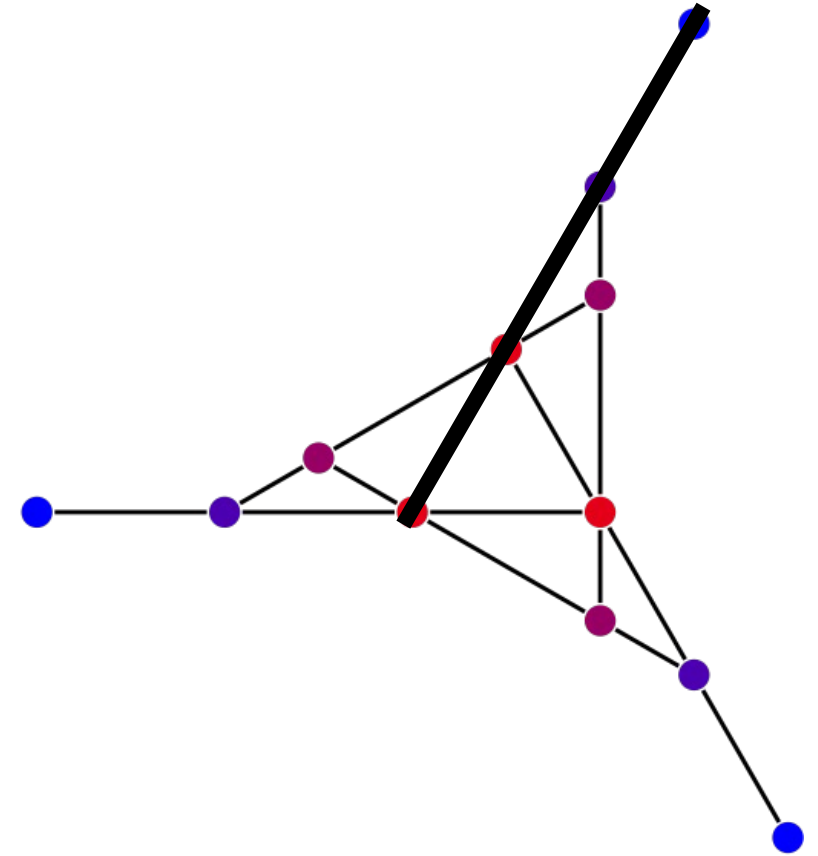


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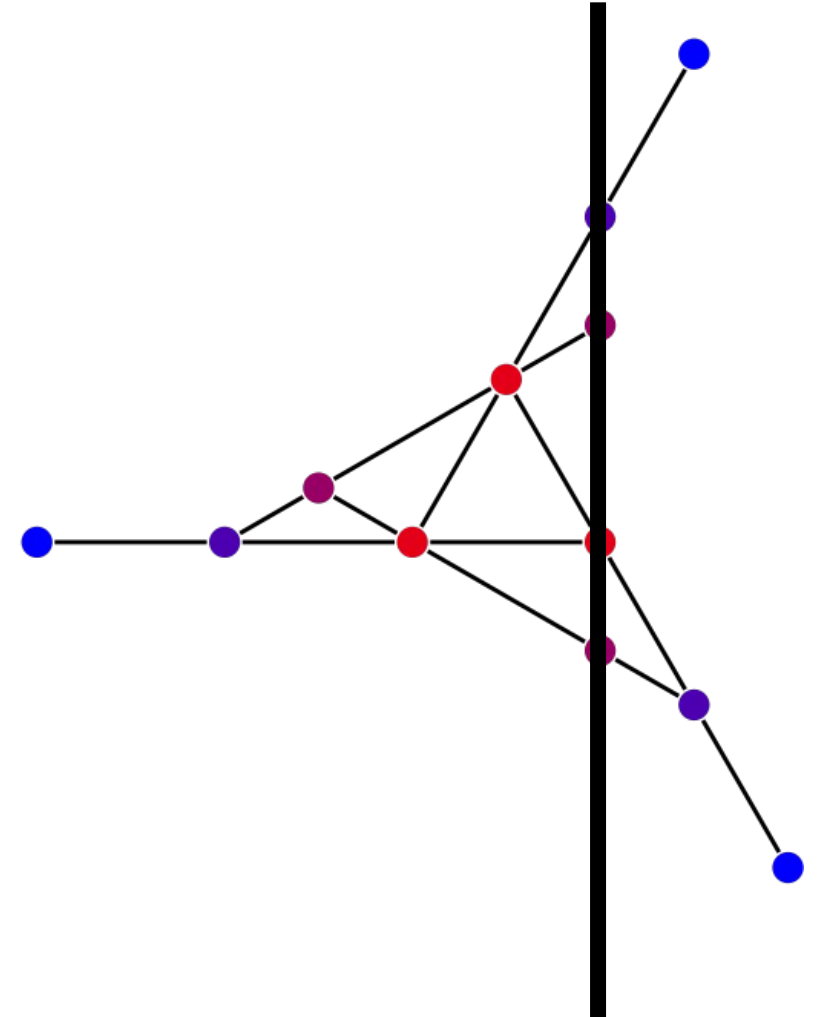


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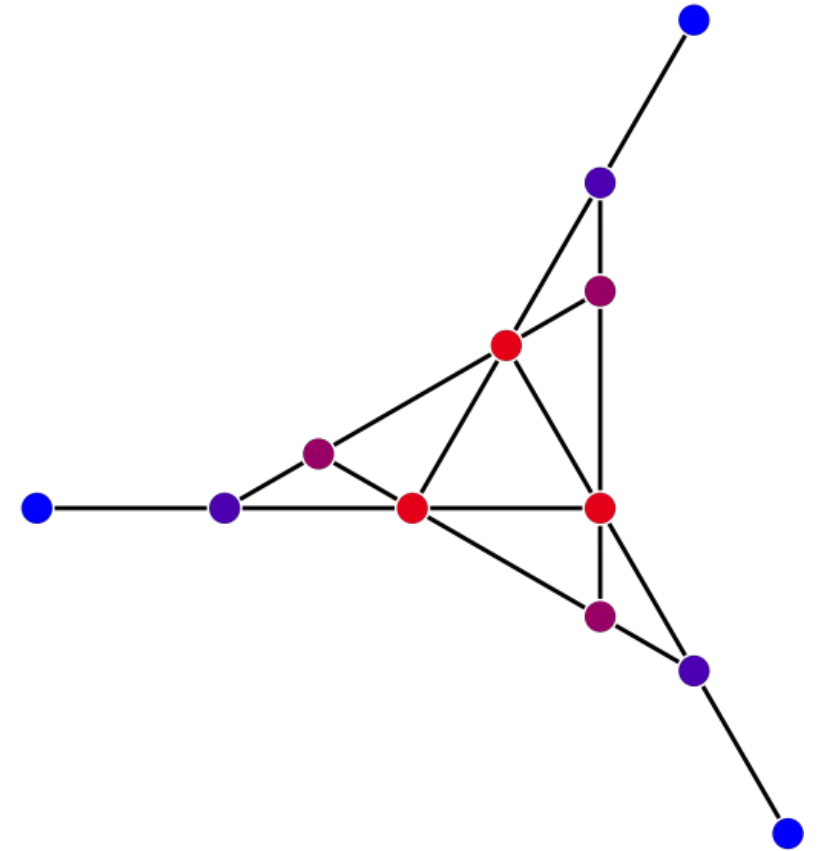
Everywhere Unbalanced Points

Alon's construction, circa 2000

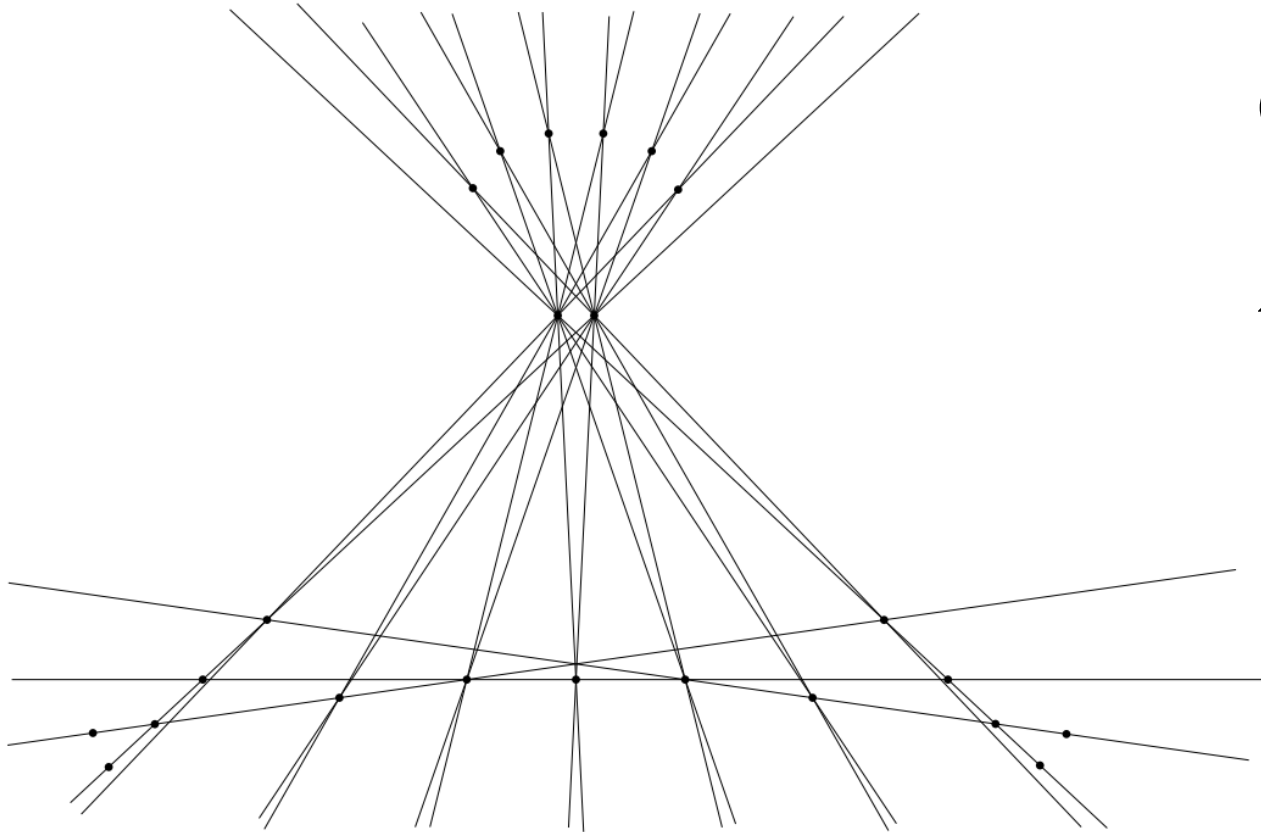
A pointset is k -EU if every line between two of its points has imbalance at least k .

Alon's construction is 2-EU, and we proved it's the smallest.

SAT encoding uses the same axioms, but now *cardinality constraints* to encode the imbalance condition.



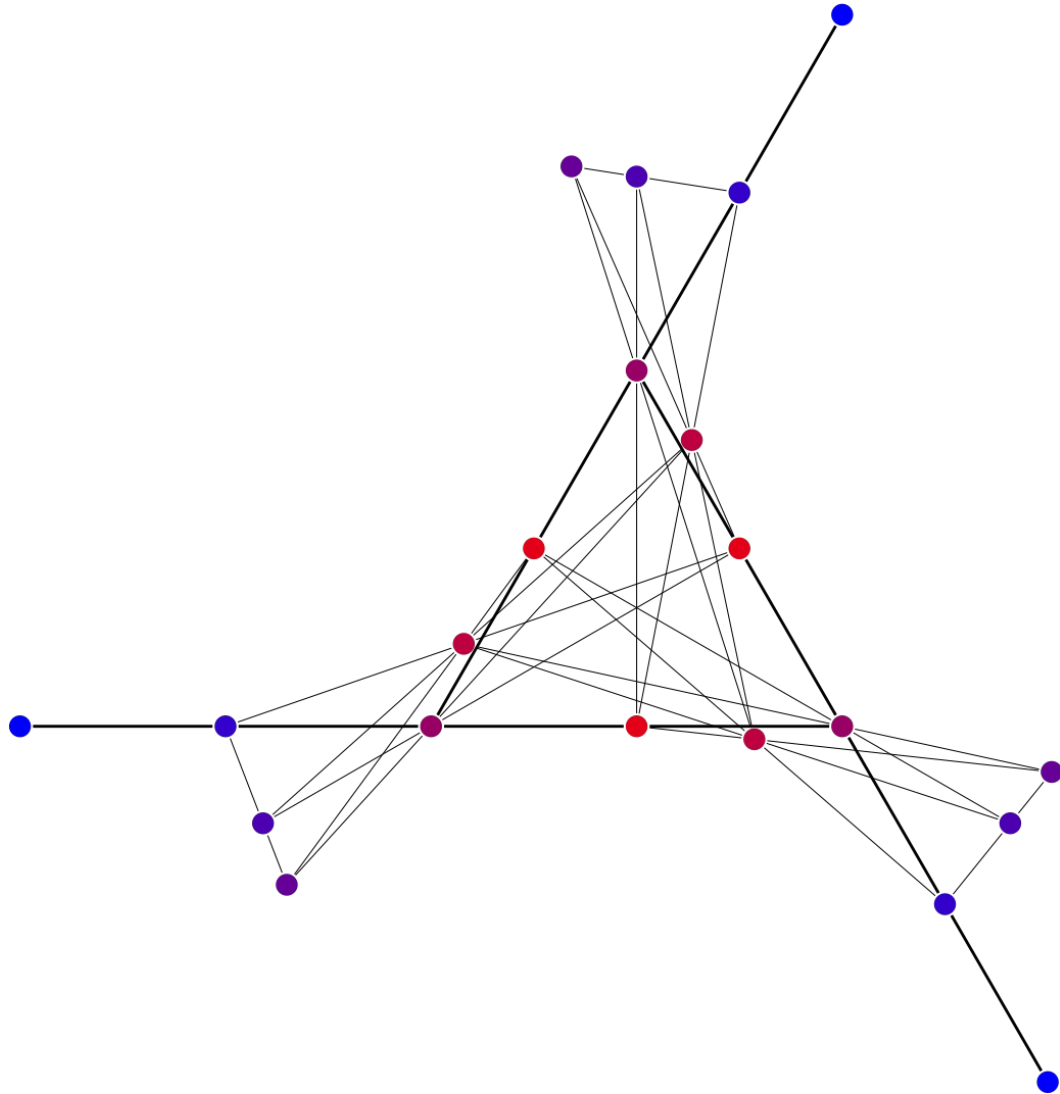
Everywhere Unbalanced Points



Conlon and Lin, 2024

23 points that are 2-EU

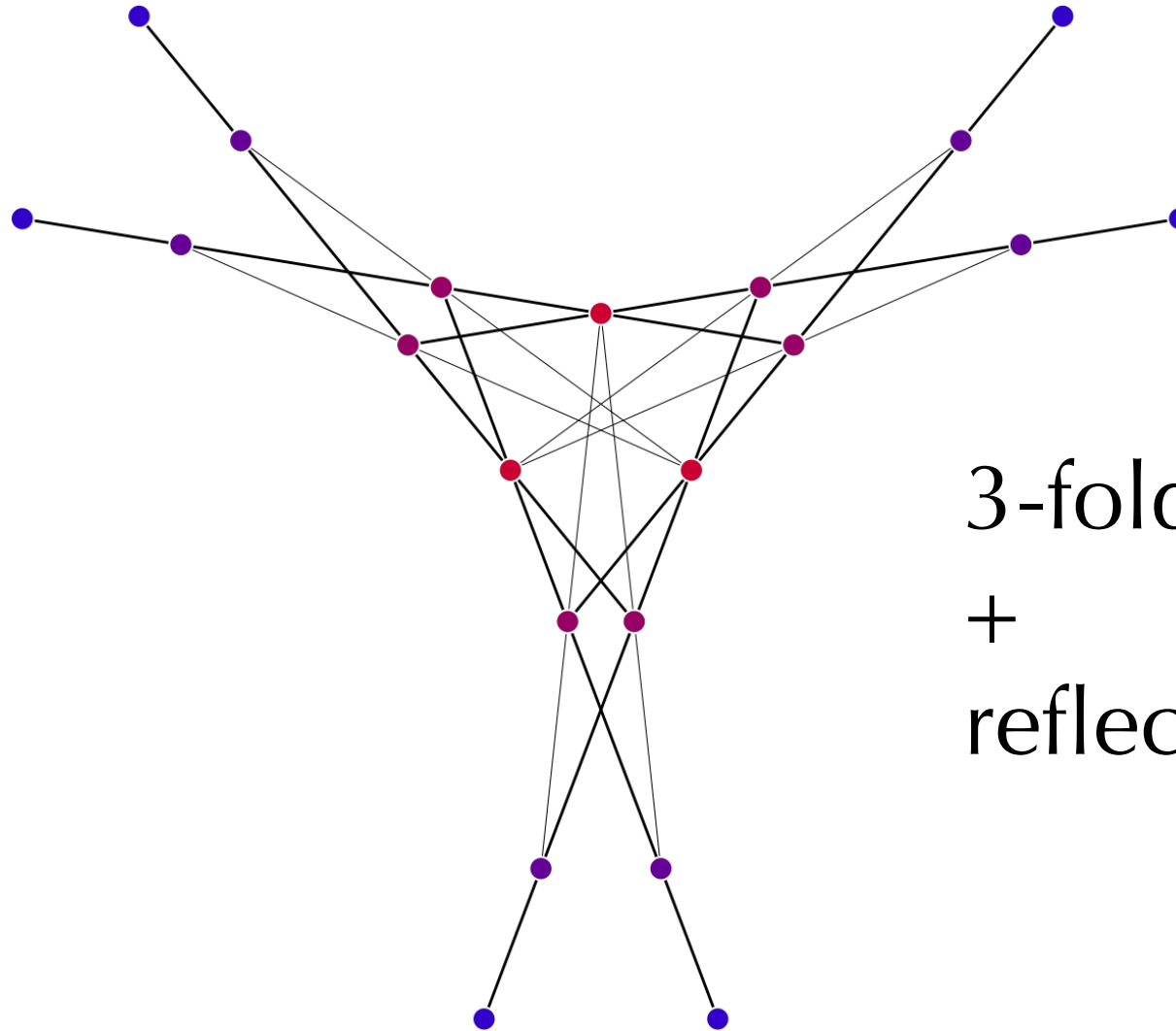
Everywhere Unbalanced Points



We find the minimum odd-size construction, with 21 points, by enforcing 3-fold symmetry.

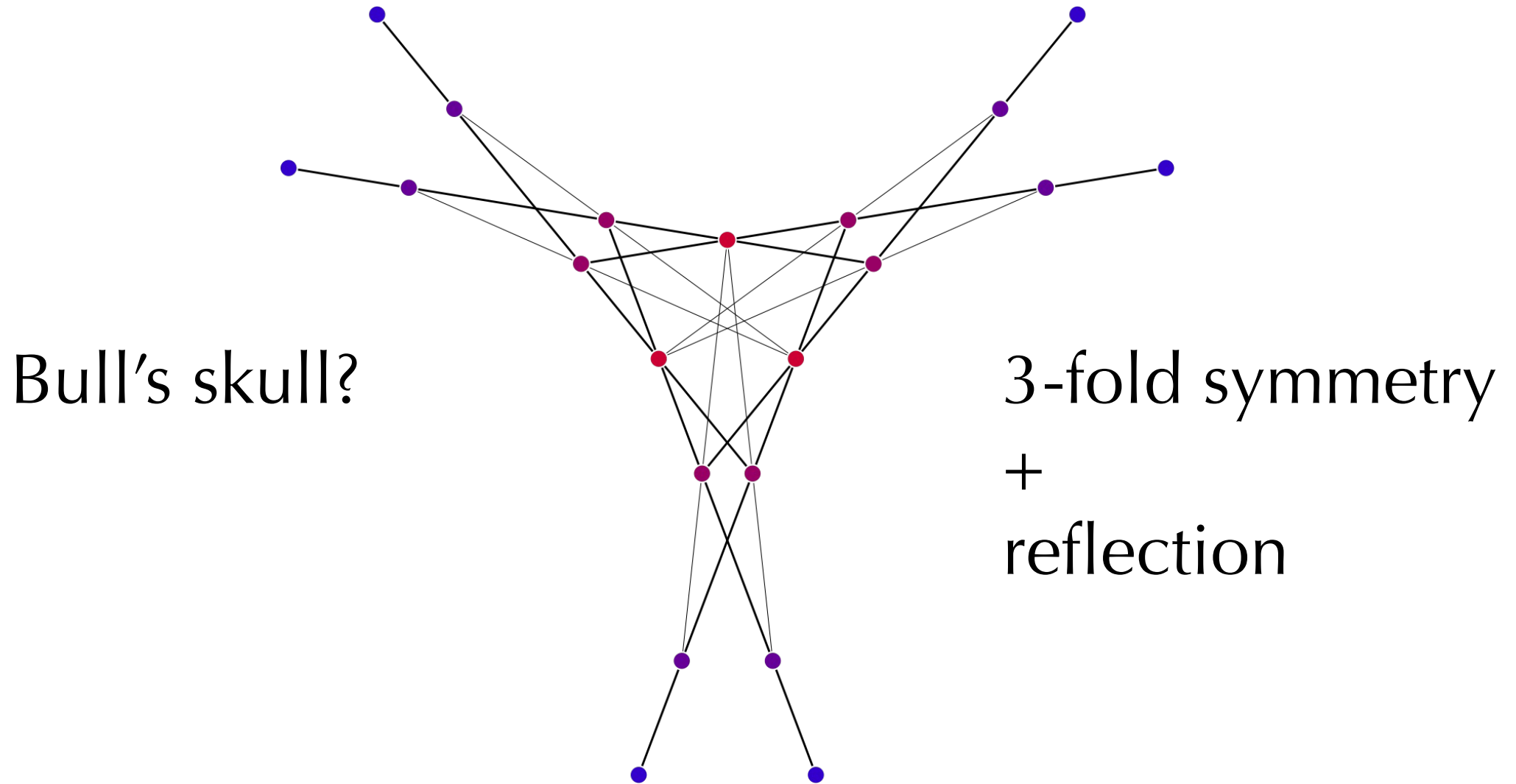
The symmetry dramatically reduces the difficulty of realizability with collinearities!

Everywhere Unbalanced Points

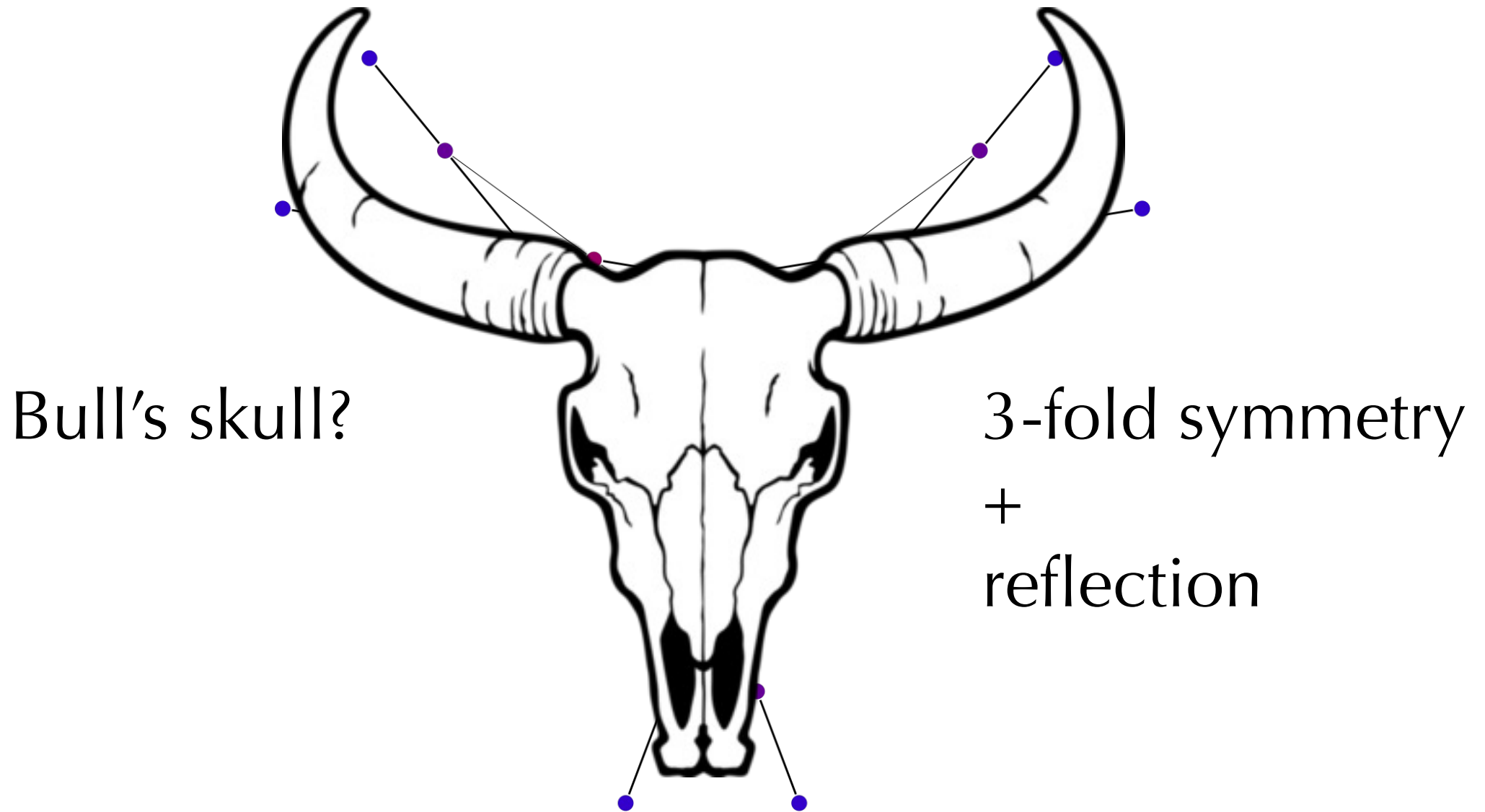


3-fold symmetry
+
reflection

Everywhere Unbalanced Points



Everywhere Unbalanced Points





Automated Symmetric Constructions in Discrete Geometry

Bernardo Subercaseaux, Ethan Mackey, Long Qian, Marijn J. H. Heule
Carnegie Mellon University